

1 Amazon Forest Structure Generates Diurnal and Seasonal 2 Variability in Light Utilization

3
4 Douglas C. Morton^{1*}, Jérémy Rubio^{1,2}, Bruce D. Cook¹, Jean-Philippe Gastellu-
5 Etchegorry², Marcos Longo³, Hyeungu Choi^{1,4}, Maria Hunter⁵, Michael Keller^{3,6}

6 [1] NASA Goddard Space Flight Center, Greenbelt, MD 20771 USA

7 [2] Centre d'Etudes Spatiales de la Biosphère (CESBIO) - UPS, CNES, CNRS, IRD,
8 Université de Toulouse, 31401 Toulouse cedex 9, France

9 [3] Embrapa Monitoramento por Satélite, CEP: 13070-115, Campinas, SP, Brazil

10 [4] Global Science & Technology Inc., Greenbelt, MD 20770 USA

11 [5] University of New Hampshire, Durham, NH 03824 USA

12 [6] USDA Forest Service, International Institute of Tropical Forestry, San Juan, PR 00926
13 USA

14 * Correspondence to: D.C. Morton (douglas.morton@nasa.gov)

15

16 Abstract

17 The complex three-dimensional (3D) structure of tropical forests generates a diversity of light
18 environments for canopy and understory trees. Understanding diurnal and seasonal changes
19 in light availability is critical for interpreting measurements of net ecosystem exchange and
20 improving ecosystem models. Here, we used the Discrete Anisotropic Radiative Transfer
21 (DART) model to simulate leaf absorption of photosynthetically active radiation (*APAR*) for
22 an Amazon forest. The 3D model scene was developed from airborne lidar data, and local
23 measurements of leaf reflectance, aerosols, and PAR were used to model *APAR* under direct
24 and diffuse illumination conditions. Simulated *APAR* under clear sky and cloudy conditions
25 was corrected for light saturation effects to estimate light utilization, the fraction of *APAR*
26 available for photosynthesis. Although the fraction of incoming PAR absorbed by leaves was
27 consistent throughout the year (0.80-0.82), light utilization varied seasonally (0.67-0.74), with
28 minimum values during the Amazon dry season. Shadowing and light saturation effects

Formatted: Font: Italic

Formatted: Font: Italic

Formatted: Font: Italic

Formatted: Font: Italic

1 moderated potential gains in forest productivity from increasing PAR during dry season
2 months when the diffuse fraction from clouds and aerosols was low. Comparisons between
3 DART and other models highlighted the role of 3D forest structure to account for seasonal
4 changes in light utilization. Our findings highlight how directional illumination and forest 3D
5 structure combine to influence diurnal and seasonal variability in light utilization, independent
6 of further changes in leaf area, leaf age, or environmental controls on canopy photosynthesis.
7 Changing illumination geometry constitutes an alternative biophysical explanation for
8 observed seasonality in Amazon forest productivity without changes in canopy phenology.

9

10 **1 Introduction**

11 Seasonal and interannual variability in vegetation productivity has profound impacts
12 on the global carbon cycle (e.g., Poulter et al., 2014; Zeng et al., 2014; Keppel-Aleks et al.,
13 2014; Le Quéré et al., 2013; Gatti et al., 2014; Schimel et al., 2015; Cleveland et al., 2015).
14 Understanding the mechanisms that link environmental variability and vegetation productivity
15 is particularly important to constrain projections of Earth system feedbacks under future
16 climate (e.g., Keppel-Aleks et al., 2014; Cox et al., 2013; Randerson, 2013; Boisier et al.,
17 2015). Under current climate conditions, few tropical forest regions experience temperature
18 limitations on biologic activity (Nemani et al., 2003), yet to the degree to which water and
19 light limit forest productivity remains controversial (e.g., Gatti et al., 2014; Morton et al.,
20 2014; Phillips et al., 2009; Restrepo-Coupe et al., 2013; Samanta et al., 2012; Doughty et al.,
21 2015; Guan et al., 2015). A detailed understanding of vegetation productivity over large
22 spatial scales has proven elusive, even using remote sensing data and ecosystem models (e.g.,
23 Cleveland et al., 2015), given limited data on how species diversity (ter Steege et al., 2013),
24 strategies for resource competition (e.g., Chave et al., 2010), and interactions between
25 human and natural systems contribute to spatial and temporal dynamics of tropical forest
26 productivity (e.g., Chen et al., 2010; Morton et al., 2013; Oliveira et al., 2007; Rap et al.,
27 2015).

28 A more detailed investigation of the underlying mechanisms of Amazon forest
29 productivity may offer new insights into the spatial and temporal variability in Amazon forest
30 functioning. Previous studies have collected detailed data on forest growth or net carbon
31 uptake to estimate seasonal (e.g., Saleska et al., 2003; Hutyyra et al., 2007; Restrepo-Coupe et
32 al., 2013; Malhi et al., 2015) or interannual variability in Amazon forest productivity

1 (Nepstad et al., 2007; Gatti et al., 2010; Gatti et al., 2014; Phillips et al., 2009; Doughty et al.,
2 2015). A separate line of analysis has analyzed satellite data on vegetation structure and
3 reflectance (e.g., Morton et al., 2014), solar-induced fluorescence (e.g., Joiner et al., 2011;
4 Parazoo et al., 2013; Lee et al., 2013; Guan et al., 2015), canopy moisture (Frolking et al.,
5 2011; Saatchi et al., 2013), rainfall (Lewis et al., 2011), terrestrial water storage (Chen et al.,
6 2013b), and fire (Chen et al., 2013a) to characterize Amazon forest responses to large-scale
7 environmental variability. Most previous studies directly compared environmental inputs and
8 forest carbon dynamics without a mechanistic approach to translate environmental variability
9 into differences in plant-available water or light (e.g., Huete et al., 2006; Lewis et al., 2011;
10 Brando et al., 2010; Doughty et al., 2015; Guan et al., 2015). In addition, few studies have
11 jointly considered seasonal dynamics in resource availability and forest responses beyond the
12 scale of forest inventory plots or tower footprints (e.g., Oliveira et al., 2007; Nepstad et al.,
13 2007; Doughty et al., 2015; da Costa et al., 2010; Asner et al., 2004; Stark et al., 2012),
14 highlighting the important role of ecosystem models for regional carbon flux estimates.

15 Improving the representation of complex forest canopies in ecosystem models is
16 essential to understand how variability in canopy illumination contributes to changes in
17 Amazon forest productivity. Many ecosystem models are structured to partition light and
18 water vertically, with only local consideration of horizontal resource competition (e.g.,
19 Moorcroft et al., 2001; Sitch et al., 2003; Krinner et al., 2005; Clark et al., 2011). The
20 influence of vertical structure on light availability is widely recognized (e.g., Moorcroft et
21 al., 2001; Clark et al., 2011; van der Tol et al., 2009; Gibelin et al., 2008; Sellers et al., 1992),
22 yet the influence of horizontal variability in forest structure on light availability is rarely
23 directly considered in dynamic global vegetation models (except see Scheiter et al., 2013).
24 Horizontal variability in forest 3D structure results from fine-scale processes of canopy
25 turnover and gap formation (e.g., Hunter et al., 2015; Asner et al., 2013); within-biome
26 variability in tropical forest structure may also reflect large-scale environmental, climatic, and
27 disturbance gradients (e.g., Morton et al., 2014; Saatchi et al., 2011; Baccini et al., 2012;
28 Malhi et al., 2006; Espírito-Santo et al., 2014). Models of intermediate complexity may not
29 fully account for shadowing and light saturation effects that alter light utilization—the degree
30 to which leaf absorbed photosynthetically active radiation (*JAPAR*) can be used for
31 photosynthesis.

32 Three-dimensional radiative transfer models offer the ability to quantify light
33 interactions in complex forest canopies at the scale of individual leaves. A range of

Formatted: Font: Italic

1 sensitivity studies highlight the importance of 3D structure for the representation of visible
2 and near-infrared scattering and absorption in forest canopies (Widlowski et al., 2011;
3 Romanczyk et al., 2013; Gastellu-Etcheberry and Trichon, 1998). Several previous studies
4 have used radiative transfer models to evaluate light absorption in tropical forest canopies and
5 the impact of structure on forest productivity (e.g., Guillevic and Gastellu-Etcheberry, 1999;
6 Alton et al., 2007a). There is broad interest in evidence for enhanced tropical forest
7 productivity under diffuse light conditions (Oliveira et al., 2007; Rap et al., 2015; Mercado et
8 al., 2009; Cirino et al., 2014; Alton et al., 2007b; Kanniah et al., 2012), but 3D radiative
9 transfer models have not been specifically used to evaluate the potential for seasonal changes
10 in tropical forest productivity based on the interactions between illumination geometry (direct
11 and diffuse) and tropical forest structure.

