Assessing MODIS Vegetation Continuous Fields tree cover product

(collection 6): performance and applicabilityneeds calibrating in tropical

forests and savannas.

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A hstract The Moderate Resolution Imaging Spectroradiometer vegetation continuous
fields (MODIS VCF) Earth observation product is widely used to estimate forest cover
changes, parameterise vegetation and Earth Systemmodels, and as a reference for
validation or calibration where field data are limited. However, while although limited
independent validations of MODIS VCF have shown that MODIS VCF's accuracy
decreases when estimating tree cover in sparsely-vegetated areas such as tropical
35 savannas, no study has yet assessed the impact this may have on the VCF-based tree cover data
used by many in their research. Using tropical forest and savanna inventory data
collected by the TROpical Biomes in Transition (TROBIT) project, we produce a series
of corrections calibrations that take into account (i) the spatial disparity between the in-
situ plot size and the MODIS VCF pixel, and (ii) the trees' spatial distribution within
in-situplots, WoTo identify if a disparity also exists in products trained using VCF, we
used a similar approach to evaluate the finer-scale Landsat Tree Canopy Cover (TCC)
product. For MODIS VCF, we then applied our corrections calibrations to areas
identified as

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40 _____forest or savanna in the International Geosphere-Biosphere Programme (IGBP) land cover mapping All IGBP classes identified as 'savanna' show substantial increases in cover after <u>correction calibration</u>, indicating that the most recent version of MODIS VCF consistently underestimates woody cover in tropical savannas. <u>We also</u> found that these biases are propagated in the finer-scale Landsat TCC. We estimate that MODIS VCF could be underestimating tropical tree cover by as much as 29-%. Models that use MODIS VCF as their benchmark could therefore be underestimating forest-savanna dynamics. Because of

indicator of where the product is potentially more or less reliable. Until more in-situ data are available to produce more accurate corrections were commend caution when using uncalibrated MODIS VCF in tropical savannas.

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1 Introduction

They are used to estimate forest cover change, biomass, and carbon stocks (Bastin et al., 2019; Giriraj et al., 2017; Saatchi et al., 2011; Song et al., 2014); help identify key areas for conservation efforts (Miles et al., 2006); and are used as a basis for climatic and vegetation modelling and model evaluation (Brovkin et al., 2013; Burton et al., 2019; Kelley et al., 2013). All this research, in turn, plays a vital role in informing local, regional, and global environmental policies (Harris et al., 2012). As such, an EO product's accuracy is important to consider, as any errors in the initial tree cover estimate can be further compounded in downstreamwork.

Only a handful of EO products provide global maps of percentage tree cover or forest and shrub cover distributions (Bartholomé and Belward, 2005; Bicheron et al., 2008), and fewer still provide information stretching over at least a decade (Friedl et al., 2002; Hansen et al., 2003)-, Sexton et al., 2013, DiMiceli, 2017). Of these, one of the products most widely used in ecological modelling is the Moderate Resolution Imaging Spectroradiometer Vegetation Continuous Fields (MODIS VCF) product (DiMiceli, 2017). MODIS VCF is a yearly product that provides percent tree cover globally at a spatial resolution of 250 m and, The most recent iteration (Collection 6) is available for the years 2000 through to 2020. Its quantitative measure of woody cover is recorded annually and is described as a percentage of ground cover, making it particularly suited for use in evaluating dynamic global models (Lasslop et al., 2018; Rabin et al., 2017), as a proxy for in-situ data that are harder to collect (Kelley et al., 2019), and to help define parameters for calculating global tree restoration potential (Bastinet al, 2019), Collection 6 is help define parameters for calculating global tree restoration potential (Bastinet al, 2019), Collection 6 is product (Sexton et al., 2013).

As the VCF producthas progressed from Collection 1 to its current Collection 6, several validations using insitu field data or higher-resolution remotely sensed data as a reference measurement have been carried out. These have been few and limited to sites within a biome (Montesano et al., 2009a), a region (Hansen et al., 2005; White et al., 2005), or within a country (Gao et al., 2014; Sexton et al., 2013). The MODIS VCF product evaluated was the most recent collection available at the time (i.e._{Tz} Hansen et al., 2005 and White et al., 2005 for Collection 3; Montesano et al., 2009a for Collection 4; and Gao et al., 2015 and Sexton et al., 2013 for

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Collection 5). To our knowledge, no such independent validation experiment has yet been conducted on Collection 6, which produces tree cover estimates in the same manner as Collection 5 but with improvements made to the upstreaminputs to enhance its accuracy (DiMiceli, 2017). Likewise, validation of the finer-scale TCC product has been limited to its penultimate version and to the taiga-tundra circumpolar region (Montesano et al., 2016).

The validations found that MODIS VCF may be less suitable for estimating tree cover in sparsely-vegetated areas. Huang & Siegert (2006) noted that MODIS VCF classified large areas of land as 'bare' where their land cover classification systemidentified it as sparsely-vegetated. Montesano et al. (2009) found that MODIS VCF data (Collection 4) overestimated cover in areas of low tree cover in taiga-tundra transition zones. Sexton et al. (2013) found that the Collection 5 product overestimated cover in areas of low cover (below 20%) and underestimated in areas of higher tree cover, while Gao et al. (2015) found that MODIS VCF can only partially discriminate between tropical forest and non--forest, struggling in areas that have greater heterogeneity. Similarly to MODIS VCF (Montesano et al., 2009), Montesano et al., (2016) revealed an overestimation of the taigatundra low tree covers in the finer-scale Landsat TCC, suggesting that using VCF as training has propagated these overestimations into the higher resolution product. What is clear from the history of these validation and comparison experiments is that MODIS VCF has accuracy issues in areas with low woody vegetation cover, which has implications when its treecover estimates are treated as accurately representative of real-world conditions. Failure to account for the product'sVCF's difficulty in estimating low woody coverscan therefore, lead to niscalibrated models and estimations that do not reflect real-world conditions. This, in turn, has knock-on effects on environmental policy-making, conservation efforts, and future ecological research, especially in areas with vegetation cover types that are most prone to error.

Tropical savannas have woody covers that fall within the range particularly affected by the reported MODIS VCF errors. A large proportion of these savannas can be found in tropical developing

Dixon, 2012), and are predicted to behome to half of the world's population by 2050 (State of the Tropics, 2020). Tropical savannas are therefore highly vulnerable to anthropogenic change. In the face of a growing population, land fragmentation, and changing climate, a savanna's ability to maintain robustecosystem functions is directly linked to the amount of woody cover present (Sank aran et al., 2006). As a result, the ability to accurately monitor the state, dynamics, and woody cover trends of tropical savannas is a vital part of understanding how and why savannas are changing in the tropics (Harris et al., 2012; Miles et al., 2006), while also improving modelled climate projections and vegetation dynamics for this complex biome.

In this study, we validate the accuracy feature MODS VCF Glection 6 intopical savannas and for the product to comesponding field data. <u>Similarly, we evaluate Landsat TCC (version 4) to explore if</u> when VCF is used as training, VCF biases are propagated. We then, for MODIS VCF, characterise the observed bias in woody covers across both savanna for stecosystems and applyour corrections calibration across the tropics to highlight the regions nost likely to be affected by these inaccuracies in We firsh by discussing the MODIS VCF productions the uncovered biases may have on tropical vegetation and terrestrial biogeochemical modelling.

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2 Methods

2.1 EO Products and Field data

We used the MODIS VCF Collection 6 product (<u>250m</u>spatial resolution of 250 m, DiMiceli, 2017) with tree cover values averaged across the years 2006 through to 2009 to reflect the range of the field data collection period. <u>MODIS</u> <u>VCF was downloaded using the modis r package (Hijmans, 2017) in R3.5.2 (R Core Team, 2018). We used the 2005 and 201030mLandsat TCC version 4 product (https://lcluc.umd.edu/metadata/global-30m-landsat-tree-canopy-version-4), and worked with the 2005 and 2010 average values. The product was downloaded manually from https://e4fil01.cr.usgs.gov/MEASURES/GFCC30TC.003/.</u>

The in-situ field data were sourced from the 'TROpical Biomes In <u>Transition'</u>In <u>Transition</u>' project (TROBIT) (<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>(<u>www.googl.oods.ac.uk/TROBIT</u>). The data we used include the corner locations and the Canopy Area Index (CAI) values for 17 forest and 31 savanna sites distributed across Australia, Brazil, Bolivia, Cameroon, and Ghana (Fig. A1 and Table A1, Fig. 2 in Torello-Raventos et al., 2013). The TROBIT field campaigns were carried out over a 3--year period, from 2006 to 2009, and the field plots used in this study are 1 hectare in size except for BFI-01 (0.5 ha), BFI-02 (0.5 ha), BFI-03 (0.5 ha), CTC-01 (0.93 ha), and VCR-01 (0.6 ha).

All the sites fall within the tropics, that is, within 23.5 degrees north and south of the equator, and were selected in regions where savannas and forests were in close proxinity and exist within ecotones or 'zones of tension'. As such, the sites sampled show a large variation in physiognomy and <u>edaphic and climatic conditions (Table S1, Veenendaalet al., 2015).</u>

135 edaphic and climatic conditions (Table S1, Veenendaal et al., 2015).

The classification of the TROBIT plotesites as either 'forest' or 'savanna' is based on the parameters described in Torello-Raventos et al. (2013) and Veenendaal et al. (2015). A 'savanna' is a natural land cover that is not a forest, bare ground, or a body of water. 'Forest' is defined as woody vegetation with an average tree height of or exceeding 6 mand a canopy area index (CAI) value of at least 0.3 for 'open forests' and 0.7 for 'forests' forests', In addition, floristic differences (i.e., dominance of 'savanna' species) are used to differentiate forests fromtallergrowing savannas that have similar CAIs and tree heights (see Fig. 9, Torello-Raventos et al., 2013).

There is some ambiguity in how 'savannas' and 'grasslands' are defined. Some modelling-based research treat the two biomes as different (Whitley et al., 2017), while studies based on plant functional traits group them together (Solofondranohatraet al., 2018; White et al., 2000). As there is some concern that MODIS VCF will struggle to pick up woody cover in areas with really sparse vegetation, in this paper we decided to treat 'grasslands' as part of the savanna domain.

2.2 Converting In-Situ Canopy Area Index to MODIS VCF / Lands at TCC percent tree cover

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CAI is defined as the sumof the projected areas of individual tree crowns divided by the ground area. In the TROBIT project (Torello-Raventos et al. (2013) and Veenendal_et al. (2015)), plot-wide CAI is made up of the sumof the upper-stratum mid-stratum and subordinate-stratum rown areas. determined by the tree's dbh (upper-stratum dbh > 10 cm, mid-stratum subordinate-stratum dbh < 2.5 cm, height> 1.5 m). About 50 trees per plot-specific allometric relations between stemdiameter and crown area (supplement B of Torello Raventos et al.-($_{20}2013$)); These were then applied to the whole plot to establish plot-level CAI. For the allometric relationships, tree crowns were treated as circles; and the individual tree projected crown area was determined using the average of crown radii measured along the four cardinal points (i.e., from the centre of the stemto the distance furthest from the stem)

CAI values do not account for within-site tree canopy distribution patterns and the overlap between individual tree canopies. We account for this by converting each CAI value into a probability distribution function incorporating the following two extreme scenarios: $\frac{44}{20}$ enforced overlap, $\frac{100}{20}$ where the location probability of individual canopies increases linearly from 0 to 1 across a site; and $\frac{44}{20}$ unenforced overlap, $\frac{100}{20}$ where individual canopies follow a uniform and omdistribution pattern and canopy overlap is not purposefully introduced (Fig. A-21). We repeated this 1000 times per CAI measurement to determine the probability distribution of expected CAI for each field plot.

Unlike CAI, which is the fraction of ground covered by tree crowns, the percent tree cover value from MODIS VCF (and so Landsat TCC) is defined as "the portion of the skylight orthogonal to the surface which is intercepted by trees" (Hansen et al. 2002). To make MODIS VCF tree cover and Landsat TCC comparable to treecover derived from IROBIT plot CAIs, we divided the MODIS VCF these product values by 0.8 as suggested by Hansen et al. (2002). This is also the standard approach innost modelling studies that use MODIS surface (e.g., Lasslopet al., 2002). Kelkyetal, 2013; Burton et al., 2019). The 0.8 value can be thought of as a gap correction factor (GCF) that accounts for within-canopy gaps. Although the GCF has been shown to vary with vegetation type (Lloyd et al., 2008; 0.34 - 0.60) and crown cover (Tang et al., 2019: 0.9670 - 0.796), we opted to use 0.8 as we found that it yielded more conservative results compared to a variable GCF. It also avoided introducing additional parameters into our analysis.

Next, to account for the difference in size between the MODIS VCF pixel (250 mx 250 m) and the smaller field plot size (100 mx 100 m), we calculated the possible percent tree cover an area the size of a TROBIT field plot could have, given the MODIS VCF percent tree cover for a MODIS-sized pixel. This was done for two extreme scenarios: "enforced clunping," where all the tree cover for the given MODIS VCF value is forcibly 'clunped' on one side of the pixel, or "unenforced clunping," where 'clumping is not enforced, and tree cover is distributed randomly within the pixel (Fig. A32). The clumping scenarios introduce possible variations in percent cover due to the area and location mismatch between a TROBIT field plot and a MODIS pixel. A probability distribution was generated for each MODIS VCF pixel by calculating percent tree cover values for 1000 samples (100 mx 100 m) randomly placed within the 250 mx 250 m MODIS VCF pixel.

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For Landsat TCC, where the Landsat TCC pixels (30mx30m) are smaller than the TROBIT field sites, we calculated a TCC percent tree cover to match the TROBIT field site size by summing the percent tree cover within the TCC pixel part found inside the TROBIT field site and then dividing the sumby the TROBIT site area. As TROBIT site orientation was not recorded, we randomized the angle between the TROBIT site and TCC pixel grid for each of the 1000 samples when generating the probability distribution. "Enforced clumping" was performed as per MODIS VCF (Fig 2), with the direction of clumping randomized.



Figure 1. Visual representation of the effects of enforcing overlap within a (100 m x 100 m) TROBIT site with a given Canopy Area Index (CAI) Left: Overlap is not enforced, and individual crowns follow a uniform random distribution. Right: Overlap is enforced by linearly increasing the probability of a canopy being located more on one side of the site (i.e. here the right side of the site) than the other. This results in tree canopies 'overlapping' to a greater extent, which affects how accurately CAI represents actual canopy over.

2.3 Calculating Uncertainty Under Different Overlap-Clumping Scenarios

We thereby compared both MODIS VCF and Landsat TCC with TROBIT under four different scenarios: 1) unenforced overlap and clumping; 2) enforce overlap and unenforced clumping; 3) unenforced overlap and enforced clumping; 4) enforced overlap and clumping. Comparisons were conducted by fitting the following logit function:

 $logit(VCF) = C_0 + \Delta \times log(C^{+}/(1 - C^{+})) - logit(Pixel) = C_0 + \Delta \times logit(Pixel) = C_0 +$

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Where $C_{0,\tau}A_{1,\tau}\tau_{2}c_{0,0}A_{\tau_{1},\tau_{2}}$ are optimised parameters and $VG=\underline{Pixel}$ and *C* are the MODIS VCF / Landsat TCC pixel (post-conversion as described in section 2.2) and TROBIT site probability distributions, respectively. This is similar to a standard linear regression of logit transformed data, accounting for maximum and minimum bounds of 0 - 100 % tree cover, with $\tau_{4,\tau}\tau_{2}\tau_{1,\tau_{2}}$ allowing for a non-symmetric transformation of tree cover. To account for the probability density of each point, we inferred the parameters in Equation 1 using a Total Least Squares Bayesian Inference technique using a Metropolis-Hastings Markov Chain Monte Carlo step. Priors were uninformed but physically bounded (i.e., $A_{1,\tau_{0}}$, $\tau_{2} > 0_{2}A$, τ_{1} , $\tau_{2} > 0$) to assume an increasing relationship between MODIS VCF / Landsat TCC and describing our conditional

distribution (Gelman et al., 2013). Each combination was run over 10 chains, with 1000 warm-up iterations and 10,000 sampling iterations. Optimisation was performed using the rstan2.19.2 (Stan Development Team, 2019) package in R3.5.2 (R Core Team, 2018). Our optimization accounts for potential errors in TROBIT cover, which includes those caused by the allometric construction of the CAI, provided that the errors are unbiased and remain roughly consistent across sites (Gelman et al., 2013). As the TROBIT plots have relatively small total errors as sociated with the allometric relationships (Table B1, Torello–Raventos et al., 2013), systematic errors are unlikely to affect our results.

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Figure 2. Left: Example of the effects of unenforced and enforced clumping in a 250 m x 250 m MODIS VCF pixel with 50 % tree cover. Clumping all the cover (green) to one side of the pixel (left bottom) affects the average canopy cover value of a 100 mx 100 m-sized TROBIT site (black boxes). Right: Example of the effects of unenforced and enforced clumping on 30 m x 30 m Landsat TCC pixels with a mix of tree covers (green) and non-tree cover (brown). White dotted lines are TCC pixel boundaries. Clumping all the cover to one side of the pixel (right bottom) affects the average canopy cover value of a 100 mx 100 m-sized TROBIT site (black boxes).

2.4 Mapping MODIS VCF Uncertainty A cross The Tropics

We evaluated the inpact of the MODIS VCF biases inferred from this correction these regression equations across the tropics by inverting our calculation of MODIS VCF bias (Fig. A4A2) as follows: first, the inverse (i.e., solving for C) of Equation equation 1 was applied to MODIS VCF values after conversion to a 100 mx100 npixel size grid (matching the field site area); then this corrected calibrated value was translated back to the original 250 mx250 mVCF pixel size. As the inverse of Equation 1 has no analytical solution, we found the rounded percent value of C that minimises the absolute difference between the left- and right-hand side of the equation. For computational feasibility, we constructed maps of the topics with conceled calibrated MODIS VCF values (Fig. 2A3) by and only sampling 5 iterations is considered as ' forest' or 'savanna' - in the 500 mMODIS Land Cover Type (MCD12Q1 - collection 6) (Sulla-Menashe and Fried1. 2018).

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We<u>then</u> used the <u>500mMODIS Lond CoverType</u> (MD12Q1-collection6) product to identify the acas of fixest and 'savarna' ans http is inter MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere-Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen Needleleaf Forests, Evergreen Broadleaf Forests, Deciduous Needleleaf Forests, Deciduous Broadleaf Forests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

Forance detailed land-coverspecific evaluation, we recompledented the control detailed allocated 250n MODIS VE pixels of a pixel values for each corresponding 500 mg/id and combined it with the MCD12Q1 pixel to construct land coverspecific MODIS VE free cover frequency distributions (Fig. A5A4). Our tree cover correction calibration by cover type (Fig. 3) for the four clumping/overlap regression combinations was then calculated by multiplying each cover type MODIS VCF frequency distribution (Fig. A5A4) with curves representing the median, 5 %, and 95 % confidence lines of the correction calibration equation ensembles.

3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforcedunenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and may instead be overestimating tree cover (50 % confidence, dashed line. Fig.

significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pixeite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered is used as where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; black for forest and savanna combined). The bol, and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent the uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18-19-21% (without enforced clumping) or $9-10\cdot11-12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overestimates tree cover where tree cover exceeds 7884% (at the 95 % confidence)

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interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds $\frac{9078}{28}$ % (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered to one side within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestination of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestination (at 95% confidence) is significant and

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consistent for covers below 75 - 80 % (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of <u>post-calibration</u> change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

Fue 2Dth/ordnowoms <u>(fr)</u> (expinently of iddDDS/CF4n(ph) the part of of a lange in the experimental of the hyperbolic temperature of the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the

(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the

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forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited. Indeed, it is allocuted bits fifty as 5 and 42 are to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95 % confidence interval)(Fig. 6). The nost substantial and significant underestimation is in the classes 'woody savannas' and 'savannas', The underestimation is the largest in woody savannas, except when clunping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clunping is the largest (i.e, at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26 - 67 %, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60 %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al., 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.</u>

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is

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insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, close shrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is minily occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A 3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm. as well as trees with a height exceeding 1.5 m when dbh <2.5 cm.

