

1 **Relationships between burned area, forest cover loss, and land cover** 2 **change in the Brazilian Amazon based on satellite data**

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7 **Abstract**

8 Fires are used as a tool in the deforestation process. Yet, the relationship between fire and
9 deforestation may vary temporally and spatially depending on the type of deforestation and
10 climatic conditions. This study evaluates spatiotemporal dynamics of deforestation and fire
11 represented by burned area over the 2002 – 2012 period in the Brazilian Legal Amazon. As a
12 first step, we compared newly available Landsat-based maps of gross forest cover loss from the
13 Global Forest Change (GFC) project with maps of deforestation extent from the Amazon
14 Deforestation Monitoring Project (PRODES) produced by the Brazilian National Institute for
15 Space Research (INPE). As a second step, we rescaled the Landsat-based data to the 500 meter
16 resolution of the Moderate Resolution Imaging Spectroradiometer (MODIS) burned area data
17 (MCD64A1) and stratified this using MODIS land cover data to study the role of burned area in
18 forest cover loss and deforestation. We found that while GFC forest cover loss and PRODES
19 deforestation generally agreed on spatial and temporal dynamics, there were several key
20 differences between the datasets. Both showed a decrease in the extent of forest cover loss or
21 deforestation after 2004, but the drop was larger and more continuous in PRODES than in GFC.
22 The observed decrease in forest cover loss or deforestation rates over our study period was
23 mainly due to lower clearing rates in the evergreen broadleaf forests in the states of Mato
24 Grosso, Pará and Rondônia. GFC indicated anomalous high forest cover loss in the years 2007
25 and 2010 not reported by PRODES. The burned area data indicated that this was predominantly
26 related to increased burned area occurring outside of the tropical forest area during these dry
27 years, mainly in Pará. This indicated that fire and forest loss dynamics in woodlands or
28 secondary forests may be equally important as deforestation in regulating atmospheric CO₂
29 concentrations. In addition to the decrease in forest cover loss rates, we also found that post-

1 deforestation fire use declined; burned area within 5 years after forest cover loss decreased from
2 54% to 39% during our study period.

3 **1 Introduction:**

4 In the tropics, fires are regularly used as a cheap means to dispose of vegetative debris on
5 deforested land for the purpose of agriculture or pasture, and to maintain an open landscape. Fire
6 represents a major source of carbon emissions in the Amazon, especially in the southern fringe of
7 the Amazon known as the arc of deforestation (Fearnside et al., 2009). Various factors contribute
8 to relatively high rates of deforestation in this region, including available infrastructure and
9 capital and relatively long dry seasons (Kirby et al., 2006). Fire susceptibility is influenced by
10 factors such as deforestation, forest management, cultivation and reforestation (Osborne, 2000).
11 Development of human infrastructure including road construction, and the spread of residential
12 settlements in general increase deforestation and fire activity in more remote regions (Nepstad et
13 al., 2001; Malhi et al., 2009). In 2005, approximately 90% of the observed burning in the
14 southeast of Rondônia and the northwest of Mato Grosso occurred in previously deforested
15 areas, thereby demonstrating the link between fire and land use and land cover change (LULCC)
16 processes (Lima et al., 2012).

17 The objectives of landowners partly govern the degree to which fire is used in the deforestation
18 process. In general, three stages can be distinguished: logging, clearing and maintenance. In
19 tropical logging, by only harvesting selected trees, bulldozers create networks of forest roads,
20 killing non-harvested trees and fragmenting the forest and its canopy (Johns et al., 1996). This
21 practice increases the amount of dead wood in the understory and the likelihood of fires. The
22 damaged canopy allows light and wind to penetrate, increasing desiccation (Holdsworth and Uhl,
23 1997; Cochrane and Laurance, 2008). If fires occur they can affect forest composition and
24 structure. Recovery is diminished due to the impact of surface fires on seed availability (Van
25 Nieuwstadt et al., 2001). For clearing, slash-and-burn is a very efficient technique to clear land
26 intentionally for cattle or pasture (Tinker et al., 1996). To prepare the field, vegetation is cut and
27 left on the ground to dry. At the end of the dry season this is subsequently burned. If biomass
28 remains after the first fire, the remains are piled up and burned again, often until all biomass is
29 removed. Once everything is cleared, planting can be done directly in the ashes. Forest edge
30 biodiversity is impoverished by grass invasion due to repetitive burning in rapid agricultural

1 expansion (Balch et al., 2009). Maintenance fires are used regularly to prevent trees from
2 invading pastures after deforestation, for nutrient recycling and to remove crop residues (van der
3 Werf et al., 2009).

4 The vast scales of fire and deforestation have become increasingly evident as a result of
5 developments in remote sensing and research that has used the continuously increasing data
6 streams associated with this. Often these deforestation-focused studies used NASA's Landsat
7 data (Tucker et al., 2004), but coarser scale Moderate Resolution Imaging Spectroradiometer
8 (MODIS) data is also used for rapid identification of deforestation location areas and trends in
9 dynamics (Morton et al., 2005). Landsat generated deforestation datasets allow accurate
10 quantification of the total area of deforestation (Broich et al., 2009). Now spanning a time frame
11 of approximately 25 years, the Brazilian National Institute for Space Research (INPE) has
12 created the Amazon Deforestation Monitoring Project (PRODES) to quantify deforestation rates
13 in the Brazilian Legal Amazon (BLA) (INPE, 2014). Initially only available in a tabular format,
14 digital maps were created using a new methodology based on the work of Shimabukuro et al.
15 (1998) from 1997 onwards (Hansen et al., 2008). PRODES has been used in many studies to
16 evaluate the effects of deforestation (e.g., Morton et al., 2006; Broadbent et al., 2008). Also
17 using Landsat satellite imagery, the Global Forest Change project (GFC) has compiled a global
18 forest cover loss dataset at the same spatial resolution as PRODES with an annual time step for
19 the period 2001 – 2013 (Hansen et al., 2013). The objectives of these two datasets diverge as
20 PRODES only focuses on primary forest loss, while GFC includes forest changes in every type
21 of vegetation taller than 5m in height and is a global data set while PRODES only focuses on the
22 BLA.