12 Here, we developed a detailed Amazon forest scene in the DART model (Gastellu-
13 Etcheberry et al., 2015) using airborne lidar data and in situ measurements of forest structure
14 and reflectance properties, PAR, and aerosols. The goal of this work was to evaluate the
15 influence of Amazon forest structure on leaf absorption and light utilization by explicitly
16 accounting for shadowing and light saturation under diurnal and seasonal variability in
17 illumination conditions. By targeting the mechanisms that link PAR availability, absorption,
18 and light utilization in a 3D forest canopy, this study evaluated the potential responses of
19 tropical forests to changing light conditions on seasonal or interannual time scales. This
20 detailed investigation of light absorption, including the distribution of sunlit and shaded
21 leaves, is also an important precursor for efforts to interpret global measurements of solar-
22 induced fluorescence (SIF, e.g., Joiner et al., 2011; Guan et al., 2015).

23

24 **2 Methods**

25 2.1 DART Model Simulations

26 Diurnal and seasonal changes in the 3D light environment of an Amazon forest were
27 simulated using the DART model (Gastellu-Etcheberry et al., 2012; Gastellu-Etcheberry et
28 al., 2015). The 3D forest scene (50 m × 50 m) was developed using high-density airborne
29 lidar data (>20 returns per m²) from the Tapajos National Forest near Santarém, Pará, Brazil,
30 and local measurements of leaf reflectance, litter reflectance, and leaf area (see Extended Data
31 Figure 1 in *Morton et al.* [2014]). Discrete return airborne scanning lidar data were thinned to
32 a consistent point density following methods described in (Leitold et al., 2015), and leaf area

1 (6 m²/m², (Asner et al., 2004)) was allocated to 1 m³ voxels based on the distribution of
2 multistop lidar returns, with LAI distributed equally among lidar returns (Figure 1). The
3 lidar digital terrain model was used to represent surface topography in the 3D scene.

4 Tree objects representing stems and branches were added for canopy trees to estimate
5 PAR absorption by non-photosynthetic canopy elements. Lidar-based estimates of tree height
6 and crown dimensions were used to scale a generic tree object to represent the stem and
7 branches for each canopy tree (Figure 1). Woody structures were represented in DART using
8 facets (triangles), so that light interception by branches and stems could be tracked separately
9 from absorption by leaves or the ground surface. The geometry of facets is independent of
10 voxel dimensions, allowing stems and woody branches to be represented at finer scales than
11 leaf voxels. Previous studies have documented the importance of large branches for
12 scattering and absorption of near-infrared energy in the forest canopy (Romanczyk et al.,
13 2013). The use of tree objects in this study builds on the sensitivity study by Romanczyk et
14 al. (2013) to investigate the impact of PAR absorption by woody branches within the canopy.
15 Branches are rarely studied in tropical forests (except see (Higuchi et al., 1998)), and many
16 radiative transfer and ecosystem models exclude light interactions with branches altogether.

17 DART simulations for cloudy and clear-sky conditions were run for five hours per day
18 and one day per month to produce 3D estimates of daily, monthly and annual leaf-absorbed
19 photosynthetically active radiation (IAPAR). DART simulations were configured to simulate a
20 repeating (infinite) scene. Illumination geometry, aerosol optical depth, and incident PAR
21 varied for each hourly simulation, but forest structure and reflectance properties were held
22 constant. Hourly estimates of incident PAR and cloud cover were based on PAR
23 measurements from the KM67 eddy flux tower within the Tapajos National Forest (8:00-9:00,
24 10:00-11:00, 12:00-1:00, 14:00-15:00, and 16:00-17:00 local time, (Hutyra et al., 2008),
25 Table S1). Cloud cover is highest during the wet season (December-~~June~~May), with higher
26 average incident PAR during dry season months (Jul~~yne~~-November). At 3°S latitude, midday
27 illumination conditions are near nadir in both March and September. Mean PAR values for
28 cloudy and clear-sky conditions were estimated based on the distribution of hourly
29 observations, with clear-sky conditions defined as hourly PAR values between 70-100% of
30 maximum PAR for each hour and month (Table S1). Mean incident PAR for cloudy
31 conditions was derived from observations <70% of maximum PAR. These thresholds are
32 similar to the approaches used in previous studies to interpret in situ measurements of incident
33 radiation when no diffuse PAR sensor was available (Oliveira et al., 2007). A monthly

1 | climatology of aerosol optical depth was developed using data from the Belterra AERONET
2 | station (Holben et al., 1998) ~~was used~~ to simulate diffuse light from aerosol scattering under
3 | clear-sky DART simulations using based on an updated atmospheric radiative transfer scheme
4 | in DART (Grau and Gastellu-Etchegorry, 2013). Cloudy conditions were simulated as 100%
5 | diffuse light. The treatment of illumination conditions as 0% or 100% cloudy is a convenient
6 | simplification that avoids the need to resolve cloud properties (e.g., optical thickness, size,
7 | altitude) and atmospheric transport—attributes that could be the basis for a further study
8 | where more detailed ground measurements are available, or using an Earth system model that
9 | simultaneously considers the impact of dynamic atmospheric processes on surface energy
10 | budgets. Combined hourly simulations were constructed using a weighted average of clear
11 | and cloudy DART model simulations for each hour and month.

12 | Model simulations tracked light interactions with leaves, woody elements, and the
13 | ground surface (Table S2). Estimates of *I*APAR ($\mu\text{mol m}^{-2} \text{s}^{-1}$) for each 1 m^3 voxel were post-
14 | processed to account for light saturation effects based on a photosynthetic light response
15 | curve from leaf-level measurements of tropical forest trees (*Anacardium*, (Kitajima et al.,
16 | 1997)). Light utilization is therefore a unitless measure of “effective *I*APAR,” based on the
17 | fraction of light absorbed by leaves that can be used for photosynthesis in the absence of
18 | constraints based on leaf temperature or moisture stress (e.g., (Doughty and Goulden, 2008).
19 | Fractional light utilization per unit leaf area decreased for light absorption $>225 \mu\text{mol m}^{-2} \text{s}^{-1}$,
20 | declining to approximately 0.8, 0.6, and 0.4 for *I*APAR values of 360, 450, and $825 \mu\text{mol m}^{-2}$
21 | s^{-1} , respectively (Figure 2). Throughout the manuscript, light saturation effects were
22 | calculated at the voxel scale and summed for the model scene based on the difference between
23 | absorbed and utilized light. Average light absorption by leaf material (turbid) in each voxel
24 | provided a conservative estimate of light saturation, since absorbed light was distributed
25 | across all leaf area in the voxel. Light saturation effects lead to lower light utilization based
26 | on the reduction in fractional light utilization above $225 \mu\text{mol m}^{-2} \text{s}^{-1}$, consistent with a shift
27 | from light to rubisco limitation of photosynthesis.

28 | 29 | 3.2 Model Comparisons

30 | DART model simulations of *I*APAR and light utilization were compared with two
31 | additional modeling approaches. *Stark et al.* (2012) used a light extinction model to estimate
32 | the vertical profile of light interception in Amazon forests. We used the vertical distribution
33 | of LAI across the DART scene and identical inputs for incident radiation (Table S1) to

1 estimate the profile of light absorption following the methods described in *Stark et al.* (2012).
2 We also used the Ecosystem Demography (ED) model (Version 2.2, ~~(Longo, 2014)~~) to
3 simulate the vertical profile of light absorption and light utilization. The vertical profile of
4 LAI was used to initialize an ED patch. Tower measurements of PAR (Hutyra et al., 2008)
5 and the site coordinates were used to simulate incident radiation. ED model simulations also
6 evaluated light absorption and utilization without moisture stress. Separate simulations
7 considered the influence of dividing the DART scene into 1, 25, and 2500 patches in ED to
8 evaluate the role of horizontal heterogeneity in forest structure on light absorption.
9 Representing the DART scene as a single ED patch ~~simulates the influence of the average~~
10 ~~forest structure is similar to a “big leaf” model.~~ ED simulations with 25 patches (100 m²) is a
11 typical representation of forest ~~size and age~~ structure in the model (Moorcroft et al., 2001),
12 while simulations with 2500 patches (1 m²) examined the potential to represent horizontal
13 heterogeneity in vertical structure using the ED modeling approach.

14

15 **3 Results**

16 Light availability in tropical forests is dynamic on diurnal and seasonal time scales.
17 Hourly distributions of *l*APAR at the voxel scale highlighted diurnal variability in leaf
18 absorption, including the fraction of leaves experiencing light-saturated conditions (Figure 33,
19 Table S2). Shadowing effects were pronounced in early morning (09:00) and late afternoon
20 (17:00) DART simulations, with most leaf voxels experiencing low *l*APAR. The degree of
21 shadowing changed seasonally, such that early morning overpass satellites (e.g., Terra
22 MODIS, GOME-2) observe large seasonal changes in shadowing and illumination of tropical
23 forests ~~(Figure 4a)~~, altering the overall distribution of light absorption at the leaf level and the
24 reflectance from sunlit and shaded leaves.

