Manuscript prepared for Hydrol. Earth Syst. Sci. with version 5.0 of the LATEX class copernicus.cls.

Date: 2 November 2014

Hydrometeorological Effects of Historical Land-Conversion in an Ecosystem-Atmosphere Model of Northern South America

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Abstract.

This work investigates how the integrated land-use of northern South America has affected the present day regional patterns of hydrology. A numerical model of the terrestrial ecosystems (Ecosystem Demography Model 2 -ED2) is combined with an atmospheric model (Brazilian Re- 35 gional Atmospheric Modeling System - BRAMS) to investigate how land conversion in the Amazon and northern South America have changed the hydrology of the region. Two realizations of the structure and composition of terrestrial vegetation are used as the sole differences in boundary condi- 40 tions that drive two simulations. One realization captures the present day vegetation condition that includes deforestation and land-conversion, the other is an estimate of the potential structure and composition of the region's vegetation without human influence. Model output is assessed for differences in 45 resulting hydrometeorology.

The simulations suggest that the history of land-conversion in northern South America is not associated with a significant precipitation bias in the northern part of the continent, but has shown evidence of a negative bias in mean regional evapotranspiration and a positive bias in mean regional runoff. As well, negative anomalies in evaporation rates showed pattern similarity with areas where deforestation has occurred. However, in the central-eastern Amazon there was an area where deforestation and abandonment had lead to an overall reduction of above ground biomass, but this was accompanied by a shift in forest composition towards early-successional functional types and grid-average

patterned increases in annual transpiration. Anomalies in annual precipitation showed mixed evidence of consistent patterning. Two focus areas were identified where more consistent precipitation anomalies formed, one in the Brazilian state of Pará where a dipole pattern formed, and one in the Bolivian Gran Chaco where a negative anomaly was identified. These locations were scrutinized to understand the basis of their anomalous hydrometeorologic response. In both cases, deforestation led to increased total surface albedo, driving decreases in net-radiation, boundary layer moist static energy and ultimately decreased convective precipitation. In the case of the Gran Chaco, decreased precipitation was also a result of decreased advective moisture transport, indicating that differences in local hydrometeorology may manifest via teleconnections with the greater region.

1 Introduction

It has been held that massive and widespread Amazonian deforestation would lead to regional reductions in precipitation, evaporation and moisture convergence, with slight increases in surface temperature (Henderson-Sellers et al., 1993; Nobre et al., 1991; Lean and Warrilow, 1989; Dickinson and Henderson-Sellers, 1998). The Amazon basin and its forest ecosystems are also an important component of the global circulation of energy (Gedney and Valdes, 2000), where changes (complete deforestation in either Amazonia or all tropical broad-leaf forests) are thought to teleconnect beyond

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the continent (Avissar and Werth, 2005; Snyder, 2010). The literature documenting Amazonian land conversion and the surrounding areas is significant, the reader is referred to a small selection of non-exhaustive references for some back- 115 ground, (Cardille and Foley, 2003; Skole and Tucker, 1993; INPE, 2003; Geist and Lambin, 2002; Laurance et al., 2001; Nepstad et al., 2001; Soares-Filho and co authors, 2006). The work presented here is motivated by a need to better understand how the history of land-conversion has influenced 120 the hydrology of the region. As will be outlined further, the mechanistic relationships between land-conversion and hydrometeorological response is complex and has benefited from study with newer generations of land-atmosphere models with increased granularity and increased complexity in 125 representing physical process.

There are several direct hydrologic mechanisms that connect changes in tropical forest structure (i.e. deforested versus intact canopies) to the regional climate system. Leaves, stems and bare earth have variable light scattering proper- 130 ties, such that intact forest canopies composed of dark vegetation typically have lower short-wave radiation albedo than areas with exposed soil (Chapin et al., 2002). This directly impacts the surface energy balance via net-radiation. Forests canopies have a complex relationship with the surface mois- 135 ture balance and mediate the transport of water in numerous ways. Model studies have shown that the representation of canopy interception can substantially impact the partitioning of evapotranspiration and surface runoff (Pitman et al., 1990; Wang et al., 2007; Crockford and Richardson, 2000). 140 Some studies have found that forest canopies increase the interception of precipitation (Asdak et al., 1998; Dietz et al., 2006), and further that crown structure influences turbulent transport and evaporation of wet leaves (Dietz et al., 2006). Yet some have indicated that canopy interception increased 145 in degraded forests (Chappell et al., 2001). Pastures and converted agricultural systems are generally associated with soil degradation such as decreased infiltration rates, nutrient loss and increased surface runoff, subject to variability and factors such as the soil texture and the existence of perennial 150 under-story vegetation (Benegas et al., 2014). Practices such as grazing and agriculture promote soil compaction and decreased infiltration (Martinez and Zinck, 2004; Lal, 1996), and intense fires used for clearing lands may reduce soil organic matter that may favor infiltration (Kennard and Gholz, 155 2001). Forests with deep rooted trees draw from deeper soil moisture pools, which have different periodicity in available water and therefore alter the timing of latent heat flux via transpiration compared to grasslands (Kleidon and Heimann, 2000; Nepstad et al., 1994). Canopy structure also influences 160 the turbulent exchange of heat, moisture and momentum with the atmosphere (Raupach et al., 1996).