23 Fire has been used as part of the clearing process since the inauguration of the Transamazon
24 Highway in 1970 (Fearnside, 2005). Different remotely sensed datasets allow the quantity and
25 intensity of fires to be estimated. Active fire algorithms (e.g., Giglio et al., 2003) detect a fire at
26 the time of satellite overpass, while burned area algorithms (e.g., Roy et al., 2008) use satellite
27 time series of vegetation indices and other information to classify whether a pixel was burned or
28 not. Burned area algorithms are less sensitive to cloud cover and smoke and are therefore more
29 reliable for spatiotemporal studies allowing better analysis of spatial association between
30 landscape attributes and burning events, while active fires are better for identifying smaller fires
31 and yield near-real time information (Roy et al., 2008).

1 After the first fire following deforestation there is generally still too much biomass on the ground
2 to enable agriculture or cattle grazing. Usually, repetitive burning is carried out to remove
3 remaining biomass. The number of fires is thought to depend on the post-deforestation land use
4 with agriculture requiring more complete combustion than pastures. For example, in a previous
5 modeling exercise pasture combustion completeness was set between 50-90%, while conversion
6 to cropland may lead to 100% combustion completeness over a period of 1-3 years (Morton et
7 al., 2008). The connection between these fires and deforestation is well established. For example,
8 according to a study focusing on one Landsat 5 TM tile (231/67 in the WR-2 indexing system) at
9 the border of Rondônia and Mato Grosso, 31% of the area burned in 2005 was deforested in one
10 of the preceding 3 years (Lima et al., 2012).

11 Deforestation and forest cover loss rates are impacted by numerous socio-economic factors, and
12 via the fire – deforestation link drought also plays a role in explaining spatio-temporal variability
13 (van der Werf et al., 2008). In general, drier than normal dry seasons see increased fire activity
14 (Chen et al., 2013) which can inhibit rainfall, creating a feedback mechanism (Nobre et al., 1991;
15 Nepstad et al., 2001). In addition, the vulnerability of these forests may increase due to climate
16 change (Aragão et al., 2008) with increased temperatures and stronger droughts (IPCC, 2007). In
17 2005, 2007, and 2010, three strong droughts occurred in the Amazon. Standardized anomalies of
18 dry-season rainfall showed that in 2005 37%, and in 2010 57%, of the total area was
19 experiencing rainfall more than 1 standard deviation below the long-term mean (Lewis et al.,
20 2011). The 2005 event increased the number of fires observed by satellite-derived monthly and
21 annual time series by 33% compared to the 1999-2005 mean (Aragão et al., 2007). Droughts are
22 associated with the El Niño – Southern Oscillation, but are also impacted by tropical Atlantic sea
23 surface temperature (SST) anomalies related to the Atlantic Multidecadal Oscillation (AMO)
24 (Marengo et al., 2008). Eastern Amazonia is more vulnerable to fire during El Niño whereas the
25 south and southwest is more linked to AMO (Chen et al., 2011). Southern and Eastern states in
26 the Amazon basin are relatively dry with lower annual rainfall and stronger dry seasons, while
27 the interior of the Amazon is more humid (Nepstad et al., 1994).

28 In this study we focus on the relationship between burned areas and forest cover loss using newly
29 available data. We discuss and highlight the patterns seen in the forest cover loss, deforestation,
30 and burned area datasets. Our main objective is to identify the role that fires play in these
31 differences, and how droughts influence the fire – tree cover loss dynamics. We also focus on the

1 spatial and temporal variability in the use of fire as a tool for deforestation for different states
2 and land cover types.

3 **2 Datasets and Methods:**

4 Our study area is the Brazilian Legal Amazon or BLA, which consists of 9 Brazilian states in the
5 northern part of Brazil (Fig. 1). Its geographic area is approximately 5 million km². 62% of the
6 Amazon forest lies within Brazilian borders. It consists predominantly of closed tropical forest
7 but also of cerrado (wooded savanna), agriculture, and pasture land (Carreiras et al., 2006).
8 Deforestation is concentrated in the arc of deforestation along the southern and eastern edges of
9 the BLA (Fearnside, 2005). We used four satellite-derived datasets characterizing forest
10 dynamics and fire: GFC forest cover loss, PRODES deforestation, MODIS burned area, and
11 MODIS land cover. After describing these datasets in more detail, we explain what steps we
12 have taken to study relationships between the various datasets.

13 The PRODES deforestation dataset is derived from Landsat data using algorithms described by
14 Shimabukuro et al. (1998). Landsat TM bands 3, 4 and 5 are used to generate the fraction of
15 vegetation, soil, and shade within each pixel using the constrained least squares method. The
16 image fraction is then classified using image segmentation followed by unsupervised image
17 classification and image editing to determine primary forest deforestation (Valeriano et al.,
18 2004). This methodology has been used since 1997 and once an area is detected as deforested, it
19 is masked out and not considered anymore in following years.

20 GFC defines forest cover loss as a stand-replacement disturbance or the complete removal of tree
21 cover canopy for the period 2001 – 2013. GFC uses pre-processed cloud-free Landsat images to
22 calculate time series spectral metrics. Over the study interval three groups of per-band metrics
23 were employed: reflectance values, mean reflectance and slope of linear regression of band
24 reflectance value versus image date. Using these time-series metrics, a decision tree was used to
25 find forest cover loss (Hansen et al., 2013). A key difference between GFC and PRODES is that
26 PRODES only quantifies primary forest loss while GFC includes changes in “all vegetation taller
27 than 5m in height”. This may include the clearing of secondary forest or certain types of cultures
28 such as palm oil production, and in general one expects that GFC flags more area than PRODES.
29 MODIS burned area (MCD64A1) is based on 500 meter MODIS imagery coupled with 1km
30 MODIS active fire observations. It produces composite imagery to detect persistent changes and
31 then uses spatial and temporal active-fire information to create probabilistic thresholds to classify

1 each pixel as burned or unburned (Giglio et al., 2009). To identify land cover changes, we used
2 the MODIS MCD12Q1 Land cover type product version 051 (Loveland and Belward, 1997;
3 Friedl et al., 2010) with the International Global Biosphere Program (IGBP) global land cover
4 classification scheme.

5 To combine forest cover loss, burned area and land cover datasets we used ArcMap10 to rescale
6 PRODES and GFC to the 500-meter MODIS grid using the most common Z value. The rescaled
7 dataset slightly overestimated the area of forest loss in the BLA by approximately 4%. However,
8 this bias was uniform over time and space, and did not change the general trend and patterns in
9 which we are interested here. We performed all analyses at 500 meter resolution and rescaled the
10 results to 0.1 degree to facilitate visual interpretation of the figures.