25 Seasonal variability in total *l*APAR at the canopy scale was driven by a combination
26 of changes in solar zenith angle and the fraction of diffuse light from clouds and aerosols
27 (Figure 34). Combined cloudy and clear-sky simulations showed highest total *l*APAR in
28 March and September, consistent with more even distribution of light under near-nadir
29 midday illumination conditions in these months. March simulations were characterized by a
30 more even distribution of *l*APAR but lower incident radiation under cloudy conditions.
31 Lower light utilization in September, based on light saturation effects under clear-sky
32 simulations, led to similar estimates of total utilized *l*APAR in both months (Figure 43).

1 These cases illustrate how different mechanisms interact with forest structure to alter light
2 availability for photosynthesis.

3 Shadowing, light saturation, and light absorption by woody branches reduced the
4 fraction of absorbed PAR available for photosynthesis. [At the scene scale](#), DART estimates of
5 monthly fractional *l*APAR varied between 0.8 and 0.85, depending on illumination conditions
6 (Figure [54b](#)). Fractional losses of 0.15 – 0.20 of incoming PAR resulted from the combined
7 influence of leaf reflectance (0.036 ± 0.014), wood absorption (0.094 ± 0.024) and light
8 reaching the ground surface (0.065 ± 0.029) (Table S2). Light saturation effects at the leaf
9 level, calculated as the difference between leaf absorption and light utilization, further
10 reduced effective *l*APAR. Mean saturation effects were lowest for simulations at 09:00
11 (0.0428 ± 0.0237) and highest in midday simulations (13:00, 0.121 ± 0.027), with a maximum
12 of 0.17 in September (Table S2, [see Figure 2](#)). At the monthly time scale, saturation effects
13 varied from 0.05-0.13, such that only 67% - 74% of incoming PAR was estimated to be
14 available for photosynthesis.

15 Light saturation effects moderated the apparent benefit of increasing PAR during dry-
16 season months (Figure [34](#), Figure [45](#)). During July-November, saturation effects under
17 midday conditions were similar in magnitude to combined APAR losses from absorption by
18 woody elements and the ground (Table S2). Together, midday light saturation effects and
19 non-leaf absorption led to the lowest light utilization during July-November (Figure [45](#)).
20 Canopy 3D structure generated a decrease in light utilization during the dry season based on
21 two competing processes. Lower cloud cover in the early dry season increased PAR at the
22 top of canopy (Table S1, Figure [4b5b](#)), but lower sun angles and more direct radiation altered
23 the distribution of light at the leaf level. Thus, monthly increases in PAR were not distributed
24 across all leaves, as some canopy leaves were light saturated while shading other parts of the
25 same crown or shorter neighboring trees.

26 This decrease in canopy light utilization can be described in terms of efficiency.
27 Increased incident PAR between June and July ($+40.3 \mu\text{mol m}^{-2} \text{s}^{-1}$) only resulted in a 50%
28 relative increase in light utilization ($+20.0 \mu\text{mol m}^{-2} \text{s}^{-1}$, Figure [43](#), Table S2), with the
29 remaining *l*APAR lost to light saturation. Even under near-nadir illumination conditions in
30 September, light saturation effects moderated the change in effective *l*APAR to 65% of the
31 relative increase in PAR between August and September. These cases highlight the need to
32 consider how forest 3D structure alters the distribution of *l*APAR at the leaf level in order to
33 estimate light utilization in tropical forests.

1 Horizontal variability in forest 3D structure generated clear spatial and temporal
2 differences in light absorption profiles and the vertical distribution of light saturation effects
3 (Figure 56). DART model results differed substantially from the depiction of diurnal
4 variability in light interception from the exponential model (non-spatial) or the pseudo-spatial
5 representation of forest structure in ED. The exponential model of light extinction
6 overestimated total absorption by leaves compared to DART or ED under midday and
7 afternoon illumination conditions (Figure 5b6b, 56d). Adding more patches in ED model
8 simulations generated light absorption profiles that were more similar to DART results, with
9 horizontal differences in forest structure lowering cumulative light absorption from ~~a big-~~
10 ~~leafsimulations with model (a single patch)~~ to a hyper-parameterization of forest structure
11 ~~with (2500 patches)~~. However, estimates of light utilization in ED exhibited different vertical
12 profiles and cumulative $IAPAR$ than in DART (Figure 5b, 5d). ED model estimates of light
13 utilization were more strongly influenced by total incident PAR than the diversity of patch
14 environments, since each patch receives the same incident PAR at the top of canopy.
15 Differences between models also reflect a more complete characterization of light interactions
16 in DART, including 3D representation of ~~shading from neighboring trees, leaf reflectance,~~
17 ~~scattering,~~ and absorption by woody elements.

18 Light-saturated leaves were distributed throughout the vertical canopy profile under
19 midday or diffuse illumination conditions in the DART simulations (Figure 65a, Movie S1).
20 Greater light penetration in DART, with light saturation effects below mean canopy height,
21 further distorted the vertical profile of DART light utilization relative to more simplified
22 representations of forest structure (Figure 65b). Evidence for greater light absorption by
23 shorter canopy trees than taller canopy trees underscores the need for a full 3D representation
24 of the tropical forest light environment to accurately estimate tropical forest responses to
25 changing light conditions. Horizontal variability in 3D structure also generated a diversity of
26 light environments at the forest floor (Movie S2). The frequency, intensity, and duration of
27 sun flecks offers a promising avenue for studies of forest regeneration and the role of gap
28 dynamics for the heterogeneity of light environments in the forest understory.

29

30 **4 Discussion**

31 Incident PAR at the top of a tropical forest canopy is an imperfect measure of light
32 availability for photosynthesis. The 3D structure of tropical forests, combined with diurnal

1 and seasonal variability in direct and diffuse illumination, alters *l*APAR and light saturation
2 effects. Together, shadowing and saturation may reduce the amount of light available for
3 photosynthesis by 11% - 23%, given estimates of mean monthly light utilization in this study
4 (0.67-0.74) and biome-wide estimates of Amazon FAPAR from satellite data (0.85-0.9;
5 (Senna et al., 2005). An accurate representation of light saturation effects in ecosystem
6 models is critical to constrain potential gains in gross primary productivity from changing
7 light levels under scenarios of future climate, including the influence of clouds and biomass
8 burning aerosols on diffuse light conditions (Rap et al., 2015; Mercado et al., 2009; Oliveira
9 et al., 2007; Cirino et al., 2014; Kanniah et al., 2012). Ecosystem models of intermediate
10 complexity may not fully account for shadowing and light saturation effects in tropical
11 forests, including illumination differences among canopy trees and light penetration to lower
12 canopy layers. For these models, it may be possible to leverage information on forest
13 structure from satellite data to account for these processes. Forest types with greater vertical
14 and horizontal heterogeneity generate stronger bidirectional reflectance effects in passive
15 optical remote sensing data (Morton et al., 2014; Nagol et al., 2015), providing a proxy for
16 fine-scale spatial variability in canopy structure.

17 Changing illumination geometry influences the distribution of light within the forest
18 canopy, and this physical mechanism may partially explain patterns of seasonal carbon uptake
19 in Amazon forests. Evidence for seasonal variability in light utilization in this study
20 constitutes an alternate biophysical explanation for Amazon forest seasonality without
21 concurrent changes in canopy phenology. Previous studies have estimated changes in
22 photosynthetic capacity (P_c) of Amazon forests based on the ratio of net ecosystem exchange
23 (NEE) to incident PAR (e.g., Hutyrá et al., 2007; Restrepo-Coupe et al., 2013). One
24 suggested mechanism for seasonal variability in P_c is forest phenology. New leaves
25 photosynthesize more efficiently than old leaves, and evidence for seasonal flushing of new
26 leaves has therefore been hypothesized to promote greater P_c from leaf demographics (Huete
27 et al., 2006; Brando et al., 2010; Restrepo-Coupe et al., 2013; Wu et al., 2016). However, P_c
28 does not account for the influence of forest 3D structure on light utilization from shadowing
29 or light saturation effects. The impact of changes in leaf age or leaf characteristics (e.g.,
30 Kitajima et al., 1997) must therefore be evaluated based on the distribution of *l*APAR at the
31 leaf level. At present, is unclear whether seasonal changes in illumination are simply aliased
32 to other seasonal phenomena in studies of P_c , including phenology (Restrepo-Coupe et al.,
33 2013; Medvigy et al., 2013; Wu et al., 2016), since these studies do not specifically separate

1 leaf demography from other mechanisms for upregulation of P_c . The results of this study
2 highlight how directional illumination and forest 3D structure combine to influence diurnal
3 and seasonal variability in light utilization, independent of further changes in leaf area, leaf
4 age, or environmental controls on canopy photosynthesis. DART model results emphasize
5 the importance of light utilization (rather than PAR, FAPAR, or even $lAPAR$) to attribute
6 changes in light availability to seasonal dynamics of Amazon forest productivity.