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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Figure 36. Percent change in tree cover alter the optical and the optical matrix (betwice: merificed change or overlap (blue); no enforced changing and enforced overlap (red); enforced changing and overlap (pink) in the 'forest' super category and the 5 savanna classes. Palest tone indicates positive change, and tone indicates net change. Err or bars denote the 5-95% confidence indicates net with a val; if the error bar extends past the x-axis, the post-correction calibration change is not considered significant.

difference between TROBIT and MODIS-VCF we would have expected an increasing underestimation in the lower height ranges. Instead₇ we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and 'closed shrublands' classes of the MCD12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or_a some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to and improve its useability.

within-field-site and field-site-pixel variation are accounted for during validation in calibration. We also fund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net)</u>.

					Conon			
		Latitud	Longitu	MODI	Canop	Average		
Site		e Latitud	deLongit	Tree	v Area	Stratum	Cover.	/
Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Fore st	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	<u>1.17</u>	1.33	0.52	12.53	Savanna	Savanna woodland
BBL02	Burlina Faso	12 73	-1.16	1.5	0.99	13.6	Savanna	Savanna woodland
pbi-02	Burnita Aaso	12.13	1.10	1.5	V .99	13.0	gavanna	Sa valilla woodlalid
BDA-01	Burkina Faso	10.94	<u></u> 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodloo
BDA-02	Burkina, Faso	10.94	<u>-</u> 3.15	4.5	0.18	14.47	Savanna	Shrub-rich sa vanna woodl
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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1 Introduction

	1 Introduction	
	Tree cover values derived from Earth observation (EO) data form a fundamental part of ecological research. They are use	Formatted
55	to estimate forest cover change, biomass, and carbon stocks (Bastin et al., 2019; Giriraj et al., 2017; Saatchi et al., 2011;	
	Song et al., 2014); help identify key areas for conservation efforts (Miles et al., 2006); and are used as a basis for climpic	
	and vegetation modelling and model evaluation (Brovkin et al., 2013; Burton et al., 2019; Kelley et al., 2013). All this	
	research, in turn, plays a vital role in informing local, regional, and global environmental policies (Harris et al., 2017). As	
60	such, an EO product's accuracy is important to consider, as any errors in the initial tree cover estimate can be further	
	compounded in downstreamwork.	
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	Only a handful of EO products provide global maps of percentage tree cover or forest and shrub cover distributions	Formatted: Normal, Space Before: 0 pt, Line spacing:
65	(Bartholomé and Belward, 2005; Bicheron et al., 2008), and fewer still provide information stretching over at least a	1.5 lines
	decade (Friedlet al., 2002: Hansen et al., 2003)-, Sexton et al., 2013, DiMiceli, 2017), Ofthese, one of the products nost	Formatted: Normal, Indent: Left: 0", Right: 0", Line spacing: 15 lines
	widely used in ecological modelling is the Moderate Resolution Imaging Spectroradiometer Vegetation Continuous Field	Formatted
	(MODIS VCF) product (DiMiceli, 2017). MODIS VCF is a yearly product that provides percent tree cover globally at a	- omattea
70	spatial resolution of 250 m and. The most recent iteration (Collection 6) is available for the years 2000 through to 2021.	
	Its quantitative measure of woody cover is recorded annually and is described as a percentage of ground cover, making it	
	particularly suited for use in evaluating dynamic global models (Lasslop et al., 2018; Rabin et al., 2017), as a proxy ter in	-
75	situ data that are harder to collect (Kelley et al., 2019), and to help define parameters for calculating global tree restoration	1
75	potential (Bastinet al., 2019), Collection 6 is the most recent iteration of the product. MODIS VCF is also used	
	to train alternative products, such as the newer finer-scale Landsat Tree Canopy Cover (TCC) product (Sexton et al., 2013)	
	L	Formatted: Font: Times New Roman, 10 pt,
80	As the VCF product has progressed from Collection 1 to its current Collection 6, several validations using in-situ field da	Strikethrough
00	or higher-resolution remotely sensed data as a reference measurement have been carried out. These have been few and	Formatted: Normal, Line spacing: 1.5 lines
	limited to sites within a biome (Montes ano et al., 2009a), a region (Hansen et al., 2005; White et al., 2005), or within a	Formatted: Font: Times New Roman, 10 pt
	country (Gao et al., 2014; Sexton et al., 2013). The MODIS VCF product evaluated was the most recent collection	Formatted: Normal, Indent: Left: 0", Right: 0", Line spacing: 15 lines
85	available at the time (i.e., Hansen et al., 2005 and White et al., 2005 for Collection 3; Montesano et al., 2009a for	Formatted
	Collection 4: and Gao et al., 2015 and Sexton et al., 2013 for Collection 5). To our knowledge, no such independent	
	validation experiment has yet been conducted on Collection 6, which produces tree cover estimates in the same manner as	
	Collection 5 but with improvements made to the upstreaminputs to enhance its accuracy (DiMiceli, 2017). Likewise.	
90	validation of the finer-scale TCC product has been limited to its penultimate version and to the taiga-tundra circump flar	
	region (Montesanoet al., 2016)	
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	The validations found that MODIS VCF may be less suitable for estimating tree cover in sparsely-vegetated areas: Huan	Formatted: Normal, Line spacing: 1.5 lines
95	& Siegert (2006) noted that MODIS VCF classified large areas of land as 'bare' where their land cover classification	Formatted: Font: Times New Roman, 10 pt
	systemidentified it as sparsely-vegetated. Montesano et al. (2009) found that MODIS VCF data (Collection 4)	Formatted
	overestimated cover in areas of low tree cover in taiga-tundra transition zones. Sexton et al. (2013) found that the	Formatted: Normal, Indent: Left: 0", Right: 0", Line
	Collection 5 product overestimated cover in areas of low cover (below 20%) and underestimated in areas of higher tree	spacing: 1.5 lines

cover, while Gao et al. (2015) found that MODIS VCF can only partially discriminate between tropical forest and pon-

forest, struggling in areas that have greater heterogeneity. <u>Similarly to MODIS VCF (Montesanoet al., 2009)</u>. <u>Montesano et al., (2016) revealed an overestimation of the taiga-tundra low tree covers in the finer-scale</u> <u>Landsat TCC, suggesting that using VCF as training has propagated these overestimations into the higher</u> <u>resolution product</u>. What is clear from the history of these validation and comparison experiments is that MODIS VCF has accuracy issues in areas with low woody vegetation cover, which has implications when its tree cover estimates are treated as accurately representative of real-world conditions. Failure to <u>accurately</u> account for <u>the product'SVCF's</u> difficulty in estimating low woody covers can, therefore, lead to miscalibrated models and estimations that do not reflect real-world conditions. This, in turn, has knock-on effects on environmental policy – making, conservation efforts, and future ecological research, especially in areas with vegetation cover types that are most prone to error.

Tropical savannas have woody covers that fall within the range particularly affected by the reported MODIS VCF errors. A large proportion of these savannas can be found in tropical developing Dixon, 2012), and are predicted to be home to half of the world's population by 2050 (State of the Tropics, 2020). Tropical savannas are therefore highly vulnerable to anthropogenic change. In the face of a growing population, land fragmentation, and changing climate, a savanna's ability to mintain robustecosystem functions is directly linked to the amount of woody cover present (Sankaran et al., 2006). As a result, the ability to accurately monitor the state, dynamics, and woody cover trends of tropical savannas is a vital part of understanding how and why savannas are changing in the tropics (Harris et al., 2012; Miles et al., 2006), while also improving modelled climate projections and vegetation dynamics for this complex biome.

In this study, we validate the course of the product to consequence MODS VCF collection 6 intropical savanues and for the product to consequence of the product of the prod

2 Methods

2.1 EO Products and Field data

We used the MODIS VCF Collection 6 product (<u>250m</u>spatial resolution of 250 m, DiMiceli, 2017) with tree cover values averaged across the years 2006 through to 2009 to reflect the range of the field data collection period. <u>MODIS</u> <u>VCF</u> was downloaded using the modis rpackage (Hijmans. 2017) in R3.5.2 (R Core Team. 2018). We used the 2005 and 2010 30mLandsat TCC version 4 product (https://lcluc.umd.edu/metadata/global-30m-landsat-tree-canopy-version-4), and worked with the 2005 and 2010 average values. The product was downloaded manually from https://e4ft101.cr.usgs.gov/MEASURES/GFCC30TC.003/,

The in-situ field data were sourced from the 'TROpical Biomes In <u>Transition'In Transition</u> project (TROBIT)

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(Lopez-Gonzalez et al., 2011; Lopez-Gonzalez et al., 2009). The data we used include the corner locations and the Canopy Area Index (CAI) values for 17 forest and 31 savanna sites distributed across Australia, Brazil, Bolivia, Cameroon, and Ghana (Fig. A1 and Table A1, Fig. 2 in Torello-Raventos et al., 2013). The TROBIT field campaigns were carried out over a 3-year period, from 2006 to 2009, and the field plots used in this study are 1 hectare in size except for BFI-01 (0.5 ha), BFI-02 (0.5 ha), BFI-03 (0.5 ha), CTC-01 (0.93 ha), and VCR-01 (0.6 ha).

All the sites fall within the tropics, that is, within 23.5 degrees north and south of the equator, and were selected in regions where savannas and forests were in close proximity and $\frac{1}{2}$ within ecotones or 'zones of tonsion.'tension'. As such, the sites sampled show a large variation in physiognomy and <u>edaphic and climatic conditions (Table S1</u>, <u>Veenendaalet al., 2015)</u>.

135 edaphic and climatic conditions (Table S1, Veenendaal et al., 2015).

The classification of the TROBIT plotesites as either 'forest' or 'savanna' is based on the parameters described in Torello-Raventos et al. (2013) and Veenendaal et al. (2015). A 'savanna' is a natural land cover that is not a forest, bare ground, or a body of water. 'Forest' is defined as woody vegetation with an average tree height of or exceeding 6 mand a canopy area index (CAI) value of at least 0.3 for 'open forests' and 0.7 for 'forests' forests', In addition, floristic differences (i.e., dominance of 'savanna' species) are used to differentiate forests fromtaller-growing savannas that have similar CAIs and tree heights (see Fig. 9, Torello-Raventos et al., 2013).

There is some ambiguity in how 'savannas' and 'grasslands' are defined. Some modelling-based research treat the two biomes as different (Whitley et al., 2017), while studies based on plant functional traits group them together (Solofondranohatraet al., 2018; White et al., 2000). As there is some concern that MODIS VCF will struggle to pick up woody cover in areas with really sparse vegetation, in this paper we decided to treat 'grasslands' as part of the savanna domain.

2.2 Converting In-Situ Canopy Area Index to MODIS VCF / Lands at TCC percent tree cover

CAI is defined as the sumof the projected areas of individual tree crowns divided by the ground area. In the TROBIT project (Torello-Raventos et al. (2013) and Veenendal et al. (2015)), plot-wide CAI is made up of the sumof the upper-stratum mid-stratum and subordinate-stratum rown areas. determined by the tree's dbh (upper-stratum dbh > 10 cm, mid-stratum subordinate-stratum dbh < 2.5 cm, height> 1.5 m). About 50 trees per

plot-specific allometric relations between stemdiameter and crown area (supplement B of Torello Raventos et al- $(\underline{x}, 2013)$). These were then applied to the whole plot to establish plot-level CAI. For the allometric relationships, tree crowns were treated as circles, and the individual tree projected crown area was determined using the average of crown radii measured along the four cardinal points (i.e., from the centre of the stemto the distance furthest from the stem).

CAI values do not account for within-site tree canopy distribution patterns and the overlap between individual tree canopies. We account for this by converting each CAI value into a probability distribution function incorporating the following two extreme scenarios: $\frac{4}{3}$ enforced overlap, $\frac{1}{3}$, where the location probability of

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individual canopies increases linearly from 0 to 1 across a site; and $\frac{1}{2}$ unenforced overlap $\frac{1}{2}$, where individual canopies follow a uniform random distribution pattern and canopy overlap is not purposefully introduced (Fig. A21). We repeated this 1000 times per CAI measurement to determine the probability distribution of expected CAI for each field plot.

Unlike CAI, which is the fraction of ground covered by tree crowns, the percent tree cover value from MODIS VCF (and so Landsat TCC) is defined as "the portion of the skylight orthogonal to the surface which is intercepted by trees" (Hansen et al. 2002). To make MODIS VCF these product values by 0.8 as suggested by Hansen et al. 2002). This is also the standard approachinmost modelling studies that use MODIS using VCF (e.g., Lasslopet al., 2020;Kelkyetal, 2013; Burton et al., 2019). The 0.8 value can be thought of as a gap correction factor (GCF) that accounts for within-canopy gaps. Although the GCF has been shown to vary with vegetation type (Lloyd et al., 2008; 0.34 - 0.60) and crown cover (Tang et al., 2019: 0.9670-0.796), we opted to use 0.8 as we found that it yielded more conservative results compared to a variable GCF. It also avoided introducing additional parameters into our analysis.

Next, to account for the difference in size between the MODIS VCF pixel (250 mx 250 m) and the smaller field plot size (100 mx 100 m), we calculated the possible percent tree cover an area the size of a TROBIT field plot could have, given the MODIS VCF percent tree cover for a MODIS-sized pixel. This was done for two extreme scenarios: "enforced clumping," where all the tree cover for the given MODIS VCF value is forcibly 'clumped' on one side of the pixel, or "unenforced clumping," where 'clumping is not enforced₇ and tree cover is distributed randomly within the pixel (Fig. A32). The clumping scenarios introduce possible variations in percent cover due to the area and location mismatch between a TROBIT field plot and a MODIS pixel. A probability distribution was generated for each MODIS VCF pixel by calculating percent tree cover values for 1000 samples (100 mx 100 m) randomly placed within the 250 mx 250 m MODIS VCF pixel.

For Landsat TCC, where the Landsat TCC pixels (30mx30m) are smaller than the TROBIT field sites, we calculated a TCC percent tree cover to match the TROBIT field site size by summing the percent tree cover within the TCC pixel part found inside the TROBIT field site and then dividing the sumby the TROBIT site area. As TROBIT site orientation was not recorded, we randomized the angle between the TROBIT site and TCC pixel grid for each of the 1000 samples when generating the probability distribution. "Enforced clumping" was performed as per MODIS VCF (Fig 2), with the direction of clumping randomized.

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Figure 1. Visual representation of the effects of enforcing overlap within a (100 m x 100 m) TROBIT site with a given <u>Canopy Area Index</u> (CAI) Left: Overlap is not enforced, and individual crowns follow a uniform random distribution. Right: Overlap is enforced by linearly increasing the probability of a canopy being located more on one side of the site (i.e. here the right side of the site) than the other. This results in tree canopies 'overlapping' to a greater extent, which affects how accurately CAI represents actual canopy cover.

2.3 Calculating Uncertainty Under Different Overlap-Clumping Scenarios

We thereby compared both MODIS VCF and Landsat TCC with TROBIT under four different scenarios: 1) unenforced

overlap and clumping; 2) enforce overlap and unenforced clumping; 3) unenforced overlap and enforced clumping; 4) enforced overlap and clumping. Comparisons were conducted by fitting the following logit function:

$\frac{logit(VCF) - C_0 + \Delta \times log(C^{\tau_1}/(1 - C^{\tau_2}))}{logit(Pixel)} = C_0 + \Delta \times log(C^{\tau_1}/(1 - C^{\tau_2}))$

Where $G_{2r}A_{1}\tau_{2r}\tau_{2r}C_{0}$, $\Delta_{1}\tau_{1}$, τ_{2} are optimised parameters and VGPixel and C are the MODIS VCF / Landsat TCC pixel (post-conversion as described in section 2.2) and TROBIT site probability distributions, respectively. This is similar to a standard linear regression of logit transformed data, accounting for maximum and minimum bounds of 0 - 100 % tree cover, with $\tau_{2r}\tau_{2}\tau_{1}$, τ_{2} allowing for a non-symmetric transformation of tree cover. To account for the probability density of each point, we inferred the parameters in Equation 1 using a Total Least Squares Bayesian Inference technique using a Metropolis-Hastings Markov Chain Monte Carlo step. Priors were uninformed but physically bounded (i.e., $T_{2r}, \tau_{2r} > 0$, $\Delta_{1}, \tau_{1}, \tau_{2} > 0$) to assume an increasing relationship between

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MODIS VCF / Landsat TCC and

describing our conditional

distribution (Gelman et al., 2013). Each combination was run over 10 chains, with 1000 warm-up iterations and 10,000 sampling iterations. Optimisation was performed using the rstan2.19.2 (Stan Development Team, 2019) package in R3.5.2 (RCore Team, 2018). Our optimization accounts for potential errors in TROBIT cover, which includes those caused by the allometric construction of the CAI, provided that the errors are unbiased and remain roughly consistent across sites (Gelman et al., 2013). As the TROBIT plots have relatively small total errors associated with the allometric relationships (Table B1, Torello–Raventos et al., 2013), systematic errors are unlikely to affect our results.



Figure 2. Left: Example of the effects of unenforced and enforced clumping in a 250 m x 250 m MODIS VCF pixel with 50 % tree cover. Clumping all the cover (green) to one side of the pixel (left bottom) affects the average canopy cover value of a 100 m x 100 m-sized TROBIT site (black boxes). Right: Example of the effects of unenforced and enforced clumping on 30 m x 30 m Landsat TCC pixels with a mix of tree covers (green) and non-tree cover (brown). White dotted lines are TCC pixel boundaries. Clumping all the cover to one side of the pixel (right bottom) affects the average canopy cover value of a 100 m x 100 m-sized TROBIT site (black boxes).

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2.4 Mapping MODIS VCF Uncertainty Across The Tropics

We evaluated the inpact of the MODIS VCF biases inferred from this correction these regression equations across the tropics by inverting our calculation of MODIS VCF bias (Fig. A4A2) as follows: first, the inverse (i.e., solving for C) of Equation equation 1 was applied to MODIS VCF values after conversion to a 100 mx 100 mpixel size grid (matching the field site area); then this corrected calibrated value was translated back to the original 250 mx 250 mVCF pixel size. As the inverse of Equation 1 has no analytical solution, we found the rounded percent value of C that minimises the absolute difference between the left- and right-hand side of the equation. For computational feasibility, we constructed mpsofile tropics with corrected calibrated MODIS VCF values (Fig. 2A3) by and only sampling 5 leadors the two enclosed calibrated as ' forest' or 'savanna'-in the 500 mMODIS Land Cover Type (MCD 12Q1 - collection 6) (Sulla-Menashe and Friedl, 2018).

Weten used the 500m MODIS Land Cover Type (MD12Q1-collection6) product to identify the acas of fixes' and 'savama' austhet pisithe MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere–Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen Needleleaf Forests, Evergreen Broadleaf Forests, Deciduous Needleleaf Forests, Deciduous Broadleaf Forests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

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3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforced unenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and may instead be overestimating tree cover (50 % confidence, dashed line, Fig.

significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pixeite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered is used as where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; black for forest and savanna combined). The bol, and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent the uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18-19-21% (without enforced clumping) or $9-10\cdot11-12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF <u>overoclimateQunderestimates</u> tree cover where tree cover exceeds 7884% (at the 95 % confidence)

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interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds $\frac{9078}{28}$ % (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered to one side within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by culmping; the horizontal error bars represent the uncertainty introduced by overlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and

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consistent for covers below 75 - 80 % (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of <u>post-calibration</u> change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

Fue 2Dth/ordnowoms <u>(fr)</u> (expinently of iddDDS/CF4n(ph) the part of of a lange in the experimental of the hyperbolic temperature of the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the

(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the

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forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited. Indeed, it is allocuted bits fifty as 5 and 42 are to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95 % confidence interval)(Fig. 6). The nost substantial and significant underestimation is in the classes 'woody savannas' and 'savannas', The underestimation is the largest in woody savannas, except when clunping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clunping is the largest (i.e, at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26 - 67 %, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60 %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al., 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.</u>

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is

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insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, close shrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is minily occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A 3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm. as well as trees with a height exceeding 1.5 m when dbh <2.5 cm.