The higher surface temperatures associated with widespread deforestation, as reported with the first generations of General Circulation models and beyond, 165 (Henderson-Sellers et al., 1993; Nobre et al., 1991; Lean

and Warrilow, 1989; Dickinson and Henderson-Sellers, 1998) are thought to be the result of losses in evaporative cooling associated with cleared vegetation. The decrease in evaporative cooling is also thought to drive reduced precipitation, and subsequently reduces the heat released to the atmosphere through condensation (Eltahir and Bras, 1993). This out-competes the effect of increased surface albedo of deforested lands (Eltahir, 1996), which suppresses net-radiation thereby promoting surface cooling and divergence (Eltahir and Bras, 1993; Lean and Warrilow, 1989). Positive surface temperature anomalies induce convergent circulations coincident with a decrease in surface pressure. Decreased precipitation heating anomalies reduce the tendency towards convergence.

In the southwestern Amazonian dry season, statistical connections have been made between pastures and higher incidents of shallow cumulus clouds, compared to intact forests where shallow clouds are less frequent yet deep precipitating convective events are more frequent (Wang et al., 2009). The higher rate of deep precipitating convection over forests was associated with larger values of convective available potential energy (CAPE) (Williams and N, 1993), which in this case was driven by increased humidity and moisture flux from intact forests canopies. The increased frequency of shallow convection was attributed to more vigorous mesoscale circulations associated with deforestation induced land-surface heterogeneity (Wang et al., 2000)(Souza et al., 2000).

Regional land-atmosphere simulations parametrize convective clouds indicate that structured land-conversion scenarios elicit shifts in mean basin precipitation, however less so than traditional coarse scale General Circulation Model studies (Silva et al., 2008). Coherent land surface patterns may strengthen convergence zones on the surface, creating vertical wind triggers to thunderstorms. For instance, Avissar and Werth introduced that coherent land surface patterns transfer heat, moisture and wave energy to the higher latitudes through thunderstorm activity. Moreover, meso-scale simulations are found to capture key cloud feedback processes which fundamentally alter the atmospheric response to land-surface heterogeneities (Medvigy et al., 2011). The meso-scale simulation's ability to represent land-use scenarios at finer resolutions can impact spatial patterning of rainfall. For example, western propagating squall-lines from the Atlantic are thought to dissipate over regions of wide-spread deforestation (Silva et al., 2008; d'Almeida et al., 2007). Evidence has also shown that convection can be driven by localized convergent air circulations triggered by land-surface heterogeneities, and that the likelihood and quality of resulting events are both dependent on the scale of heterogeneity and the position relative to disturbed and intact landscapes (Pielke, 2001; Dalu et al., 1996; Baldi et al., 2008; Anthes, 1984; Knox et al., 2011; Wang et al., 2009).

Regional scale coupled land-atmosphere models can capture feed-backs resulting from land conversion at the scale of

tens of kilometers and lower, particularly through improved resolution and the parametrization of atmospheric physics (such as convection and radiation scattering). At the land-surface, there is variability in canopy structure at the gap 220 (size of a single large tree crown) scale and below. Processes that occur at these scales may be important to predicting ecosystem response and land-atmosphere exchange, as discussed above. Physics-based land surface models have non-linear representations of hydrologic and thermodynamic process, therefore using average canopy structure (such as 225 in the "big-leaf" approach) to represent processes uniformly may provide different results compared to explicitly capturing these processes at their native scales.

This research uses the Brazilian Regional Atmospheric Modeling System (BRAMS, a variant of the Regional At-230 mospheric Modeling System (RAMS) (Cotton et al., 2003)) coupled with the Ecosystem Demography Model 2 (ED2 or EDM2, (Moorcroft et al., 2001; Medvigy et al., 2009)), to explore the sensitivity of hydrologic climate of northern South America in response to present day land conversion. This 235 modeling system can explicitly represent the processes of energy and mass transfer in the canopy and soil system, with sub-grid variability along ecosystem age-structured and vegetation size-structured axes. An experiment is conducted by comparing simulations that singularly differ in their repre-240 sentation of regional vegetation cover, one which captures the present day vegetation condition that includes deforestation and land-conversion, the other being an estimate of the potential structure and composition of the region's vegetation without human influence. Section 2 of this manuscript will detail experiment design of the coupled model experiment. The model system and experiment design is verified by com-245 parison of model output with observations, see Appendices A2, A3, A4, A5 and A6.

In section 3 we evaluate hydrometeorological response to the changes in land-use history in a regional context. In section 4 we evaluate the processes underlying the observed 250 changes in hydrometeorology in two focus areas. A discussion and conclusion of the results follows.

2 Experiment Design

The main task of this experiment is to conduct two regional simulations of the South American biosphere and atmosphere. The defining difference between the two simulations 260 is how the land-surface model (ED2) represents the structure (the distribution of plant sizes) and composition (the distribution of plant types) of the region's terrestrial ecosystems, as a consequence of two different disturbance regimes. In one simulation, the vegetation reflects a structure and composi- 265 tion that has no effects of human land-use, i.e. a *Potential Vegetation* (PV) condition. In the other simulation, the model will incorporate an estimate of modern (e.g. 2008) human

land-use, i.e. an *Actual Vegetation* (AV) condition. The procedure is broken down into steps and elaborated upon.