7 Whether the differences between DART and other models represent an offset or a bias
8 depends on whether forest structure influences PAR absorption and utilization in consistent
9 ways across seasons, latitudes, and forest types. In this study, $lAPAR$ varied diurnally, but
10 midday simulations (11:00 and 13:00) were more consistent across months, suggesting that
11 shadowing may be less important than light saturation effects for estimates of midday
12 photosynthesis in tropical forests. Saturation losses of 13-17% in midday simulations with
13 DART underscore the need for leaf-level information to convert leaf absorption to light
14 utilization. ED model simulations overestimated midday light saturation losses compared to
15 DART, and underestimated light saturation effects at lower sun angles, likely because all
16 patches receive the same incident PAR at the top of canopy (no shadowing from neighboring
17 patches). Tropical forests present particular challenges for ecosystem models; regional
18 differences in Amazon forest structure (Morton et al., 2014) interact with seasonal and
19 interannual differences in diffuse illumination cloud cover and biomass burning aerosols.
20 These challenges point to the potential benefits of developing more robust, 3D ecosystem
21 models to estimate forest productivity under direct and diffuse illumination conditions.

22 Illumination conditions differ dramatically between the tropics and higher latitudes;
23 temperate and boreal forests may never experience near-nadir illumination conditions
24 approximated using one-dimensional light extinction profiles (Guillevic and Gastellu-
25 Etchegorry, 1999), except under full diffuse illumination conditions. Seasonal variability in
26 shadowing and light interception by woody elements may therefore be more important for
27 understanding photosynthesis in these systems. Importantly, radiative transfer models such as
28 DART must be coupled with ecosystem models to estimate how temperature and moisture
29 stress reduce the utilization of $lAPAR$ for midday, clear-sky simulations.

30 Light interactions at the leaf level are the basis for remote sensing approaches to
31 monitor vegetation productivity. This study highlighted how horizontal variability in forest
32 3D structure altered $lAPAR$ and light absorption by woody elements and the ground surface.
33 The influence of diurnal and seasonal variability in illumination is one factor that contributes

1 to variability in surface reflectance estimates over Amazon forests (Morton et al., 2014; Nagol
2 et al., 2015). Changes in the fraction of sunlit and shaded leaves, along with differences in
3 the degree of light saturation, likely contributes to seasonal variability in SIF measurements
4 from satellite platforms (e.g., Joiner et al., 2011; Guan et al., 2015), especially given the
5 early morning overpass time of satellites such as GOME-2 (09:30, see Figure 54a). To date,
6 models and remote sensing approaches do not account for the role of local heterogeneity in
7 forest structure as a mechanism for SIF variability (Joiner et al., 2011; Guan et al., 2015;
8 Zhang et al., 2014).

9 The growing availability of lidar-based measurements of forest structure opens several
10 important avenues for ecosystem model development. One under-represented element of
11 forest structure in ecosystem models is branches. In tropical forests, branches may account
12 for 1/3 of the total aboveground biomass (Higuchi et al., 1998), yet few ecosystem models
13 realistically account for the roles of branches and stem material for light interception or
14 canopy turnover from branch falls. New measurement capabilities from terrestrial lidar
15 scanning (TLS) systems, including multispectral instruments, offer new insights into the
16 contribution of branches to 3D structure and canopy reflectance in visible and near-infrared
17 wavelengths. TLS data also provide a detailed depiction of the vertical distribution of leaf
18 area and the forest understory environment that could improve model parameterization.
19 Finally, model simulation studies offer the potential to run simple or complex scenarios; in a
20 future study, the single light saturation curve, leaf angle distribution, and leaf reflectance
21 properties in this analysis could be modified based on new regional measurements to evaluate
22 the influence of plant trait diversity on light responses in tropical forests.

23 In addition to data on branch structure, new field data are needed to constrain the
24 influence of plant trait variability on canopy reflectance and light utilization. Recent studies
25 highlight the potential for leaf demography to alter leaf reflectance on a seasonal basis
26 (Chavana-Bryant et al., 2016; Wu et al., 2016; Brando et al., 2010). Without a broader
27 sample of Amazon tree species, and additional data on transmittance and absorptance, it is
28 unclear whether subtle and short-term changes in leaf reflectance properties (Chavana-Bryant
29 et al., 2016) are sufficient to alter PAR availability for canopy and understory trees. New data
30 are also needed to model differences in light saturation among species, canopy positions, and
31 leaf ages. Subsequent studies that combine forest 3D structure (including branches) with leaf-
32 level variability in light saturation could extend the work in this paper on the contributions

1 | [from shadowing and light saturation to seasonal variability in light utilization in tropical](#)
2 | [forests.](#)

3 This study illustrates the importance of realistic, 3D representations of the forest
4 canopy for accurate simulations of light availability in tropical forests. DART model results
5 have important implications for both modeling and remote sensing of tropical forest
6 ecosystems, including how the vertical and horizontal distributions of light saturation effects
7 influence remote sensing measurements and model estimates of forest productivity. Radiative
8 transfer models provide an important link between top-down estimates from remote sensing
9 platforms and bottom-up estimates of forest structure and carbon fluxes from field and tower-
10 based instruments. The growing availability of airborne lidar data offers the potential to
11 investigate complex ecosystem interactions using DART or similar models to improve the
12 representation of light utilization in ecosystem models.

14 **Acknowledgements**

15 This research was funded by NASA's Terrestrial Ecology and Carbon Monitoring System
16 programs and Brazil's National Council on Scientific Development & Technology (CNPq)
17 Science Without Borders Fellowship Program. Funding for lidar data collection was provided
18 by the U.S. Department of State, USAID, and the US SilvaCarbon Program. Lidar data are
19 available from the Sustainable Landscapes Project:

20 <http://mapas.cnpm.embrapa.br/paisagens sustentaveis/>

1 References

- 2 Alton, P. B., Ellis, R., Los, S. O., and North, P. R.: Improved global simulations of gross
3 primary product based on a separate and explicit treatment of diffuse and direct sunlight,
4 *Journal of Geophysical Research: Atmospheres*, 112, n/a-n/a, 10.1029/2006JD008022, 2007a.
- 5 Alton, P. B., North, P. R., and Los, S. O.: The impact of diffuse sunlight on canopy light-use
6 efficiency, gross photosynthetic product and net ecosystem exchange in three forest biomes,
7 *Global Change Biology*, 13, 776-787, 10.1111/j.1365-2486.2007.01316.x, 2007b.
- 8 Asner, G. P., Nepstad, D., Cardinot, G., and Ray, D.: Drought stress and carbon uptake in an
9 amazon forest measured with spaceborne imaging spectroscopy, *Proceedings of the National*
10 *Academy of Sciences of the United States of America*, 101, 6039-6044, 2004.
- 11 Asner, G. P., Kellner, J. R., Kennedy-Bowdoin, T., Knapp, D. E., Anderson, C., and Martin,
12 R. E.: Forest canopy gap distributions in the southern peruvian amazon, *PLoS ONE*, 8,
13 e60875, 10.1371/journal.pone.0060875, 2013.
- 14 Baccini, A., Goetz, S. J., Walker, W. S., Laporte, N. T., Sun, M., Sulla-Menashe, D., Hackler,
15 J., Beck, P. S. A., Dubayah, R., Friedl, M. A., Samanta, S., and Houghton, R. A.: Estimated
16 carbon dioxide emissions from tropical deforestation improved by carbon-density maps,
17 *Nature Clim. Change*, 2, 182-185, 2012.
- 18 Boisier, J. P., Ciais, P., Ducharne, A., and Guimberteau, M.: Projected strengthening of
19 amazonian dry season by constrained climate model simulations, *Nature Clim. Change*, 5,
20 656-660, 10.1038/nclimate2658
- 21 <http://www.nature.com/nclimate/journal/v5/n7/abs/nclimate2658.html> - [supplementary-](#)
22 [information](#), 2015.
- 23 Brando, P. M., Goetz, S. J., Baccini, A., Nepstad, D. C., Beck, P. S. A., and Christman, M. C.:
24 Seasonal and interannual variability of climate and vegetation indices across the amazon,
25 *Proceedings of the National Academy of Sciences*, 107, 14685-14690, 2010.
- 26 Chavana-Bryant, C., Malhi, Y., Wu, J., Asner, G. P., Anastasiou, A., Enquist, B. J., Cosio
27 Caravasi, E. G., Doughty, C. E., Saleska, S. R., Martin, R. E., and Gerard, F. F.: Leaf aging of
28 amazonian canopy trees as revealed by spectral and physiochemical measurements, *New*
29 *Phytologist*, n/a-n/a, 10.1111/nph.13853, 2016.
- 30 Chave, J., Navarrete, D., Almeida, S., Álvarez, E., Aragão, L. E. O. C., Bonal, D., Châtelet,
31 P., Silva-Espejo, J. E., Goret, J. Y., von Hildebrand, P., Jiménez, E., Patiño, S., Peñuela, M.
32 C., Phillips, O. L., Stevenson, P., and Malhi, Y.: Regional and seasonal patterns of litterfall in
33 tropical south america, *Biogeosciences*, 7, 43-55, 10.5194/bg-7-43-2010, 2010.
- 34 Chen, Y., Randerson, J. T., Van Der Werf, G. R., Morton, D. C., Mu, M., and Kasibhatla, P.
35 S.: Nitrogen deposition in tropical forests from savanna and deforestation fires, *Global*
36 *Change Biology*, 16, 2024-2038, 10.1111/j.1365-2486.2009.02156.x, 2010.
- 37 Chen, Y., Morton, D. C., Jin, Y., Collatz, G. J., Kasibhatla, P. S., van der Werf, G. R.,
38 DeFries, R. S., and Randerson, J. T.: Long-term trends and interannual variability of forest,
39 savanna and agricultural fires in south america, *Carbon Management*, 4, 617-638,
40 10.4155/cmt.13.61, 2013a.
- 41 Chen, Y., Velicogna, I., Famiglietti, J. S., and Randerson, J. T.: Satellite observations of
42 terrestrial water storage provide early warning information about drought and fire season

1 severity in the amazon, Journal of Geophysical Research: Biogeosciences, 118, 495-504,
2 10.1002/jgrg.20046, 2013b.