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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Figure 36. Percent change in tree cover alter the optical and the optical matrix (betwice: merificed change or overlap (blue); no enforced changing and enforced overlap (red); enforced changing and overlap (pink) in the 'forest' super category and the 5 savanna classes. Palest tone indicates positive change, and tone indicates net change. Err or bars denote the 5-95% confidence indicates net with a val; if the error bar extends past the x-axis, the post-correction calibration change is not considered significant.

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower height ranges. Instead₇ we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and

' closed shrublands' classes of the MCD12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or_a some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to the future use of this product to the prove its useability.

within-field-site and field-site-pixel variation are accounted for during validation incalibration. We asofund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net)</u>.

					Canon			
		Latitud	Longitu	MODI S VCF	WCanon	Average		
Site		e Latitud	deLongit	Tree	v Area	Stratum	Cover.	
Name	Country 2	<u>e</u> ,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	<u>∗</u> 1.17	1.33	0.52	12.53	Savanna	Savanna woodland
BBI-02	Burkina Faso	12.73	- 1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	- 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl
BDA-02	Burkina Faso	10.94	<u>-</u> 3.15	4.5	0.18	14.47	Savanna	Shrub-rich savanna woodl
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall sa vanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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Tropical savannas have woody covers that fall within the range particularly affected by the poted MODS NCF errors A large proportion of these savannas can be found in tropical developing	Formatted: Normal, Space Before: 0 pt, Line spacing: 1.5 lines
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countries (Boval and Dixon, 2012), and are predicted to be home to half of the world's population by 2050 (State of the Tropics, 2020). Tropical savannas are therefore highly vulnerable to anthropogenic change. In the face of a growing population, land fragmentation, and changing climate, a savanna's ability to maintain robust ecosystemfunctions is directly linked to the amount of woody cover present (Sankaran et al., 2006). As a result, the ability to accurately monitor the state, dynamics, and woody cover trends of tropical savannas is a vital part of understanding how and why savannas are changing in the tropics (Harris et al., 2012; Miles et al., 2006), while also improving modelled climate projections and vegetation dynamics for this complex biome.

In this study, we validate the accuracy feeduce MODS VCF Ollection 6 intropical savanues and for the product to comesponding field data. <u>Similarly, we evaluate Landsat TCC (version 4) to explore if</u> when VCF is used as training, VCF biases are propagated. We then, for MODIS VCF, characterise the observed bias in woody covers across both savanua for stecosystems and applyour corrections calibration across the tropics to highlight the regions nost likely to be affected by these inaccuracies in We finish by discussing the MODS VCF productions the uncovered biases may have on tropical vegetation and terrestrial biogeochemical modelling.

2 Methods

2.1 EO Products and Field data

We used the MODIS VCF Collection 6 product (<u>250m</u>spatial resolution of 250 m, DiMiceli, 2017) with tree cover values averaged across the years 2006 through to 2009 to reflect the range of the field data collection period. <u>MODIS</u> <u>VCF</u> was downloaded using the modis rpackage (Hijmans, 2017) in R3.5.2 (R Core Team, 2018). We used the 2005 and 201030mLandsat TCC version 4 product (https://lcluc.umd.edu/metadata/global-30m-landsat-tree-canopy-version-4), and worked with the 2005 and 2010 average values. The product was downloaded manually from https://e4ft101.cr.usgs.gov/MEASURES/GFCC30TC.003/,

The in-situ field data were sourced from the 'TROpical Biomes In <u>Transition'</u> project (TROBIT) (<u>www.geogl.eeds.ac.uk/TROBIT (wwwgogleds.ac.uk/TROBIT</u> Toello-Rwentosetal, 2013) and accessed via the Forestplots net database (Lopez-Gonzalez et al., 2011; Lopez-Gonzalez et al., 2009). The data we used include the corner locations and the Canopy Area Index (CAI) values for 17 forest and 31 savanna sites distributed across Australia, Brazil, Bolivia, Cameroon, and Ghana (Fig. A1 and Table A1, Fig. 2 in Torello-Raventos et al., 2013). The TROBIT field campaigns were carried out over a 3--year period, from 2006 to 2009, and the field plots used in this study are 1 hectare in size except for BFI-01 (0.5 ha), BFI-02 (0.5 ha), BFI-03 (0.5 ha), CTC-01 (0.93 ha), and VCR-01 (0.6 ha).

All the sites fall within the tropics, that is, within 23.5 degrees north and south of the equator, and were selected in regions where savannas and forests were in close proximity and Oxid within ecotones or 'zones of tension'. As such, the sites sampled show a large variation in physiognomy and <u>edaphic and climatic conditions (Table S1, Veenendaal et al., 2015).</u>

135 edaphic and climatic conditions (Table S1, Veenendaal et al., 2015).

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The classification of the TROBIT plots_ites as either ' forest' or 'savanna' is based on the parameters described in Torello-Raventos et al. (2013) and Veenendaal et al. (2015). A 'savanna' is a natural land cover that is not a forest, bare ground, or a body of water. 'Forest' is defined as woody vegetation with an average tree height of or exceeding 6 mand a canopy area index (CAI) value of at least 0.3 for 'open forests' and 0.7 for 'forests' in addition, floristic differences (i.e., dominance of 'savanna' species) are used to differentiate forests from taller-growing savannas that have similar CAIs and tree heights (see Fig. 9, Torello-Raventos et al., 2013).

There is some ambiguity in how 'savannas' and 'grasslands' are defined. Some modelling-based research treat the two biomes as different (Whitley et al., 2017), while studies based on plant functional traits group them together (Solofondranohatraet al., 2018; White et al., 2000). As there is some concern that MODIS VCF will struggle to pick up woody cover in areas with really sparse vegetation, in this paper we decided to treat 'grasslands' as part of the savanna domain.

2.2 Converting In-Situ Canopy Area Index to MODIS VCF / Lands at TCC percent tree cover

CAI is defined as the sumof the projected areas of individual tree crowns divided by the ground area. In the TROBIT project (Torello-Raventos et al. (2013) and Veenendal_et al. (2015)), plot-wide CAI is made up of the sumof the upper-stratum mid-stratum and subordinate-stratumcrown areas. determined by the tree's dbh (upper-stratum dbh > 10 cm, mid-stratum subordinate-stratum dbh < 2.5 cm, height > 1.5 m). About 50 trees per plot-specific allometric relations between stemdiameter and crown area (supplement B of Torello Raventos et al. (2013));). These were then applied to the whole plot to establish plot-level CAI. For the allometric relationships, tree crowns were treated as circles; and the individual tree projected crown area was determined using the average of crown radii measured along the four cardinal points (i.e;₇₂ from the centre of the stemto the distance furthest from the stem).

CAI values do not account for within-site tree canopy distribution patterns and the overlap between individual tree canopies. We account for this by converting each CAI value into a probability distribution function incorporating the following two extreme scenarios: $\frac{544}{2}$ enforced overlap, $\frac{100}{2}$, where the location probability of individual canopies increases linearly from 0 to 1 across a site; and $\frac{144}{2}$ unenforced overlap, $\frac{100}{2}$, where individual canopies follow a uniform random distribution pattern and canopy overlap is not purposefully introduced (Fig. A.2.1). We repeated this 1000 times per CAI measurement to determine the probability distribution of expected CAI for each field plot.

Unlike CAI, which is the fraction of ground covered by tree crowns, the percent tree cover value from MODIS VCF (and so Landsat TCC) is defined as "the portion of the skylight orthogonal to the surface which is intercepted by trees" (Hansen et al. 2002). To make MODIS VCF these cover and Landsat TCC comparable to treecover derived from IROBIT plot CAIs, we divided the MODIS VCF these product values by 0.8 as suggested by Hansen et al. 2002). This is also the standard approachinmost modelling studies that use MODIS using VCF (e.g., Lasslopet al., 2020;Kelkyetal, 2013; Burton et al., 2019). The 0.8 value can be thought of as a gap correction factor (GCF) that accounts for within-canopy gaps. Although the GCF has been shown to vary with vegetation type (Lloydet al., 2008; 0.34 -

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0.60) and crown cover (Tang et al., 2019: 0.9670 - 0.796), we opted to use 0.8 as we found that it yielded more conservative results compared to a variable GCF. It also avoided introducing additional parameters into our analysis.

Next, to account for the difference in size between the MODIS VCF pixel (250 mx 250 m) and the smaller field plot size (100 mx 100 m), we calculated the possible percent tree cover an area the size of a TROBIT field plot could have, given the MODIS VCF percent tree cover for a MODIS-sized pixel. This was done for two extreme scenarios: "enforced clunping," where all the tree cover for the given MODIS VCF value is forcibly 'clunped' on one side of the pixel, or "unenforced clunping," where 'clumping is not enforced₇ and tree cover is distributed randomly within the pixel (Fig. A32). The clumping scenarios introduce possible variations in percent cover due to the area and location mismatch between a TROBIT field plot and a MODIS pixel. A probability distribution was generated for each MODIS VCF pixel by calculating percent tree cover values for 1000 samples (100 mx 100 m) randomly placed within the 250 mx 250 m MODIS VCF pixel.

For LandsatTCC, where the LandsatTCC pixels (30mx30m) are smaller than the TROBIT field sites, we calculated a TCC percent tree cover to match the TROBIT field site size by summing the percent tree cover within the TCC pixel part found inside the TROBIT field site and then dividing the sumby the TROBIT site area. As TROBIT site orientation was not recorded, we randomized the angle between the TROBIT site and TCC pixel grid for each of the 1000 samples when generating the probability distribution. "Enforced clumping" was performed as per MODIS VCF (Fig 2), with the direction of clumping randomized.

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Figure 1. Visual representation of the effects of enforcing overlap within a (100 m x 100 m) TROBIT site with a given <u>Canopy Area Index</u> (CAI) Left: Overlap is not enforced, and individual crowns follow a uniform random distribution. Right: Overlap is enforced by linearly increasing the probability of a canopy being located more on one side of the site (i.e. here the right side of the site) than the other. This results in tree canopies 'overlapping' to a greater extent, which affects how accurately CAI represents actual canopy cover.

2.3 Calculating Uncertainty Under Different Overlap-Clumping Scenarios

We thereby compared both MODIS VCF and Landsat TCC with TROBIT under four different scenarios: 1) unenforced

overlap and clumping; 2) enforce overlap and unenforced clumping; 3) unenforced overlap and enforced clumping; 4) enforced overlap and clumping. Comparisons were conducted by fitting the following logit function:

$\frac{logit(VCF) - C_0 + \Delta \times log(C^{\tau_1}/(1 - C^{\tau_2}))}{logit(Pixel)} = C_0 + \Delta \times log(C^{\tau_1}/(1 - C^{\tau_2}))$

Where $G_{2r}A_{1}\tau_{2r}\tau_{2r}C_{0}$, $\Delta_{1}\tau_{1}$, τ_{2} are optimised parameters and VGPixel and C are the MODIS VCF / Landsat TCC pixel (post-conversion as described in section 2.2) and TROBIT site probability distributions, respectively. This is similar to a standard linear regression of logit transformed data, accounting for maximum and minimum bounds of 0 - 100 % tree cover, with $\tau_{2r}\tau_{2}\tau_{1}$, τ_{2} allowing for a non-symmetric transformation of tree cover. To account for the probability density of each point, we inferred the parameters in Equation 1 using a Total Least Squares Bayesian Inference technique using a Metropolis-Hastings Markov Chain Monte Carlo step. Priors were uninformed but physically bounded (i.e., $T_{2r}, \tau_{2r} > 0$, $\Delta_{1}, \tau_{1}, \tau_{2} > 0$) to assume an increasing relationship between

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MODIS VCF / Landsat TCC and

describing our conditional

distribution (Gelman et al., 2013). Each combination was run over 10 chains, with 1000 warm-up iterations and 10,000 sampling iterations. Optimisation was performed using the rstan2.19.2 (Stan Development Team, 2019) package in R3.5.2 (RCore Team, 2018). Our optimization accounts for potential errors in TROBIT cover, which includes those caused by the allometric construction of the CAI, provided that the errors are unbiased and remain roughly consistent across sites (Gelman et al., 2013). As the TROBIT plots have relatively small total errors associated with the allometric relationships (Table B1, Torello–Raventos et al., 2013), systematic errors are unlikely to affect our results.



Figure 2. Left: Example of the effects of unenforced and enforced clumping in a 250 m x 250 m MODIS VCF pixel with 50 % tree cover. Clumping all the cover (green) to one side of the pixel (left bottom) affects the average canopy cover value of a 100 m x 100 m-sized TROBIT site (black boxes). Right: Example of the effects of unenforced and enforced clumping on 30 m x 30 m Landsat TCC pixels with a mix of tree covers (green) and non-tree cover (brown). White dotted lines are TCC pixel boundaries. Clumping all the cover to one side of the pixel (right bottom) affects the average canopy cover value of a 100 m x 100 m-sized TROBIT site (black boxes).

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2.4 Mapping MODIS VCF Uncertainty Across The Tropics

We evaluated the inpact of the MODIS VCF biases inferred from this correction these regression equations across the tropics by inverting our calculation of MODIS VCF bias (Fig. A4A2) as follows: first, the inverse (i.e., solving for C) of Equation equation 1 was applied to MODIS VCF values after conversion to a 100 mx 100 mpixel size grid (matching the field site area); then this corrected calibrated value was translated back to the original 250 mx 250 mVCF pixel size. As the inverse of Equation 1 has no analytical solution, we found the rounded percent value of C that minimises the absolute difference between the left- and right-hand side of the equation. For computational feasibility, we constructed mpsofile tropics with corrected calibrated MODIS VCF values (Fig. 2A3) by and only sampling 5 incritors the two encloses are as ' forest' or 'savanna'-in the 500 mMODIS Land Cover Type (MCD 12Q1 - collection 6) (Sulla-Menashe and Friedl, 2018).

Weten used the 500m MODIS Land Cover Type (MD12Q1-collection6) product to identify the acas of fixes' and 'savama' austhet pisithe MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere–Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen Needleleaf Forests, Evergreen Broadleaf Forests, Deciduous Needleleaf Forests, Deciduous Broadleaf Forests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

Forame detailed land-coverspecific evaluation, we recently the control of the con

3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforced unenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and may instead be overestimating tree cover (50 % confidence, dashed line, Fig.

significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pixeite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered is used as where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; black for forest and savanna combined). The bol, and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent the uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18-19-21% (without enforced clumping) or $9-10\cdot11-12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF <u>overoclimateQunderestimates</u> tree cover where tree cover exceeds 7884% (at the 95 % confidence)

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interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds $\frac{9078}{28}$ % (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered to one side within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by culmping; the horizontal error bars represent the uncertainty introduced by overlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and

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consistent for covers below 75 - 80 % (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of <u>post-calibration</u> change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

Fue 2Dth/ordnowoms <u>(fr)</u> (expinently of iddDDS/CF4n(ph) the part of of a lange in the experimental of the hyperbolic temperature of the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the for scenarios) (PA) is slowed as the slowed or slowed in the precedent in the foreign of field sites used as references and proceed as the precedent in the precedent in

(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the

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forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited. Indeed, it is allocuted bits fifty as 5 and 42 are to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95% confidence interval)(Fig. 6). The nost substantial and significant underestimation is in the classes 'woody savannas' and 'savannas', The underestimation is the largest in woody savannas, except when clumping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clumping is the largest (i.e., at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26-67%, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60 %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al. 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.</u>

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is

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insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, close shrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm as well as trees with a height exceeding 1.5 m when dbh <2.5 cm

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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37014 calbration (cbckwie: no enforced clumping or overlap (black); enforced Figure 36. Percent change in tree cover clumping and no enforced overlap (blue); no enforced clumping and enforced overlap (red); enforced clumping and overlap (pinl) in the 'forest' supercategory and the 5 savanna classes. Palest tone indicates positive change, mid-tone indicates negative change, and the darkest tone indicates net change. Error bars denote the 5-95% confidence $interval; if the \, error \, bar \, extends \, past \, the \, x-axis, the \, post- {}^{\rm corroction} {}^{\rm calibration} \, change \, is \, not \, considered \, significant.$

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower height ranges. Instead, we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 mheight in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Rutten et al., 2015), (Rutten et al., 2015), more research needstobedone with more insituheight data

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m

always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and ' closed shrublands' classes of the MCD 12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or_a some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to and improve its useability.

within-field-site and field-site-pixel variation are accounted for during validation incalibration. We asofund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net)</u>.

Site		Latitud ØLatitud	Longitu de <u>Longit</u>	MODI S VCF Tree	Canop <u>yCanop</u> yArea	Average Upper Stratum	Cover	
Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	∗ 1.17	1.33	0.52	12.53	Savanna	Savannawoodland
BBI-02	Burkina Faso	12.73	, 1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	* 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl. 10
BDA-02	Burkina Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Şavanna	Shrub-rich sa vanna woodl n
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman Not Expanded by
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	All the sites fall within the tropics, that is, within 23.5 degrees north and south of the equator, and were selected		Formatted: Normal, Space Before: 0 pt, Line spacing	1:
	in regions where savannas and forests were in close proximity and exist within ecotones or 'zones of		1.5 lines	
	tonsion.'tension', As such, the sites sampled show a large variation in physiognomy and edaphic and climatic	\geq	Formatted	
	conditions (Table S1, Veenendaal et al., 2015).			
	135 edaphic and climatic conditions (Table S1, Veenendaal et al., 2015).			
	The classification of the TROBIT olots ites as either 'forest' or 'savanna' is based on the parameters described in Tr	orell	Formatted: Font: Times New Roman, 10 pt	
	Raventos et al. (2013) and Veenendaal et al. (2015). A 'savanna' is a natural land cover that is not a forest, have grou	ind, o	Formatted: Normal Indept: Left: 0" Right: 0" Space	
140	a body of water. 'Forest' is defined as woody vegetation with an average tree height of or exceeding 6 m and a cano	DY	Before: 0 pt, Line spacing: 1.5 lines	
	area index (CAI) value of at least 0.3 for 'open forests' and 0.7 for 'forests'. In addition, floristic difference		Formatted	
	(i e_ dominance of savanna's necies) are used to differentiate forests from taller-growing savannas that have similar		le le	
	and trachaights (son Fig. 0. Torallo Pavantas at al. 2012)		L3	
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	There is some applicative in how 'seven ness' and 'arresplands' are defined. Some modelling based research tract the tract	\leq	Formatted: Font: Times New Roman, 10 pt	$ \rightarrow$
	biorms as different (Whitley et al. 2017) while studies based on plant functional traits group them together		Formatted: Normal, Space Before: 0 pt, Line spacing 1.5 lines	J:
450	(Solofondranohatra et al. 2018; White et al. 2000) As there is some concern that MODIS VCE will struggle to nick		Formatted: Normal. Indent: Left: 0". Right: 0". Line	\equiv
190	woody cover in areas with really sparse vegetation in this paper we decided to treat 'areas lands' as part of the savan	na	spacing: 1.5 lines	
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155	22 Conventing In Site Conomy Area Indexts MODIS VCE / Londsot TCC noncontinue correct	\sim	Formatted: Font: Times New Roman, 14 pt	\dashv
100	2.2 Converting HPSite Canopy Area index to MODIS VCF / Landsat ICC percent the Cover		Formatted: Normal, Space Before: 0 pt	
	CAI is defined as the sumof the projected areas of individual tree crowns divided by the ground area. In the		Formatted	
	TROBIT project (Torello-Raventos et al. (2013) and Veenendal et al. (2015)), plot-wide CAI is, made up of the	1		
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	2.5 cm < dbh < 10 cm, and subordinate-stratum dbh < 2.5 cm, height > 1.5 m). About 50 trees per		Formettad	

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stratumper plot were measured to derive plot-specific allometric relations between stemdiameter and crown area (supplement B of Torello Raventos et al.-($_{a.}2013$))- $_{b.}$ These were then applied to the whole plot to establish plot-level CAI. For the allometric relationships, tree crowns were treated as circles; and the individual tree projected crown area was determined using the average of crown radii measured along the four cardinal points (i.e., from the centre of the stemto the distance furthest from the stem).