2.1 Description of the Vegetation Model - ED2

The Ecosystem Demography Model 2 predicts the changes in terrestrial vegetation structure, as modulated by the physically based conservation of water, carbon and enthalpy. Its central design philosophy assumes that the stochastic representation of plant communities integrated over a large sample can be portrayed deterministically as land fractions and plant groups, with explicit size (of the plants) and age (time since a patch of soil housing the plants has experienced major disturbance) structure. By discretely representing the distribution of plant sizes and types it can estimate vertical canopy structure, which directly impacts radiation scattering, throughfall interception, and in-canopy transport of scalars. By discretely representing variable disturbance history, the model can also explicitly simulate energy balance over a wide array of canopy types (closed canopies, recovering forests, grasslands, etc.) that exist within the footprint of driving meteorological data. In this experiment, the ED2 model resolves 5 different relevant tropical plant functional types (PFT); C4 grasses, early-successional trees, mid-successional trees, late-successional trees and tropical c2 grasses. In the ED2 system, PFTs are used as sets of attributes that can be applied to numerous explicitly resolved plant groups of different size and in different parts of the disturbance strata.

2.2 Generation of Surface Boundary Conditions

The creation of the vegetation initial conditions used a "spinup" process. The spin-up process is an off-line dynamic ED2 simulation, where the driving atmospheric information comes from a pre-compiled forcing data set. The vegetation is initialized with an equal assortment of newly recruited (saplings) plant types. The off-line model is integrated over several centuries by sampling from the climate data set as the vegetation reaches an equilibrium. We identify equilibrium when the total biomass of each plant functional type in a grid-cell does not change more than 0.5% over a period of 40 years. If equilibrium within this threshold was not achieved, the spin-up was allowed to continue to 508 years before stopping. For reference, ED2 simulations in old-growth central Amazonian forests take roughly 250 years to reach equilibrium biomass. The ED2 vegetation structure and composition at the end of the multi-century simulation was saved as the Potential Vegetation (PV) initial condition.

A summary of the simulation conditions in the spin-up is covered in Table 01. The model soil textures were derived from a combination of databases. Within the Amazon basin soils data were retrieved from (Quesada et al., 2011), outside the basin soils data were retrieved from a combination of RADAMBRASIL and IGBP-DIS(Scholes et al., 1995; Rossato, 2001). The climate data used to drive the

spin-up process was derived from the UCAR DS314 product (Sheffield et al., 2006)¹. The DS314 is based on the National 315 Center for Environmental Prediction's Reanalysis Product (NCEP) and maintains the same global and temporal coverage period but has bias corrections and increased resolution based on the assimilation of composite data sets. The DS314 surface precipitation record was further processed such that 320 grid cell average precipitation was down-scaled to reflect the point-scale statistical qualities of local rain-gauges. This technique used methods of Lammering and Dwyer (Lammering and Dwyer, 2000), and is explained in more detail in (Knox, 2012). The native NCEP reanalysis and European 325 Center for Medium Range Weather Forecasting (ECMWF) 40 year Reanalysis (ERA-40) were also tested as driver data sets. The down-scaled DS314 was ultimately chosen due to better agreement of estimated equilibrium biomass with observations (not shown).

The Actual Vegetation (AV) was created by continuing the simulation that produced the Potential Vegetation, and assumed that the starting year was 1900. This simulation was continued for another 108 years (until 2008) while incorporating human driven land-use change. Throughout the 508 335 year potential vegetation spin-up, as well as the 108 year continuation with human land-use, the atmospheric carbon dioxide concentration was held constant at 378 ppm, which approximates concentrations at the turn of the millennium (present day).

The model applies human land-use by reading an externally compiled data set of land-use transition matrices (Albani et al., 2006) that defines the area fractions of which various land-covers types will change to another type over the course of the year. Two external data sets are used to create 345 the land-use transition matrices, the Global Land-Use data set (GLU) (Hurtt et al., 2006) and the SIMAMAZONIA-1 data set (Soares-Filho and co authors, 2006). The GLU data set incorporates the SAGE-HYDE 3.3.1 data set and provides land-use transitions in its native format globally, on a 1-degree grid from the years 1700-1999². The SIMAMA-ZONIA 1 product provides a more intensive assessment of 350 forest cover and deforestation focused in the Amazon basin, starting in the year 2000. The data is formatted as yearly 1kilometer forest cover grids (forest,non-forest,natural grasslands). The fraction of forest and non-forest cells that fall within each ED2 model simulation grid-cell are counted for 355 each year. This enables the calculation of a rate of change

equivalent to the transition matrix format of the GLU data set. The transitions from the GLU data set from 1990-1999 were linearly scaled to have continuity with the SIMAMA-ZONIA data set that is introduced in 2000. Land-use reported in the GLU data prior to 1900 were lumped into a single combined transition and applied at the year 1900. A map of the fraction of the land surface containing human land-use is provided, see Figure 01.