3 Cirino, G. G., Souza, R. A. F., Adams, D. K., and Artaxo, P.: The effect of atmospheric
4 aerosol particles and clouds on net ecosystem exchange in the amazon, Atmos. Chem. Phys.,
5 14, 6523-6543, 10.5194/acp-14-6523-2014, 2014.

6 Clark, D. B., Mercado, L. M., Sitch, S., Jones, C. D., Gedney, N., Best, M. J., Pryor, M.,
7 Rooney, G. G., Essery, R. L. H., Blyth, E., Boucher, O., Harding, R. J., Huntingford, C., and
8 Cox, P. M.: The joint uk land environment simulator (jules), model description – part 2:
9 Carbon fluxes and vegetation dynamics, Geosci. Model Dev., 4, 701-722, 10.5194/gmd-4-
10 701-2011, 2011.

11 Cleveland, C. C., Taylor, P., Chadwick, K. D., Dahlin, K., Doughty, C. E., Malhi, Y., Smith,
12 W. K., Sullivan, B. W., Wieder, W. R., and Townsend, A. R.: A comparison of plot-based
13 satellite and earth system model estimates of tropical forest net primary production, Global
14 Biogeochemical Cycles, 29, 626-644, 10.1002/2014GB005022, 2015.

15 Cox, P. M., Pearson, D., Booth, B. B., Friedlingstein, P., Huntingford, C., Jones, C. D., and
16 Luke, C. M.: Sensitivity of tropical carbon to climate change constrained by carbon dioxide
17 variability, Nature, 494, 341-344,
18 <http://www.nature.com/nature/journal/v494/n7437/abs/nature11882.html> - [supplementary-](#)
19 [information](#), 2013.

20 da Costa, A. C. L., Galbraith, D., Almeida, S., Portela, B. T. T., da Costa, M., de Athaydes
21 Silva Junior, J., Braga, A. P., de Gonçalves, P. H. L., de Oliveira, A. A. R., Fisher, R.,
22 Phillips, O. L., Metcalfe, D. B., Levy, P., and Meir, P.: Effect of 7 yr of experimental drought
23 on vegetation dynamics and biomass storage of an eastern amazonian rainforest, New
24 Phytologist, 187, 579-591, 10.1111/j.1469-8137.2010.03309.x, 2010.

25 Doughty, C. E., and Goulden, M. L.: Are tropical forests near a high temperature threshold?,
26 Journal of Geophysical Research: Biogeosciences, 113, n/a-n/a, 10.1029/2007JG000632,
27 2008.

28 Doughty, C. E., Metcalfe, D. B., Girardin, C. A. J., Amezquita, F. F., Durand, L., Huaraca
29 Huasco, W., Silva-Espejo, J. E., Araujo-Murakami, A., da Costa, M. C., da Costa, A. C. L.,
30 Rocha, W., Meir, P., Galbraith, D., and Malhi, Y.: Source and sink carbon dynamics and
31 carbon allocation in the amazon basin, Global Biogeochemical Cycles, 29, 645-655,
32 10.1002/2014GB005028, 2015.

33 Espírito-Santo, F. D. B., Gloor, M., Keller, M., Malhi, Y., Saatchi, S., Nelson, B., Junior, R.
34 C. O., Pereira, C., Lloyd, J., Frohling, S., Palace, M., Shimabukuro, Y. E., Duarte, V.,
35 Mendoza, A. M., López-González, G., Baker, T. R., Feldpausch, T. R., Brienen, R. J. W.,
36 Asner, G. P., Boyd, D. S., and Phillips, O. L.: Size and frequency of natural forest
37 disturbances and the amazon forest carbon balance, Nat Commun, 5, 10.1038/ncomms4434,
38 2014.

39 Frohling, S., Milliman, T., Palace, M., Wisser, D., Lammers, R., and Fahnestock, M.:
40 Tropical forest backscatter anomaly evident in seawinds scatterometer morning overpass data
41 during 2005 drought in amazonia, Remote Sensing of Environment, 115, 897-907, 2011.

42 Gastellu-Etchegorry, J.-P., Yin, T., Lauret, N., Cajgfinger, T., Gregoire, T., Grau, E., Feret, J.-
43 B., Lopes, M., Guilleux, J., Dedieu, G., Malenovsky, Z., Cook, B., Morton, D., Rubio, J.,
44 Durrieu, S., Cazanave, G., Martin, E., and Ristorcelli, T.: Discrete anisotropic radiative

1 transfer (dart 5) for modeling airborne and satellite spectroradiometer and lidar acquisitions of
2 natural and urban landscapes, *Remote Sensing*, 7, 1667, 2015.

3 Gastellu-Etchegorry, J. P., and Trichon, V.: A modeling approach of par environment in a
4 tropical rain forest in sumatra: Application to remote sensing, *Ecol. Model.*, 108, 237-264,
5 [http://dx.doi.org/10.1016/S0304-3800\(98\)00032-5](http://dx.doi.org/10.1016/S0304-3800(98)00032-5), 1998.

6 Gastellu-Etchegorry, J. P., Grau, E., and Lauret, N.: Dart: A 3d model for remote sensing
7 images and radiative budget of earth surfaces, in: *Modeling and simulation in engineering*,
8 edited by: Alexandru, C., InTech, ISBN: 978-953-951-0012-0016, 2012.

9 Gatti, L. V., Miller, J. B., D'Amelio, M. T. S., Martinewski, A., Basso, L. S., Gloor, M. E.,
10 Wofsy, S., and Tans, P.: Vertical profiles of co2 above eastern amazonia suggest a net carbon
11 flux to the atmosphere and balanced biosphere between 2000 and 2009, *Tellus B*, 62, 581-
12 594, 2010.

13 Gatti, L. V., Gloor, M., Miller, J. B., Doughty, C. E., Malhi, Y., Domingues, L. G., Basso, L.
14 S., Martinewski, A., Correia, C. S. C., Borges, V. F., Freitas, S., Braz, R., Anderson, L. O.,
15 Rocha, H., Grace, J., Phillips, O. L., and Lloyd, J.: Drought sensitivity of amazonian carbon
16 balance revealed by atmospheric measurements, *Nature*, 506, 76-80, 10.1038/nature12957,
17 2014.

18 Gibelin, A.-L., Calvet, J.-C., and Viovy, N.: Modelling energy and co2 fluxes with an
19 interactive vegetation land surface model-evaluation at high and middle latitudes, *Agricultural
20 and Forest Meteorology*, 148, 1611-1628, <http://dx.doi.org/10.1016/j.agrformet.2008.05.013>,
21 2008.

22 Grau, E., and Gastellu-Etchegorry, J.-P.: Radiative transfer modeling in the earth-atmosphere
23 system with dart model, *Remote Sensing of Environment*, 139, 149-170,
24 <http://dx.doi.org/10.1016/j.rse.2013.07.019>, 2013.

25 Guan, K., Pan, M., Li, H., Wolf, A., Wu, J., Medvigy, D., Caylor, K. K., Sheffield, J., Wood,
26 E. F., Malhi, Y., Liang, M., Kimball, J. S., Saleska, S. R., Berry, J., Joiner, J., and Lyapustin,
27 A. I.: Photosynthetic seasonality of global tropical forests constrained by hydroclimate,
28 *Nature Geosci*, 8, 284-289, 10.1038/ngeo2382
29 <http://www.nature.com/ngeo/journal/v8/n4/abs/ngeo2382.html> - [supplementary-information](#),
30 2015.

31 Guillevic, P., and Gastellu-Etchegorry, J. P.: Modeling brf and radiation regime of boreal and
32 tropical forest: Ii. Par regime, *Remote Sensing of Environment*, 68, 317-340,
33 [http://dx.doi.org/10.1016/S0034-4257\(98\)00120-5](http://dx.doi.org/10.1016/S0034-4257(98)00120-5), 1999.

34 Higuchi, N., dos Santos, J., Ribeiro, R., Minette, L., and Biot, Y.: Biomass da parte aérea da
35 vegetação da floresta tropical úmida de terra-firme da amazônia brasileira, *Acta Amazon.*, 28,
36 153-166, 1998.