CAI values do not account for within-site tree canopy distribution patterns and the overlap between individual tree canopies. We account for this by converting each CAI value into a probability distribution function incorporating the following two extreme scenarios: $\frac{44}{2}$ enforced overlap, $\frac{100}{2}$ where the location probability of individual canopies increases linearly from 0 to 1 across a site; and $\frac{14}{2}$ unenforced overlap, $\frac{100}{2}$ where individual canopies follow a uniform random distribution pattern and canopy overlap is not purposefully introduced (Fig. A-21). We repeated this 1000 times per CAI measurement to determine the probability distribution of expected CAI for each field plot.

Unlike CAI, which is the fraction of ground covered by tree crowns, the percent tree cover value from MODIS VCF (and so Landsat TCC) is defined as "the portion of the skylight orthogonal to the surface which is intercepted by trees" (Hansen et al. 2002). To make MODIS VCF tree cover and Landsat TCC comparable to tree cover derived frontiROBIT plot CAIs, we divided the MODIS VCF these product values by 0.8 as suggested by Hansen et al. 2002). This is also the standard approach innost modelling studies that use MODIS surface (e.g., Lasslopet al., 2002;Kelkyetal, 2013; Burton et al., 2019). The 0.8 value can be thought of as a gap correction factor (GCF) that accounts for within-canopy gaps. Although the GCF has been shown to vary with vegetation type (Lloyd et al., 2008; 0.34 -0.60) and crown cover (Tang et al., 2019: 0.9670-0.796), we opted to use 0.8 as we found that it yielded more conservative results compared to a variable GCF. It also avoided introducing additional parameters into our analysis.

Next, to account for the difference in size between the MODIS VCF pixel (250 mx 250 m) and the smaller field plot size (100 mx 100 m), we calculated the possible percent tree cover an area the size of a TROBIT field plot could have, given the MODIS VCF percent tree cover for a MODIS-sized pixel. This was done for two extreme scenarios: "enforced clunping," where all the tree cover for the given MODIS VCF value is forcibly 'clumped' on one side of the pixel, or "unenforced clumping," where 'clumping is not enforced₇ and tree cover is distributed randomly within the pixel (Fig. A32). The clumping scenarios introduce possible variations in percent cover due to the area and location mismatch between a TROBIT field plot and a MODIS pixel. A probability distribution was generated for each MODIS VCF pixel by calculating percent tree cover values for 1000 samples (100 mx 100 m) randomly placed within the 250 mx 250 m MODIS VCF pixel.

For Landsat TCC, where the Landsat TCC pixels (30mx30m) are smaller than the TROBIT field sites, we calculated a TCC percent tree cover to match the TROBIT field site size by summing the percent tree cover within the TCC pixel part found inside the TROBIT field site and then dividing the sumby the TROBIT site area. As TROBIT site orientation was not recorded, we randonized the angle between the TROBIT site and

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TCC pixel grid for each of the 1000 samples when generating the probability distribution. "Enforced clumping" was performed as per MODIS VCF (Fig 2), with the direction of clumping randonized.



Figure 1. Visual representation of the effects of enforcing overlap within a (100 m x 100 m) TROBIT site with a given Canopy Area Index (CAI) Left: Overlap is not enforced, and individual crownsfollow a uniform random, distribution. Right: Overlap is enforced by linearly increasing the probability of a canopy being located more on one, side of the site (i.e. here the right side of the site) than the other. This results in tree canopies 'overlapping' to a greater extent, which affects how accurately CAI represents actual canopy cover.

2.3 Calculating Uncertainty Under Different Overlap-Clumping Scenarios

We thoroby compared <u>both</u> MODIS VCF and <u>Landsat TCC with</u> TROBIT under four different scenarios: 1) unenforced overlap and clumping; 2) enforce overlap and unenforced clumping; 3) unenforced overlap and enforced clumping; 4) enforced overlap and clumping. Comparisons were conducted by fitting the following logit function:

 $\frac{logit(VCF) = C_0 + \Delta \times log(C^{\tau_1}/(1 - C^{\tau_2}))}{logit(Pixel)} = C_0 + \Delta \times log(C^{\tau_1}/(1 - C^{\tau_2}))$

Where $C_{ar}A_{r}\tau_{ar}\tau_{c}C_{0}$, $A_{r}\tau_{1}$, τ_{2} are optimised parameters and VGPixel and C are the MODIS VCF / Landsat TCC pixel (post-conversion as described in section 2.2) and TROBIT site probability distributions, respectively. This is similar to a standard linear regression of logit transformed data, accounting for maximum and minimum bounds of 0 - 100 % tree cover, with $\tau_{ar}+z\tau_{1}$, τ_{2} allowing for a non-symmetric transformation of tree cover. To account for the probability density of each point, we inferred the parameters in Equation 1 using a Total Least

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Squares Bayesian Inference technique using a Metropolis-Hastings Markov Chain Monte Carlo step. Priors were uninformed but physically bounded (i.e., $4, \tau_0, \tau_2 > 0, \Delta, \tau_1, \tau_2 > 0$) to assume an increasing relationship between MODIS VCF /LandsatTCC and describer our provide the set of the set

describing our conditional

distribution (Gelman et al., 2013). Each combination was run over 10 chains, with 1000 warm-up iterations and 10,000 sampling iterations. Optimisation was performed using the rstan2.19.2 (Stan Development Team 2019) package in R3.5.2 (R Core Team, 2018). Our optimization accounts for potential errors in TROBIT cover, which includes those caused by the allometric construction of the CAI, provided that the errors are unbiased and remain roughly consistent across sites (Gelman et al., 2013). As the TROBIT plots have relatively small total errors associated with the allometric relationships (Table B1, Torello–Raventos et al., 2013), systematic errors are unlikely to affect our results.



Figure 2. Left: Example of the effects of unenforced and enforced clumping in a 250 m x 250 m MODIS VCF pixel with 50 % tree cover. Clumping all the cover (green) to one side of the pixel (left bottom) affects the average canopy cover value of a 100 m x 100 m-sized TROBIT site (black boxes). Right: Example of the effects of unenforced and enforced clumping on 30 m x 30 m Landsat TCC pixels with a mix of tree covers (green) and non-tree cover (brown). White dotted lines are TCC pixel boundaries. Clumping all the cover to one side of the pixel (right bottom) affects the average canopy cover value of a 100 m x 100 m-sized TROBIT site (black boxes).

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2.4 Mapping MODIS VCF Uncertainty Across The Tropics

We evaluated the inpact of the MODIS VCF biases inferred from this correction these regression equations across the tropics by inverting our calculation of MODIS VCF bias (Fig. A4A2) as follows: first, the inverse (i.e., solving for C) of Equation equation 1 was applied to MODIS VCF values after conversion to a 100 mx 100 mpixel size grid (matching the field site area); then this corrected calibrated value was translated back to the original 250 mx 250 mVCF pixel size. As the inverse of Equation 1 has no analytical solution, we found the rounded percent value of C that minimises the absolute difference between the left- and right-hand side of the equation. For computational feasibility, we constructed mpsofile tropics with corrected calibrated MODIS VCF values (Fig. 2A3) by and only sampling 5 invations the two considered as ' forest' or 'savanna', in the 500 mMODIS Land Cover Type (MCD 12Q1 - collection 6) (Sulla-Menashe and Friedl, 2018).

We<u>ten</u>usedthe<u>500mMODSLandCoverType</u>(MD12Q1-collection6) productoidentify the acas of fixest and 'savarn' anshter is inter MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere–Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen NeedleleafForests, Evergreen BroadleafForests, Deciduous NeedleleafForests, Deciduous BroadleafForests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

Foramore detailed land-coverspecific evaluation, we **Exampled** <u>evaluated</u> the <u>control d</u> <u>alhated</u> <u>250nMODIS VE picels 0 apiel values</u> <u>for each corresponding</u> 500 mgrid and <u>combined it with the MCD12Q1 productive</u> to construct and coverspecific MODIS VE tree cover frequency distributions (Fig. <u>A5A4</u>). Our tree cover <u>correction calibration</u> by cover type (Fig. 3) for the four clumping/overlap regression combinations was then calculated by multiplying each cover type MODIS VCF frequency distribution (Fig. <u>A5A4</u>) with curves representing the redian, 5 %, and 95 % confidence lines of the <u>correction calibration</u> equation ensembles.

3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforced unenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and may instead be overestimating tree cover (50 % confidence, dashed line, Fig. significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pickite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered in one area of the pixel. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; blackfor forest and savanna combined). <u>The), and the thin lines</u> represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 48—19–21% (without enforced

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clumping) or $9 - 10 \cdot 11 \cdot 12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overestimates underestimates the cover where the cover exceeds 7884 % (at the 95 % confidence interval) when neither overlap nor clumping is enforced, and overestimates where the cover exceeds 9078% (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixels, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by coverlap.

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Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and consistent for covers below 75 - 80% (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of post-calibration change in tropical tree cover

and savanna land cover classes we identified accounting other forest of savanna, justinga botter faithing hard in the forest savanna six factor of savanna specific control (range curve, Fig. 1). We did not use 3) instead of using the savanna-onlysites frasavanna specific control (range curve, Fig. 1). This is because there were few TROBIT sites representing savanna with MODIS VCF tree cover values exceeding 40%, and global land cover maps disagree on the distribution of savannas within the forest-savanna ecotone (Herold et al., 2008).

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(Fig. 2A3), and the regions where all four scenarios agree on the direction of change (positive and negative) are

substantial-<u>(Fig.5)</u>. However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help inprove confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited in the second definitive. Indeed, it is a limited number of regions, it is important to not that the maps shown in Figure 2 are not definitive. Indeed, it is a limited number of the second definitive in the second definitity in the

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95% confidence interval)(Fig. 6). The most substantial and significant underestimation is in the classes 'woody savannas' and 'savannas' The underestimation is the largest in woody savannas, except when clumping and overlap are enforced (in purple, Fig_6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clumping is the largest (i.e. at about 20% tree cover, see Fig_A4), while the peak in cover distribution for woody savannas (26 - 67%, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60</u> %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al., 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.

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We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced, Sinilarly, the net change is insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, closes hrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm as well as trees with a height exceeding 1.5 m when dbh <2.5 cm

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 mthreshold to our observed underestimation.

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37014 calbration (cbckwie: no enforced clumping or overlap (black); enforced Figure 36. Percent change in tree cover clumping and no enforced overlap (blue); no enforced clumping and enforced overlap (red); enforced clumping and overlap (pinl) in the 'forest' supercategory and the 5 savanna classes. Palest tone indicates positive change, mid-tone indicates negative change, and the darkest tone indicates net change. Error bars denote the 5-95% confidence $interval; if the \, error \, bar \, extends \, past \, the \, x-axis, the \, post- {}^{\rm corroction} {}^{\rm calibration} \, change \, is \, not \, considered \, significant.$

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower height ranges. Instead, we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 mheight in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Rutten et al., 2015), (Rutten et al., 2015), more research needstobedone with more insituheight data.

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and

' closed shrublands' classes of the MCD 12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or, some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to and improve its useability.

within-field-site and field-site-pixel variation are accounted for during validation in calibration. We also fund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net)</u>.

Site		Latitud ØLatitud	Longitu de <u>Longit</u>	MODI S VCF Tree	Canop <u>yCanop</u> yArea	Average Upper Stratum	Cover	
Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	∗ 1.17	1.33	0.52	12.53	Savanna	Savannawoodland
BBI-02	Burkina Faso	12.73	, 1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	* 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl. 10
BDA-02	Burkina Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Şavanna	Shrub-rich sa vanna woodl n
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Fore st	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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Next, to account for the difference in size between the MODIS VCF pixel (250 mx 250 m) and the smaller field plot size (100 mx 100 m), we calculated the possible percent tree cover an area the size of a TROBIT field plot could have, given the MODIS VCF percent tree cover for a MODIS-sized pixel. This was done for two extreme scenarios: "enforced clumping," where all the tree cover for the given MODIS VCF value is forcibly 'clumped' on one side of the pixel, or "unenforced clumping," where 'clumping is not enforced₇ and tree cover is distributed randomly within the pixel (Fig. A32). The clumping scenarios introduce possible variations in percent cover due to the area and location mismatch between a TROBIT field plot and a MODIS pixel. A probability distribution was generated for each MODIS VCF pixel by calculating percent tree cover values for 1000 samples (100 mx 100 m) randomly placed within the 250 mx 250 m MODIS VCF pixel.

For Landsat TCC, where the Landsat TCC pixels (30mx30m) are smaller than the TROBIT field sites, we calculated a TCC percent tree cover to match the TROBIT field site size by summing the percent tree cover within the TCC pixel part found inside the TROBIT field site and then dividing the sumby the TROBIT site area. As TROBIT site orientation was not recorded, we randomized the angle between the TROBIT site and TCC pixel grid for each of the 1000 samples when generating the probability distribution. "Enforced clumping" was performed as per MODIS VCF (Fig 2), with the direction of clumping randomized.



Figure 1. Visual representation of the effects of enforcing overlap within a (100 m x 100 m) TROBIT site with a given Canopy Area Index (CAI), Left: Overlap is not enforced, and individual crownsfollow a uniform random distribution. Right: Overlap is enforced by linearly increasing the probability of a canopy being located more on one Formatted: Font: Times New Roman, Bold

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side of the site (i.e. here the right side of the site) than the other. This results in tree canopies 'overlapping' to a greater extent, which affects how accurately CAI represents actual canopy cover.

2.3 Calculating Uncertainty Under Different Overlap-Clumping Scenarios

We thoroby compared <u>both</u> MODIS VCF and <u>Landsat TCC with</u> TROBIT under four different scenarios: 1) unenforced overlap and clumping; 2) enforce overlap and unenforced clumping; 3) unenforced overlap and enforced clumping; 4) enforced overlap and clumping. Comparisons were conducted by fitting the following logit function:

$logit(VCF) = C_0 + \Delta \times log(C^{\tau_1}/(1 - C^{\tau_2})) - logit(Pixel) = C_0 + \Delta \times log(C^{\tau_1}/(1 - C^{\tau_2})) - (Equation 1)$

Where $C_{0r}\Delta_{r}\tau_{2r}\tau_{2}C_{0}$, $\Delta_{r}\tau_{1}$, τ_{2} are optimised parameters and $\downarrow C = Pixel$ and C are the MODIS VCF / Landsat TCC pixel (post-conversion as described in section 2.2) and TROBIT site probability distributions, respectively. This is similar to a standard linear regression of logit transformed data, accounting for maximum and minimum bounds of 0 - 100 % tree cover, with $\tau_{2r}\tau_{2}\tau_{1}$, τ_{2} allowing for a non-symmetric transformation of tree cover. To account for the probability density of each point, we inferred the parameters in Equation 1 using a Total Least Squares Bayesian Inference technique using a Metropolis-Hastings Markov Chain Monte Carlo step. Priors were uninformed but physically bounded (i.e., $\Delta_{r}\tau_{2r}\tau_{2r} > 0, \Delta, \tau_{1}, \tau_{2} > 0$) to assume an increasing relationship between MODIS VCF / Landsat TCC and describing our conditional

distribution (Gelman et al., 2013). Each combination was run over 10 chains, with 1000 warm-up iterations and 10,000 sampling iterations. Optimisation was performed using the rstan2.19.2 (Stan Development Team, 2019) package in R3.5.2 (R Core Team, 2018). Our optimization accounts for potential errors in TROBIT cover, which includes those caused by the allometric construction of the CAI, provided that the errors are unbiased and remain roughly consistent across sites (Gelman et al., 2013). As the TROBIT plots have relatively small total errors as sociated with the allometric relationships (Table B1, Torello–Raventos et al., 2013), systematic errors are unlikely to affect our results.

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Figure 2. Left: Example of the effects of unenforced and enforced clumping in a 250 m x 250 m MODIS VCF pixel with 50 % tree cover. Clumping all the cover (green) to one side of the pixel (left bottom) affects the average canopy cover value of a 100 mx 100 m-sized TROBIT site (black boxes). Right: Example of the effects of unenforced and enforced clumping on 30 m x 30 m Landsat TCC pixels with a mix of tree covers (green) and non-tree cover (brown). White dotted lines are TCC pixel boundaries. Clumping all the cover to one side of the pixel (right bottom) affects the average canopy cover value of a 100 mx 100 m-sized TROBIT site (black boxes).

2.4 Mapping MODIS VCF Uncertainty A cross The Tropics

We evaluated the inpact of the MODIS VCF biases inferred from this correction these regression equations across the tropics by inverting our calculation of MODIS VCF bias (Fig. A4A2) as follows: first, the inverse (i.e., solving for C) of Equation equation 1 was applied to MODIS VCF values after conversion to a 100 mx 100 mpixel size grid (matching the field site area); then this corrected calibrated value was translated back to the original 250 mx 250 mVCF pixel size. As the inverse of Equation 1 has no analytical solution, we found the rounded percent value of C that minimises the absolute difference between the left- and right-hand side of the equation. For computational feasibility, we constructed mpsofile tropics with corrected calibrated MODIS VCF values (Fig. 2A3) by and may sampling 5 incritors the two considered as ' forest' or 'savanna' - in the 500 m MODIS Land Cover Type (MCD 12Q1 - collection 6) (Sulla-Menashe and Fried1. 2018).

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We<u>then</u> used the <u>500mMODIS Lond CoverType</u> (MD12Q1-collection6) product to identify the acas of fixest and 'savarna' ans http is inter MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere-Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen Needleleaf Forests, Evergreen Broadleaf Forests, Deciduous Needleleaf Forests, Deciduous Broadleaf Forests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

Forance detailed land-coverspecific evaluation, we recompledented the control detailed allocated 250n MODIS VE pixels of a pixel values for each corresponding 500 mg/id and combined it with the MCD12Q1 pixel to construct land coverspecific MODIS VE free cover frequency distributions (Fig. A5A4). Our tree cover correction calibration by cover type (Fig. 3) for the four clumping/overlap regression combinations was then calculated by multiplying each cover type MODIS VCF frequency distribution (Fig. A5A4) with curves representing the median, 5 %, and 95 % confidence lines of the correction calibration equation ensembles.

3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforcedunenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and may instead be overestimating tree cover (50 % confidence, dashed line. Fig.

significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pixeite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered is used as where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; black for forest and savanna combined). The bol, and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent the uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18-19-21% (without enforced clumping) or $9-10\cdot11-12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overodimateQuadeustimates tree cover where tree cover exceeds 7884% (at the 95 % confidence)

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interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds $\frac{9078}{28}$ % (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixels, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and

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consistent for covers below 75 - 80 % (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of <u>post-calibration</u> change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

Fue 2Dth/ordnowoms <u>(fr)</u> (expinently of iddDDS/CF4n(ph) the part of of a lange in the experimental of the hyperbolic temperature of the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the

(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the

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forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited. Indeed, it is allocuted bits fifty as 5 and 42 are to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95 % confidence interval)(Fig. 6). The nost substantial and significant underestimation is in the classes 'woody savannas' and 'savannas', The underestimation is the largest in woody savannas, except when clunping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clunping is the largest (i.e, at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26 - 67 %, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60 %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al. 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.</u>

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is

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insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, close shrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is minily occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A 3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm. as well as trees with a height exceeding 1.5 m when dbh <2.5 cm.