Regional maps of above ground biomass for the Potential Vegetation (PV) scenario and the differences between the two scenarios (Actual Vegetation-Potential Vegetation, or AV-PV) are provided in Figure 02. The majority of above ground biomass in the *Potential* simulation is concentrated in the Amazon basin and the Atlantic Forest of southern Brazil. Late successional broad-leaf evergreens comprise most of the above ground biomass in these regions. Early successional broad-leaf evergreens are a prevalent but secondary contributor to biomass in the Amazon basin. The early succession's contribute the majority of biomass in Cerrado (savanna like ecosystem, mixed open canopy forests with grasses) ecotones found roughly in central Brazil on the southern border of the Amazon rain-forest. This is consistent with their competition and resource niche which emphasizes fast growth and colonization of disturbed areas (such as fire and drought prone Cerrado). Human land-use is concentrated in the Arc of Deforestation, which is the southern and eastern edge of the Amazon's tropical forests accessible by highways and heavy equipment. Human land-use encapsulated by the GLU data sets has degraded the Atlantic Forest. The model estimated equilibrium above ground biomass (AGB) and basal area (BA) that represent the initial condition are compared with a collection of census measurements in (Baker et al., 2004a,b), see Appendix A2.

2.3 Land-Atmosphere Coupled Simulations

The two coupled land-atmosphere simulations were conducted over four years, from January 2002 through December 2005. These four years were chosen because of the availability of lateral boundary conditions and validation data sets. With the exception of differing vegetation structure at the lower boundary, the lateral boundary conditions, model parameters, initialization of the atmospheric state, and timing are all identical between the two. The lateral boundary conditions (air temperature, specific humidity, geopotential, meridional wind speed, zonal wind-speed) are taken from the European Centre for Medium Range Weather Forecasting's Interim Reanalysis (ERA-Interim) product (Dee and co authors, 2011). The data is interpolated from the ERA-Interim's model native Reduced Gaussian Grid (N128, which has an equatorial horizontal resolution of 0.75°).

The *Actual* and *Potential* boundary conditions utilized a dynamic model process to generate structure and composition. However, when applied to the coupled simulations, the land-surface dynamics including the processes of mortality,

¹Original data sets used in the DS314 are from the Research Data Archive (RDA) which is maintained by the Computational and Information Systems Laboratory (CISL) at the National Center for Atmospheric Research (NCAR). The original Sheffield/DS314 data are available from the RDA (http://dss.ucar.edu) in data set number ds314.0.

²The use of the SAGE-HYDE 3.3.1 Global Land-use Data set acknowledges the University of New Hampshire, EOS-WEBSTER Earth Science Information Partner (ESIP) as the data distributor for ³⁶⁵ this data set.

recruitment and growth are turned off and only phenology is left to vary in time. The motivation for this decision is to create a more simple comparison and efficient simulation. Further, the length of the coupled simulations are not long $_{\rm 420}$ enough to generate large changes in above ground biomass. As an example, Lewis and Baker et al. estimate that in recent decades, the Amazon has sequestered approximately $0.6\pm0.2~{\rm Mg~C~ha^{-1}~yr^{-1}}$. Over a course of four years, this is less than 3 Mg C ha $^{-1}$, which is on the order of 1-2% of $_{\rm 425}$ total forest biomass.

A group of modeling parameters associated with convective parametrization and the radiation scattering of convective clouds were tuned using a manual binary search procedure. The parameters of mean cloud radius, mean cloud 430 depth, cumulus convective trigger mechanism, dynamic control method and the condensate to precipitation conversion efficiency were calibrated against the Tropical Rainfall Measurement Mission 3B43 product and the surface radiation from the Global Energy and Water Cycle Experiment - Sur-435 face Radiation Budget (SRB) product version 2.5. Fitness metrics include monthly mean spatial bias, mean squared error and the variance ratios (i.e. the spatial variance of mean model output over the spatial variance of the observations). Manual binary search calibration was chosen because of the 440 complexity of the parameter space, the need for human supervision and sanity checks, and the non-trivial computation requirements for each simulation. Fifty four iterations were performed, utilizing a reduced domain in the first group of iterations to facilitate a more rapid calibration. A table of 445 the finalized coupled model run-time conditions is provided in Table 02. Model output was then compared with observations of atmospheric thermodynamic variables, mean regional surface fluxes (precipitation, radiation, latent heat flux and sensible heat flux) and mean cloud cover profiles; these 450 comparisons are provided in Appendices A3, A4, A5 and A6 respectively.

3 Regional Analysis of the Actual and Potential Scenar- 455 ios

The following analysis of results will repeatedly refer to anomalies, here defined as the subtracted differences of the *Potential* vegetation scenario model output from the *Actual* vegetation scenario model output (or alternatively, AV-PV). 460

3.1 Emergence of Patterning and Continental Biases

The annual precipitation accumulations for each simulation were mapped in space and the anomalies between the two were compared for consistent patterning, see Figure 03. Each 465 anomaly appears to feature a dipole structure with a positive lobe (more *Actual* scenario precipitation) in the north and west of the Amazon Delta and negative in the south and east of the Amazon delta region. Pattern differences also appear

on the Peruvian-Bolivian border, although whether or not it can be considered a dipole is left to the reader. In each year, the precipitation anomaly shows increases on the foothills of the Andes Mountains in southern Peru and the northern tip of Bolivia. There is also a negative anomaly in precipitation in southern and central Bolivia. However, in each year there is also noise among the pattern. For instance in 2002, 2003 and 2005 there are locations in southern Bolivia that show increases in the precipitation anomaly adjacent to the area of decrease.

