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Impacts of Climate Change and the End of Deforestation on Land Use in the Brazilian Legal Amazon

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ABSTRACT: Climate change scenarios vary considerably over the Amazon region, with an extreme scenario projecting a dangerous (from the human

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perspective) increase of 3.8°C in temperature and 30% reduction in precipitation by 2050. The impacts of such climate change on Amazonian land-use dynamics, agricultural production, and deforestation rates are still to be determined. In this study, the authors make a first attempt to assess these impacts through a systemic approach, using a spatially explicit modeling framework to project crop yield and land-use/land-cover changes in the Brazilian Amazon by 2050. The results show that, without any adaptation, climate change may exert a critical impact on the yields of crops commonly cultivated in the Amazon (e.g., soybean yields are reduced by 44% in the worst-case scenario). Therefore, following baseline projections on crop and livestock production, a scenario of severe regional climate change would cause additional deforestation of 181 000 km² (+20%) in the Amazon and 240 000 km² (+27%) in the Cerrado compared to a scenario of moderate climate change. Putting an end to deforestation in the Brazilian Amazon forest by 2020 (and of the Cerrado by 2025) would require either a reduction of 26%–40% in livestock production until 2050 or a doubling of average livestock density from 0.74 to 1.46 head per hectare. These results suggest that (i) climate change can affect land use in ways not previously explored, such as the reduction of yields entailing further deforestation, and (ii) there is a need for an integrated/multidisciplinary plan for adaptation to climate change in the Amazon.

**KEYWORDS:** Tropical agriculture; Cattle ranching; Climate change adaptation; Integrated assessment; Cerrado

### 1. Introduction

The Amazon has been recognized as a region particularly vulnerable to climate change over this century (Lenton et al. 2008; Malhi et al. 2008). Although climate change scenarios for the region differ considerably (Li et al. 2006), the high end of projections show a temperature increase of 3.8°C and up to 30% reduction in precipitation by 2050 (Figure 1). The impacts of such regional climate change and of the projected “forest dieback” on the vegetation dynamics, water and carbon cycle, and feedbacks with the global climate system have been extensively investigated in the last decade (Cox et al. 2000; Cox et al. 2004; Cramer et al. 2001; Huntingford et al. 2004; Huntingford et al. 2008; Sitch et al. 2008; Lapola et al. 2009a). In addition, field observations (Gash and Nobre 1997) as well as modeling studies (Nobre et al. 1991; Costa and Foley 2000; Sampaio et al. 2007) have shown that there is considerable change in the local and regional climate after the replacement of forest by pasture or crops. On the other hand, considerably less research has been done to assess the effects of future climate on land-use and land-cover dynamics in the Amazon region.

Recent extreme climate events, like the 1997/98 El Niño drought (Nepstad et al. 1999a) or the droughts of 2005 (Marengo et al. 2008) and 2010 (Lewis et al. 2011), brought considerable reductions in crop/pasture productivity and food shortage, among a variety of other relevant impacts inside and outside the Amazon (Nepstad et al. 1999; Nepstad et al. 2001; Moran et al. 2006; Bronzicio and Moran 2008; Lenton et al. 2009). Modeling studies by Cox et al. (Cox et al. 2000; Cox et al. 2004; Cox et al. 2008) project a future in which the Amazon would be exhibited in a permanent El Niño–like climate after 2040 and that events like the 2005 drought will increase in frequency from a 1-in-20-yr event to a 16-in-20-yr event by 2050.
Therefore, in light of the impact extreme climate events had on agriculture in the past and considering that these events might come close to the future “norm” (Battisti and Naylor 2009), the impacts of future climate change on land-use and land-cover change are highly relevant (see Lambin and Geist 2006, p. 174).

Agricultural activities are now solidly established in the Brazilian Amazon (Nepstad et al. 2006), especially the lucrative soybean farming, which had an increase in area from 16 000 km$^2$ in 1990 to 60 000 km$^2$ in 2008 (IBGE 2010). Nearly 36% of the Brazilian cattle herd and pasture area is currently located in the Legal Amazon,$^1$ the only region in the country that has experienced an increase in pasture area in the last two decades (IBGE 2010; Barreto et al. 2008). Moreover, the Legal Amazon currently contributes 15% of the national agricultural gross domestic product (GDP) and had a (total) GDP growth of 6.6% yr$^{-1}$ in the 1999–2008 period, compared to the national average of 3.4% yr$^{-1}$ (Tomazela 2007;

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$^1$ The Brazilian states of Acre, Amapá, Amazonas, Mato Grosso, Rondônia, Roraima, Tocantins, and (part of) Maranhão. The Legal Amazon comprises 61% of the national territory, roughly 62% of the Amazon forest area (Soares-Filho et al. 2006), and has a population of 23 million people (IBGE 2010).
Salomon 2008). On the other hand, this surge of the Amazon economy was accompanied by increasing conservation concerns. For example, more than 75% of the area under strict protection in the Brazilian Amazon has been enacted after 1990 (ISA 2010), and since 2002 the protected area network has increased by 6400 km², now covering 51% of the remaining forest (Soares-Filho et al. 2010). In 2008, the Brazilian government made a formal announcement within the United Nations climate treaty framework of reducing Amazon deforestation by 80% compared to the historical rate of 19 500 km² yr⁻¹ by 2020 (Government of Brazil 2008; Nepstad et al. 2009). The interplay between these two apparently antagonistic issues (high growth of agricultural economy and the end of deforestation) in view of future climate change and growing demands for land (for food, feed, and biofuel production) calls for in-depth scientific research to provide a sound foundation for decision making.

Here, we applied a spatially explicit modeling framework to assess the impacts of climate change and conservation targets on land-use and land-cover changes (LUCC) in the Legal Amazon by 2050, taking into account projected levels of crop and livestock production. In this study, LUCC are affected by climate change via crop/pasture productivity. Two different scenarios of climate change are used, namely, moderate and extreme regional climate change. Additionally, we also investigate how 2050 crop and livestock production demands could be conciliated with the end of deforestation in the Brazilian Amazon forest and Cerrado savanna in the 2020s (Nepstad et al. 2009).

2. Methods

2.1. Model description

The central feature of our modeling framework is the Land Simulation to Harmonize and Integrate Freshwater Availability and the Terrestrial Environment (LandSHIFT) model, which simulates land-use and land-cover change on a 5-arc-min spatial resolution (Schaldach and Koch 2009). By using a “land-use systems” approach, it describes the interplay between anthropogenic and environmental system components as drivers for land-use change in three major land-use activities (settlement, crop cultivation, and grazing) and their competition for land resources. Moreover, LandSHIFT’s livestock module simulates not only the occurrence of pastures but also the intensity of grazing. The model has been applied and evaluated in assessments of the impact of grazing management in the Jordan River region (Koch et al. 2008), the quantification of future LUCC and water use by agriculture in Africa (Weiß et al. 2009), and LUCC associated with increased production of biofuels in Brazil (Lapola et al. 2010) and India (Schaldach et al. 2011).