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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Figure 36. Percent change in tree cover alter the optical and the optical matrix (betwice: merificed change or overlap (blue); no enforced changing and enforced overlap (red); enforced changing and overlap (pink) in the 'forest' super category and the 5 savanna classes. Palest tone indicates positive change, and tone indicates net change. Err or bars denote the 5-95% confidence indicates net with a val; if the error bar extends past the x-axis, the post-correction calibration change is not considered significant.

difference between TROBIT and MODIS-VCF we would have expected an increasing underestimation in the lower height ranges. Instead₇ we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and 'closed shrublands' classes of the MCD12Q1 product (Fig. A5A4), even though the height range for these classes

is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or_a some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to and improve its useability.

within-field-site and field-site-pixel variation are accounted for during validation incalibration. We asofund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net)</u>.

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		Latitud	Longitu	MODI S VCF	WCanon	Average		
Site		e Latitud	deLongit	Tree	v Area	Stratum	Cover.	
Name	Country 2	<u>e</u> ,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	<u>∗</u> 1.17	1.33	0.52	12.53	Savanna	Savanna woodland
BBI-02	Burkina Faso	12.73	- 1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	- 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl
BDA-02	Burkina Faso	10.94	<u>-</u> 3.15	4.5	0.18	14.47	Savanna	Shrub-rich savanna woodl
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCP 01	Austrolio	17.02	145 6	65 67	0.71	22.51	Saucana	Tall sayanna woodland	
DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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Next, to account for the difference in size between the MODIS VCF pixel (250 mx 250 m) and the smaller field plot size (100 mx 100 m), we calculated the possible percent tree cover an area the size of a TROBIT field plot could have, given the MODIS VCF percent tree cover for a MODIS-sized pixel. This was done for two extreme scenarios: "enforced clumping," where all the tree cover for the given MODIS VCF value is forcibly 'clumped' on one side of the pixel, or "unenforced clumping," where 'clumping is not enforced₇ and tree cover is distributed randomly within the pixel (Fig. A32). The clumping scenarios introduce possible variations in percent cover due to the area and location mismatch between a TROBIT field plot and a MODIS pixel. A probability distribution was generated for each MODIS VCF pixel by calculating percent tree cover values for 1000 samples (100 mx 100 m) randomly placed within the 250 mx 250 m MODIS VCF pixel.

For Landsat TCC, where the Landsat TCC pixels (30mx30m) are smaller than the TROBIT field sites, we calculated a TCC percent tree cover to match the TROBIT field site size by summing the percent tree cover within the TCC pixel part found inside the TROBIT field site and then dividing the sumby the TROBIT site area. As TROBIT site orientation was not recorded, we randomized the angle between the TROBIT site and TCC pixel grid for each of the 1000 samples when generating the probability distribution. "Enforced clumping" was performed as per MODIS VCF (Fig 2), with the direction of clumping randomized.



Figure 1. Visual representation of the effects of enforcing overlap within a (100 m x 100 m) TROBIT site with a given Canopy Area Index (CAI), Left: Overlap is not enforced, and individual crownsfollow a uniform random distribution. Right: Overlap is enforced by linearly increasing the probability of a canopy being located more on one Formatted: Font: Times New Roman, Bold

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side of the site (i.e. here the right side of the site) than the other. This results in tree canopies 'overlapping' to a greater extent, which affects how accurately CAI represents actual canopy cover.

2.3 Calculating Uncertainty Under Different Overlap-Clumping Scenarios

We thoroby compared <u>both</u> MODIS VCF and <u>Landsat TCC with</u> TROBIT under four different scenarios: 1) unenforced overlap and clumping; 2) enforce overlap and unenforced clumping; 3) unenforced overlap and enforced clumping; 4) enforced overlap and clumping. Comparisons were conducted by fitting the following logit function:

$logit(VCF) = C_0 + \Delta \times log(C^{\tau_1}/(1 - C^{\tau_2})) - logit(Pixel) = C_0 + \Delta \times log(C^{\tau_1}/(1 - C^{\tau_2})) - (Equation 1)$

Where $C_{0r}\Delta_{r}\tau_{2r}\tau_{2}C_{0}$, $\Delta_{r}\tau_{1}$, τ_{2} are optimised parameters and $\downarrow C = Pixel$ and C are the MODIS VCF / Landsat TCC pixel (post-conversion as described in section 2.2) and TROBIT site probability distributions, respectively. This is similar to a standard linear regression of logit transformed data, accounting for maximum and minimum bounds of 0 - 100 % tree cover, with $\tau_{2r}\tau_{2}\tau_{1}$, τ_{2} allowing for a non-symmetric transformation of tree cover. To account for the probability density of each point, we inferred the parameters in Equation 1 using a Total Least Squares Bayesian Inference technique using a Metropolis-Hastings Markov Chain Monte Carlo step. Priors were uninformed but physically bounded (i.e., $\Delta_{r}\tau_{2r}\tau_{2r} > 0, \Delta, \tau_{1}, \tau_{2} > 0$) to assume an increasing relationship between MODIS VCF / Landsat TCC and describing our conditional

distribution (Gelman et al., 2013). Each combination was run over 10 chains, with 1000 warm-up iterations and 10,000 sampling iterations. Optimisation was performed using the rstan2.19.2 (Stan Development Team, 2019) package in R3.5.2 (R Core Team, 2018). Our optimization accounts for potential errors in TROBIT cover, which includes those caused by the allometric construction of the CAI, provided that the errors are unbiased and remain roughly consistent across sites (Gelman et al., 2013). As the TROBIT plots have relatively small total errors as sociated with the allometric relationships (Table B1, Torello–Raventos et al., 2013), systematic errors are unlikely to affect our results.

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Figure 2. Left: Example of the effects of unenforced and enforced clumping in a 250 m x 250 m MODIS VCF pixel with 50 % tree cover. Clumping all the cover (green) to one side of the pixel (left bottom) affects the average canopy cover value of a 100 mx 100 m-sized TROBIT site (black boxes). Right: Example of the effects of unenforced and enforced clumping on 30 m x 30 m Landsat TCC pixels with a mix of tree covers (green) and non-tree cover (brown). White dotted lines are TCC pixel boundaries. Clumping all the cover to one side of the pixel (right bottom) affects the average canopy cover value of a 100 mx 100 m-sized TROBIT site (black boxes).

2.4 Mapping MODIS VCF Uncertainty A cross The Tropics

We evaluated the inpact of the MODIS VCF biases inferred from this correction these regression equations across the tropics by inverting our calculation of MODIS VCF bias (Fig. A4A2) as follows: first, the inverse (i.e., solving for C) of Equation equation 1 was applied to MODIS VCF values after conversion to a 100 mx 100 mpixel size grid (matching the field site area); then this corrected calibrated value was translated back to the original 250 mx 250 mVCF pixel size. As the inverse of Equation 1 has no analytical solution, we found the rounded percent value of C that minimises the absolute difference between the left- and right-hand side of the equation. For computational feasibility, we constructed mpsofile topics with corrected calibrated MODIS VCF values (Fig. 2A3) by and my sampling 5 iterations for considered as ' forest' or 'savanna' - in the 500 mMODIS Land Cover Type (MCD12Q1 - collection 6) (Sulla-Menashe and Fried1. 2018).

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We<u>then</u> used the <u>500mMODIS Lond CoverType</u> (MD12Q1-collection6) product to identify the acas of fixest and 'savarna' ans http is inter MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere-Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen Needleleaf Forests, Evergreen Broadleaf Forests, Deciduous Needleleaf Forests, Deciduous Broadleaf Forests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

Forance detailed land-coverspecific evaluation, we recompledent and the control of all hand 250n MODIS VE pixeled a pixel values for each corresponding 500 mg/id/and combined it with the MCD12Q1 productive to construct and coverspecific MODIS VE tree cover frequency distributions (Fig. A5A4). Our tree cover correction calibration by cover type (Fig. 3) for the four clumping/overlap regression combinations was then calculated by multiplying each cover type MODIS VCF frequency distribution (Fig. A5A4) with curves representing the redian, 5 %, and 95 % confidence lines of the correction calibration equation ensembles.

3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforcedunenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and may instead be overestimating tree cover (50 % confidence, dashed line. Fig.

significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pixeite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered is used as where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; black for forest and savanna combined). The bol, and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent the uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18-19-21% (without enforced clumping) or $9-10\cdot11-12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overestimates tree cover where tree cover exceeds 7884% (at the 95 % confidence)

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interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds $\frac{9078}{28}$ % (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixels, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and

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consistent for covers below 75 - 80 % (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of <u>post-calibration</u> change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

Fue 2Dth/ordnowoms <u>(fr)</u> (expinently of iddDDS/CF4n(ph) the part of of a lange in the experimental of the hyperbolic temperature of the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the

(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the

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forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50%, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited. Indeed, it is allocuted bits fifty as 5 and 42 are to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95% confidence interval)(Fig. 6). The nost substantial and significant underestimation is in the classes 'woody savannas' and 'savannas', The underestimation is the largest in woody savannas, except when clumping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clumping is the largest (i.e., at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26-67%, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60 %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al., 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.</u>

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is

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insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, close shrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm as well as trees with a height exceeding 1.5 m when dbh <2.5 cm

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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Figure 36. Percent change in tree cover alter the optical and the optical matrix (betwice: merificed change or overlap (blue); no enforced changing and enforced overlap (red); enforced changing and overlap (pink) in the 'forest' super category and the 5 savanna classes. Palest tone indicates positive change, and tone indicates net change. Err or bars denote the 5-95% confidence indicates net with a val; if the error bar extends past the x-axis, the post-correction calibration change is not considered significant.

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower height ranges. Instead₇ we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and

' closed shrublands' classes of the MCD12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or, some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, accounted accounted by a state of the sevent accounted by a state of the s

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to and improve its useability.

within-field-site and field-site-pixel variation are accounted for during validation incalibration. We asofund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net)</u>.

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Site		e Latitud	deLongit	Tree	v Area	Stratum	Cover.	/
Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Fore st	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	<u>1.17</u>	1.33	0.52	12.53	Savanna	Savanna woodland
BBL02	Burlina Faso	12 73	-1.16	1.5	0.99	13.6	Savanna	Savanna woodland
pbi-02	Burnita Aaso	12.13	1.10	1.5	V .99	13.0	gavanna	Sa valilla woodlalid
BDA-01	Burkina Faso	10.94	<u></u> 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodloo
BDA-02	Burkina, Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Savanna	Shrub-rich sa vanna woodl
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by
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probability of observations for a given parameter combination by a normal distribution (Gelman et al., 2013). Each combination was run over 10 chains, with 1000 warm-up iterations and 10,000 sampling iterations. Optimisation was performed using the rstan2.19.2 (Stan Development Team, 2019) package in R3.5.2 (R Core Team, 2018). Our optimization accounts for potential errors in TROBIT cover, which includes those caused by the allometric construction of the CAI, provided that the errors are unbiased and remain roughly consistent across sites (Gelman et al., 2013). As the TROBIT plots have relatively small total errors are unlikely to affect our results.



Figure 2. Left: Example of the effects of unenforced and enforced clumping in a 250 m x 250 m MODIS VCF pixel with 50 % tree cover. Clumping all the cover (green) to one side of the pixel (left bottom) affects the average canopy cover value of a 100 m x 100 m-sized TROBIT site (black boxes). Right: Example of the effects of unenforced and enforced clumping on 30 m x 30 m Landsat TCC pixels with a mix of tree covers (green) and non-tree cover (brown). White dotted lines are TCC pixel boundaries. Clumping all the cover to one side of the pixel (right bottom) affects the average canopy cover value of a 100 m x 100 m-sized TROBIT site (black boxes).

2.4 Mapping MODIS VCF Uncertainty Across The Tropics

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We evaluated theinpact of the MODIS VCF biases inferred from this correction these regression equations across the tropics by inverting our calculation of MODIS VCF bias (Fig. A4A2) as follows: first, the inverse (i.e., solving for C) of Equation equation 1 was applied to MODIS VCF values after conversion to a 100 mx 100 mpixel size grid (matching the field site area); then this corrected calibrated value was translated back to the original 250 mx 250 mVCF pixel size. As the inverse of Equation 1 has no analytical solution, we found the rounded percent value of C that minimises the absolute difference between the left- and right-hand side of the equation. For computational feasibility, we constructed mpsofile tropics with corrected calibrated MODIS VCF values (Fig. 2A3) by and only sampling 5 intations the considered as ' forest' or 'savanna' - in the 500 m MODIS Land Cover Type (MCD 12Q1 - collection 6) (Sulla-Menashe and Friedl, 2018).

We<u>then</u> used the <u>500mMODISLandCoverType</u> (MD12Q1-collection6) product to identify the acas of firest and 'savarna' ans http is inter MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere–Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen Needleleaf Forests, Evergreen Broadleaf Forests, Deciduous Needleleaf Forests, Deciduous Broadleaf Forests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

Foranoe detailed land-coverspecific evaluation, we recompled extracted the compled extracted the control of early and the early and t

3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforcedunenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and may instead be overestimating tree cover (50 % confidence, dashed line, Fig.

significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pixeite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered is used as where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; black for forest and savanna combined). The bol, and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent the uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18-19-21% (without enforced clumping) or $9-10\cdot11-12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF <u>overoclimateQunderestimates</u> tree cover where tree cover exceeds 7884% (at the 95 % confidence)

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interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds $\frac{9078}{28}$ % (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered to one side within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by culmping; the horizontal error bars represent the uncertainty introduced by overlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and

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consistent for covers below 75 - 80 % (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of <u>post-calibration</u> change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

Fue 2Dth/ordnowoms <u>(fr)</u> (expinently of iddDDS/CF4n(ph) the part of of a lange in the experimental of the hyperbolic temperature of the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the

(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the

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forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited. Indeed, it is allocuted bits fifty as 5 and 42 are to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95% confidence interval)(Fig. 6). The nost substantial and significant underestimation is in the classes 'woody savannas' and 'savannas', The underestimation is the largest in woody savannas, except when clumping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clumping is the largest (i.e., at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26-67%, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60 %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al. 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.</u>

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is

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insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, close shrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A 3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm. as well as trees with a height exceeding 1.5 m when dbh <2.5 cm.

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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Figure 36. Percent change in tree cover after the proceeding the approximation of the provided t

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower heightranges. Instead, we found a low R^2 and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and 'closed shrublands' classes of the MCD12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or, some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to the future use of this product to the prove its useability.

within-field-site and field-site-pixel variation are accounted for during validation in calibration. We also fund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net)</u>.

Site.Latitud OLatitudLongitu doLongitu doLongituMODI SVCFCanop YCanopAverage Upper, Tree, VAreaAverage YCanopIndexStratum Upper, Tree, VAreaCover, VareaCover, 					-				
ALC-01 Brazil -2.53 -54.91 12.5 0.32 6.56 Savanna Savannawoodland ALF-01 Brazil -9.6 -55.94 77 2.31 37.02 Forest Tallforest ALF-02 Brazil -9.58 -55.92 76 2.65 41.32 Forest Tallforest ASU-01 Ghana 7.14 -2.45 41.33 2.54 45.27 Forest Tallforest BBI-01 Burkina, Faso 12.73 1.17 1.33 0.52 12.53 Savanna Savannawoodland BBI-02 Burkina, Faso 12.73 1.16 1.5 0.99 13.6 Savanna Savannawoodland BDA-01 Burkina, Faso 10.94 13.15 6.17 0.3 14.53 Savanna Shrub-rich savannawoodland BDA-02 Burkina, Faso 10.94 13.15 4.5 0.18 14.47 Savanna Shrub-rich savannawoodland BF1-01 Ghana 7.71 -1.69 15 1.22 29.67 Savanna Tallclosed woodland BF1-02	Site Name	Country	Latitud 9Latitud ६	Longitu deLongit ude,	MODI S VCF Tree Cover (%)	<mark>Canop</mark> ¥Canop yArea Jndex	Average Upper Stratum Height (m)	Cover. Type	TROBIT Site Description
ALF-01 Brazil -9.6 -55.94 77 2.31 37.02 Forest Tallforest ALF-02 Brazil -9.58 -55.92 76 2.65 41.32 Forest Tallforest ASU-01 Ghana 7.14 -2.45 41.33 2.54 45.27 Forest Tallforest BBI-01 Burkina Faso 12.73 1.17 1.33 0.52 12.53 Savanna Savannawoodland BBI-02 Burkina Faso 12.73 1.16 1.5 0.99 13.6 Savanna Savannawoodland BDA-01 Burkina Faso 10.94 .3.15 6.17 0.3 14.53 Savanna Shrub-rich savannawoodland BDA-02 Burkina Faso 10.94 .3.15 4.5 0.18 14.47 Savanna Shrub-rich savannawoodland BFI-01 Ghana 7.71 -1.69 1.22 29.67 Savanna Tallforest BFI-02 Ghana 7.71 -1.69 1.28 1.08	ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-02 Brazil -9.58 -55.92 76 2.65 41.32 Forest Tallforest ASU-01 Ghana 7.14 -2.45 41.33 2.54 45.27 Forest Tallforest BBI-01 Burkina,Faso 12.73 1.17 1.33 0.52 12.53 Savanna Savannawoodland BBI-02 Burkina,Faso 12.73 1.16 1.5 0.99 13.6 Savanna Savannawoodland BDA-01 Burkina,Faso 10.94 .3.15 6.17 0.3 14.53 Savanna Shrub-rich savannawoodland BDA-02 Burkina,Faso 10.94 .3.15 4.5 0.18 14.47 Savanna Shrub-rich savannawoodland BF1-01 Ghana 7.71 -1.69 1.5 1.22 29.67 Savanna Tallforest woodland BF1-02 Ghana 7.71 -1.69 1.283 1.08 28.2 Savanna Tallforest woodland BF1-03 Ghana 7.71 -1.69 12.83 1.08 28.2 Savanna Tallforest GH1-03	ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
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BBI-02Burkina Faso12.731.161.50.9913.6Savanna </td <td>BBI-01</td> <td>Burkina Faso</td> <td>12.73</td> <td><u>∗</u>1.17</td> <td>1.33</td> <td>0.52</td> <td>12.53</td> <td>Savanna</td> <td>Savannawoodland</td>	BBI-01	Burkina Faso	12.73	<u>∗</u> 1.17	1.33	0.52	12.53	Savanna	Savannawoodland
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BFI-03 Ghana 7.71 -1.7 25.83 2.54 45.07 Savanna Tallsavannawoodland CTC-01 Australia -16.1 145.45 72.67 2.35 40.37 Forest Tallforest	BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
CTC-01 Australia -16.1 145.45 72.67 2.35 40.37 Forest Tall forest	BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
	CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01 Australia -17.02 145.58 21.67 1.67 27.19 Savanna Tallsavannawoodland	DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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Figure 2. Left: Example of the effects of unenforced and enforced clumping in a 250 m x 250 m MODIS VCF pixel with 50 % tree cover. Clumping all the cover (green) to one side of the pixel (left bottom) affects the average canopy cover value of a 100 mx 100 m-sized TROBIT site (black boxes). Right: Example of the effects of unenforced and enforced clumping on 30 m x 30 m Landsat TCC pixels with a mix of tree covers (green) and non-tree cover (brown). White dotted lines are TCC pixel boundaries. Clumping all the cover to one side of the pixel (right bottom) affects the average canopy cover value of a 100 mx 100 m-sized TROBIT site (black boxes).

2.4 Mapping MODIS VCF Uncertainty Across The Tropics

We evaluated theinpact of the MODIS VCF biases inferred from this correction these regression equations across the tropics by inverting our calculation of MODIS VCF bias (Fig. A4A2) as follows: first, the inverse (i.e., solving for C) of Equation a wave splice to MODIS VCF values after conversion to a 100 mx 100 mpixel size grid (matching the field site area); then this corrected calibrated value was translated back to the original 250 mx 250 mVCF pixel size. As the inverse of Equation 1 has no analytical solution, we found the rounded percent value of C that minimises the absolute difference between the left- and right-hand side of the equation. For computational feasibility, we constructed mapsofile topics with corrected calibrated MODIS VCF values (Fig. 2A3) by and my sampling 5 iterations that were reacted as from each of our 10 optimisation chains (50 in total) and masking out pixels with cover types not considered as 122

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'forest' or 'savanna'. <u>in the 500 mMODIS Land Cover Type (MCD 1201 - collection 6) (Sulla-Menashe and</u> Friedl, 2018).