The patterning in down-welling short-wave surface radiation showed opposite behavior to that of precipitation (not shown), the response is strongly influenced by increased cloud optical depth where convective precipitation has increased, and vice-versa. The atmospheric model did not incorporate dynamics of aerosols or atmospheric gases other than multi-phase water, therefore variability in multi-phase water explains the variability and differences seen in optical depth. Maximum mean annual differences in surface irradiance peak at about $10~{\rm W\,m^{-2}}$, and are strongest over the dipole associated with the precipitation anomaly, as well as over the eastern Brazilian dry lands (41^oW) .

The mean annual continental bias in accumulated precipitation, evapotranspiration and total runoff is presented in Figure 04. There is little evidence of an overall continental bias in accumulated precipitation. However, the human land-use scenario generated a negative continental evapotranspiration anomaly and a positive runoff anomaly in each of the four years. Consistent patterning in evapotranspiration and transpiration anomaly were also evident, see Figure 05. Generally speaking, a negative anomaly in transpiration and total evapotranspiration is evident over the "arc of Amazonian deforestation" (starting at 48^oW2^oS going clockwise to $62^{\circ}W10^{\circ}S$, also see forest biomass differences in Figure 02). The spatial correlation between the biomass and evapotranspiration anomalies ($R^2 = 0.4$), suggests that the variability in evapotranspiration cannot be explained purely by first order effects from changes to forest structure. Second order effects and complex system feed-backs account for a portion of the variability. These effects include differences in precipitation, and potentially the effects of differences in surface heating and turbulent transport of scalars (heat and

3.2 Connecting Hydrologic Anomalies and Ecosystem Response

The availability of root-zone soil moisture, photosynthetically active radiation (PAR) and nutrients are examples of resource limitations that can potentially mediate the response of vegetation to changes in climate. Light and water are critical limiters of plant growth, disturbance (particularly through fire), and mortality (which can be functionally related to growth). However the significance of these limiters in how they may drive ecosystem response is dependent on vari-

ous factors other than the mean, such as the consistency of change (inter-annual variance), when the changes occur (seasonality) and how large the differences are relative to the total. A standard score " ζ " is one way to evaluate consistency, 515 calculated as the inter-annual mean difference (denoted by brackets "<") divided by the first standard deviation of the normalized difference η of variable x for year t (mean annual precipitation or down-welling shortwave radiation).

$$\eta_{(t)} = \frac{x_{AV,(t)} - x_{PV,(t)}}{0.5\left(x_{AV,(t)} + x_{PV,(t)}\right)} \tag{1}$$

$$\zeta = \frac{\langle \eta \rangle}{\sigma_{\eta}} \tag{2}$$

The spatial maps of the standard scores for precipitation and shortwave radiation anomalies are provided in the upper panels of Figure 06. For reference, a standard difference of 1 suggests that the normalized difference is equal to its interannual standard deviation. The maps indicate that the differences in precipitation and radiation from the two scenarios 530 are relatively consistent at the two locations previously identified (Pará Brazil and northern Bolivia).

They also indicate that the negative precipitation anomaly, and the positive radiation anomaly over the regions of intense deforestation (i.e. the Arc of Deforestation) are consistent.

The susceptibility of ecosystems to anomalous precipitation forcing may be derived from an ED2 model mechanic called the "moisture stress index" msi. This metric is simply the fraction of time that ED2 vegetation cohorts (plants) are actively keeping their stomata closed due to water limi-540 tations. For an ecosystem with N plant groups (also known as cohorts) indexed i, the mean land-surface moisture stress index is calculated by the leaf area index LAI weighting of the open-fraction f'_o of stomata for each plant group in the community. Brackets "<>" denote an averaging in space 545 and time. The stomatal open fraction f'_o , is based on the ratio of the plant's demand for root zone soil moisture, and the supply of water the roots are capable of extracting at that time. The demand requirement is driven by the maximum transpiration the plant would generate given the exist-550 ing light, carbon and vapor pressure conditions with unlimited soil moisture.

$$msi = 1 - \langle \frac{\sum_{i=1}^{N} LAI_{(i)} f'_{o(i)}}{\sum_{i=1}^{N} LAI_{(i)}} \rangle$$
 (3)

$$f'_{o(i)} = \frac{1}{1 + \frac{Demand}{Sumly}} \tag{4}$$

Vegetation communities that have experienced high moisture stress indices in the past are more likely to respond structurally to changes in precipitation, because subsequent changes in soil moisture availability will have immediate impacts on photosynthesis and the assimilation of carbon. The lower left panel of Figure 06 shows the mean moisture stress index for the *actual* vegetation scenario. The lower right panel of Figure 06 shows a map of above ground biomass as a reference to the extents of the modeled Amazon tropical forests. Moisture stress is low in areas where there is copious precipitation (the supply term). Note that in the interior of the Amazon basin, moisture stress has little to no influence on stomatal regulation (and subsequently photosynthesis). The open canopy dry forests in southern Brazil an Bolivia, as well as the Cerrado, have higher moisture stress.