The International Model for Policy Analysis of Agricultural Commodities and Trade (IMPACT) (Rosegrant et al. 2008) provides future projections of crop/livestock production in the Legal Amazon, and the International Futures (IFs) model (Hughes 1999) projects population growth. Because the latter projects population growth on country level, we assumed the Legal Amazon to entire Brazil population ratio of 0.12 in 2007 to be constant until 2050. The Lund–Potsdam–Jena (LPJ) Dynamic Global Vegetation Model for Managed Lands (LPJmL) is used to calculate crop and grassland potential productivity on a 0.5° resolution grid.
Bondeau et al. (2007). These three models (IMPACT, IFs, and LPJmL) provide inputs to LandSHIFT, even though they are not dynamically coupled to LandSHIFT. However, though the exchange of information between the models is unidirectional (from IMPACT and LPJmL to LandSHIFT), the assumptions used to generate their outputs are compatible. For example, the baseline scenario of IMPACT is consistent with the Special Report on Emission Scenarios (SRES) A2 emission scenario used to compute future crop yields with LPJmL (in terms of projected population and economical growth; see next section).

Starting from an initial land-use map (see appendix A), the spatial allocation of different land uses in subsequent time steps is based on a multicriteria suitability analysis following the equation

\[
\psi_k = \sum_{i=1}^{n} w_i p_{i,k} \times \prod_{j=1}^{m} c_{j,k}, \quad \text{with} \quad \sum_{i} w_i = 1, \quad \text{and} \quad p_{i,k}, c_{j,k} \in [0, 1],
\]

where the factor weight \( w_i \) determines the importance that each suitability factor \( p_i \) has at grid cell \( k \), whereas \( c_j \) represents possible constraints for changing the land-use type at that given cell. In this study, \( p_i \) includes slope, distance to paved roads, distance to all roads, vegetation type (for sources, see Soares-Filho et al. 2006), potential crop/grassland yield (from the LPJmL model), proximity to cropland, attraction to national markets (see below), and distance to deforested land. The latter is used only for grazing because it has the same effect as proximity to cropland in crop cultivation. Therefore, \( n = 7 \) for crop cultivation and \( n = 8 \) for grazing. Paved roads are updated following the road paving schedule in the study by Soares-Filho et al. (Soares-Filho et al. 2006). Secondary roads are updated using the outputs of the road constructor submodel of the “SimAmazonia 1” model of land-cover changes (Soares-Filho et al. 2006) under a “business as usual” (BAU) scenario, which is consistent with the aforementioned paving schedule. The attraction to national markets factor represents the influence of the Brazilian cities that are the biggest consumers of Amazonian agricultural products, especially meat. These cities are located in southeast and northeast Brazil (Barreto et al. 2008). The index is calculated using a unidirectional gravity-type model,

\[
NMa_k = \sum_{v} \frac{\text{Pop}_v}{d_k^2},
\]

where \( NMa_k \) is the national markets attraction exerted in the grid cell \( k \), determined by summing up the population of the five most populous cities in southeast and northeast Brazil (\( v = 5 \): São Paulo, Rio de Janeiro, Belo Horizonte, Salvador, and Fortaleza) weighted by the distance of these cities to the cell \( k \). Soil type is not considered as a \( p_i \) factor because of its spatial correlation with the factors crop/grass productivity and vegetation type.

Weights \( w_i \) were determined by using the analytic hierarchy process (AHP) test (Saaty 1980). Determination of the relative importance of each \( p_i \) factor in relation to the others \( R_{i,AHP} \), used as an entry to the AHP test, was determined by the normalized difference between the average of \( p_i \) over areas with and without land-use changes \( \bar{e}_i \).
where $\alpha_i$ is the average value of variable $p_i$ in the grid cells where land-use change has occurred in the 2001–06 period and $\lambda_i$ is the average value of variable $p_i$ in the grid cells of the 2001 map where land-use change has not occurred (excluding the land-use activity in question: e.g., crops).

Therefore, the higher the $e_i$ value, the higher the difference between the $\alpha_i$ and $\lambda_i$ averages and the importance of that $p_i$ factor. Then, $\text{RI}_{\text{AHP}}$ is determined with a pairwise comparison of $e_i$ from all $p_i$ factors. The procedure was repeated for the three major land-use activities considered here: crop cultivation, well-managed (WM) grazing, and poorly managed (PM) grazing (Figure 2). Overall, this procedure showed that the distance to roads and the distance to previously deforested areas are the most important factors for explaining current patterns of land-use change, in agreement with the analysis by Soares-Filho et al. (Soares-Filho et al. 2006). However, other factors contribute as well to explain the different land-use activities. For example, slope has a higher importance for the location of PM pastures than for other land-use activities. Interestingly, potential crop/grass productivity does not play an important role for the location of croplands and pastures.
The future land-use patterns modeled by LandSHIFT in this study follow the projected changes in these factors (e.g., the road paving schedule mentioned above). Although prices are not considered as one \( p_i \) factor influencing the spatial location of land uses, it is indirectly considered in this method. For example, the land located closer to paved roads obviously has a higher associated value or opportunity cost (see Nepstad et al. 2009) than that located farther from paved roads. The same can be deducted also from factors like slope and potential crop/pasture yields.

Constraints \( c_j \) comprise conservation areas and land-use transition. The level of constraint for each category of conservation area (strict protection is 0.19, sustainable use is 0.66, indigenous reserve is 0.54, military reserve is 0.01, and not protected is 1.0) was derived from the analyses by Soares-Filho et al. (Soares-Filho et al. 2006; Soares-Filho et al. 2010). In this study, the land-use transition constraints all have a value of 1.0 (i.e., no constraint), except the conversion from urban to other land use, which has a value of 0.0. The transition from forest to soybean is reduced to 0.1 after 2006 (until 2050) to simulate the soybean moratorium introduced in that year, which almost completely stopped deforestation directly caused by soybean (ABIOVE 2009).

The allocation algorithm assumes that crop cultivation takes place generally but not always in the most suitable cells for each crop/pasture type and calculates a “quasi optimum” spatial crop distribution. The Multiobjective Land Allocation (MOLA) heuristic used here seeks pattern stability and keeps previous land uses even if another crop/pasture type has a higher suitability in that cell. LPJmL potential yields are applied a crop-specific factor to match current crop yields with statistics from the study area (Schaldach and Koch 2009; IBGE 2010). These factors, which are calculated at the first simulation time step, account for uncertainties due to crop management, (e.g., multicropping) or discrepancies due to the aggregation of crop types to LPJmL crop functional types (e.g., LPJmL pulses represent extratropical pulses such as lentils). Crop production of a given grid cell \( k \) is defined as the potential crop yield at \( k \) multiplied by the area of \( k \) that is not covered by settlement.