11 Because of the annual time step in the GFC and PRODES data, we have also merged the
12 monthly burned area data into annual files to compare that layer with deforestation and forest
13 cover loss. We have simplified the land cover classification scheme to three classes with the
14 dominant vegetation types in the BLA: Evergreen Broadleaf forest (EB), Low Tree Cover (LT,
15 which includes shrublands and savannas) and Cropland (C, including croplands interspersed with
16 natural vegetation). Other land cover type, such as grassland, needle leaf forest or permanent
17 wetland, have been merged (Other) but excluded in our analyses as they only represent a very
18 small portion of BLA land cover. We further analysed the land use for each land cover group in
19 the two states most affected by fires: Mato Grosso and Pará. We used the Landsat-based
20 TerraClass land use dataset from INPE, rescaled at 500-meter, available for 2008 and 2010 for
21 this exercise (de Almeida et al., 2009).

22 Finally, we aimed to not only better understand how burned area and forest cover loss are
23 related, but also how frequently burned area is observed after land cover transition and whether
24 this varies between post tree cover loss land use. For each evergreen broadleaf pixel where GFC
25 indicated forest cover loss, we summed burned area occurrence during the 5 years following this
26 event. This 5 years time period was chosen to balance the number of years we had for analysis
27 (which decreases when we lengthen our time period) and the time required for conversion. The 5
28 year time window was also used by Lima et al. (2012), who showed that 15% of the fires in their
29 study area in 2005 occurred in land deforested in 2000. We tested time periods from 2 to 6 years
30 and found that shorter time periods lowers this number somewhat (e.g., 3 years yields 37%)
31 while longer time periods make a marginal difference. Because our time series ended in 2012 we

1 could only do this for 2002 through 2007. In the process, we kept track of the post tree cover loss
2 land use defined by the land cover type 5 years after tree cover loss. For each year of evergreen
3 broadleaf forest cover loss, we calculated the fraction burned in the subsequent 5 years, and
4 associated them with the initial year of tree cover loss. Finally, we stratified these summed
5 fractions by post tree loss land cover type to evaluate potential trends in conversion and the use
6 of fire in them.

7 **3 RESULTS:**

8 During our study period, within the BLA, both tree cover loss and burned areas were
9 concentrated along the arc of deforestation (Fig. 2). The largest difference between GFC and
10 PRODES on one hand and burned area on the other hand is that forest loss and deforestation are
11 concentrated along the edge of the Amazon forest while most fire activity is observed further
12 away from the deforestation frontier. In addition to the arc of deforestation, forest cover loss and
13 deforestation occurs along the Amazon River and close to roads. A state-level breakdown shows
14 that Mato Grosso (both GFC and PRODES indicate 33% of the total BLA loss), Pará (32% for
15 GFC, 36% for PRODES) and Rondônia (12%, 13%) were the states with most forest cover loss
16 over our full study period. There is spatial overlap between forest cover loss / deforestation and
17 burned area, mainly in the Southeast of Pará and in Eastern Mato Grosso. However, regions with
18 more shrubland or savannas, further away from the arc of deforestation (e.g. Tocantins and
19 Maranhão) have the highest burned area.

20 Over our study period, PRODES and GFC annual forest cover loss increased up to the year 2004
21 and decreased afterwards (Fig. 3). While both datasets demonstrated this post-2004 decline, the
22 extent differed. PRODES showed a continuous decrease except for 2008, while GFC plateaued
23 towards the end of the study period at a higher level. In both datasets, all states showed a
24 decrease between 2004 (highest year of deforestation during our study) and 2012. Mato Grosso,
25 Pará and Rondônia showed the largest reduction between 2004 and 2012 in both datasets,
26 respectively 63%, 50% and 56% in GFC and, 94%, 80% and 80% with PRODES. While all other
27 states had a decrease of at least 40% according to PRODES, GFC does not show any other state
28 having more than a 30% decrease. In addition, GFC has annual peaks in forest cover loss during
29 dry years, especially in 2010, while PRODES does not show this variability in deforestation.

1 Partitioning GFC forest cover loss by land cover, we found that until 2006 most loss occurred in
2 evergreen broadleaf forests (Fig. 4a). From 2002 to 2009, the amount of annual evergreen
3 broadleaf forest loss dropped from approximately 16,000 km² to almost 5,000 km² and stayed
4 relatively constant in the following years. The most significant decrease came again from Mato
5 Grosso and Rondônia with an absolute forest cover loss reduction in evergreen broadleaf forests
6 of 93% and 81% respectively (Fig. 4b). The decrease in forest cover loss started earlier in Mato
7 Grosso (in 2004) than in Rondônia (in 2005) probably related to the strong 2005 drought, which
8 was most prominent in the western BLA. The state with on average the second highest amount of
9 forest cover loss, Pará, showed a slower diminution in evergreen broadleaf forest loss (60%).
10 Total loss in other types of land cover remained relatively stable during our study period (Fig.
11 4c-d). Mato Grosso is the only state with a significant decrease in clearing in the low tree cover
12 biome, halving the amount of shrubland and savannas cleared between 2004 and 2012.

13 In contrast to these trends in deforestation and forest cover loss, total burned area remained
14 relatively stable over our study period except in drier years, such as 2007 and 2010, when burned
15 area increased markedly, especially in low tree vegetation. These peaks mainly occurred in
16 southern and eastern states such as Mato Grosso, Pará, Maranhão and Tocantins (Fig. 5a,b). In
17 2008, pastures as classified by TerraClass covered 15% of Mato Grosso's and 12% of Pará's
18 surface. For a large part, pastures overlapped with our low tree cover class (80% and 65% of the
19 pastures were in that class for those two states). In Mato Grosso pasture fires represented only
20 7% of all burned area in 2008 and 9% in 2010, while most of the burned area occurred in the
21 "unmanaged land use" classes (respectively 86% and 87%). In Pará, pasture fires comprised
22 19% and 34% of the burned area in 2008 and 2010, also much lower than burned area in
23 unmanaged areas (69% to 56%).