4 Hydrometeorological Focus Areas - Pará Brazil and the Gran Chaco

Two locations that coincide with the pattern differences in precipitation are highlighted in Figure 06. Each location shows decreases in normalized precipitation and increases in down-welling short-wave radiation associated with the *Actual Vegetation* scenario. The vegetation of these locations also show a degree of seasonal moisture stress according to the MSI metric presented in section 3.2. One site is centered on $4.5^{\circ}S$ $50.5^{\circ}W$ in the Brazilian state of Pará. The other site is centered on $19.5^{\circ}S$ $63.5^{\circ}W$ in the northwestern part of the Bolivian Gran Chaco where it meets the Andes mountains (sometimes referred to as the Montane Gran Chaco). These two locations are chosen as areas of focused evaluation of hydrology and hydrometeorology. For simplicity, these will be referred to as the *Pará* and *Gran Chaco* focus areas.

A representation of the vegetation demographics at the centroids of the two focus areas, as estimated by the ED2 model, are provided in Figure 07. The natural landscapes at the Pará focus areas are dominated by tropical evergreen forests, and are close to (but not within) the ecotone transition between tropical forests and Cerrado. The offline model spinup of the Actual vegetation scenario imposed pastures on approximately 1/3 of the land-cover. Roughly 10% of the landscape contains old-growth forests that have gone 200 years since the last disturbance. The focus area in the Gran Chaco is located in a region influenced by the outlet of the South American Low Level Jet. The continental precipitation recycling ratio in this area is relatively high compared to the rest of the continent (Eltahir and Bras, 1994). This exact location in the Gran Chaco is a dry forest ecosystem that borders adjacent ecotones of tropical rainforests to the north, montane ecosystems to the west and grasslands to the south. The ED2 model estimated a Potential vegetation demographic that is fairly consistent with the depiction of dry forests, a sparse cover of short trees with grasses in the under-story. The Actual vegetation simulation of the Gran Chaco, driven by the GLU data set (Hurtt et al., 2006), forced 25% of the natural

landscape to pasture (grasses), with an accompanying 20% 615 of abandoned and degraded lands. Human land-use, as represented in the ED2 model at this specific location, led to a collapse of the estimated tree cover, which includes natural landscapes. This specific site is undoubtedly a more aggressive representation of the differences between the *Actual* and 620 *Potential* scenario ecosystems in this region. Human land-conversion has not lead to a collapse of the Gran Chaco's dry-forest ecosystems for the region as a whole.

4.1 Canopy Water and Energy Balance - Pará

Simulated annual precipitation at the Pará focus area was typically around 1500 millimeters per year, the surface energy flux was dominated by leaf evaporation and transpiration. Transpiration dominated vapor flux in the dry season 630 (May-November). Runoff in the form of drainage through the lower soil column occurred mostly during the wet season. The time series water and energy balance at the land-surface is summarized in Figure 08. Accumulated water fluxes from the Potential vegetation scenario are shown in the upper left 635 panel (A), anomalous accumulations are shown in the upper right panel (B). The Actual vegetation scenario experienced roughly 10% less surface precipitation at the Pará focus area. However, the site experienced a small net increase $(30mmyr^{-1})$ in precipitation through-fall, due to a propor- 640 tionally stronger decrease in leaf interception surfaces. There is also increased drainage in the actual vegetation scenario, which appears to be symptomatic of both increased throughfall and the decrease in the root-zone soil moisture sink from transpiration.

The *Actual* vegetation scenario receives more total shortwave and long-wave radiation $(R_{SD}+R_{LD})$, which is di-645 rectly attributable to the decrease in mean convective cloud albedo associated with the decrease in convective rainfall at the site. Although the site receives more total incoming radiation in the *Actual* vegetation simulation, the surface albedo increases with the conversion of forests to pasture. This re-650 sults in more reflected radiation and a decrease in combined sensible and latent heat flux (H+L), see the bottom right panel (D) of Figure 08.

4.2 Canopy Water and Energy Balance - Gran Chaco 655

The annual precipitation at Gran Chaco in the *Potential* simulation ranged from 500 to nearly 1000 millimeters. Annual precipitation was roughly 15% lower in the *Actual* vegetation simulation. A summary of the hydrologic response and 660 the anomalies are shown in Figure 09. Like the Pará site, land-conversion drove a decrease in leaf area, and therefore a significnat decrease in leaf interception of precipitation in the *Actual* (AV) simulation. However in this case, the relative decrease in interception surfaces due to deforestation 665 was less affecting than the decrease in total precipitation due to land-atmosphere feedbacks. Therefore, the *Actual* vegeta-

tion simulation experienced a *Decrease* in total precipitation throughfall. Soil evaporation acounted for half of the water losses, while leaf evaporation and transpiration equally combined to represent the other half. The relatively low preciptation rates promoted almost no detectable runoff. Transsiration decreased by 20% in the *Actual* simulation, which is a direct consequence of decreased stomatal density and precipitation throughfall.

Notwithstanding the decreased precipitation throughfall in the *Actual* simulation, surface evaporation unexpectedly increased. Despite decreased precipitation through-fall, upper soil column moisture from rain events has a longer residence time in the root zone, as shown in Figure 010. This is an effect of decreased transpiration, and thus moisture in the grass root zone lasts comparatively longer into the dry season. Note that the relative reduction in precipitation throughfall nearly balanced the reduction in transpiration. There is little evidence to suggest that increased soil evaporation rates would maintain indefinitely in the *Actual* scenario, which could alternatively be associated with transient changes in soil column storage.