Allocation of both types of pasture depends on the potential productivity of grass in the grid cells, based on a livestock feed supply–demand logic. Forage supply is calculated by summing up the grass productivity of every pasture cell multiplied by the fraction of biomass that is utilized by livestock (grazing efficiency \( g_e \)). Here \( g_e \) is equal to 0.37 in WM pastures and is 0.12 in PM pastures, meaning that WM pastures have a higher carrying capacity than PM pastures. These values of \( g_e \) are based on literature (Rueda et al. 2003; Camarão et al. 2000) and calibration (only in terms of total pasture area) against the initial land-use maps. Forage demand is determined by the multiplication of the total livestock herd by the average forage consumption per livestock unit (10 kg of dry matter per day; Krausmann et al. 2008). In this study, the word “livestock” refers to bovine species such as cattle and buffaloes, which represent by far the majority of the grazing livestock herd in the Legal Amazon. By overlaying the initial land-use map (appendix A) and the map of livestock density (LD) by FAO (FAO 2007a), we estimated that approximately 14% of the Legal Amazon livestock herd is located in PM pastures. Therefore, in the simulations in which PM pastures persist in the future (see next section), we assign a constant value of 14% of the total livestock herd to be
allocated in PM pastures (and 86% in WM pastures). A total of 95% of the livestock feed demand is fulfilled by forage from pastures, and the rest is from feed grains or crop residues (Krausmann et al. 2008). If forage demand is higher than supply, then new pasture cells are allocated, starting from grid cells with the highest suitability for grazing until demand is fulfilled. Average LD is calculated by dividing the livestock herd by the pasture area. Allocation of land-use activities follows the hierarchical order: settlement, crop cultivation, well-managed grazing, and poorly managed grazing. Only one land-use type can occur in a grid cell.

2.2. Input data and modeling protocol

LandSHIFT is initialized with a land-use/land-cover map for the year 2006, a map of population density (Goldewijk 2005), and national statistics of crop production and livestock herd (IBGE 2010). Socioeconomic projections include future demands for food (Rosegrant et al. 2008) and population growth (Hughes 1999) under a baseline scenario.

2.2.1. Population and economics

The human population in the study area increases from 24.2 million people in 2006 to 32.6 million people in 2050, representing an average annual growth of 0.8% yr\(^{-1}\). Brazil GDP increases from $954 \times 10^9$ U.S. dollars (USD; year 2000 dollars) in 2006 to $7,226 \times 10^9$ USD (year 2000 dollars) in 2050, with an average growth rate of 4.42% yr\(^{-1}\), which is comparable to those projected by the Intergovernmental Panel on Climate Change (IPCC) SRES A2 and A1 for Latin America (3.8% and 5.5% yr\(^{-1}\), respectively) (Nakicenovic and Swart 2000).

2.2.2. Agricultural production

The projections of IMPACT consider price effects that come from dynamics on both the supply and demand side of food and feed commodities (Rosegrant et al. 2008). Prices are calculated internally in IMPACT, to satisfy a market-clearing condition: that is, when world supply and demand for agricultural products are in balance and world net trade equals zero (Rosegrant et al. 2008). The IMPACT baseline scenario projects that the demand for agricultural products, especially cereals and animal products, will increase worldwide until 2050 driven by population growth and also by other factors like increasing demands for crop-based biofuels. Importantly, it projects considerable increases in the annual demand for meat in East Asia (+55%), South Asia (+133%), and Sub-Saharan Africa (+63%) and little changes in Latin America and other parts of the world (for region definition, see Rosegrant et al. 2008). Such an increase in the demand for food products pushes global food prices up, with beef prices increasing from $1,912 USD Mg\(^{-1}\) in 2000 to $2,504 USD Mg\(^{-1}\) in 2050 (+30%), rice prices increasing from $184 to $323 USD Mg\(^{-1}\) (+75%), maize prices increasing from $87 to $132 USD Mg\(^{-1}\) (+52%), and soybean prices increasing from $205 to $328 USD Mg\(^{-1}\) (+60%). The production (supply) of agricultural goods also increases worldwide, with a remarkable increase in the production of cereals in North America, Europe, and
Central Asia. In terms of livestock production, the largest share of global production (421 × 10^6 Mg)\(^2\) in 2050 is in East Asia (33%), followed by North America and Europe (28%) and Latin America (11%). Latin America’s livestock product output increases by 180% over the 2006–50 period (3.6% yr\(^{-1}\)). A similar growth is also projected for South Asia, although it comprises a smaller share of global production (7%) in 2050.

Crop production in the Legal Amazon increases by 93%, with soybean production increasing by 11% (Table 1). The livestock herd of Legal Amazon grows from 75.7 million head in 2006 to 152.9 million head in 2050, with an average increase of 2.3% yr\(^{-1}\). This growth rate is far below the average growth of 9.3% yr\(^{-1}\) observed in the 1974–2006 period in the region (IBGE 2010) and reflects the effect of livestock’s own price and the price of competing commodities in the future (shown above). Moreover, it would be too difficult to sustain the high growth observed in the last 30 years in the long term. Official statistics show that livestock growth rate is reducing with years (e.g., 6.1% yr\(^{-1}\) in the 1990–2006 period) because of, among other factors, an increase in the slaughtering rate (Barreto et al. 2008; Gouvello et al. 2010). More information on the IMPACT baseline scenario can be found in the study by Msangi and Rosegrant (Msangi and Rosegrant 2009).

### 2.2.3. Crop/pasture productivity

Potential crop/grass yields were calculated with the LPJmL model, which simulates global terrestrial vegetation dynamics, agricultural productivity, and the associated carbon and water cycles in a 0.5° spatial resolution (Sitch et al. 2003; Gerten et al. 2004; Bondeau et al. 2007). The LPJmL model calculations are based on physiological processes such as photosynthesis, autotrophic respiration, evapotranspiration, and effects of soil moisture and drought stress, as well as on plant’s functional and allometric rules, phenology, and growth parameterizations. Full model description as well as extensive validation against observed data of sowing dates, fraction of photosynthetically active absorbed radiation, seasonal CO\(_2\) flux exchanges, and crop yields can be found in the studies by Bondeau et al. (Bondeau et al. 2007) and Lapola et al. (Lapola et al. 2009b).