24 Focusing on burned area coinciding with GFC forest cover loss (so observed in the same grid
25 cell in the same year), we found that they occur mainly in evergreen broadleaf forest and that this
26 is only a small fraction of the total burned area in the BLA (Fig. 5c,d). The general trend is
27 similar to total GFC although the decrease starts after 2005 due to a drought in the southwest of
28 the BLA, mainly in Acre and Amazonas. Looking into more detail at 4 years (2004, 2005, 2007
29 and 2010) with above-average tree cover loss and/or burned area, we found that the spatial
30 distribution of coinciding forest cover loss and burning has changed (Fig. 6). Especially in 2004
31 but also in 2005, most of this was observed in northern Mato Grosso and Rondônia. These

1 regions featured less prominently in the latter part of this analysis during the dry years of 2007
2 and 2010. In contrast, Pará had very little burned area in 2004, but started to have a major
3 hotspot in the southeast in 2005 and in following dry years. The central region of Pará had also
4 more activity from 2005 onwards

5 Using burned area and GFC to evaluate the time interval between forest cover loss and fire we
6 found that a substantial part (46% when averaged over the full study period) of cleared evergreen
7 broadleaf forest was burned within 5 years after forest cover loss, with the use of fire decreasing
8 over time. Of all forest loss grid cells, 24.5% was converted to shrubland or savannas, 18% was
9 converted to cropland and the remaining 3.5% had no permanent conversion (Fig. 7). We also
10 found that 10% of the land converted to shrubland or savanna was burned the same year as forest
11 cover loss was detected in these grid cells (we could not identify whether it was the cause of
12 conversion), while conversions to cropland were burned in the first year in about 7.5% of the
13 grid cells. Grid cells identified as forest cover loss but remaining evergreen broadleaf forest had
14 little burned area, between 0-2%.

15 Strikingly, when looking at the temporal trend of burned area after evergreen broadleaf GFC
16 loss, we found that the use of fire after conversions to shrubland, savannas or cropland decreased
17 over our study period (Fig. 8). The fraction that burned for conversion to shrubland or savannas
18 decreased from over 31% between 2002 and 2007 to 17% between 2007 and 2012. Combining
19 all conversion types, post tree cover loss burned area was most frequent for the grid cells that had
20 forest loss in 2003 where 60% burned in the following 5 years, decreasing to 29% for 2006, and
21 increasing again to 39% in 2007. This trend came mainly from grid cells that underwent any kind
22 of conversion. In the grid cells without permanent conversion the use of fire actually increased
23 somewhat (Fig. 8)

24 **4 Discussion**

25 Multiple lines of evidence demonstrate the successful decrease of deforestation within the BLA
26 under the 2004 Action Plan for the Prevention and Control of Deforestation in the Legal
27 Amazon. This plan introduced innovative procedures for monitoring, environmental control and
28 territorial management (Assunção et al., 2012). In the same year, the collaboration between
29 INPE and the Center for Environment Monitoring (CEMAM) from the Brazilian institute of
30 environment and renewable natural resources (IBAMA), created the System for Detection of

1 Deforested areas in Real Time (DETER). This new program generated maps of areas with
2 critical forest cover changes in near real time which proved to be an important tool for law
3 enforcement (May et al., 2011). In both datasets used in this study, the post-2004 drop in tree
4 cover loss and deforestation can be clearly identified. According to GFC, the resulting downward
5 trend is mainly occurring in states within the arc of deforestation. In 2006, the soy moratorium
6 was implemented to prohibit the trade of soybean produced in deforested areas (Meyfroidt and
7 Lambin, 2011). After 5 year of application, Rudorff et al. (2011) concluded that although it was
8 premature to attribute the decrease in deforestation in the Amazon biome to this moratorium, it
9 had undoubtedly created an inhibitory effect on the expansion of soybean in this biome. After
10 this moratorium, 90% of the soybean plantation expansion occurred on land previously used as
11 rangeland (Lapola et al., 2010).

12 In the future, the Brazilian government is willing to significantly increase the production of
13 biofuel (Bergmann et al., 2013). One consequence could be the migration of cattle ranchers to
14 other regions and thereby increasing deforestation (Nepstad et al., 2006). According to GFC,
15 forest cover loss ceased to decrease shortly after the implementation of this moratorium. This
16 study cannot connect these two facts, but one hypothesis could be that more non-primary forest
17 has been cleared to allow the soybean extension or the resettlement of cattle ranchers without
18 breaching the moratorium agreements.

19 Pará ranked second with regard to forest cover loss and deforestation in 2002, behind Mato
20 Grosso. Mato Grosso drastically reduced its tree clearing rates, and so did Pará but to a lesser
21 extent. Mato Grosso had only 44% of its state covered by evergreen broadleaf forest in 2002
22 compared to 80% for Pará. The availability of land accessible to clear may explain part of the
23 difference in absolute trend strength. We observed that some regions in Pará that were not
24 substantially impacted by clearing fires early in our study period became hotspots towards the
25 end. For example, in the region of Tucumã in central Pará, fire and clearing were not collocated
26 in 2004, but became really active in 2005, 2007 and 2010. A highly fragmented forest region
27 may potentially explain this, as a close access to the Rio Xingu, tributary of the Amazon River,
28 facilitates tree removal. Further north in Pará, we noted that strong tree cover loss was occurring
29 without burned area being observed. This could be due to higher annual rainfall, which can limit
30 the use of fire in forest clearing.

1 During our study period, the BLA faced severe drought years in 2005, 2007 and 2010. According
2 to PRODES, these events did not influence the rate of deforestation, except in 2008 when there
3 was a small increase possibly related to the 2007 drought. GFC shows more sensitivity to
4 droughts. This is most evident in 2010 when there was an increase in forest cover loss compared
5 to the previous year, the first time since 2004. But also 2005 and 2007 showed less of a decline
6 compared to the previous year in GFC than in PRODES. These dry years had also a significant
7 effect on fire in southern and western states. Maranhão and Tocantins had a very strong burning
8 peak, but this did not influence forest cover loss, probably because savannas are by far the
9 dominant land cover here. On the other hand, Mato Grosso and Pará have also seen elevated
10 burning activity increasing the burning of evergreen broadleaf forest, which could explain part of
11 GFC plateau towards the end of our study period.