37 Holben, B. N., Eck, T. F., Slutsker, I., Tanré, D., Buis, J. P., Setzer, A., Vermote, E., Reagan,
38 J. A., Kaufman, Y. J., Nakajima, T., Lavenu, F., Jankowiak, I., and Smirnov, A.: Aeronet—a
39 federated instrument network and data archive for aerosol characterization, *Remote Sensing
40 of Environment*, 66, 1-16, [http://dx.doi.org/10.1016/S0034-4257\(98\)00031-5](http://dx.doi.org/10.1016/S0034-4257(98)00031-5), 1998.

41 Huete, A. R., Didan, K., Shimabukuro, Y. E., Ratana, P., Saleska, S. R., Huttyra, L. R., Yang,
42 W., Nemani, R. R., and Myneni, R. B.: Amazon rainforests green-up with sunlight in dry
43 season, *Geophysical Research Letters*, 33, doi:10.1029/2005GL025583, 2006.

1 Hunter, M. O., Keller, M., Morton, D., Cook, B., Lefsky, M., Ducey, M., Saleska, S., de
2 Oliveira, R. C., Jr., and Schiatti, J.: Structural dynamics of tropical moist forest gaps, PLoS
3 ONE, 10, e0132144, 10.1371/journal.pone.0132144, 2015.

4 Hutyra, L. R., Munger, J. W., Saleska, S. R., Gottlieb, E., Daube, B. C., Dunn, A. L., Amaral,
5 D. F., de Camargo, P. B., and Wofsy, S. C.: Seasonal controls on the exchange of carbon and
6 water in an amazonian rain forest, *Journal of Geophysical Research: Biogeosciences*, 112,
7 n/a-n/a, 10.1029/2006JG000365, 2007.

8 Hutyra, L. R., Wofsy, S. C., and Saleska, S. R.: Lba-eco cd-10 co2 and h2o eddy fluxes at km
9 67 tower site, tapajos national forest, in, ORNL Distributed Active Archive Center, 2008.

10 Joiner, J., Yoshida, Y., Vasilkov, A. P., Yoshida, Y., Corp, L. A., and Middleton, E. M.: First
11 observations of global and seasonal terrestrial chlorophyll fluorescence from space,
12 *Biogeosciences*, 8, 637-651, 10.5194/bg-8-637-2011, 2011.

13 Kanniah, K. D., Beringer, J., North, P., and Hutley, L.: Control of atmospheric particles on
14 diffuse radiation and terrestrial plant productivity: A review, *Progress in Physical Geography*,
15 36, 209-237, 2012.

16 Keppel-Aleks, G., Wolf, A. S., Mu, M., Doney, S. C., Morton, D. C., Kasibhatla, P. S.,
17 Miller, J. B., Dlugokencky, E. J., and Randerson, J. T.: Separating the influence of
18 temperature, drought, and fire on interannual variability in atmospheric co2, *Global*
19 *Biogeochemical Cycles*, 28, 1295-1310, 10.1002/2014GB004890, 2014.

20 Kitajima, K., Mulkey, S. S., and Wright, S. J.: Seasonal leaf phenotypes in the canopy of a
21 tropical dry forest: Photosynthetic characteristics and associated traits, *Oecologia*, 109, 490-
22 498, 10.1007/s004420050109, 1997.

23 Krinner, G., Viovy, N., de Noblet-Ducoudré, N., Ogée, J., Polcher, J., Friedlingstein, P.,
24 Ciais, P., Sitch, S., and Prentice, I. C.: A dynamic global vegetation model for studies of the
25 coupled atmosphere-biosphere system, *Global Biogeochemical Cycles*, 19, n/a-n/a,
26 10.1029/2003GB002199, 2005.

27 Le Quére, C., Andres, R. J., Boden, T., Conway, T., Houghton, R. A., House, J. I., Marland,
28 G., Peters, G. P., van der Werf, G. R., Ahlström, A., Andrew, R. M., Bopp, L., Canadell, J.
29 G., Ciais, P., Doney, S. C., Enright, C., Friedlingstein, P., Huntingford, C., Jain, A. K.,
30 Jourdain, C., Kato, E., Keeling, R. F., Klein Goldewijk, K., Levis, S., Levy, P., Lomas, M.,
31 Poulter, B., Raupach, M. R., Schwinger, J., Sitch, S., Stocker, B. D., Viovy, N., Zaehle, S.,
32 and Zeng, N.: The global carbon budget 1959-2011, *Earth Syst. Sci. Data*, 5, 165-185,
33 10.5194/essd-5-165-2013, 2013.

34 Lee, J.-E., Frankenberg, C., van der Tol, C., Berry, J. A., Guanter, L., Boyce, C. K., Fisher, J.
35 B., Morrow, E., Worden, J. R., Asefi, S., Badgley, G., and Saatchi, S.: Forest productivity and
36 water stress in amazonia: Observations from gosat chlorophyll fluorescence, *Proceedings of*
37 *the Royal Society of London B: Biological Sciences*, 280, 2013.

38 Leitold, V., Keller, M., Morton, D. C., Cook, B. D., and Shimabukuro, Y. E.: Airborne lidar-
39 based estimates of tropical forest structure in complex terrain: Opportunities and trade-offs for
40 redd+, *Carbon Balance Manag.*, 10, 3, 10.1186/s13021-015-0013-x, 2015.

41 Lewis, S. L., Brando, P. M., Phillips, O. L., van der Heijden, G. M. F., and Nepstad, D.: The
42 2010 amazon drought, *Science*, 331, 554-554, 2011.

43 Longo, M.: Amazon forest response to changes in rainfall regime: Results from an individual-
44 based dynamic vegetation model, PhD, Harvard University, Cambridge, MA, 2014.

1 Malhi, Y., Wood, D., Baker, T. R., Wright, J., Phillips, O. L., Cochrane, T., Meir, P., Chave,
2 J., Killeen, T. J., Laurance, S. G., Laurance, W. F., Vargas, P. N., Pitman, N. C. A., Quesada,
3 C. A., Salomão, R., Silva, J. N., Lezama, A. T., Terborgh, J., Martínez, R. V., and Vinceti, B.:
4 The regional variation of aboveground live biomass in old-growth amazonian forests, *Global*
5 *Change Biology*, 12, 1-32, 2006.

6 Malhi, Y., Doughty, C. E., Goldsmith, G. R., Metcalfe, D. B., Girardin, C. A. J., Marthews, T.
7 R., del Aguila-Pasquel, J., Aragão, L. E. O. C., Araujo-Murakami, A., Brando, P., da Costa,
8 A. C. L., Silva-Espejo, J. E., Farfán Amézquita, F., Galbraith, D. R., Quesada, C. A., Rocha,
9 W., Salinas-Revilla, N., Silvério, D., Meir, P., and Phillips, O. L.: The linkages between
10 photosynthesis, productivity, growth and biomass in lowland amazonian forests, *Global*
11 *Change Biology*, 21, 2283-2295, 10.1111/gcb.12859, 2015.

12 Medvigy, D., Jeong, S.-J., Clark, K. L., Skowronski, N. S., and Schäfer, K. V. R.: Effects of
13 seasonal variation of photosynthetic capacity on the carbon fluxes of a temperate deciduous
14 forest, *Journal of Geophysical Research: Biogeosciences*, 118, 1703-1714,
15 10.1002/2013JG002421, 2013.

16 Mercado, L. M., Bellouin, N., Sitch, S., Boucher, O., Huntingford, C., Wild, M., and Cox, P.
17 M.: Impact of changes in diffuse radiation on the global land carbon sink, *Nature*, 458, 1014-
18 1017, http://www.nature.com/nature/journal/v458/n7241/supinfo/nature07949_S1.html,
19 2009.

20 Moorcroft, P. R., Hurtt, G. C., and Pacala, S. W.: A method for scaling vegetation dynamics:
21 The ecosystem demography model (ed), *Ecol. Monogr.*, 71, 557-586, 2001.

22 Morton, D. C., Le Page, Y., DeFries, R. S., Collatz, G. J., and Hurtt, G. C.: Understorey fire
23 frequency and the fate of burned forests in southern amazonia, *Philosophical Transactions of*
24 *the Royal Society B*, 368, 20120163, 2013.

25 Morton, D. C., Nagol, J., Carabajal, C. C., Rosette, J., Palace, M., Cook, B. D., Vermote, E.
26 F., Harding, D. J., and North, P. R. J.: Amazon forests maintain consistent canopy structure
27 and greenness during the dry season, *Nature*, 506, 221-224, 10.1038/nature13006
28 <http://www.nature.com/nature/journal/v506/n7487/abs/nature13006.html> - [supplementary-](#)
29 [information](#), 2014.

30 Nagol, J. R., Sexton, J. O., Kim, D.-H., Anand, A., Morton, D., Vermote, E., and Townshend,
31 J. R.: Bidirectional effects in landsat reflectance estimates: Is there a problem to solve?,
32 *ISPRS Journal of Photogrammetry and Remote Sensing*, 103, 129-135,
33 <http://dx.doi.org/10.1016/j.isprsjprs.2014.09.006>, 2015.

34 Nemani, R. R., Keeling, C. D., Hashimoto, H., Jolly, W. M., Piper, S. C., Tucker, C. J.,
35 Myneni, R. B., and Running, S. W.: Climate-driven increases in global terrestrial net primary
36 production from 1982 to 1999, *Science*, 300, 1560-1563, 2003.