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\(^2\) Production of beef, pork, lamb, and poultry.
Crop yields for the 1990s, used as baseline yields in LandSHIFT, were calculated using the Climate Research Unit-Time Series 2.1 (CRU-TS2.1) climate dataset, a monthly climatology of meteorological variables, and atmospheric CO$_2$ concentration for the 1901–2003 period (Österle et al. 2003; Keeling and Whorf 2010). LPJmL transient simulations are preceded by a 1000-yr spinup period during which the first 30 years of the climate dataset are repeated cyclically to bring all carbon pools into equilibrium. Future crop yields (2036–65 mean) were calculated using the outputs from two IPCC AR4 general circulation models (GCMs), both under the SRES A2 emission scenario: the third climate configuration of the Met Office Unified Model (HadCM3) and the National Center for Atmospheric Research (NCAR) Community Climate System Model, version 3 (CCSM3) (Meehl et al. 2007). Climate anomalies (Figure 1) were defined as the differences from the 1961–90 mean of the CRU-TS2.1 dataset. Besides being among the GCMs that best represent the current climate over the Amazon (Li et al. 2006), these two GCMs project highly distinct climatic changes for the twenty-first century in the Amazon. HadCM3 projects an average increase of 3.8°C and a 30% decrease in precipitation over the Legal Amazon in the 2036–65 period (hereafter the severe climate change scenario), whereas CCSM3 projects a smaller temperature increase of 1.8°C and no changes in average precipitation (hereafter the moderate climate change scenario). Atmospheric CO$_2$ concentration increases from an average 333 ppmv in the 1961–90 period to 537 ppmv in the 2036–65 period. Improvement of yields through technological changes (e.g., plant breeding, increased use of fertilizer, and irrigation) are not considered in our simulations.

2.2.4. Experimental design

The effects of CO$_2$ fertilization on crop productivity are still poorly understood, especially in the tropics, and seem to be overestimated by most vegetation models currently available, including LPJmL (Slingo et al. 2005; Ainsworth and Long 2005). Therefore, we consider the upper limit of the effect of climate change on crop/grass productivity to be the HadCM3 climate scenario without the effects of CO$_2$ fertilization. The lower limit is then considered to be the yields calculated with CCSM3 climate and with the CO$_2$ fertilization effect. Four scenario variations are modeled with LandSHIFT, all of them with road paving and IMPACT projections on crop and livestock production for 2050:

(i) CCSM3 climate + CO$_2$ fertilization (moderate-BAU);

(ii) HadCM3 climate, no CO$_2$ fertilization (severe-BAU);

(iii) CCSM3 climate + CO$_2$ fertilization, suppression of PM pastures, and deforestation of the Amazon (Cerrado) gradually reduced to zero until 2020 (2025) (moderate-CONSERV); and

(iv) HadCM3 climate, no CO$_2$ fertilization, suppression of PM pastures, and deforestation of the Amazon (Cerrado) gradually reduced to zero until 2020 (2025) (severe-CONSERV).

PM pastures are gradually replaced by WM pastures until 2025 in the variations in which a suppression of PM pastures is assumed. The intensification of grazing needed to meet the feed demands of future livestock production in variations moderate-CONSERV and severe-CONSERV is determined by increasing the
grazing efficiency $g_e$ factor to the level at which demands are met, though keeping $g_e$ below the maximum reported value of 0.7 (Difante et al. 2009). The definition of deforestation used within LandSHIFT refers only to primary vegetation, and it means that all the vegetation of a given grid cell (forest or savanna) is cleared.

3. Results

3.1. Potential yields

Figure 3 shows the simulated changes in crop/grass yields relative to their values in the 1990s. Average (between all crop/grass types) yield changes range from $-11\%$ with HadCM3 climate to $+14\%$ with CCSM3 climate when the CO$_2$ fertilization effect is considered. However, crop yields are $-31\%$ (HadCM3) to $-8\%$ (CCSM3) lower compared to the 1990s if we consider that the CO$_2$ fertilization effect will have no influence on future crop yields. The reductions by $44\%$ and $10\%$ in the yields of soybean and grassland, respectively, under the severe-BAU scenario are particularly relevant for the Legal Amazon (besides considerable reduction in the yields of maize, rice, and other crops under that scenario). Soybean yield decreases by $1.8\%$ and grass yield increases by $4.5\%$ in the moderate-BAU scenario. Tropical roots functional type (cassava) is the only crop that experiences an increase of yields in every scenario because, in LPJmL, this crop type benefits from the increase in temperature. In general, the most pronounced yield reductions are found in the northern portion of the Legal Amazon (especially in Pará and Maranhão) because both HadCM3 and CCSM3 climate model project reductions in precipitation in that region (Figure 1).
3.2. Land-use change with business as usual

Under a moderate climate change scenario (and ignoring the target of halting deforestation in the Amazon) deforestation of the Brazilian Amazon would amount to 928 000 km² by 2050 (20 173 km² yr⁻¹ in the 2006–50 period) in our simulations (Figure 4a and Table 2). On the other hand, under a severe climate change scenario (severe-BAU) the forest would be reduced by 1 109 000 km² (24 108 km² yr⁻¹). Therefore, in these BAU simulations Amazon deforestation would be 20% higher under severe climate change compared to a scenario of moderate climate change. Deforestation of the Cerrado simulated by LandSHIFT would amount to 88 000 km² (1913 km² yr⁻¹) and 328 000 km² (7130 km² yr⁻¹) under the moderate-BAU and severe-BAU climate change scenarios, respectively (Figure 4b and Table 2). Thus, in these BAU simulations deforestation of the Cerrado would be 273% higher with severe climate change compared to a scenario with moderate climate change.

Altogether, crops would need a 45% larger area under the severe scenario compared to the moderate scenario to meet the 2050 demands projected by IMPACT for the Amazon. Soybean alone would occupy a 49% larger area in the severe scenario compared to the moderate scenario (94 000 versus 63 000 km²). WM pasture (PM pasture) would have its area increased by 615 000 (311 000) km² in the moderate scenario and by 838 000 (423 000) km² in the severe scenario. Difference in total area of both types of pastures between the climate scenarios would be of 18%. LD decreases in both climate scenarios, although this decrease is more pronounced in the severe climate change scenario. That is because, even though average grass productivity increases from 2006 to 2050 in the moderate-BAU, it decreases punctually in the regions where new pastures are established until 2050, north and northeast Legal Amazon. Abandoned area increases by 16 000 and 9000 km² in the moderate and severe climate change scenarios, respectively (Table 2), because of a shift in cropland location driven by local climate change (e.g., reduction in precipitation in western Mato Grosso with CCSM3 causes some soybean fields to shift to southeastern Mato Grosso).