12 Previous work has shown a strong correlation between deforestation and fire occurrences
13 (Aragão et al., 2008; Chen et al., 2013). Although deforestation and forest cover loss have
14 declined sharply since 2004, burned area did not follow this trend because only a small fraction
15 of burned area is associated with forest cover loss and because droughts have a strong impact on
16 the inter-annual variability in burned area. Although the effects of droughts are not uniform
17 between all states, general impacts are evident. Burned area always exceeded forest cover loss
18 area due to the fact that fires in grassland and on agricultural grounds are not part of
19 deforestation or forest cover loss. States with a relatively small fraction of evergreen broadleaf
20 forest but a larger portion of cerrado (low tree cover), such as Mato Grosso, Tocantins and
21 Maranhão have in general more fire activity in the conversion to agriculture or grazing process
22 (Mistry, 1998). In Mato Grosso and Pará, we found that relatively speaking, less burned area in
23 our low tree cover class occurred in pastures compared to non-managed land. This indicates that
24 management in general decreases burned area for this land cover class.

25 Our results on time intervals between forest cover loss and burned area show that although large
26 fire years occurred throughout our study period, the overall rate of burned area per unit forest
27 loss is decreasing. In other words, over our study period the post-conversion use of fire
28 decreased. About 54% of the evergreen broadleaf forest that was cleared in 2002 burned in the
29 following 5 years, dropping to 29% for the evergreen broadleaf forest cleared in 2006. However,
30 this number was somewhat higher (39%) in 2007, which was a dry year and the second highest
31 fire year during our study period. Longer time series will enable us to see whether the increase in

1 post-forest loss fire use in 2007 was a temporary result because of the drought or that the
2 downward trend is increasing again. Lima et al. (2012) showed in one Landsat 5 TM tile located
3 between Rondônia and Mato Grosso that 53% of the 2005 fires occurred in land deforested
4 within the prior 5 years. Aragão and Shimabukuro (2010), however, found that between 2000
5 and 2007, 59% of areas experiencing reduction in deforestation rates had an increase in fire
6 events. Using 0.25-degree spatial resolution, they explain that this reverse trend could come from
7 secondary forest clearing, not detected by PRODES. Our results on 500-meter resolution indicate
8 that within the same time period (2000-2007), more than 50% of the cleared vegetation taller
9 than 5m in height was burned within a few years but this number dropped below 40% after 2004.
10 This shows that burned area per unit forest loss area decreased after Aragão and Shimabukuro
11 (2010) study period.

12 By combining the three different datasets we have aimed not only to understand the differences
13 between them, but also to highlight uncertainties associated with relying on one dataset. Our
14 study shows that, at least for the Brazilian Amazon, secondary forest loss and fire dynamics
15 outside the primary tropical forest domain may be equally important in regulating atmospheric
16 CO₂ concentrations as the much better studied dynamics within the primary forest. The exact
17 impact of these processes requires taking into account the biomass density (e.g., Baccini et al.
18 (2012)), which will be the subject of a follow-up study. These types of fires are also important in
19 other regions outside the BLA (e.g. Chen et al., 2014) and expanding the current focus on
20 deforestation regions to also include woodland dynamics may be an important step in better
21 understanding biosphere-atmosphere interactions and biodiversity issues. For example, large
22 scale carbon budget studies often use deforestation rates in their assessments of sources and sinks
23 of carbon (Le Quéré et al., 2013).

24 One key uncertainty we have not addressed here is the use of burned area to study fire extent.
25 The burned area algorithms are often geared towards detecting relatively large fires and not all
26 fires in tree cover loss or deforestation zones may be flagged as burned by them. In addition, it is
27 not well known how reliable the burned area data is in tree cover loss or deforestation regions.
28 The use of active fires instead of or in combination with burned area may mitigate some of these
29 issues (Randerson et al., 2012). However, these approaches rely on statistics between active fires
30 and burned area that can only be derived from coarse scale analyses; in the study by Randerson
31 et al. (2012) a 0.25° spatial resolution was used which is far too coarse for our purposes. On

1 native resolution, active fires are still 4 times coarser than burned area but the key issue is that
2 active fires can be associated both with a small agricultural burn and with a large fire consuming
3 all biomass in the grid cell. Because of this lack of quantitative information we refrained from
4 using active fires. In the future, higher resolution observations from for example VIIRS
5 (Schroeder et al., 2014) may enable simultaneous use of burned area and active fires in native
6 resolution applications like ours.

7 **5 Conclusions**

8 We used satellite observation of deforestation (PRODES), forest cover loss (GFC), burned area
9 (MCD64A1) and land cover (MCD12Q1) between 2002 and 2012 in the Brazilian Legal
10 Amazon (BLA) to study relationships between the different datasets. The PRODES dataset
11 indicated that deforestation has sharply declined since 2004, something also seen in the GFC
12 forest cover loss data but to a smaller extent. The states with the strongest decline were in the
13 south of the BLA. Most tree cover loss and deforestation occurred on the edge of the Amazon
14 forest while more burned area was located on the outskirts of the arc of deforestation. Looking at
15 coinciding forest cover loss and burned area, we found variability in the spatial distribution
16 during our study period. This was most predominantly found in the state of Pará, where burned
17 area and forest cover loss were not linked up to 2004, which changed later when large patches of
18 burned area coinciding with forest cover loss were observed in the southwest and the centre of
19 the state during the dry years of 2007 and 2010. This indicates that droughts not only enhance
20 burned area, but also increase forest cover loss in for example secondary forests. These dynamics
21 are not necessarily monitored by PRODES, which excludes grid cells that have been deforested
22 before, but are important to be included in large-scale carbon cycle studies.

23 Early in our study period, more than 50% of cleared evergreen broadleaf forest was burned
24 throughout the subsequent 5 years, reducing to below 40% during the last time segment studied
25 (2007-2012). In other words, besides the well-documented decrease in deforestation rates we
26 also showed that burned area per unit forest loss area decreased over our study period. More
27 research is required to better understand these dynamics; one potential reason could be that with
28 lower deforestation rates the previously deforested lands are becoming more intensively
29 managed. This is partly supported by our finding that in Mato Grosso and Pará burned area in in
30 managed pasture is on average 5 times less frequent than in unmanaged land.