37 Nepstad, D. C., Tohver, I. M., Ray, D., Moutinho, P., and Cardinot, G.: Mortality of large
38 trees and lianas following experimental drought in an amazon forest, *Ecology*, 88, 2259-2269,
39 2007.

40 Oliveira, P. H. F., Artaxo, P., Pires, C., De Lucca, S., ProcÓpio, A., Holben, B., Schafer, J.,
41 Cardoso, L. F., Wofsy, S. C., and Rocha, H. R.: The effects of biomass burning aerosols and
42 clouds on the co2 flux in amazonia, *Tellus B*, 59, 338-349, 10.1111/j.1600-
43 0889.2007.00270.x, 2007.

1 Parazoo, N. C., Bowman, K., Frankenberg, C., Lee, J.-E., Fisher, J. B., Worden, J., Jones, D.
2 B. A., Berry, J., Collatz, G. J., Baker, I. T., Jung, M., Liu, J., Osterman, G., O'Dell, C.,
3 Sparks, A., Butz, A., Guerlet, S., Yoshida, Y., Chen, H., and Gerbig, C.: Interpreting seasonal
4 changes in the carbon balance of southern amazonia using measurements of xco2 and
5 chlorophyll fluorescence from gosat, *Geophysical Research Letters*, 40, 2829-2833,
6 10.1002/grl.50452, 2013.

7 Phillips, O. L., Arag o, L. E. O. C., Lewis, S. L., Fisher, J. B., Lloyd, J., L pez-Gonz lez,
8 G., Malhi, Y., Monteagudo, A., Peacock, J., Quesada, C. A., van der Heijden, G., Almeida, S.,
9 Amaral, I. d., Arroyo, L., Aymard, G., Baker, T. R., B nki, O., Blanc, L., Bonal, D., Brando,
10 P., Chave, J., de Oliveira,  . C. A., Cardozo, N. D. v., Czimczik, C. I., Feldpausch, T. R.,
11 Freitas, M. A., Gloor, E., Higuchi, N., Jim nez, E., Lloyd, G., Meir, P., Mendoza, C.,
12 Morel, A., Neill, D. A., Nepstad, D., Pati o, S., Pe uela, M. C., Prieto, A., Ram rez, F.,
13 Schwarz, M., Silva, J., Silveira, M., Thomas, A. S., Steege, H. t., Stropp, J., V squez, R.,
14 Zelazowski, P., D vila, E. A., Andelman, S., Andrade, A., Chao, K.-J., Erwin, T., Di Fiore,
15 A., C. E. d. H., Keeling, H., Killeen, T. J., Laurance, W. F., Cruz, A. P. a., Pitman, N. C. A.,
16 Vargas, P. N. e., Ram rez-Angulo, H., Rudas, A. n., Salam o, R., Silva, N., Terborgh, J.,
17 and Torres-Lezama, A.: Drought sensitivity of the amazon rainforest, *Science*, 323, 1344-
18 1347, 2009.

19 Poulter, B., Frank, D., Ciais, P., Myneni, R. B., Andela, N., Bi, J., Broquet, G., Canadell, J.
20 G., Chevallier, F., Liu, Y. Y., Running, S. W., Sitch, S., and van der Werf, G. R.:
21 Contribution of semi-arid ecosystems to interannual variability of the global carbon cycle,
22 *Nature*, 509, 600-603, 10.1038/nature13376, 2014.

23 Randerson, J. T.: Climate science: Global warming and tropical carbon, *Nature*, 494, 319-320,
24 2013.

25 Rap, A., Spracklen, D. V., Mercado, L., Reddington, C. L., Haywood, J. M., Ellis, R. J.,
26 Phillips, O. L., Artaxo, P., Bonal, D., Restrepo Coupe, N., and Butt, N.: Fires increase amazon
27 forest productivity through increases in diffuse radiation, *Geophysical Research Letters*, 42,
28 4654-4662, 10.1002/2015GL063719, 2015.

29 Restrepo-Coupe, N., da Rocha, H. R., Hutyr , L. R., da Araujo, A. C., Borma, L. S.,
30 Christoffersen, B., Cabral, O. M. R., de Camargo, P. B., Cardoso, F. L., da Costa, A. C. L.,
31 Fitzjarrald, D. R., Goulden, M. L., Kruijt, B., Maia, J. M. F., Malhi, Y. S., Manzi, A. O.,
32 Miller, S. D., Nobre, A. D., von Randow, C., S , L. D. A., Sakai, R. K., Tota, J., Wofsy, S. C.,
33 Zanchi, F. B., and Saleska, S. R.: What drives the seasonality of photosynthesis across the
34 amazon basin? A cross-site analysis of eddy flux tower measurements from the brasil flux
35 network, *Agricultural and Forest Meteorology*, 182-183, 128-144,
36 <http://dx.doi.org/10.1016/j.agrformet.2013.04.031>, 2013.

37 Romanczyk, P., van Aardt, J., Cawse-Nicholson, K., Kelbe, D., McGlinchy, J., and Krause,
38 K.: Assessing the impact of broadleaf tree structure on airborne full-waveform small-footprint
39 lidar signals through simulation, *Canadian Journal of Remote Sensing*, 39, S60-S72,
40 10.5589/m13-015, 2013.

41 Saatchi, S., Asefi-Najafabady, S., Malhi, Y., Arag o, L. E. O. C., Anderson, L. O., Myneni,
42 R. B., and Nemani, R.: Persistent effects of a severe drought on amazonian forest canopy,
43 *Proceedings of the National Academy of Sciences*, 110, 565-570, 2013.

44 Saatchi, S. S., Harris, N. L., Brown, S., Lefsky, M., Mitchard, E. T. A., Salas, W., Zutta, B.
45 R., Buermann, W., Lewis, S. L., Hagen, S., Petrova, S., White, L., Silman, M., and Morel, A.:

1 Benchmark map of forest carbon stocks in tropical regions across three continents,
2 Proceedings of the National Academy of Sciences, 2011.

3 Saleska, S. R., Miller, S. D., Matross, D. M., Goulden, M. L., Wofsy, S. C., da Rocha, H. R.,
4 de Camargo, P. B., Crill, P., Daube, B. C., de Freitas, H. C., Hutyyra, L. R., Keller, M.,
5 Kirchhoff, V., Menton, M., Munger, J. W., Pyle, E. H., Rice, A. H., and Silva, H.: Carbon in
6 amazon forests: Unexpected seasonal fluxes and disturbance-induced losses, *Science*, 302,
7 1554-1557, 2003.

8 Samanta, A., Knyazikhin, Y., Xu, L., Dickinson, R. E., Fu, R., Costa, M. H., Saatchi, S. S.,
9 Nemani, R. R., and Myneni, R. B.: Seasonal changes in leaf area of amazon forests from leaf
10 flushing and abscission, *J. Geophys. Res.*, 117, G01015, 10.1029/2011jg001818, 2012.

11 Scheiter, S., Langan, L., and Higgins, S. I.: Next-generation dynamic global vegetation
12 models: Learning from community ecology, *New Phytologist*, 198, 957-969,
13 10.1111/nph.12210, 2013.

14 Schimel, D., Pavlick, R., Fisher, J. B., Asner, G. P., Saatchi, S., Townsend, P., Miller, C.,
15 Frankenberg, C., Hibbard, K., and Cox, P.: Observing terrestrial ecosystems and the carbon
16 cycle from space, *Global Change Biology*, 21, 1762-1776, 10.1111/gcb.12822, 2015.

17 Sellers, P. J., Berry, J. A., Collatz, G. J., Field, C. B., and Hall, F. G.: Canopy reflectance,
18 photosynthesis, and transpiration. Iii. A reanalysis using improved leaf models and a new
19 canopy integration scheme, *Remote Sensing of Environment*, 42, 187-216,
20 [http://dx.doi.org/10.1016/0034-4257\(92\)90102-P](http://dx.doi.org/10.1016/0034-4257(92)90102-P), 1992.

21 Senna, M. C. A., Costa, M. H., and Shimabukuro, Y. E.: Fraction of photosynthetically active
22 radiation absorbed by amazon tropical forest: A comparison of field measurements, modeling,
23 and remote sensing, *Journal of Geophysical Research: Biogeosciences*, 110, n/a-n/a,
24 10.1029/2004JG000005, 2005.

25 Sitch, S., Smith, B., Prentice, I. C., Arneth, A., Bondeau, A., Cramer, W., Kaplan, J. O.,
26 Levis, S., Lucht, W., Sykes, M. T., Thonicke, K., and Venevsky, S.: Evaluation of ecosystem
27 dynamics, plant geography and terrestrial carbon cycling in the lpj dynamic global vegetation
28 model, *Global Change Biology*, 9, 161-185, 10.1046/j.1365-2486.2003.00569.x, 2003.