Most of the deforestation would still occur in the southern and eastern Amazon and along the paved highways (Figure 5), which is explained by the weights \( w_i \) given to the \( p_i \) factors shown in Figure 2. Cropland expansion would take place...
Table 2. Land-use and land-use change (relative to 2006) according to different modeled scenarios in the Brazilian Legal Amazon in 2050 (urban areas increase from 3151 km² in 2006 to 3949 km² in 2050 in all scenarios). Here, cropland includes soybean. Moderate-BAU is CCSM3 climate + CO₂ fertilization; severe-BAU is HadCM3 climate and no CO₂ fertilization; moderate-CONSERV is CCSM3 climate + CO₂ fertilization, suppression of PM pastures, and deforestation of the Amazon (Cerrado) gradually reduced to zero until 2020 (2025); and severe-CONSERV is HadCM3 climate, no CO₂ fertilization, suppression of PM pastures, and deforestation of the Amazon (Cerrado) gradually reduced to zero until 2020 (2025). (Well managed (WM), poorly managed (PM), and livestock density (LD).)

<table>
<thead>
<tr>
<th>Year and scenario</th>
<th>Forest (×1000 km²)</th>
<th>Cerrado (×1000 km²)</th>
<th>Cropland (×1000 km²)</th>
<th>Soybean (×1000 km²)</th>
<th>Abandoned (×1000 km²)</th>
<th>WM pasture (×1000 km²)</th>
<th>PM pasture (×1000 km²)</th>
<th>LD (WM pasture) (head ha⁻¹)</th>
<th>LD (PM pasture) (head ha⁻¹)</th>
</tr>
</thead>
<tbody>
<tr>
<td>2006, —</td>
<td>3336</td>
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<td>135</td>
<td>59</td>
<td>110</td>
<td>585</td>
<td>294</td>
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<tr>
<td>2050, moderate-BAU</td>
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<td>560</td>
<td>208</td>
<td>63</td>
<td>126</td>
<td>1200</td>
<td>605</td>
<td>1.10</td>
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<tr>
<td>2050, severe-BAU</td>
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<td>320</td>
<td>302</td>
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<tr>
<td>2050, moderate-CONSERV</td>
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<td>209</td>
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<td>2</td>
<td>1063</td>
<td>0</td>
<td>see Table 3</td>
<td>—</td>
</tr>
<tr>
<td>2050, severe-CONSERV</td>
<td>3320</td>
<td>447</td>
<td>303</td>
<td>93</td>
<td>2</td>
<td>1036</td>
<td>0</td>
<td>see Table 3</td>
<td>—</td>
</tr>
</tbody>
</table>
mostly in Mato Grosso and Tocantins. WM pastures would be widespread along the deforestation arc. Because PM pasture is the last in the hierarchical allocation of major land-use activities in LandSHIFT, this land-use type is relegated to more remote and less productive areas. Figure 6 highlights how different climate change scenarios could result in distinct deforestation patterns, impacting both the extent and location of future LUCC. Pastures expand deeper into the western Amazon forest and especially in the Cerrado in the severe scenario compared to the moderate one because of the pronounced decrease in precipitation projected by HadCM3 in northern Legal Amazon.

3.3. Land-use change with the end of deforestation

From 2006 to 2050 Amazon deforestation would amount to 29 000 km² (2230 km² yr⁻¹) and 16 000 km² (1230 km² yr⁻¹) under moderate-CONSERV
and severe-CONSERV scenarios, respectively, before it ends by 2020 (Figure 4a and Table 2). On the other hand, the Cerrado loses 121 000 km$^2$ (9307 km$^2$ yr$^{-1}$) and 201 000 km$^2$ (15 461 km$^2$ yr$^{-1}$) of its native vegetation by 2025 in the moderate-CONSERV and severe-CONSERV scenarios, respectively (Figure 4b and Table 2). Halting Amazon deforestation by 2020 but still allowing deforestation of the Cerrado until 2025 explains these highly different deforestation rates when compared to the BAU scenarios.

Future cropland area would be roughly the same as in the BAU scenarios because crops have priority over pastures in LandSHIFT in this study; that is, crops are allowed to displace pastures. Therefore, the area of pasture would increase in approximately 171 000 km$^2$, by replacing natural vegetation, specifically by occupying PM pastures and abandoned areas (Figure 5). Then, at this point, two options are considered here to conciliate agricultural production and conservation targets in the Legal Amazon (Table 3):

(i) Livestock production is reduced; that is, from 26% (moderate-CONSERV) to 40% (severe-CONSERV) of the Legal Amazon livestock production projected for the year 2050 cannot be produced there.

(ii) Livestock production is ensured/kept up with the intensification of livestock in the Legal Amazon. In that case, livestock density needs to roughly double from 0.74 head per hectare (average between PM and WM pastures) in 2006 to $\sim 1.46$ head per hectare (1.44 head per hectare in moderate-CONSERV and 1.48 head per hectare in severe-CONSERV). Grazing efficiency $g_e$ is increased to 0.47, a value still far from the maximum of 0.7 reported by Difante et al. (Difante et al. 2009) under rotational stocking management.

Figure 5 shows the land-use pattern in 2050 in the moderate-CONSERV and severe-CONSERV scenarios. Some deforestation of the Amazon is projected in northeast
Pará and along the highways BR-163 in Mato Grosso and BR-364 in Acre. Most of the deforestation of the Cerrado takes place in Maranhão and Tocantins in areas that are not specified as conservation units, and it is more pronounced in severe-CONSERV than in moderate-CONSERV. Compared to BAU scenarios, there would be an avoided deforestation of 899 000 km² of Amazon forest in the moderate-CONSERV climate scenario and of 1 093 000 km² in the severe-CONSERV scenario. In the Cerrado, there would be an additional deforestation of 33 000 km² in the moderate-CONSERV scenario and an avoided deforestation of 120 000 km² in the severe-CONSERV scenario (Figure 6).

4. Discussion

From our simulations, we can infer that in the Legal Amazon,

- Future climate change may influence LUCC in ways that have previously remained unexplored. Severe climate change in some regions can shift the deforestation frontier. For example, the harsh climate projected by HadCM3 in central and eastern Amazon increases human pressure in the Cerrado and western Amazon.
- Ambitious conservation targets and increased agricultural production can be conciliated even under a scenario of severe climate change, but it will require either a more intensive use of the land or a slowdown in the growing production of meat.

These two major findings are discussed below.