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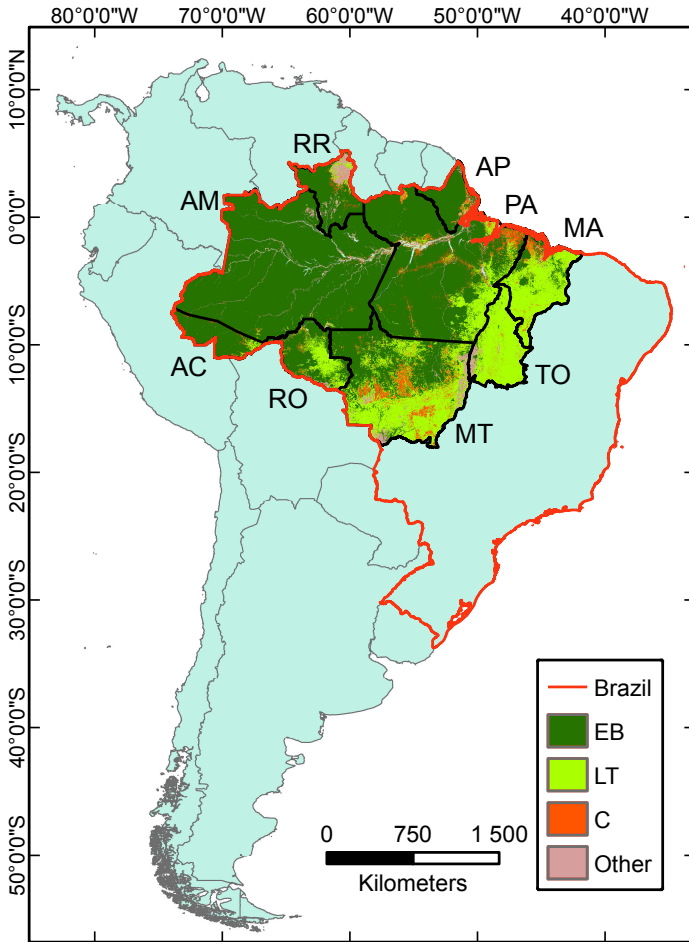
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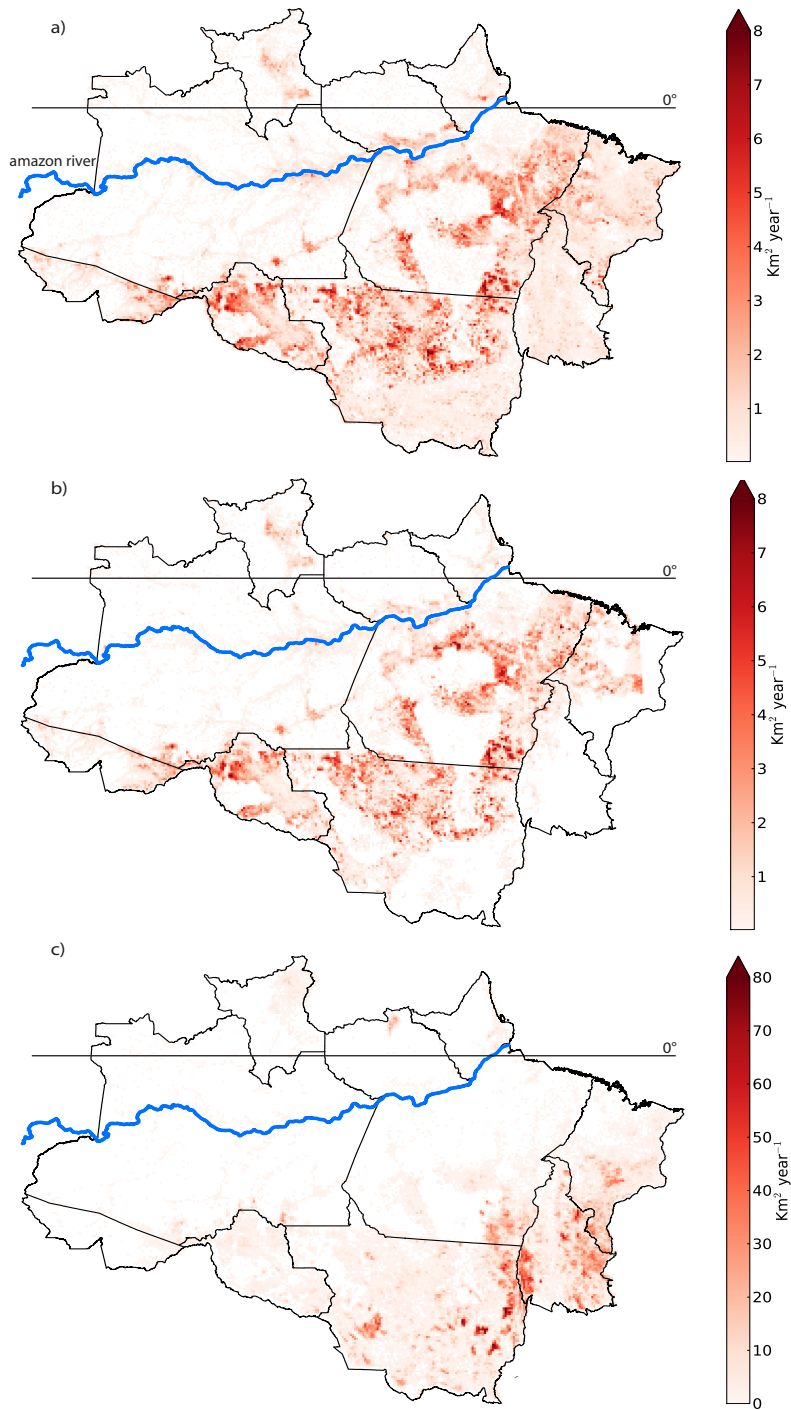
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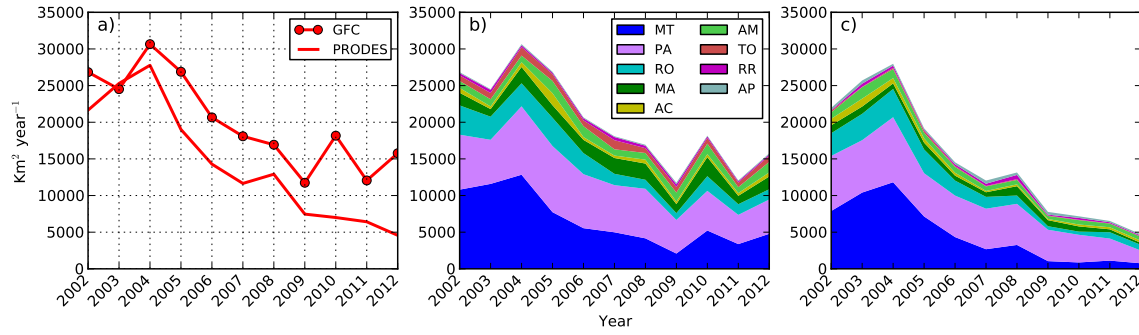