Like the Pará site, land-conversion at Gran Chaco also drove an increase in total surface albedo, a direct effect due to the loss of dark foliage. More incident radiation is reflected which reduces net-radiation. Unlike the Pará focus area, the albedo effect is stronger than the increase in incident radiation, which leads to decreased sensible heat flux (see Figure 09) and slightly cooler annual surface temperatures (not shown).

4.3 Land-Atmosphere Coupling - Pará

A box is constructed around this site for the month of September 2003 that contains the extents of a continuous space with negative precipitation anomaly, see Figure 011. Table 03 shows a selection of spatio-temporal mean indicators from the bounded domain. To summarize the differences in surface fluxes, the results are consistent with the single site time-series analysis, where the *Actual* vegetation scenario experienced a decrease in net-radiation ($-10~{\rm W\,m^{-2}}$, despite increased incident shortwave radiation) and an increased mean surface albedo.

The decreased precipitation and surface energy flux of the *Actual* scenario are accompanied by a boundary layer with lower equivalent potential temperature, see Figure 012. The decrease in boundary layer equivalent potential temperature has a strong physical connection to explaining the decrease in precipitation, particuarly since the vast majority of the precipitation was generated through the convective parameterization (data not shown). This was verified by recording a log of failures in deep convection (precipitating convective clouds) generated by the convective parameterization. All of the bias in these convective failures occuring in the *Actual* simulation was attributed to the generation of convection that resulted in clouds that were too thin to be classified

as precipitating deep convection clouds. This is indicative of the how much convective available potential energy (CAPE) can be released through convective bouyancy, which is controlled by the moist static energy in the surface parcels as well as the moist static energy of the mean atmosphere over the depth of the troposphere. Alternatively, there was no positive bias in the logs associated with the inability to trigger parcel bouyancey.

It is questioned if the driving force behind the reduced equivalent potential temperature profiles of the *Actual* scenario is solely the result of local surface fluxes or caused by changes in the regional energy circulation. Both scenarios net a negative moisture convergence (divergent) budget for the month (total integrated water mass flux through the box boundaries, normalized by the box area). The *Potential* vegetation scenario loses more water $(-51.32~{\rm kg\,m^{-2}})$ through its lateral boundaries than the AV scenario $(-37.14~{\rm kg\,m^{-2}})$, see Table 03. This can be visualized by flux vectors as well, see Figure 013. In the *Potential* vegetation case shown in left panel, the flow vectors run east-to-west and up-gradient, which means the advecting air mass is gaining moisture and is consistent with the net water divergence described in Table 03.

4.4 Land-Atmosphere Coupling - Gran Chaco

Similar to the case study in Pará, a bounding box was constructed around the Gran Chaco site in April 2003 that captures a spatially continuous negative anomaly in precipita- $_{745}$ tion. The boundaries are shown in black against the mean total and mean anomaly in monthly precipitation and evapotranspiration, see Figure 014. Mean statistics are shown in Table 04. The *Actual* scenario experienced less than half as much precipitation (41 kg m $^{-2}$ compared to $85 \ \rm kg \, m^{-2}$). The $_{750}$ evaporation anomaly between the two scenarios was not as strong (111 kg m $^{-2}$ in the *Potential* scenario compared to $83 \ \rm kg \, m^{-2}$ for *Actual*).

The Actual vegetation scenario experienced more downwelling shortwave radiation yet less net surface radiation, 755 which was influenced by an increased surface albedo. Like the Pará case, the Actual scenario experienced a lower equivalent potential temperature over the boundary layer, see Figure 015. The convective parameterization logs accounted that the Actual scenario experienced a great deal more failed 760 convective events associated with an inability to generate deep clouds (the same reason as the Pará case). Note that the model's cloud depth parameterization is controlled by convective available potential energy. However in this case, about 25% of the bias in failures was also explained by an in-765 ability to trigger convection. This is interesting because this case was slightly different from the Pará in that the Actual scenario did not experience higher levels of turbulent kinetic energy over the boundary layer.

An analysis of moisture convergence and advective flux $_{770}$ was used to better understand the local versus regional controls that drive convective precipitation. Both scenarios showed negative moisture convergence, typical during the onset of the dry season in this region, refer to Table 04. The *Potential* scenario showed less moisture divergence (-37_{775} kg m $^{-2}$) than the *Actual* scenario (-52 kg m $^{-2}$). The moisture advected into this region comes via northerly winds from the moist Amazonian air mass, see the left panel of Figure 016. Moisture transport from the north decreases in the *Actual* scenario, see the right panel of Figure 016.