4.1. Climate change effects on land use

There is now extensive documentation about the impacts of regional or continental extreme climatic events on agriculture and livestock production. Excellent examples are the hot summer of 1972 in the southwest of the former Soviet Union and its consequences in world cereal markets (Dronin and Bellinger 2005); the record yield drops and livestock stress in Europe during the anomalous heat in the
summer of 2003 (UNEP 2004); the 2005 drought in the Amazon and associated agricultural losses in many parts of Brazil (Lenton et al. 2009); and several studies on the impacts of the El Niño–Southern Oscillation on crop/pasture productivity and food security in the Amazon (Moran et al. 2006; Brondizio and Moran 2008), in Indonesia (Keil et al. 2008), or worldwide (Ferris 1999). Nonetheless, currently such climatic events have a relatively long return interval [El Niño: ~7 yr (Cobb et al. 2003); 2005-like drought: 20 years (Cox et al. 2008)] and are not yet the climatic norm, as, for example, would be the case of the permanent El Niño events projected by HadCM3. One of the single documented references to recent long-term climate change and its effects on yields and LUCC is the prolonged drier conditions found in the Sahel from the late 1960s until the early 1990s which caused the abandonment of crop and grazing fields, besides massive migration and countless hunger- and battle-related deaths (Kandji et al. 2006; Burke et al. 2009). However, most of the examples of long-term climate change impacts on LUCC stem from archeological/historical records, as is the case for the theory of the collapse of the Maya in the Yucatán Peninsula in the late tenth century (Turner et al. 2003) or the effects of the onset of the Little Ice Age (sixteenth century) on the agriculture of the Iberian Peninsula (Puigdefábregas 1998).

These catastrophic experiences reveal that the impacts of climate change on LUCC are always, though not solely, mediated by changes in crop/grass productivity, which is the way LUCC is affected by climate in this study. In view of that, we can consider the method used here for assessing the impacts of climate change on LUCC as reasonable, even though it does not consider other ways in which climate change could indirectly affect LUCC in the Legal Amazon. Difficulties for navigation if the level of rivers is too low, decrease of fish stocks (which is one of the main sources of protein of the Amazonians), spread of diseases, potable water shortage, and higher frequency of floods and fires—all these examples represent pathways through which climate change could affect farmers’ and other people’s living conditions and, consequently, LUCC in the region.

The simulated range of changes in crop/grass productivity lies within the range projected in other studies for Brazil (Assad and Pinto 2008; Lobell et al. 2008) and the whole globe (Tebaldi and Lobell 2008). That is particularly true for the projections in which the CO₂ fertilization effect does affect crop yields in the future. On the other hand, LPJmL yield projections with HadCM3 climate and no CO₂ fertilization are much lower than what has been projected in the studies mentioned above but should not be considered as less probable because the uncertainties regarding the effects of rising CO₂ on future crop yields are still large (Ainsworth and Long 2005; Long et al. 2006; Lobell and Field 2008). The pronounced decrease in the yields of soybean, slight decrease of maize and rice, as well as the increase of cassava yields are particularly in agreement with the projections by Assad and Pinto (Assad and Pinto 2008), using a regional climate model for entire Brazil. Nevertheless, the authors of that study point out a reduction of up to 25% of pasture productivity (for entire Brazil) as compared to the 10% projected with LPJmL–HadCM3 for the Legal Amazon. Although in this study we calculate the LUCC resulting from yield changes with the HadCM3 climate (Figures 5b,d), we believe it is unlikely that in reality crop cultivation would continue after such a reduction of yields, especially in large-scale farming systems. It is more reasonable to think that such decreasing yields would,
in the long term, reduce the profitability of agriculture in the region. That could lead in turn to an encroachment of the cultivated area and certainly to a shift of that agricultural production to other more lucrative areas of Brazil (probably generating more deforestation of the Cerrado as suggested by our results and the study by Gouvello et al. 2010) or even to other parts of the world. This might in turn have serious consequences for the economy of the Legal Amazon and for the food security of its inhabitants.

Technological improvements of yields are, on purpose, not considered in our simulations so one can regard the projections shown in Figure 3, especially those calculated with HadCM3 climate, as an outlook on the magnitude of adaptation needed by the agriculture of the Legal Amazon over the next decades. So, for example, to avoid the soybean yield reduction caused by an extreme climate change scenario (i.e., to keep soybean yields at their current values at least), a yield increment rate of 23 kg ha\(^{-1}\) yr\(^{-1}\) would be needed until 2050, which is far lower than the soybean yield enhancement rate of 39 kg ha\(^{-1}\) yr\(^{-1}\) observed in the last two decades in Brazil (FAO 2010; Lapola et al. 2010). For maize, this yield adaptation would be 11 kg ha\(^{-1}\) yr\(^{-1}\), compared to 78 kg ha\(^{-1}\) yr\(^{-1}\) yield enhancement observed in the last two decades in Brazil (FAO 2010; Lapola et al. 2010). This adaptation of cropping and livestock systems could come in the form of better management of water resources, change in sowing dates, development of heat-tolerant crop varieties, infrastructure to minimize heat-stress-related reductions of livestock productivity, or even altering the location of cropping/livestock activities (Howden et al. 2007). However, all these actions would obviously demand financial investments. As a consequence, adaptation seems more feasible to large-scale farmers than for smallholder or subsistence farmers because of the former's easier access to credit. A recent survey revealed that, although smallholder agriculture occupies only 24% of the total farmed area in Brazil, it is responsible for 87% of the national production of cassava, 70% of dry beans, 46% of maize, 36% of rice, and 58% of milk (IBGE 2009). As presumed from our results, this agricultural production (its share in the Legal Amazon) might be compromised in the future assuming no intervention and/or support from the government or other bodies to develop adaptation strategies for the sector (Morton 2007). Such a strategy should take into account the sociocultural and environmental diversity of the Amazonian small-scale farmers and, importantly, institutionalize the translation of large-scale projections, like the one in this study, into local actions (Brondizio and Moran 2008).

4.2. The end of deforestation and land use

Our results also show that a combination of ambitious conservation targets (in the way suggested by Nepstad et al. 2009) with increased agricultural production is feasible even under a scenario of severe climate change. However, adaptation of agriculture, especially the intensification of cattle ranching, which is the main land use in Legal Amazon, is a sine qua non condition to achieve both targets. Brazil’s recent economic growth has boosted people’s monetary access to meat, and the country today is the fourth biggest consumer of meat per capita in the world (Barreto et al. 2008; Friends of the Earth 2009). Considering these current trends of
changes in life style, it seems more likely that the mentioned conservation targets might be achieved via intensification of livestock production rather than via reduction of livestock production and consequent meat consumption.