6 Figure 1: MODIS MCD12Q1 (v051) Land cover in the Brazilian Legal Amazon (BLA). EB =
7 Evergreen Broadleaf forest, LT = Low Tree cover (shrubland and savannas) and C = Cropland
8 and Cropland/Natural Vegetation Mosaic. Other includes all remaining land cover such as
9 grassland, needle leaf forest, permanent wetlands or urban areas. Letter codes for the states in the
10 BLA are clockwise: AP, Amapá; PA, Pará; MA, Maranhão; TO, Tocantins; MT, Mato Grosso;
11 RO, Rondônia; AC, Acre; AM, Amazonas; RR, Roraima.

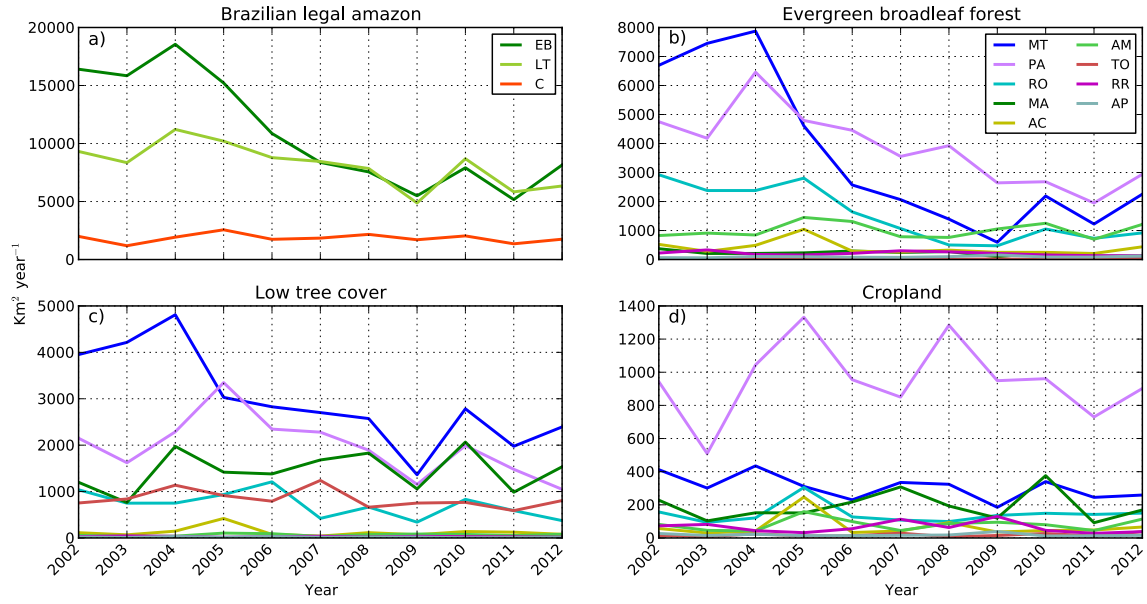
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1
 2 Figure 2: 2002-2012 mean forest loss (a, based on GFC), deforestation (b, based on PRODES),
 3 and burned area (c, based on MCD64A1). All data sets are regridded to 0.1° resolution. Note the
 4 order of magnitude difference between forest loss and deforestation one hand and burned area on
 5 the other hand.

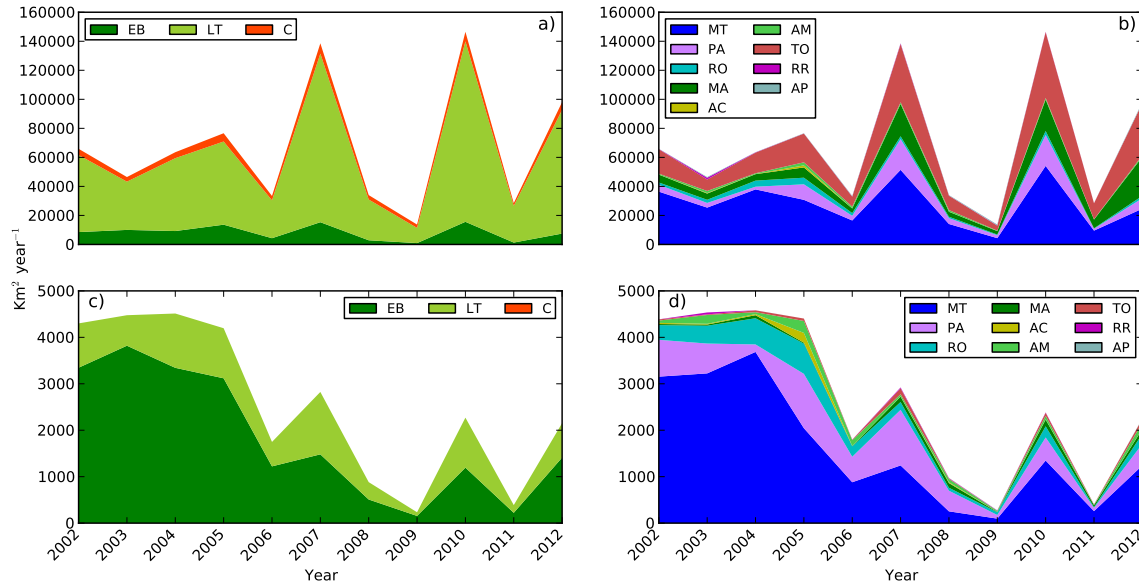


1
 2 Figure 3: Annual forest loss estimates from the Global Forest Change (GFC) project and
 3 deforestation estimates from the Amazon Deforestation Monitoring Project (PRODES) for 2002-
 4 2012 in all of the BLA (a), and partitioned on state level for GFC (b) and PRODES (c).
 5