We<u>ten</u>usedthe<u>500mMODSLandCoverType</u>(MD12Q1-collection6) productoidentify the acas of fixest and 'savarn' arcshterisishe MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere–Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen NeedleleafForests, Evergreen BroadleafForests, Deciduous NeedleleafForests, Deciduous BroadleafForests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

Foranoe detailed land-coverspecific evaluation, we **Exampled_stated** the **control d_alhated** 250nMODIS VE **piceled o picel values** <u>for each corresponding</u> 500 mgrid and combined it with the MED 12Q1 productive to construct and coverspecific MODIS VE tree cover frequency distributions (Fig. A5A4). Our tree cover correction calibration by cover type (Fig. 3) for the four clumping/overlap regression combinations was then calculated by multiplying each cover type MODIS VCF frequency distribution (Fig. A5A4) with curves representing the median, 5 %, and 95 % confidence lines of the correction calibration equation ensembles.

3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforcedunenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and may instead be overestimating tree cover (50 % confidence, dashed line. Fig. significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one

side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pixeite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered is used as where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; black for forest and savanna combined). The bol, and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent the uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18-19-21% (without enforced clumping) or $9-10\cdot11-12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overestimates tree cover where tree cover exceeds 7884% (at the 95 % confidence)

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interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds $\frac{9078}{28}$ % (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed within the field site; (3) with overlap and no clumping, where tree canopies are clustered in one area of the pixels, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest: orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by coverlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestination of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestination (at 95% confidence) is significant and

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consistent for covers below 75 - 80 % (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of <u>post-calibration</u> change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

Fue 2Dth/ordnowoms <u>(fr)</u> (expinently of iddDDS/CF4n(ph) the part of of a lange in the experimental of the hyperbolic temperature of the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the

(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the

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forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50%, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited. Indeed, it is allocuted bits fifty as 5 and 42 are to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95 % confidence interval)(Fig. 6). The nost substantial and significant underestimation is in the classes 'woody savannas' and 'savannas', The underestimation is the largest in woody savannas, except when clunping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clunping is the largest (i.e, at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26 - 67 %, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60 %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al. 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.</u>

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is

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insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, close shrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A 3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm. as well as trees with a height exceeding 1.5 m when dbh <2.5 cm.

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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Figure 36. Percent change in tree cover alter the optical and the optical matrix (betwice: merificed change or overlap (blue); no enforced changing and enforced overlap (red); enforced changing and overlap (pink) in the 'forest' super category and the 5 savanna classes. Palest tone indicates positive change, and tone indicates net change. Err or bars denote the 5-95% confidence indicates net with a val; if the error bar extends past the x-axis, the post-correction calibration change is not considered significant.

difference between TROBIT and MODIS-VCF we would have expected an increasing underestimation in the lower height ranges. Instead₇ we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and

' closed shrublands' classes of the MCD 12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or, some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, accounting the states and forests.

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to the future use of this product to the prove its useability.

within-field-site and field-site-pixel variation are accounted for during validation incalibration. We asofund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net)</u>.

Site		Latitud 9Latitud	Longitu de <u>Longit</u>	MODI S VCF Tree	Canop <u>yCanop</u> yArea	Average Upper Stratum	Cover	
Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	∗ 1.17	1.33	0.52	12.53	Savanna	Savannawoodland
BBI-02	Burkina Faso	12.73	, 1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	* 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl. 10
BDA-02	Burkina Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Şavanna	Shrub-rich sa vanna woodl n
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65 67	0.71	22 51	Sayanna	Tall sayanna woodland	
EVD 01	Austrolio	18.07	145.00	42.5	0.74	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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Figure 2. Left: Example of the effects of unenforced and enforced clumping in a 250 m x 250 m MODIS VCF pixel with 50 % tree cover. Clumping all the cover (green) to one side of the pixel (left bottom) affects the average canopy cover value of a 100 mx 100 m-sized TROBIT site (black boxes). Right: Example of the effects of unenforced and enforced clumping on 30 m x 30 m Landsat TCC pixels with a mix of tree covers (green) and non-tree cover (brown). White dotted lines are TCC pixel boundaries. Clumping all the cover to one side of the pixel (right bottom) affects the average canopy cover value of a 100 mx 100 m-sized TROBIT site (black boxes).

2.4 Mapping MODIS VCF Uncertainty Across The Tropics

We evaluated theinpact of the MODIS VCF biases inferred from this correction these regression equations across the tropics by inverting our calculation of MODIS VCF bias (Fig. A4A2) as follows: first, the inverse (i.e., solving for C) of Equation a wave splice to MODIS VCF values after conversion to a 100 mx 100 mpixel size grid (matching the field site area); then this corrected calibrated value was translated back to the original 250 mx 250 mVCF pixel size. As the inverse of Equation 1 has no analytical solution, we found the rounded percent value of C that minimises the absolute difference between the left- and right-hand side of the equation. For computational feasibility, we constructed mapsofile topics with corrected calibrated MODIS VCF values (Fig. 2A3) by and my sampling 5 iterations that were reacted as from each of our 10 optimisation chains (50 in total) and masking out pixels with cover types not considered as 136

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'forest' or 'savanna'. <u>in the 500 mMODIS Land Cover Type (MCD 1201 - collection 6) (Sulla-Menashe and</u> Friedl, 2018).

We<u>ten</u>usedthe<u>500mMODSLandCoverType</u>(MD12Q1-collection6) productoidentify the acas of fixest and 'savarn' arcshterisishe MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere–Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen NeedleleafForests, Evergreen BroadleafForests, Deciduous NeedleleafForests, Deciduous BroadleafForests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

Foranoe detailed land-coverspecific evaluation, we **Exampled_stated** the **control d_alhated** 250nMODIS VE **piceled o picel values** <u>for each corresponding</u> 500 mgrid and combined it with the MED 12Q1 productive to construct and coverspecific MODIS VE tree cover frequency distributions (Fig. A5A4). Our tree cover correction calibration by cover type (Fig. 3) for the four clumping/overlap regression combinations was then calculated by multiplying each cover type MODIS VCF frequency distribution (Fig. A5A4) with curves representing the median, 5 %, and 95 % confidence lines of the correction calibration equation ensembles.

3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforcedunenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and may instead be overestimating tree cover (50 % confidence, dashed line. Fig. significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one

side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pixeite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered is used as where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; black for forest and savanna combined). The bol, and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent the uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18-19-21% (without enforced clumping) or $9-10\cdot11-12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overestimates tree cover where tree cover exceeds 7884% (at the 95 % confidence)

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interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds $\frac{9078}{28}$ % (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixels, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and

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consistent for covers below 75 - 80 % (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of <u>post-calibration</u> change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

Fue 2Dth/ordnowoms <u>(fr)</u> (expinently of iddDDS/CF4n(ph) the part of of a lange in the experimental of the hyperbolic temperature of the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the

(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the

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forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited. Indeed, it is allocuted bits fifty as 5 and 42 are to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95 % confidence interval)(Fig. 6). The nost substantial and significant underestimation is in the classes 'woody savannas' and 'savannas', The underestimation is the largest in woody savannas, except when clunping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clunping is the largest (i.e, at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26 - 67 %, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60 %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al. 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.</u>

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is

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insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, close shrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A 3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm. as well as trees with a height exceeding 1.5 m when dbh <2.5 cm.

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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Figure 36. Percent change in tree cover after the place of the approximation (checkie: menforced change or overlap (hack enforced changing and enforced overlap (red); enforced changing and overlap (ned); enforced changing and overlap (nel); in the 'forest' super category and the 5 savanna classes. Palest tone indicates positive change, mid-tone indicates negative change, and the darkest tone indicates net change. Err or bars denote the 5-95 % confidence interval; if the error bar extends past the x-axis, the post-corrections allocations change is not consider red significant.

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower heightranges. Instead, we found a low R^2 and a mixture of under and overestimations in heights between 0 and 10 m (Fig. A6A5). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and 'closed shrublands' classes of the MCD12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or_a some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to and improve its useability.

within-field-site and field-site-pixel variation are accounted for during validation incalibration. We asofund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net).</u>

Site		Latitud 9Latitud	Longitu de <u>Longit</u>	MODI S VCF Tree	Canop <u>yCanop</u> yArea	Average Upper Stratum	Cover	
Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	∗ 1.17	1.33	0.52	12.53	Savanna	Savannawoodland
BBI-02	Burkina Faso	12.73	1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	* 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl. 10
BDA-02	Burkina Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Şavanna	Shrub-rich sa vanna woodl n
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Fore st	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	
	Tustunu	10.07	145.77	45.5	0.74	20.15	Suvunnu		Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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We<u>then</u> used the <u>500mMODIS Lond CoverType</u> (MD12Q1-collection6) product to identify the acas of fixest and 'savarna' ans http is inter MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere-Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen Needleleaf Forests, Evergreen Broadleaf Forests, Deciduous Needleleaf Forests, Deciduous Broadleaf Forests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

Forance detailed land-coverspecific evaluation, we recompledent and the compledent land 250n MODIS VE pixeled a pixel values for each consequence of the construct land coverspecific MODIS VE the cover frequency distributions (Fig. A5A4). Our tree cover correction calibration by cover type (Fig. 3) for the four clumping/overlap regression combinations was then calculated by multiplying each cover type MODIS VCF frequency distribution (Fig. A5A4) with curves representing the median, 5 %, and 95 % confidence lines of the correction calibration equation ensembles.

3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforcedunenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and hav instead be overestimating tree cover (50 % confidence, dashed line. Fig.

significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pixeite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered is used as where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; black for forest and savanna combined). The bol, and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent the uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18-19-21% (without enforced clumping) or $9-10\cdot11-12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overestimates tree cover where tree cover exceeds 7884% (at the 95 % confidence)

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interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds $\frac{9078}{28}$ % (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixels, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and

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consistent for covers below 75 - 80 % (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of post-calibration change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

Fue 2Dth/ordnowoms <u>(fr)</u> (expinently of iddDDS/CF4n(ph) the part of of a lange in the experimental of the hyperbolic temperature of the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the

(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the

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forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited. Indeed, it is allocuted bits fifty as 5 and 42 are to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95% confidence interval)(Fig. 6). The nost substantial and significant underestimation is in the classes 'woody savannas' and 'savannas', The underestimation is the largest in woody savannas, except when clumping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clumping is the largest (i.e., at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26-67%, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60 %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al. 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.</u>

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is

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insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, close shrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A 3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm as well as trees with a height exceeding 1.5 m when dbh <2.5 cm

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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Figure 36. Percent change in tree cover after the proceeding the approximation of the provided t

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower heightranges. Instead, we found a low R^2 and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and 'closed shrublands' classes of the MCD12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or, some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, accounted accounted actually be and actually be actual

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to and improve its useability.

within-field-site and field-site-pixel variation are accounted for during validation incalibration. We asofund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net).</u>

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Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Fore st	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	<u>1.17</u>	1.33	0.52	12.53	Savanna	Savanna woodland
BBL02	Burlina Faso	12 73	-1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BBI-02	Burnita Aaso	12.13	1.10	1.5	V .99	13.0	gavanna	Sa valilla woodlalid
BDA-01	Burkina Faso	10.94	<u></u> 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodloo
BDA-02	Burkina, Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Savanna	Shrub-rich sa vanna woodl
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by
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We<u>then</u> used the <u>500mMODIS Lond CoverType</u> (MD12Q1-collection6) product to identify the acas of fixest and 'savarna' ans http is inter MODIS VCF product. MCD12Q1 is widely used by the global land surface modelling community (e.g., Sellar et al., 2019; Wiltshire et al., 2020) and describes land cover in terms of 17 global land cover classes as per the International Geosphere-Biosphere Programme (IGBP, Table 3 in Sulla-Menashe and Friedl, 2018). The product is based on the same spectroradiometer (MODIS) and temporal resolution as the VCF product. Referring to the definition of 'savanna' of Veenendaal et al. (2015), the following land cover classes were chosen to represent 'savanna': Closed Shrubland, Open Shrubland, Woody Savanna, Savanna, and Grassland, while 'forest' encompasses: Evergreen Needleleaf Forests, Evergreen Broadleaf Forests, Deciduous Needleleaf Forests, Deciduous Broadleaf Forests, and Mixed Forests. We subset MCD12Q1 to the tropical zone between _+/- 30° North and took the median class for the 2006 to 2009 period, matching the field data collection period.

Forance detailed land-coverspecific evaluation, we **Decempled** <u>evaluated</u> the <u>compled</u> <u>evaluated</u> <u>interact</u> <u>250nMODIS VE pixel values</u> <u>for each consequence</u> <u>500 mgid and combined it with the MCD12Q1 productive</u> to construct <u>and coverspecific MODIS VE the ecover</u> frequency distributions (Fig. <u>A5A4</u>). Our tree cover <u>compled evaluation</u> by cover type (<u>Fig. 3</u>) for the four clumping/overlap regression combinations was then calculated by multiplying each cover type MODIS VCF frequency distribution (Fig. <u>A5A4</u>) with curves representing the median, 5 %, and 95 % confidence lines of the <u>completion calibration</u> equation ensembles.

3 Results

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforcedunenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and hav instead be overestimating tree cover (50 % confidence, dashed line. Fig.

significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).

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Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field pixeite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered is used as where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or ange for savanna; black for forest and savanna combined). The bol, and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent the uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18-19-21% (without enforced clumping) or $9-10\cdot11-12$ % (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overestimates tree cover where tree cover exceeds 7884% (at the 95 % confidence)

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interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds $\frac{9078}{28}$ % (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixels, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and

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consistent for covers below 75 - 80 % (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of <u>post-calibration</u> change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

Fue 2Dth/ordnowoms <u>(fr)</u> (expinently of iddDDS/CF4n(ph) the part of of a lange in the experimental of the hyperbolic temperature of the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration leading to an increase or decrease in tree cover, respectively) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) acrossed in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the post-calibration change in tree cover, calculated as the 90th per centile (maximum of the for scenarios) (PA) in sl(the precedent in the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the for scenarios) (PA) is slowed as the slowed of the

(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the

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forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited. Indeed, it is allocuted bits fifty as 5 and 42 are to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clunping (95 % confidence interval)(Fig. 6). The nost substantial and significant underestimation is in the classes 'woody savannas' and 'savannas', The underestimation is the largest in woody savannas, except when clunping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clunping is the largest (i.e, at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26 - 67 %, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60 %) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see Table S6 in Sulla-Menashe et al. 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.</u>

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is

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insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, close shrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A3) (i.e. when there is a uniform random distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm. as well as trees with a height exceeding 1.5 m when dbh <2.5 cm.

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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Figure 36. Percent change in tree cover after the proceeding the approximation of the provided t

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower heightranges. Instead, we found a low R^2 and a mixture of under and overestimations in heights between 0 and 10 m (Fig. A6A5). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and 'closed shrublands' classes of the MCD12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or, some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, accounting the states and forests.

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to and improve its useability.

within-field-site and field-site-pixel variation are accounted for during validation in calibration. We also fund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net).</u>

Site		Latitud ØLatitud	Longitu de <u>Longit</u>	MODI S VCF Tree	Canop <u>yCanop</u> yArea	Average Upper Stratum	Cover	
Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	∗ 1.17	1.33	0.52	12.53	Savanna	Savannawoodland
BBI-02	Burkina Faso	12.73	, 1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	* 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl. 10
BDA-02	Burkina Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Şavanna	Shrub-rich sa vanna woodl n
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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ROBIT field sites	Formatted: Indent: Left: 0", Space Before: 12 pt, After: 12 pt, Pattern: Clear (White), Tab stops: Not at 0.58"
tree cover values do not	Formatted: Font: Times New Roman
e cover (50 % confidence.	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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3 Results

250 3.1. Comparing MODIS VCF and Landsat TCC to tree cover from TROBIT field sites

MODIS VCF underestimates tree cover within the 19 % to 81 % range across all four combinations of enforcedunenforced overlap and clumping (black line, Fig. 3). Below 12 %, MODIS VCF tree cover values do not significantly disagree with TROBIT field data, and may instead be overestimating tree cover (50 % confidence, dashed line, Fig.

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TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).



Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field plotsite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor for est and savanan combined). The), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and

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consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 48 - 19 - 21% (without enforced clumping) or 9 - 40 - 11 - 12% (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overestimates tree cover where tree cover exceeds 7884% (at the 95 % confidence interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds 9078% (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly. distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed within the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixels, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or range for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 %.

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<u>confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by</u> <u>clumping: the horizontal error bars represent the uncertainty introduced by overlap.</u>

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and consistent for covers below 75 - 80% (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of post-calibration change in tropical tree cover

and swama and cover classes we identified as being either forest or bavenne,) using a benefit is 1 min back not use 3) instead of using the savanna-onlysites fra savanna specific control (cange curve, Fig. 1). We did not use 3) instead of using the savanna-onlysites fra savanna specific control (cange curve, Fig. 1). This is because there were few TROBIT sites representing savanna with MODIS VCF tree cover values exceeding 40%, and global land cover maps disagree on the distribution of savannas within the forest-savanna ecotone (Herold et al., 2008).

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(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-<u>savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.</u>

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited in the control of the site of the

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clumping (95% confidence interval)(Fig. 6). The most substantial and significant underestimation is in the classes 'woody savannas' and 'savannas' The underestimation is the largest in woody savannas, except when clumping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clumping is the largest (i.e., at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26 - 67%, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60</u> <u>%) matching the range where MODIS VCF nost underestimates tree cover (26 - 67 % cover). The discrepancy</u> <u>may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see</u> <u>Table S6 in Sulla-Menashe et al., 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is</u> <u>therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that</u> <u>are more conservative.</u>

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We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced, Sinilarly, the net change is insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, closes hrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or, some combination of both.

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More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

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should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to and improve its useability.

within-field-site and field-site-pixel variation are accounted for during validation incalibration. We asofund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net).</u>

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Site		e Latitud	deLongit	Tree	v Area	Stratum	Cover.	/
Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Fore st	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	<u>1.17</u>	1.33	0.52	12.53	Savanna	Savanna woodland
BBL02	Burlina Faso	12 73	-1.16	1.5	0.99	13.6	Savanna	Savanna woodland
pbi-02	Burnita Aaso	12.13	1.10	1.5	V .99	13.0	gavanna	Sa valilla woodlalid
BDA-01	Burkina Faso	10.94	<u></u> 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodloo
BDA-02	Burkina, Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Savanna	Shrub-rich sa vanna woodl
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by
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3). A similar pattern is seen when tree cover exceeds 84 %: MODIS VCF does not differ significantly from TROBIT when there is enforced overlap (i.e. when tree canopies are clustering towards one side increasing the degree of canopy overlap - Fig. 1 right), but may underestimate tree cover when overlap is not enforced (i.e. tree canopies are spaced randomly within the site - Fig. 1 left).



Figure 43. MODIS VCF percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a MODIS pixel and/or field plotsite. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within both pixel and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixel, and randomly distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixel, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within a pixel, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor for est and savanan combined). The), and the thin lines represent the 5 and 95 % confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and

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consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 48 - 19 - 21% (without enforced clumping) or 9 - 40 - 11 - 12% (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overestimates tree cover where tree cover exceeds 7884% (at the 95 % confidence interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds 9078% (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly. distributed throughout the field site; (3) with overlap and no clumping, where tree canopies are clustered in one area of the pixels, and randomly. distributed within the field site; (3) with overlap and no clumping, where tree canopies are randomly distributed within the pixels, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; or range for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 %.

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<u>confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by</u> <u>clumping: the horizontal error bars represent the uncertainty introduced by overlap.</u>

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestimation of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and consistent for covers below 75 - 80% (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of post-calibration change in tropical tree cover

and swama and cover classes we identified as being either forest or bavenne,) using a benefit is 1 min back not use 3) instead of using the savanna-onlysites fra savanna specific control (cange curve, Fig. 1). We did not use 3) instead of using the savanna-onlysites fra savanna specific control (cange curve, Fig. 1). This is because there were few TROBIT sites representing savanna with MODIS VCF tree cover values exceeding 40%, and global land cover maps disagree on the distribution of savannas within the forest-savanna ecotone (Herold et al., 2008).