29 Stark, S. C., Leitold, V., Wu, J. L., Hunter, M. O., de Castilho, C. V., Costa, F. R. C.,
30 McMahon, S. M., Parker, G. G., Shimabukuro, M. T., Lefsky, M. A., Keller, M., Alves, L. F.,
31 Schiatti, J., Shimabukuro, Y. E., Brandão, D. O., Woodcock, T. K., Higuchi, N., de Camargo,
32 P. B., de Oliveira, R. C., and Saleska, S. R.: Amazon forest carbon dynamics predicted by
33 profiles of canopy leaf area and light environment, *Ecology Letters*, 15, 1406-1414,
34 10.1111/j.1461-0248.2012.01864.x, 2012.

35 ter Steege, H., Pitman, N. C. A., Sabatier, D., Baraloto, C., Salomão, R. P., Guevara, J. E.,
36 Phillips, O. L., Castilho, C. V., Magnusson, W. E., Molino, J.-F., Monteagudo, A., Núñez
37 Vargas, P., Montero, J. C., Feldpausch, T. R., Coronado, E. N. H., Killeen, T. J., Mostacedo,
38 B., Vasquez, R., Assis, R. L., Terborgh, J., Wittmann, F., Andrade, A., Laurance, W. F.,
39 Laurance, S. G. W., Marimon, B. S., Marimon, B.-H., Guimarães Vieira, I. C., Amaral, I. L.,
40 Brienen, R., Castellanos, H., Cárdenas López, D., Duivenvoorden, J. F., Mogollón, H. F.,
41 Matos, F. D. d. A., Dávila, N., García-Villacorta, R., Stevenson Diaz, P. R., Costa, F., Emilio,
42 T., Levis, C., Schiatti, J., Souza, P., Alonso, A., Dallmeier, F., Montoya, A. J. D., Fernandez
43 Piedade, M. T., Araujo-Murakami, A., Arroyo, L., Gribel, R., Fine, P. V. A., Peres, C. A.,
44 Toledo, M., Aymard C, G. A., Baker, T. R., Cerón, C., Engel, J., Henkel, T. W., Maas, P.,
45 Petronelli, P., Stropp, J., Zartman, C. E., Daly, D., Neill, D., Silveira, M., Paredes, M. R.,

1 Chave, J., Lima Filho, D. d. A., Jørgensen, P. M., Fuentes, A., Schöngart, J., Cornejo
2 Valverde, F., Di Fiore, A., Jimenez, E. M., Peñuela Mora, M. C., Phillips, J. F., Rivas, G., van
3 Andel, T. R., von Hildebrand, P., Hoffman, B., Zent, E. L., Malhi, Y., Prieto, A., Rudas, A.,
4 Ruschell, A. R., Silva, N., Vos, V., Zent, S., Oliveira, A. A., Schutz, A. C., Gonzales, T.,
5 Trindade Nascimento, M., Ramirez-Angulo, H., Sierra, R., Tirado, M., Umaña Medina, M.
6 N., van der Heijden, G., Vela, C. I. A., Vilanova Torre, E., Vriesendorp, C., Wang, O.,
7 Young, K. R., Baider, C., Balslev, H., Ferreira, C., Mesones, I., Torres-Lezama, A., Urrego
8 Giraldo, L. E., Zagt, R., Alexiades, M. N., Hernandez, L., Huamantupa-Chuquimaco, I.,
9 Milliken, W., Palacios Cuenca, W., Pauletto, D., Valderrama Sandoval, E., Valenzuela
10 Gamarra, L., Dexter, K. G., Feeley, K., Lopez-Gonzalez, G., and Silman, M. R.:
11 Hyperdominance in the amazonian tree flora, *Science*, 342, 2013.

12 van der Tol, C., Verhoef, W., Timmermans, J., Verhoef, A., and Su, Z.: An integrated model
13 of soil-canopy spectral radiances, photosynthesis, fluorescence, temperature and energy
14 balance, *Biogeosciences*, 6, 3109-3129, 10.5194/bg-6-3109-2009, 2009.

15 Widlowski, J. L., Pinty, B., Clerici, M., Dai, Y., De Kauwe, M., de Ridder, K., Kallel, A.,
16 Kobayashi, H., Lavergne, T., Ni-Meister, W., Olchev, A., Quaife, T., Wang, S., Yang, W.,
17 Yang, Y., and Yuan, H.: Rami4pilps: An intercomparison of formulations for the partitioning
18 of solar radiation in land surface models, *Journal of Geophysical Research: Biogeosciences*,
19 116, n/a-n/a, 10.1029/2010JG001511, 2011.

20 Wu, J., Albert, L. P., Lopes, A. P., Restrepo-Coupe, N., Hayek, M., Wiedemann, K. T., Guan,
21 K., Stark, S. C., Christoffersen, B., Prohaska, N., Tavares, J. V., Marostica, S., Kobayashi, H.,
22 Ferreira, M. L., Campos, K. S., da Silva, R., Brando, P. M., Dye, D. G., Huxman, T. E.,
23 Huete, A. R., Nelson, B. W., and Saleska, S. R.: Leaf development and demography explain
24 photosynthetic seasonality in amazon evergreen forests, *Science*, 351, 972-976, 2016.

25 Zeng, N., Zhao, F., Collatz, G. J., Kalnay, E., Salawitch, R. J., West, T. O., and Guanter, L.:
26 Agricultural green revolution as a driver of increasing atmospheric co2 seasonal amplitude,
27 *Nature*, 515, 394-397, 10.1038/nature13893, 2014.

28 Zhang, Y., Guanter, L., Berry, J. A., Joiner, J., van der Tol, C., Huete, A., Gitelson, A., Voigt,
29 M., and Köhler, P.: Estimation of vegetation photosynthetic capacity from space-based
30 measurements of chlorophyll fluorescence for terrestrial biosphere models, *Global Change*
31 *Biology*, 20, 3727-3742, 10.1111/gcb.12664, 2014.

32

1 **Figure Legends**

2 Figure 1. DART model scene of an Amazon forest (50 m × 50 m): a) nadir view of canopy
3 height, b) oblique view of 1 m³ leaf voxels and woody architecture, and c) tree objects for
4 canopy tree crowns and stems.

5
6 Figure 2. Probability distribution of average leaf absorbed PAR (*l*APAR, red) and absolute
7 light utilization (dashed black) for the September 13:00 DART simulations. ~~Light saturation~~
8 ~~effects reduce the effective leaf absorption >225 μmol m⁻² s⁻¹.~~ Fractional light utilization
9 (blue) for different *l*APAR values is plotted on the right y-axis, based on leaf measurements
10 of light saturation from *Kitajima et al.* [1997]. Absolute light utilization (dashed black) is the
11 product of *l*APAR (red) and fractional light utilization (blue). Light saturation reduces the
12 effective leaf absorption for voxels with average *l*APAR >225 μmol m⁻³ s⁻¹ LAI⁻¹.

Formatted: Font: Italic

Formatted: Font: Italic

Formatted: Not Superscript/
Subscript

13
14 Figure 3. Illumination geometry alters the distribution of light absorption by leaves on a
15 diurnal and seasonal basis. Simulation results for June, September, and December illustrate
16 the distribution of fractional *l*APAR across the model scene under direct illumination
17 conditions, where diffuse light is modeled using observations of aerosol optical depth from
18 Aeronet. Fractional *l*APAR exceeds 1 for some voxel columns with high interception of
19 incoming PAR, especially with low sun angles in the morning (09:00 LT) and late afternoon
20 (17:00 LT).Figure 3.

Formatted: Font: Italic

Formatted: Font: Italic

21
22 Figure 4. Cloudy (blue) and clear-sky DART simulations (red) were corrected for light
23 saturations effects. Combined results (black line) are a weighted average of light utilization
24 estimates from clear and cloudy simulations for five hours per day and one day per month.

25
26 Figure 45. PAR (solid black) and modeled light utilization (dashed black) for DART
27 simulations at 09:00 (a, top) and monthly average values (b, bottom), based on a weighted
28 average of hourly DART simulations. Simulated monthly values of *l*APAR (solid red) and
29 fractional light utilization (dashed red) are plotted on the right-hand axis. Gray shading
30 indicates dry season months (July-November).

31
32 Figure 56. Forest 3D structure alters total leaf absorption, light utilization, and the vertical
33 distribution of light saturation effects compared to more simplified representations of the

1 | Amazon forest scene. a) Vertical profiles of leaf area density (black) and light saturation
2 | effect (red, difference between absorbed and utilized light) for September 13:00 DART
3 | simulations. b) DART cumulative *l*APAR (black) and light utilization (dashed black);
4 | differences between light absorption and light utilization in DART simulations are plotted as
5 | the red curve in panel “a” to illustrate the vertical distribution of light saturation effects
6 | through the profile of canopy leaf area. DART results –were compared to an exponential
7 | model of light extinction (blue, following *Stark et al.*, [2012]) and ED2 model simulations
8 | (green). Solid and dashed green lines depict cumulative leaf absorption and cumulative light
9 | utilization, respectively, for ED2 simulations with 1, 25, and 2500 patches. c, d) Same as ‘a’
10 | and ‘b’ for September 17:00 illumination conditions.