It is well known that the oxisols and ultisols of the Amazon, dominant in over 75% of the basin, make it difficult to keep a high productivity of pastures for more than ~5 years without active management (Walker et al. 2000). However, other factors such as land tenure (e.g., in many cases LD is kept at a minimum level only to guarantee ownership over public land), and ongoing policies of “perverse” subsidies (e.g., animal acquisition is heavily subsidized in Brazilian cattle ranching but nearly no incentives are provided specifically for the recovery of degraded pastures and intensification of grazing) also have a decisive influence on the widespread low LD across the Legal Amazon (Hecht 1985; Fearnside 2002; Nepstad et al. 2006; Friends of the Earth 2009). As discussed by Lapola et al. (2010), an increase in livestock density in the Legal Amazon, such as the +0.72 head per hectare proposed here, is perfectly possible from a biophysical point of view with the enhancement of grass productivity and adoption of some simple management practices (FAO 2007b; Assad and Pinto 2008). Nevertheless, this intensification seems to be impossible without a concerted effort in terms of providing adequate subsidies (Friends of the Earth 2009), increasing land tenure in the region (Fearnside 2008) and excluding deforesters from the livestock supply chain (Nepstad et al. 2009).

Particularly for the Cerrado, this study calls attention for the lack of protected areas of that habitat in the Legal Amazon (viz., in east Mato Grosso, northeast Tocantins, and south Maranhão), and the potential consequences this might have in the future in case we have the climate projected by HadCM3 by 2050. Likewise, it suggests that stopping deforestation concomitantly in the Amazon and Cerrado might be important to prevent the end of deforestation in the Amazon forest causing an escalation of deforestation in this highly biodiverse but threatened savanna (currently 39% of the native cover of Cerrado has already been deforested compared to 15% of the Amazon forest; Sano et al. 2007; PRODES 2009).

4.3. Caveats and future research

The main caveat of our simulation is that there are no feedbacks between the models comprised in our framework. Without a real coupling between the models, we are unable to assess, for example, the feedbacks between climate change, crop yields, and crop market prices. In this study, we consider both projections of crop prices and production as not dependent on climate or conservation targets. However, it is probable that a future El Niño–like climate (e.g., projected by HadCM3) would drive crop prices up in the Legal Amazon, which in the long term could lead to a shift of the regional agricultural production to other parts of Brazil or even of the world. That could in turn lead to changes in the regional economy and compromise the local provision of food. Likewise, it could also occur that the agricultural production projected by IMPACT is reduced over time with the establishment of stricter conservation targets (e.g., those suggested by Nepstad et al. 2009) via constraints in the availability of arable land.

An improved and fully coupled modeling framework could also help understanding other key questions about the Amazon system. For example, what would
be the impacts of a climate-driven forest dieback (Cox et al. 2004) on the deforestation rates and land-use pattern in the Amazon? How would year-to-year climate variability influence future LUCC and food security? What are the probabilities of the impacts (e.g., assessed with ensemble runs)? These questions remain to be pursued.

5. Conclusions

The modeling exercise presented in this paper demonstrates some plausible impacts that climate change and conservation strategies might have on land use in the Brazilian Amazon. Without any adaptation, climate change would exert a critical impact on the productivity of Amazonian crops: for example, the 44% reduction in soybean yields in the worst climate scenario. Moreover, the severity of climate change may influence the location and magnitude of future LUCC. Our study suggests that agriculture, especially cattle ranching, will need to adapt to these two upcoming shifts in the Amazonian system (climate change and the end of deforestation). Importantly, however, it also suggests that both the identification of impacts and the adaptation to them should be tackled in a multidisciplinary and integrated manner, considering conservation strategies and projections on population growth, changes in lifestyle, and agricultural production.

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Appendix A

Land-Use Maps of the Legal Amazon in 2001 and 2006

For the production of the two “observed” land-use maps employed in this study (for model initialization and/or validation), we used land-cover maps of the Legal Amazon in 2001 and 2006, produced by using Monitoramento da Floresta Amazônica por Satélite (PRODES) satellite data (PRODES 2009), a 2000 vegetation map of South America (Eva et al. 2004), and a classified Moderate Resolution Imaging Spectroradiometer (MODIS) vegetation continuous field (Hansen et al. 2002). The maps’ forest category encompasses subtypes such as transitional forest, whereas the category Cerrado encompasses all the phytosociological communities of the Brazilian savannas (Cerradão, Campo Limpo, etc). These land-cover maps were degraded to the resolution of 5 arc-min and were divided into 32 regions, as suggested by Garcia et al. (Garcia et al. 2007) and Soares-Filho et al. (Soares-Filho et al. 2006). Although these regions did not change over time, this division was used to minimize the erroneous allocation or nonallocation of land uses anywhere in the Legal Amazon (the smaller the region for making the downscaling of census data, the smaller the potential error in determining the location of cropland/pasture within that particular region). Each of these subunits had their own crop and
pasture area determined from the Instituto Brasileiro de Geografia e Estatística (IBGE) municipal agricultural production database for the given years (IBGE 2010). Because data on pasture area are not available for the year 2001, it was estimated, through linear interpolation, from the 1996 and 2006 data. A total of 13 crop types and 2 pasture types were considered in the confection of the land-use maps. Only areas depicted as deforested or as Cerrado (because land-cover changes of this latter are not tracked by satellites as the deforestation of the Amazon) could have the assignment of crops or pasture. Crops had priority over pasture for occupation of grid cells, whereas only one dominant land-use type can occur in one grid cell. The allocation procedure followed a preference list of grid cells, which was built based on a 2000 map on the geographical distribution of crop/pasture areas, also on 5-arc-min resolution (Monfreda et al. 2008; Ramankutty et al. 2008). Grid cells with higher fraction of a given crop type in the map by Monfreda et al. (Monfreda et al. 2008) had preference for assignment of that crop type in our land-use map. Disambiguation within one crop type (e.g., when the Monfreda et al. map for soybeans had several grid cells with the same area) or between different crop types (i.e., when Monfreda et al. maps for two or more different crop types had exactly the same value in a given grid cell) was performed using a multicriteria analysis (MCA) of slope, potential productivity of the given crop type (or grassland for pasture), distance from settlements, soil type, and distance from paved roads (for data sources, see Soares-Filho et al. 2006). However, this MCA was needed only in a minor fraction (<1%) of the grid cells that later were assigned as crop or pasture. Therefore, the maps of Monfreda et al. (for crops) and Ramankutty et al. (for pastures) played the major role in the allocation of land uses in our base maps. Urban areas were assigned to those grid cells having a population density higher than 2000 people per square kilometer (Erb et al. 2007), using the History Database of the Global Environment (HYDE) map of population distribution (Goldewijk 2005), with no distinction between the years 2001 and 2006.