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(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-<u>savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.</u>

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that the maps shown in Figure 2 are not definited in the control of the site of the

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clumping (95% confidence interval)(Fig. 6). The most substantial and significant underestimation is in the classes 'woody savannas' and 'savannas' The underestimation is the largest in woody savannas, except when clumping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clumping is the largest (i.e., at about 20% tree cover, see Fig. A4), while the peak in cover distribution for woody savannas (26 - 67%, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60</u> <u>%) matching the range where MODIS VCF nost underestimates tree cover (26 - 67 % cover). The discrepancy</u> <u>may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see</u> <u>Table S6 in Sulla-Menashe et al., 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is</u> <u>therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that</u> <u>are more conservative.</u>

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border), Left: (No border), Right: (No border), Between : (No border), Tab stops: 3.13", Centered + 6.27", Right We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced, Sinilarly, the net change is insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, closes hrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randomness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm as well as trees with a height exceeding 1.5 m when dbh <2.5 cm

cover ranges for both MODIS VCF and Landsat TCC. In fact Montes ano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 mthreshold to our observed underestimation.

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Figure 36. Percent change in tree cover alter the optical and the optical matrix (betwice: merificed change or overlap (blue); no enforced changing and enforced overlap (red); enforced changing and overlap (pink) in the 'forest' super category and the 5 savanna classes. Palest tone indicates positive change, and tone indicates net change. Err or bars denote the 5-95% confidence indicates net with a val; if the error bar extends past the x-axis, the post-correction calibration change is not considered significant.

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower height ranges. Instead₇ we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net).</u>

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Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
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рск-02 ЕКР-01	Australia Australia	-17.03 -18.07	145.6 145.99	65.67	0.71	22.51	Savanna	Tall cave nne woodland	
EKP-01	Australia	-18.07	145.99	125			Savanna	Tali sa valilla woodlalid	Formatted: Font: Times New Roman
				43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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There is a clear difference in how accurately MODIS VCF estimates treecover in forested areas (in green, Figure 4Fig. 3) as opposed to areas identified as savannas (in orange, Fig. 43). In savanna sites, MODIS VCF significantly and consistently underestimates tree cover regardless of the amount of overlap and clumping. Significant underestimation (at 95 % confidence) occurs when *in-situ* tree cover exceeds 18—19–21% (without enforced clumping) or 9-40-11-12% (with enforced clumping). In forest sites, MODIS VCF does not show the same pattern of systematic underestimation. Divergence does occur at high covers, depending on the enforcement of overlap or clumping. MODIS VCF overestimates underestimates tree cover where tree cover exceeds 7884 % (at the 95 % confidence interval) when neither overlap nor clumping is enforced, and overestimates where tree cover exceeds 9078% (at 5% confidence interval) when both overlap and clumping are enforced.



Figure 4. Landsat TCC percent tree cover versus percent tree cover from TROBIT field data, taking into account uncertainties associated with tree cover spatial distributions within a TCC pixel and/or field site. The 4 combinations are: (1) no overlap and no clumping, where tree canopies are randomly distributed within both pixels and site; (2) no overlap and maximum clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed within the field site; (3) with overlap and no clumping, where tree canopies are clustered in one area of the pixels, and randomly distributed within the pixels, but overlap substantially within the field site; and (4) with overlap and maximum clumping, where

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tree canopies are clustered to one side within the pixels, and overlap substantially within the site. The bolded dashed line in black shows the 1:1 relationship. The solid lines represent the median of the respective regressions (green for forest; orange for savanna; blackfor forest and savanna combined), and the thin lines represent the 5 and 95 %. confidence interval of their respective regression lines. The vertical error bars represent uncertainty introduced by clumping; the horizontal error bars represent the uncertainty introduced by overlap.

Sinilar patterns can be observed with Landsat TCC (black line, Fig. 4). There is a significant underestination of tree cover in the lower cover ranges up to 59% when there is enforced overlap, and up to 82% when overlap is not enforced. In savanna sites (orange line, Fig. 4) the underestimation (at 95% confidence) is significant and consistent for covers below 75 - 80% (without enforced overlap) or below 52 - 60% (with enforced overlap). In forest sites (green line, Fig. 4) there is no systematic difference.

3.2 Global estimates of post-calibration change in tropical tree cover

and swama land cover classes we identified as being either forest or based or based

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(Fig. <u>2A3</u>), and the regions where all four scenarios agree on the direction of change (positive and negative) are substantial-(<u>Fig.5</u>). However, there are some differences caused by the uncertainty introduced by different extents of overlap and clumping. While we see a significant increase in tree cover across all clumping-overlap combinations in many regions of tropical savannas and grasslands (Pennington et al., 2018), such as in the forest-savanna mosaics that surround Congolian rainforests, we do not see the same pattern in the Cerrado of Brazil. This is likely because the African forest-<u>savanna regions fall within the range of MODIS VCF values that consistently undergo a positive calibration (~30-50 %, see Fig. A2), while the Cerrado of Brazil does not.</u>

analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help improve confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

As our calibrations were based on a limited number of sites in a limited number of regions, it is important to not that then as shown in Figure 2 are not definited. Indeed, it to the used bit of figure 5 and A2 are to fine R instrument in the Sahel post-calibration in multiple scenarios, which runs counter to the results of Brandt et al. (2020) who found that tree cover was underestimated in this region. This disparity may be due to our lack of field sites in these more arid regions, further highlighting the importance of more in-situ data for more accurate and precise calibration. Therefore, our calibrations are most useful in identifying areas where MODIS VCF estimates may be more or less reliable.

When looking at our calibration in more detail, we see that MODIS VCF significantly underestimates tree cover in all the IGBP land cover classes that we considered, regardless of overlap or clumping (95% confidence interval)(Fig. 6). The most substantial and significant underestimation is in the classes 'woody savannas' and 'savannas' The underestimation is the largest in woody savannas, except when clumping and overlap are enforced (in purple, Fig. 6). This is because the peak in the tree cover frequency distribution for savannas aligns with where the calibration for maximum overlap and clumping is the largest (i.e., at about 20% tree cover, see <u>Fig. A4</u>), while the peak in cover distribution for woody savannas (26 - 67%, Fig. A4) aligns with the cover range that undergoes the greatest change (Fig. 6) in the other clumping and overlap scenarios.

<u>'Open shrublands' only show a small underestimation of tree cover, despite its woody cover definition (10 - 60</u> <u>%) matching the range where MODIS VCF most underestimates tree cover (26 - 67 % cover). The discrepancy</u> <u>may be because the majority of the 'open shrublands' class commission error is with the 'grasslands' class (see</u> <u>Table S6 in Sulla-Menashe et al., 2019). The MODIS VCF tree cover in areas classified as 'open shrublands' is</u>

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border), Left: (No border), Right: (No border), Between : (No border), Tab stops: 3.13", Centered + 6.27", Right therefore likely to be lower than the IGBP definition would suggest (see Fig. A4), resulting in calibrations that are more conservative.

We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, closes hrubland and open shrubland).

4 Discussion

While MODIS VCF is a powerful and accessible tool to map tree cover, our field data-based calibrations indicate that the latest MODIS VCF collection 6 is missing a lot of woody cover, even when uncertainty introduced by site canopy overlap and clumping within the MODIS VCF pixel are accounted for. The Landsat TCC product, which may be viewed as an alternative with a higher spatial resolution, behaves in a similar manner. Our map (Fig. 5, top) highlights that this potential underestimation of woody cover is mainly occurring in tropical savannas. Moreover, the highest underestimation in the savanna classes occurs when there is no enforced overlap (Fig A3) (i.e. when there is a uniformrandom distribution of trees) which is the scenario that most likely reflects the TROBIT savanna plots. This is evidenced by work done by Veenendaal et al. (2015), where TROBIT plots were tested for complete spatial randonness and only minor indications of overlap were found. Woody savannas, as an example, may have their tree cover underestimated by up to 32 % (95 % confidence) when neither clumping nor overlap is enforced (grey tones, Fig. 6). If our results are representative of the tropics, then overall, MODIS VCF may be underestimating tropical tree cover by between 7 - 29 % for unenforced clumping and overlap or 0 - 21 % for when either clumping or overlap are enforced (5 - 95 % confidence).

An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al. 2016). According to its definition. MODIS VCF only maps trees that are 5 m or taller (Hansen et al. 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm as well as trees with a height exceeding 1.5 m when dbh <2.5 cm

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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Figure 36. Percent change in tree cover alter the optical and the optical matrix (betwice: merificed change or overlap (blue); no enforced changing and enforced overlap (red); enforced changing and overlap (pink) in the 'forest' super category and the 5 savanna classes. Palest tone indicates positive change, and tone indicates net change. Err or bars denote the 5-95% confidence indicates net with a val; if the error bar extends past the x-axis, the post-correction calibration change is not considered significant.

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower height ranges. Instead₇ we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Ruttern et al., 2015); (Ruttern et al., 2015), more research needs to be done with res

class definition of the MCD12Q1 product (Fig. A_{5}), A4) which again suggests that the 5 m

always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and 'closed shrublands' classes of the MCD12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or, some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, accounted accounted actually be and actually be actual

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net).</u>

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savanna regions fall within the range of MODIS VCF values that consistently undergo a positivecorrection (~ 30 - 50 %, see Fig. A4), while the Cerrado of Brazil does not.

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analysis when our calibration is broadly applied across the highly variable tropical forest-savanna ecotone. By multiplying the uncertainty range of our calibrations with the geographical distance to the closest sampled TROBIT site, we identified priority regions for further field surveying (Fig. 5 bottom). We found Southeast Asia, Central America, and Mexico are areas where additional in-situ observations would greatly help inprove confidence. Field data from the northwestern region of South America, the southeast of the African continent, and Madagascar would also help.

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We found significant increases in tree cover for ' forests' in every calibration scenario, though net change is not significant (95 % confidence) when overlap is enforced. This can be explained by the presence of both negative and positive calibrations in the higher ranges of tree cover when overlap is enforced. Similarly, the net change is insignificant across all clumping and overlap scenarios for the IGBP classes matching the lower ranges of tree cover (grassland, closes hrubland and open shrubland).

4 Discussion

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An overestimation at the lower end of the cover (<20 %) (Hansen et al., 2002; Sexton et al., 2013) and underestimation in the lower to middle range of cover (20 % - 60 %) have been identified in validations of previous MODIS VCF collections (Gross et al., 2018; Yang and Crews, 2019) and Landsat TCC version (Montes ano et al., 2016). According to its definition, MODIS VCF only maps trees that are 5 m or taller (Hansen et al., 2003), while the TROBIT CAL includes all trees with a minimum dbh of 2.5 cm as well as trees with a height exceeding 1.5 m when dbh <2.5 cm

cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.

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37014 calbration (cbckwie: no enforced clumping or overlap (black); enforced Figure 36. Percent change in tree cover clumping and no enforced overlap (blue); no enforced clumping and enforced overlap (red); enforced clumping and overlap (pinl) in the 'forest' supercategory and the 5 savanna classes. Palest tone indicates positive change, mid-tone indicates negative change, and the darkest tone indicates net change. Error bars denote the 5-95% confidence $interval; if the \, error \, bar \, extends \, past \, the \, x-axis, the \, post- {}^{\rm corroction} {}^{\rm calibration} \, change \, is \, not \, considered \, significant.$

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower height ranges. Instead, we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 mheight in the TROBIT inventory does not fully explain the observed underestimation.

subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Rutten et al., 2015), (Rutten et al., 2015), more research needstobedone with more insituheight data

class definition of the MCD12Q1 product (Fig. $\frac{A5}{A4}$) which again suggests that the 5 m

always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and ' closed shrublands' classes of the MCD 12Q1 product (Fig. A5A4), even though the height range for these classes

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is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or_a some combination of both.

MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to the future use of this product to the prove its useability.

within-field-site and field-site-pixel variation are accounted for during validation incalibration. We asofund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net).</u>

Site. Name	Country	Latitud 9 <u>Latitud</u> ६	Longitu deLongit ude,	MODI S VCF Tree Cover.(%)	<mark>Canop</mark> ¥Canop yArea Jndex	Average Upper, Stratum, Height (m)	Cover Type	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tallforest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	<u>∗</u> 1.17	1.33	0.52	12.53	Savanna	Savanna woodland
BBI-02	Burkina Faso	12.73	1 .16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	* 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl ou
BDA-02	Burkina Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Savanna	Shrub-rich sa vanna woodl
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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4 Discussion

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MODIS VCF could be the between-class onission and commission errors (Fig. 4, and Table S6 in Sulla-Menashe et al., 2019). For example, the accuracy for 'closed shrublands' is particularly low. It is mainly confused with 'openshublands', 'woodycavanas' and 'cavanas' and 'cavanas'. The miority of the 'openshublands' class commission error is with the 'grasslands' class₇ and there is confusion to a lesser extent between 'open chubland,'shrubland', 'woody savannas' and 'cavanas'. Also, the 'cropland/naturalvegetation nusaics' classisofen mapped as 'closed chrubland,'shrubland', 'woody cavanas', 'savannas', 'savannas', 'savannas' or 'grasslands'.

More work needs to be done to evaluate how effective both MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Landsat TCC</u> and MCD12Q1 are used for global model calibration or validation.

despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

We suggest that, while MODIS VCF <u>giveSand landsat TCC give</u> a good overview of tree cover on a global scale, <u>itom</u> should be <u>coalised before it is used as a construction of the product stocharacterise</u>, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of <u>the product these products</u>. The ecosystem functions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al., 2013) and even slight errors may create is sues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the

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canying capacity of a unit of land <u>may culd</u> be geater than previously thought <u>The MODIS VCF control on Culturation</u> could also esultina nore uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF on a largescale using fielddatasa reference do prosentpresents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g-, Eial at al., 2006; K orhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or LandsatTCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore artigitiges of the general state of the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002)</u>. Datacharcterising these atthe plot level would help interface of the African continent.

recommended (Gross et al., 2018; Lary and Lait, 2006, <u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to and improve its useability.

within-field-site and field-site-pixel variation are accounted for during validation incalibration. We asofund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net).</u>

Site		Latitud ØLatitud	Longitu de <u>Longit</u>	MODI S VCF Tree	Canop <u>yCanop</u> yArea	Average Upper Stratum	Cover	
Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	∗ 1.17	1.33	0.52	12.53	Savanna	Savannawoodland
BBI-02	Burkina Faso	12.73	, 1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	* 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl. 10
BDA-02	Burkina Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Şavanna	Shrub-rich sa vanna woodl n
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Fore st	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tallsavannawoodland		Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland		Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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Landsat TCC. In fact Montesanoet al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 mthreshold to our observed underestimation.



Figure <u>36</u>. Percent change in tree cover after the gric description decourt of the set of the set

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower height ranges. Instead₇ we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>AGA5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

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subordinate strata composition (and canopy cover thereof) varies widely depending on factors including ecosystemtype and altitude (Rutten et al., 2015); (Rutten et al., 2015), more research needs to be done with more insidule in

class definition of the MCD12Q1 product (Fig. <u>A5), A4</u>) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and 'closed shrublands' classes of the MCD12Q1 product (Fig. <u>A5A4</u>), even though the height range for these classes is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or_a some combination of both.

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More work needs to be done to evaluate how effective both MODIS VCF. <u>Lands at TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Lands at TCC</u> and MCD12Q1 are used for global model calibration or validation.

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2013) and even slight errors may create issues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

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north-western region of South America, the southeast of the African continent, and Madagascar would also help, the accuracy of remotely-sensed products like MODIS VCF <u>and Landsat TCC</u> (Hansen et al., 2003; Huete et <u>al.,</u> <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002).</u> Datacharcerising these atthe plot level would help identify potential confounding factors affecting MODIS VCF performance, and so help further constrain uncertainties.

recommended (Gross et al., 2018; Lary and Lait, 2006<u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to the future use and improve its useability.

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within-field site and field site pixel variation are accounted for during validation in calibration. We also fund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation

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Appendix

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Site.Latitud OLatitudLongitu doLongitu doLongituMODI SVCFCanop YCanopAverage Upper, Tree, VAreaAverage YCanopIndexStratum Upper, Tree, VAreaCover, VareaCover, 					-				
ALC-01 Brazil -2.53 -54.91 12.5 0.32 6.56 Savanna Savannawoodland ALF-01 Brazil -9.6 -55.94 77 2.31 37.02 Forest Tallforest ALF-02 Brazil -9.58 -55.92 76 2.65 41.32 Forest Tallforest ASU-01 Ghana 7.14 -2.45 41.33 2.54 45.27 Forest Tallforest BBI-01 Burkina, Faso 12.73 1.17 1.33 0.52 12.53 Savanna Savannawoodland BBI-02 Burkina, Faso 12.73 1.16 1.5 0.99 13.6 Savanna Savannawoodland BDA-01 Burkina, Faso 10.94 13.15 6.17 0.3 14.53 Savanna Shrub-rich savannawoodland BDA-02 Burkina, Faso 10.94 13.15 4.5 0.18 14.47 Savanna Shrub-rich savannawoodland BF1-01 Ghana 7.71 -1.69 15 1.22 29.67 Savanna Tallclosed woodland BF1-02	Site Name	Country	Latitud 9Latitud ६	Longitu deLongit ude,	MODI S VCF Tree Cover (%)	<mark>Canop</mark> ¥Canop yArea Jndex	Average Upper Stratum Height (m)	Cover. Type	TROBIT Site Description
ALF-01 Brazil -9.6 -55.94 77 2.31 37.02 Forest Tallforest ALF-02 Brazil -9.58 -55.92 76 2.65 41.32 Forest Tallforest ASU-01 Ghana 7.14 -2.45 41.33 2.54 45.27 Forest Tallforest BBI-01 Burkina Faso 12.73 1.17 1.33 0.52 12.53 Savanna Savannawoodland BBI-02 Burkina Faso 12.73 1.16 1.5 0.99 13.6 Savanna Savannawoodland BDA-01 Burkina Faso 10.94 .3.15 6.17 0.3 14.53 Savanna Shrub-rich savannawoodland BDA-02 Burkina Faso 10.94 .3.15 4.5 0.18 14.47 Savanna Shrub-rich savannawoodland BFI-01 Ghana 7.71 -1.69 1.22 29.67 Savanna Tallforest BFI-02 Ghana 7.71 -1.69 1.28 1.08	ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-02 Brazil -9.58 -55.92 76 2.65 41.32 Forest Tallforest ASU-01 Ghana 7.14 -2.45 41.33 2.54 45.27 Forest Tallforest BBI-01 Burkina,Faso 12.73 1.17 1.33 0.52 12.53 Savanna Savannawoodland BBI-02 Burkina,Faso 12.73 1.16 1.5 0.99 13.6 Savanna Savannawoodland BDA-01 Burkina,Faso 10.94 .3.15 6.17 0.3 14.53 Savanna Shrub-rich savannawoodland BDA-02 Burkina,Faso 10.94 .3.15 4.5 0.18 14.47 Savanna Shrub-rich savannawoodland BF1-01 Ghana 7.71 -1.69 1.5 1.22 29.67 Savanna Tallforest woodland BF1-02 Ghana 7.71 -1.69 1.283 1.08 28.2 Savanna Tallforest woodland BF1-03 Ghana 7.71 -1.69 12.83 1.08 28.2 Savanna Tallforest GH1-03	ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ASU-01 Ghana 7.14 -2.45 41.33 2.54 45.27 Forest Tallforest BBI-01 Burkina,Faso 12.73 1.17 1.33 0.52 12.53 Savanna Savannawoodland BBI-02 Burkina,Faso 12.73 1.16 1.5 0.99 13.6 Savanna Savannawoodland BDA-01 Burkina,Faso 10.94 .3.15 6.17 0.3 14.53 Savanna Shrub-rich savannawoodland BDA-02 Burkina,Faso 10.94 .3.15 6.17 0.3 14.53 Savanna Shrub-rich savannawoodland BDA-02 Burkina,Faso 10.94 .3.15 4.5 0.18 14.47 Savanna Shrub-rich savannawoodland BFI-01 Ghana 7.71 -1.69 1.22 29.67 Savanna Tallforest BFI-02 Ghana 7.71 -1.69 1.283 1.08 28.2 Savanna Tallforest BFI-03 Ghana 7.71 -1.7 25.83	ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Fore st	Tall forest
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BBI-02Burkina Faso12.731.161.50.9913.6Savanna </td <td>BBI-01</td> <td>Burkina Faso</td> <td>12.73</td> <td><u>∗</u>1.17</td> <td>1.33</td> <td>0.52</td> <td>12.53</td> <td>Savanna</td> <td>Savannawoodland</td>	BBI-01	Burkina Faso	12.73	<u>∗</u> 1.17	1.33	0.52	12.53	Savanna	Savannawoodland
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BFI-01 Ghana 7.71 -1.69 15 1.22 29.67 Savanna Tallelosed woodland BFI-02 Ghana 7.71 -1.69 12.83 1.08 28.2 Savanna Talleavannawoodland BFI-03 Ghana 7.71 -1.7 25.83 2.54 45.07 Savanna Talleavannawoodland CTC-01 Australia -16.1 145.45 72.67 2.35 40.37 Forest Tallforest	BDA-02	Burkina Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Şavanna	Shrub-rich savanna woodl
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DCK-02	Australia	-17.03	145.6	65.67	0./1	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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This could explain our observed underestimation in the lower tree cover ranges for both MODIS VCF and Landsat TCC. In fact Montesano et al. (2016) showed an improved match between Landsat TCC and their lidar-derived tree cover reference data when reducing the height threshold from 5 m to 2 m. However, because of how our field reference CAI is derived, we were not able to conclusively link the 5 m threshold to our observed underestimation.