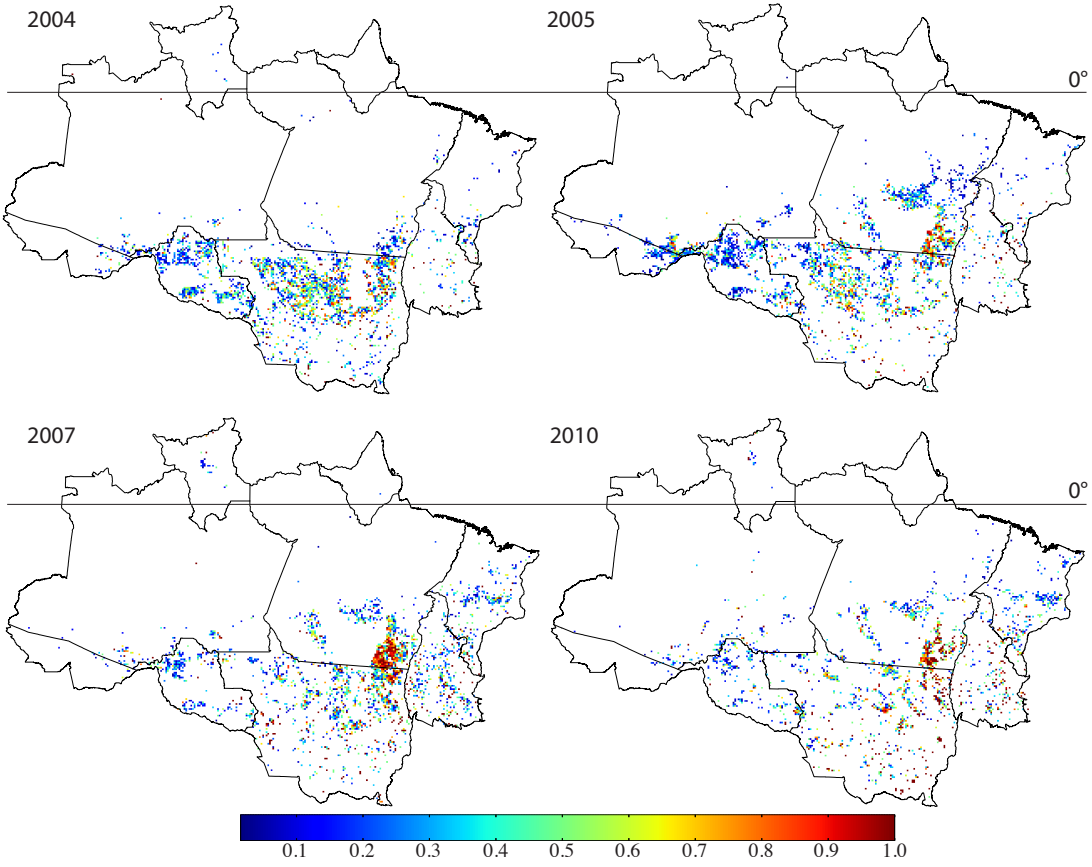


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2 Figure 4: Annual state-level GFC forest loss for the Brazilian legal amazon and three land cover
 3 classes defined at the year of forest loss.



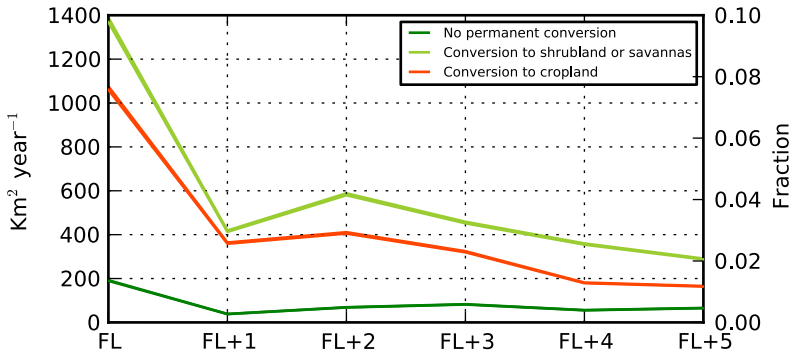
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 2 Figure 5: Burned area stratified by land cover (a, c) and by state (b, d). Top panels (a and b)
 3 indicate total burned area; bottom panels show burned area coinciding with GFC forest loss.



1

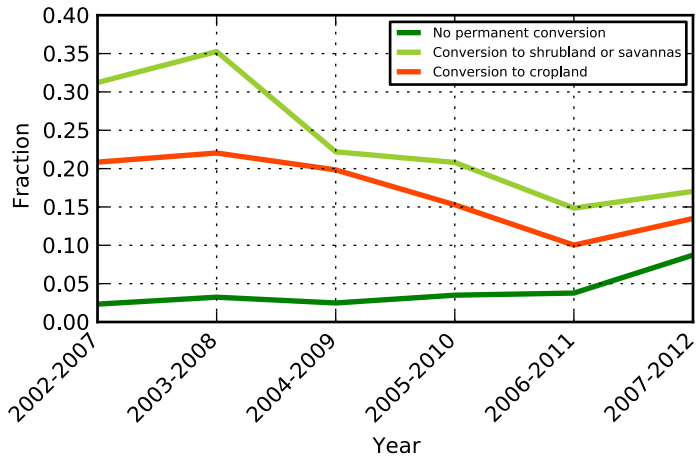
2 Figure 6: Fraction of GFC forest loss coinciding with burned area for the 4 highest fire and/or
 3 tree cover loss years in our study period, averaged on a 0.1° resolution.

4



1

2 Figure 7: The degree to which fire is used in the deforestation process for different post-forest
 3 loss land cover, shown for the year of forest loss (FL) and 5 years afterwards. Data is averaged
 4 for the 6-year periods starting between 2002 and 2007. The right y-axis indicates the burned area
 5 as a percentage of 2002-2007 averaged evergreen broadleaf forest loss ($14,211 \text{ km}^2 \text{ year}^{-1}$) and in
 6 total sums up to 46% of total forest loss.



1

2 Figure 8: Fraction of forest loss grid cells where burned area is observed in the 5 years following
 3 forest loss. Results are shown for three different post-forest loss land cover types defined as the
 4 land cover type observed 5 years after the forest loss event.