A first assessment of the land-use maps revealed that the area assigned as abandoned was too large (350 000 km² in 2001), surpassing any estimate on the extent of land currently abandoned in the Legal Amazon, which ranges from 61 000 to 106 000 km² (several datasets analyzed by Campbell et al. 2008). In fact, the very concept of abandoned land is quite variable and can, for example, refer to temporal characteristics (e.g., set aside), soil conditions (e.g., degraded), or management (e.g., poorly managed) of the land use. Here, the land-use type abandoned is considered to be simply land with no occurrence of any other land-use type. Therefore, considering that PRODES provides trustworthy numbers for the extent of the Amazon forest and that the extent of Cerrado in our maps is in fair agreement with latest surveys (Sano et al. 2007)—besides the fact that most of the geographical subunits with abandoned lands did not have Cerrado within it limits (e.g., Paragominas)—we argue that the IBGE data for pasture area in the Legal Amazon might be underestimated, at least in some regions (see Ramankutty et al. 2008). Thus, to correct this discrepancy, after IBGE area requirement for pasture is fulfilled in our maps (i.e., all crop and pasture areas were allocated at this stage), we assign the poorly managed pasture type to all the remaining grid cells that are covered by pasture in Ramankutty et al. (Ramankutty et al. 2008) map and were, until this stage, set as abandoned in our land-use map. That reduces the area of the abandoned land-use type to 102 000 km², in better agreement with data available
for comparison. This type of pasture is meant to represent pastures with a lower intensity of use, with lower livestock density (compared to WM pastures), and mixed with degraded/secondary vegetation (cf. locations with INPE 2009). The other type of pasture is then referred to as well-managed pasture. The study by Gouvello et al. (Gouvello et al. 2010) indicates that roughly 30% of the Legal Amazon pastures are low-productivity pastures, in agreement with the fraction of poorly managed pastures in the 2006 land-use map (Figure A1).

In general, the methods used to obtain these maps are not as comprehensive as, for example, the one used by Cardille and Foley (Cardille and Foley 2003) to produce land-use maps of the Brazilian Amazon for 1980 and 1995 (e.g., our maps have only one land use per grid cell instead of fractional coverage), even though they are in accordance to official statistics (IBGE). Most of all, however, one should consider that the maps presented here were produced as to fit their use in the LandSHIFT model.

Appendix B

Model Evaluation

LandSHIFT has been thoroughly evaluated in terms of the quantity of change in other studies (Koch et al. 2008; Schaldach et al. 2011), including a study in which the model was applied for entire Brazil (Lapola et al. 2010). However, the model has not been consistently evaluated in terms of the location of changes mainly because of the lack of independent time series of “observed” land-use maps generated based on the same methodology. Therefore, taking advantage of the two
independent maps of land use employed here (appendix A), a LandSHIFT run from 2001 to 2006 was performed to evaluate the model performance. The model was initialized with the 2001 land-use map of the Legal Amazon and was driven with reported statistics on crop and livestock production for 2006 (IBGE 2010). Because the 2006 map inherits the spatial pattern of the 2001 map, we assess the spatial fit only between the maps of changes. Thus, the resulting modeled map of LUCC from 2001 to 2006 was compared with the observed map of changes for that period. To reduce the dependency between the datasets used for comparison (the observed maps were used for deriving the $w_i$ weights of LandSHIFT), the evaluation was done only in four selected regions of the Legal Amazon (Figure A1). These regions were selected as to cover locations that experienced pronounced deforestation or other LUCC encompassing the three major land-use categories considered here (cropland, WM pasture, and PM pasture) in the 2001–06 period. The four regions and the dominant land-use transitions that were observed from 2001 to 2006 are central Pará (forest to PM pasture, to WM pasture, and to cropland), southeast Pará (forest to PM pasture), south Mato Grosso (Cerrado to cropland), and south Rondônia (PM pasture to WM pasture and forest to cropland). Combined, these four regions represent only 10% of the area that experienced LUCC in the 2001–06 period and ~4% of the Legal Amazon.

Both observed and modeled 2001–06 LUCC maps were reclassified into three categories for the comparison: natural vegetation, cropland, and pasture. Conversion from any land use to natural vegetation is excluded from our analysis because LandSHIFT does not simulate natural vegetation regrowth. The maps of changes were subject to the fuzzy vicinity-based comparison method developed by Hagen (Hagen 2003) (K-fuzzy method) and modified by Almeida et al. (Almeida et al. 2008) [reciprocal fuzzy comparison (RFC)]. This method takes into account the nature of LUCC models to justify a vicinity-based comparison (i.e., LUCC location is fuzzy). An exponential decay function is employed to weigh the distance of a cell in one map to its counterpart in the second map. Map comparison is carried out in a two-way manner and at multiple spatial resolutions. However, only the minimum similarity value is used to avoid an artificially high fit, which is characteristic of univocal comparison of random maps. Figure B1 shows the results of this RFC analysis over the four evaluation regions. The model does a reasonable job in capturing the right location of transitions because the average curve reaches up to 60% of similarity with a search radius of only two grid cells and peaks in 71% after five grid cells. If the average is weighted by the size of each of the four analyzed regions, then similarity reaches the value of 60% after three grid cells but peaks have a higher value of 75% after five grid cells. From the original kappa classification (Monserud and Leemans 1992), a 60% similarity is classified as a good degree of agreement. Lowest similarity is found in central Pará, because the model does not capture well the transition from forest to WM pasture. On the other hand, the highest fit is found in southeast Pará because the model simulates correctly the forest to PM pasture transition which, according to the maps presented in appendix A, responded for 56% of the Amazon deforestation in the 2001–06 period.

Cropland is overestimated by 8%, as in the study by Lapola et al. (Lapola et al. 2010) for entire Brazil. The area of pastures was calibrated with the $g_e$ factor; therefore, its fit to the observed data is nearly perfect. The modeled rate of Amazon deforestation for 2001–06 is underestimated by 11%: 20 851 km² yr⁻¹ versus
This underestimation is because LandSHIFT does not simulate the direct transition from forest to abandoned land, as is the case in some areas of the observed maps (a forestry module is currently being developed in LandSHIFT and could account for this kind of land-use transition in the future). Moreover, one should also consider that 2001–06 was a period with above-average deforestation rate. For example, average deforestation rate was 18,700 km$^2$ yr$^{-1}$ in the 1996–2000 period and 10,833 km$^2$ yr$^{-1}$ in the 2007–09 period. Deforestation of Cerrado is underestimated by 18%: 6366 km$^2$ yr$^{-1}$ versus 5206 km$^2$ yr$^{-1}$.

Nevertheless, there is high uncertainty associated with deforestation rates of the Cerrado because land-cover changes in the Cerrado are much more difficult to detect by remote sensors than in the Amazon (appendix A).

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