Figure <u>36</u>. Percent change in tree cover after the gric description decourt of the set of the set

difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower height ranges. Instead_T we found a low R² and a mixture of under and overestimations in heights between 0 and 10 m (Fig. <u>A6A5</u>). This suggests that the inclusion of trees below 5 m height in the TROBIT inventory does not fully explain the observed underestimation.

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More work needs to be done to evaluate how effective both MODIS VCF. <u>Lands at TCC</u> and MCD12Q1 are at implementing the height thresholds in their respective 'tree' definitions, as this may have implications when MODIS VCF. <u>Lands at TCC</u> and MCD12Q1 are used for global model calibration or validation.

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We suggest that, while MODIS VCF <u>givesand landsat TCC give</u> a good overview of tree cover on a global scale, it<u>oth</u>shouldber collocated before it is used control of the product stocharacterise, as solitary trees in the landscape tend to be missed by global tree cover products (Jung et al., 2006, Brandt et al., 2020). The poor performance of MODIS VCF <u>and Landsat TCC</u> in savannas in particular (Gaughan et al., 2013; Gross et al., 2018; Kumar et al., 2019) emphasises the importance of continuous independent validation and re-calibration of the product_these products. The ecos ystemfunctions of savannas can vary drastically with just a slight difference in tree cover (Gaughan et al.,

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2013) and even slight errors may create issues in how we interpret the state and dynamics of the biome, which in turn affects how the land is managed.

VCF to estimate tree cover in agricultural land. As this tree cover is likely to have been underestimated substantially, the derived available land space for replanting may be less than projected, with the restoration potential overestimated. However, our results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the canying capacity of a unit of land may could be greater than projected in the restoration could also results also indicate an underestimated tree cover in woodier savannas and forests. Accounting for this, the restoration potential could actually be greater than anticipated, as because the canying capacity of a unit of land may could be greater than provided by the greater than a transition could also resulting more uniform cover distribution across regions, producing a more gradual transition between low-cover savannas and high-cover forests. This could have implications for work that, for example, uses MODIS VCF to study forest-savanna dynamics and bi-stability (Lasslop et al., 2018; Wuyts et al., 2017; Xu et al., 2016).

should be calibrated for use in the target region. However, calibrating MODIS VCF-on a largescale using fielddatas a reference do prosent presents several challenges. Firstly, different in-situ measurement techniques tend to measure different types of tree cover (e.g., Fiala et al., 2006; Korhonen et al., 2006; Rautiainen et al., 2005) and each will require a specific conversion method to enable direct comparison with MODIS VCF-or Landsat TCC. For example, Montesano et al. (2016) 's comparison did not acknowledge VCF's and thus TCC's 'within canopy gaps,' which may explain their observed underestimation in covers above 80%. In our case, to account for gaps between tree crowns, we applied the

and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

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north-western region of South America, the southeast of the African continent, and Madagascar would also help, the accuracy of remotely-sensed products like MODIS VCF <u>and Landsat TCC</u> (Hansen et al., 2003; Huete et <u>al.,</u> <u>1997; Snith et al., 2002al., 1997; Smith et al., 2002).</u> Datacharcerising these atthe plot level would help identify potential confounding factors affecting MODIS VCF performance, and so help further constrain uncertainties.

recommended (Gross et al., 2018; Lary and Lait, 2006<u>Montesano et al. 2016</u>), though without a large-scale effort to re-calibrate MODIS VCF and products trained using VCF like Landsat TCC, the question of how appropriate MODIS VCF is for use in both forests and savannas in the tropics will remain. By highlighting the extent to which MODIS VCF struggles to estimate tree cover in tropical forests and savannas, we hope to inform the future use of this product to the future use and improve its useability.

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within-field site and field site pixel variation are accounted for during validation in calibration. We also fund that using MODE VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of ecological research including vegetation

estimating restoration potential, and identifying forest-savanna bimodality, we stress that more independent work on validating and re-calibrating is required before its tree cover estimates can be relied upon in the tropics.

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Appendix

Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net).</u>

Site. Name	Country	Latitud 9 <u>Latitud</u> ६	Longitu deLongit ude,	MODI S VCF Tree Cover.(%)	<mark>Canop</mark> ¥Canop yArea Jndex	Average Upper, Stratum, Height (m)	Cover Type	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tallforest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	<u>∗</u> 1.17	1.33	0.52	12.53	Savanna	Savanna woodland
BBI-02	Burkina Faso	12.73	1 .16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	* 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl ou
BDA-02	Burkina Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Savanna	Shrub-rich sa vanna woodl
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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Site		Latitud 9Latitud	Longitu de <u>Longit</u>	MODI S VCF Tree	Canop <u>yCanop</u> yArea	Average Upper Stratum	Cover	
Name	Country	e,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	∗ 1.17	1.33	0.52	12.53	Savanna	Savannawoodland
BBI-02	Burkina Faso	12.73	, 1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	* 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl. 10
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BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Fore st	Tall forest
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On the other hand, when looking at the relationship between TROBIT's upper stratum canopy height and the difference between TROBIT and MODIS VCF we would have expected an increasing underestimation in the lower height ranges. Instead, we found a low R^2 and a mixture of under and overestimations in heights between 0 and 10 m (Fig. AGA5). This suggests that the inclusion of trees below 5 mheight in the TROBIT inventory does not fully explain the observed underestimation.

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class definition of the MCD12Q1 product (Fig. A5), A4) which again suggests that the 5 m always apply in MODIS VCF. For example, MODIS VCF recorded tree cover in the 'open shrublands' and 'closed shrublands' classes of the MCD12Q1 product (Fig. A5A4), even though the height range for these classes is 1 - 2 m For the 'savannas' class, MODIS VCF yields a percent tree cover range that matches closely with the 'savannas' class definition (between 10% and 30%), despite the differing tree thresholds for MODIS VCF and IGBP (5 m minimum for MODIS VCF, and 2 m minimum for IGBP). These discrepancies suggestone of the following three things: 'open/closed shrublands' and 'savannas' contain trees taller than 5 m, MODIS VCF is distinguishing trees below the 5 m threshold; or_a some combination of both.

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ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	∗ 1.17	1.33	0.52	12.53	Savanna	Savannawoodland
BBI-02	Burkina Faso	12.73	, 1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	* 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl. 10
BDA-02	Burkina Faso	10.94	<u></u> 3.15	4.5	0.18	14.47	Şavanna	Shrub-rich sa vanna woodl n
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Fore st	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by
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despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for</u> MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

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and resulting tree cover could vary widely on a plot-by-plot basis (Lloyd et al., 2008). With further in-situ data that describe tropical vegetation type-specific GCF variation, we may be able to incorporate site-specific GCFs into our analysis.

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Figure A1. Location of sampling sites in Africa, Australia, and South America from the TROBIT Project (based on Fig. 2, Torello-Raventos et al., 2013) shown on MODIS VCF (DiMiceli, 2017). Of the 63 field sites, only the 48 sites with available GPS coordinates were selected-<u>(https://www.forestplots.net).</u>

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Overall, our results suggest that the biases found in the previous collections may have persisted in collection 6, despite reported improvement in accuracy (DiMiceli et al., 2017). This indicates that the biases introduced by binning the training data (Gerard et al. 2017) and using a CART (Classification and Regression Tree) model (Hanan et al., 2013) are inherent and still present within this version of MODIS VCF. <u>Similar results for MODIS VCF and Landsat TCC also suggest that by training TCC with VCF tree cover these biases have been propagated into the finer-scale product</u>. Models calibrated using MODIS VCF (Brandt et al., 2017; Lasslop et al., 2020; Burton et al., 2019; Kelley et al., 2019, 2020/2021) also risk inheriting these biases and should therefore be validated using other sources of data.

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		Latitud	Longitu	MODI S VCF	WCanon	Average		
Site		e Latitud	deLongit	Tree	v Area	Stratum	Cover.	
Name	Country 2	<u>e</u> ,	ude,	Cover (%)	Index	Height (m)	Туре	TROBIT Site Description
ALC-01	Brazil	-2.53	-54.91	12.5	0.32	6.56	Savanna	Savannawoodland
ALF-01	Brazil	-9.6	-55.94	77	2.31	37.02	Forest	Tall forest
ALF-02	Brazil	-9.58	-55.92	76	2.65	41.32	Forest	Tall forest
ASU-01	Ghana	7.14	-2.45	41.33	2.54	45.27	Forest	Tall forest
BBI-01	Burkina Faso	12.73	<u>∗</u> 1.17	1.33	0.52	12.53	Savanna	Savanna woodland
BBI-02	Burkina Faso	12.73	- 1.16	1.5	0.99	13.6	Savanna	Savanna woodland
BDA-01	Burkina Faso	10.94	<u>-</u> 3.15	6.17	0.3	14.53	Savanna	Shrub-rich sa vanna woodl
BDA-02	Burkina Faso	10.94	<u>-</u> 3.15	4.5	0.18	14.47	Savanna	Shrub-rich savanna woodl
BFI-01	Ghana	7.71	-1.69	15	1.22	29.67	Savanna	Tall closed woodland
BFI-02	Ghana	7.71	-1.69	12.83	1.08	28.2	Savanna	Tall savanna woodland
BFI-03	Ghana	7.71	-1.7	25.83	2.54	45.07	Savanna	Tall savanna woodland
CTC-01	Australia	-16.1	145.45	72.67	2.35	40.37	Forest	Tall forest
DCR-01	Australia	-17.02	145.58	21.67	1.67	27.19	Savanna	Tall savanna woodland

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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman, Not Expanded by / Condensed by
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There is also the uncertainty associated with the field data collection. In our case, the site-specific CAI standard errors (supplement B in Torello-Raventos et al., 2013) are small and show no systematic bias and are therefore

north-western region of South America, the southeast of the African continent, and Madagascar would also help. the accuracy of remotely-sensed products like MODIS VCF <u>and Landsat TCC</u> (Hansen et al., 2003; Huete et <u>al.</u> <u>1997; Snith et al.</u> 2002al., 1997; Smith et al., 2002); Datacharacterising these atthe plot level would help identify potential confounding factors affecting MODIS VCF performance, and so help further constrain uncertainties.

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Site. Name	Country	Latitud 9 <u>Latitud</u> ६	Longitu deLongit ude,	MODI S VCF Tree Cover.(%)	<mark>Canop</mark> ¥Canop yArea Jndex	Average Upper, Stratum, Height (m)	Cover Type	TROBIT Site Description
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DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woodland	Formatted: Font: Times New Roman
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Finally, factors such as cloud cover, landscape heterogeneity, phenology, vegetation type, and soil type affect the accuracy of remotely-sensed products like MODIS VCF and Landsat TCC (Hansen et al., 2003; Huete et al., 1997; Snith et al., 2002al., 1997; Snith et al., 2002b). Datacharcerising these atheptotevel would help identify potential confounding factors affecting MODIS VCF performance, and so help further constrain uncertainties.

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We found that MODIS VCF significantly underestimates tree cover in tropical forests and savannas, even when within field site and field site-pixel variation are accounted for during validation in calibration. We also found that using MODIS VCF for training likely propagates these biases, even in the finer-scale Landsat TCC. As MODIS VCF is a product that is commonly used in a wide variety of cological research including vegetation	Formatted: Normal, Indent: Left: 0", Space Before: 12 pt, After: 12 pt, Pattern: Clear (White) Formatted									
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	DCR-02	Australia	-17.03	145.6	65.67	0.71	22.51	Savanna	Tall savanna woo	odland	P	Formatted	
	EKP-01	Australia	-18.07	145.99	43.5	0.74	28.13	Savanna	Tall savanna woo	odland		Formatted	
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FLO-01	Brazil	-12.81	-51.85	65.67	2.4	28.21	Forest		Forest	-	(Formatted	
FMS-01	Australia	-18.09	144.84	7.67	0.32	20.03	Savanna	Shrub-rio	ch sa vanna woodland		{	Formatted	
FMS-02	Australia	-18.11	144.82	44.17	1.21	16.69	Forest	Stunte	d shrub-rich forest			Formatted	
HOM-01	Mali	15.34	-1.47	0.5	0.05	3.87	Savanna	Sav	anna grassland		Ľ,	Formatted	
HOM-02	Mali	15.33	-1.55	0.83	0.16	6.13	Savanna	Sav	anna grassland			Formatted	
IBG-01	Brazil	-15.95	-47.87	20.83	0.22	7.48	Savanna	S	crub sa vanna			Formatted	<u></u>
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TRG-03	Brazil	-15.93	-47.87	20.5	0.12	8.01	Savanna	S	crub savanna		1	Formatted	
IBG-04	Brazil	-15.94	-47.86	27.17	0.77	12.65	Savanna	Sav	anna woodland			Formatted	
KBL-01	Australia	-17.77	145.54	75	1.69	39.5	Forest		Tallforest			Formatted	
KBL-02	Australia	-17.85	145.53	61.17	0.81	29.2	Savanna	Talls	avanna woodland		ł	Formatted	
KBL-03	Australia	-17.69	145.53	79.5	3	36.62	Fore st		Tallforest			Formatted	<u></u>
KCR-01	Australia	-17 11	145.6	78.83	2 44	42.37	Forest		Tall forest			Formatted	<u></u>
LFB-03	Bolivia	-14.6	-60.85	28.17	0.39	9.93	Savanna	Shrub-rio	ch sa vanna woodland	-		Formatted	<u></u>
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MDJ-01	Cameroon	0.17	12.85	42	5.24	43	Folest	_	Tainforest	-		Formatted	
MDJ-02	Cameroon	6.16	12.82	18.67	0.44	16.13	Savanna	Lon	g-grass savanna	-		Formatted	
MDJ-03	Cameroon	5.98	12.87	64.67	2.97	36.53	Forest	Stunte	d_shrub-rich_forest	-		Formatted	
MDJ-04	Cameroon	6	12.87	15	0.37	18.93	Savanna	Lon	g-grass savanna			Formatted	<u>_</u>
MDJ-05	Cameroon	5.98	12.87	70.33	2.85	21.27	Forest	Stunte	d shrub-rich forest	-	ľ	Formatted	
MDJ-06	Cameroon	6	12.89	20.5	0.68	15.27	Savanna	Lon	g-grass savanna	-		Formatted	
MDJ-07	Cameroon	6.01	12.89	57.33	1.75	42.67	Fore st		Tallforest			Formatted	
MDJ-08	Cameroon	6.21	12.75	15	0.48	18	Savanna	Lon	g-grass savanna			Formatted	 [
MLE-01	Ghana	9.3	-1.86	10	0.34	14.67	Savanna	Sav	anna woodland			Formatted	
NXV-02	Brazil	-14.7	-52.35	20.83	1.82	15.76	Savanna	Tall	closed woodland	-		Formatted	
RSC-01	Australia	-20.16	146.54	28	1.15	13.14	Forest	s	Stunted forest			Formatted	
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SMT-02	Brazil	-12.82	-51.77	41.5	1.44	14.64	Savanna	Sav	anna woodland			Formatted	C
SMT-03	Brazil	-12.83	-51.77	19.33	0.53	11.19	Savanna	Sav	anna woodland	-		Formatted	
TUC-01	Bolivia	-18.52	-60.81	50.33	1.29	14.9	Forest	S	Stunted forest			Formatted	
TUC-02	Bolivia	-18.53	-60.63	21.67	0.81	12.05	Savanna	Shru	b-rich woodland			Formatted	
TUC-03	Bolivia	-18.19	-60.86	10.83	0.37	14.11	Savanna	Sav	anna woodland			Formatted	
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Table A1. Site names, locations, Canopy Area Index values, MODIS VCF percent tree cover values, cover type, and TROBIT site descriptions of the 48 TROBIT Project plots used in this study. TROBIT site descriptions are based on Table S1 of Veenendaal et al., 2015.

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Figure A2. Visual representation of the effects of enforcing overlap within a (100 m x 100 m) TROBIT site with a



Figure A3: The change in tree cover post-calibration for all four scenarios. Blackdots indicate areas where the postcalibration values have a 95% certainty of being positive (increasing cover) or negative (decreasing cover) calibrations. These uncertainty maps are indicators of areas where MODIS VCF estimates may be more or less reliable, and cannot be used as definitive calibrations due to the limited number of field sites used as reference.

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Figure A4. Frequency distributions of percent tree cover value as estimated by MODIS VCF across the 'forest' supercategory and the following IG BP classes that by our definition count as part of the 'savanna' domain: Closed Shrublands, Open Shrublands, Woody Savannas, Savannas, and Grasslands. Specific class definitions as per the User Guide for the MODIS Land Cover Product (Sulla-Menashe and Friedl, 2018).

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Figure A6A5: TROBIT plot upper stratumheight versus the difference between MODIS VCF and TROBIT percent tree cover infor the four clumping-and over lap scenarios. Upper and lower bars represent the uncertainty range's 10th and 90th percentile, respectively, based on the convolution of MODIS VCF and TROBIT cover uncertainties from Fig. 4-3.

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Code/Data Availability.

https://github.com/douglask3/VCF_vs_sites-revisionnumber fdda3ff

TROBIT and corresponding MODIS VCF values. DK and NDC.G. collated Sexton et al. (2013) values. D.K. and N.D. performed regression analysis and constructed global maps. RAR.A. wrote the first draft of the paper with input for D4DK and FGD4TG. DK plotted the figures MTR, EV, TRF, Q.P.S. FSHT, BM/TD, AGD, and J.K. carried out the extensive

responsible for field dataquality checking and digitising. RA, DK, FGRA, D.K., FG. and NDND. contributed to the final manuscript.

Earth SystemModelling Project (UKESM, grant no. NE/N017951/1), FGF.G. was supported by the SUNRISE project (grant no. NE/R000131/1). <u>NDN.D.</u> was supported by the Australia Research Council (DP170103410). <u>BSMB.S.M.</u> was supported by the National Council for Scientific and Technological Development (CNPq, 301153/2018-3).

2017-) in R 3.2.0 (R Core Team, 2015). Coastlines were obtained from maps v3.1.0 (Becker et al., 2016).

in the Australian field sites, and to Azemiyah Abdul Rahimfor her support during the production of this manuscript.

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