5 Discussion

5.1 Secondary Forests and Evapotranspiration Pattern-785 ing

The maps of evapotranspiration and transpiration anomaly showed pattern similarity with the differences in above ground biomass, as compared to a lack of pattern similarity 790 between precipitation and above ground biomass anomalies. The stronger correlation between the evapotranspiration and forest biomass anomalies can be rationalized by understanding how the ED2 model resolves canopy hydrologic process. In ED2, closed canopy tropical broadleaf evergreen forests 795

have higher leaf area indices than grasses, with possible exceptions during drought deciduous leaf drop. Deforestation of closed canopy forests will therefore decrease total leaf area. Rainfall that is intercepted in ED2 has two outcomes: it can either re-evaporate or drip to the land-surface (one shortcoming of this assumption is that epiphytes may store water directly from leaf interception). Throughfall precipitation has multiple outcomes: it can evaporate from the surface, become stored in the soil and vegetation indefinately, leave via transpiration or leave via runoff. The evaporation rates between the leaf and soil surfaces with the canopy air-space are regulated by two factors, the aerodynamic resistance and the effective vapor pressure deficit between the respective surfaces with the air-space. In the ED2 model formulation, which scales in-canopy wind-speeds following Massman, the aerodynamic resistance in the forest canopy will attenuate from top to bottom as wind-speeds monotonically decrease. Moreover, water that becomes bound in the soil matrix will have a decreased vapor pressure deficit with the adjacent air (compared to leaf water which is not bound, and assumed saturated) due to the effects of pore spaces at the soil surface (Lee and Pielke, 1992). Therefore in ED2, precipitation that is intercepted in the canopy has both an extra opportunity (simply considering the order of process) to evaporate back to the atmosphere, but also has a tendency towards both decreased aerodynamic resistance and increased vapor pressure deficit that drives evaporation rate.

However this explanation of process only considers the immediate structural effects of deforestation, which is not static but also has phases of recovery when left un-managed. During the recovery cycle of tropical forests following natural disturbance, successfull new growth in the canopy is typically dominated by pioneer species. Pioneer species have lower wood density, higher maximum photosynthetic capacity and quicker vertical growth than late successional species (Laurance et al., 2004; Poorter et al., 2006; Chave et al., 2006). The canopy leaf area may flush to previous levels within a century, yet it may take several centuries for total forest biomass to rebound. This type of behavior was observed in the model spin-up, where newly disturbed patches of land in the central Amazon reached maximum leaf area over a span of a few decades, compared to the length of time (more than a century) it took for biomass to stabilize.

The results presented here support that secondary forests undergoing recovery from deforestation can drive detectable pattern increases in total transpired water across the region. It is rationalized that at these locations, photosynthetic capacity is scaled by leaf area, as well as a distribution of members skewed towards rapid growth and high photosynthetic capacity. If there is sufficient available soil water there would be an expected increase in total transpiration. In the *Actual* model scenario containing deforestation effects, the model estimated an increase in early successional tropical evergreens (pioneers) in the recovering forests of northern Pará and eastern Amazonas (centered on 5S 58W), see Figure 02

second row right panel. There is an increase in regional transpiration here (Figure 05) that has a strong pattern match with 850 the increase in early successional biomass, moreover there is little evidence of influence from pattern precipitation here (see Figure 03).

5.2 Regional Surface Water Balance and Runoff Gener- 855 ation

There was a clear and consistent bias in the total annual evapotranspiration (negative) and runoff (positive) estimated in the Actual scenario when integrated over the entire domain of northern South America. There is some rudimen-860 tary explanation of the first order biases in evapotranspiration explained above. The regional runoff bias appears to have several potential explanations and remaining questions. It is clear that in the Actual scenario, decreased canopy interception promotes a first order effect of increased canopy throughfall. However we saw in comparing the Pará and Gran 865 Chaco case studies, that changes in canopy interception can be offset by changes in incident precipitation. Therefore increased canopy throughfall from deforestation is not ubiquitously associated with regional increases in runoff. In a regional water balance analysis, (d'Almeida et al., 2007) also 870 observed that wide-spread regional deforestation promoted decreased evapotranspiration and increased runoff. Similar to this study, they also found that precipitation feedback response to deforestation had the potential to impact the water cycling on par with direct effects of surface hydrologic 875 parameters (although in there results, bi-directionally weakening or strengthening the water-cycle depending on heteogeneity and land-cover fractionation).

In the simulations presented here, the increased continental runoff from the *Actual* scenario is driven by higher mean 880 annual soil moisture. The regional mean soil moisture depth simulated in the *Actual* scenario oscilated around a interseasonal mean of 1.40 meters (over an 8 meter medium), averaging 5 centimeters greater than the *Potential* case. As increased runoff has a negative feedback on increased soil 885 water, and there was no consistent bias in precipitation, it is most likely that the positive shift in mean annual moisture in the *Actual* scenario is driven by the decreases in regional evapotranspiration.

This experiment highlighted the use of relatively sophis-890 ticated vegetation biophysical processes, which incoporated variable vegetation structure, composition, rooting depth and uptake. However the modeling framework did not incorporate lateral transport of any surface moisture. Therefore, these results must be interpreted with the understanding that lateral re-infiltration, lateral vados zone flow, interflow and water table dynamics could not influence soil moisture dy-895 namics. In light of this, this experiment provides a gauge on the strength of the control that evaporation response to deforestation can have on regional water balance and runoff generation. There has been some evidence that soil hydrologic

properties can be affected by land-conversion in the tropics, (Zimmermann et al., 2006) found that both infiltrability and upper root-zone saturated hydraulic conductivity was highest in intact rainforest compared to pasture and tree plantations. Decreased infiltration in pastures has been related to increased runoff generation as well (Munoz-Villers and McDonnell, 2013). But there seems less certainty in the literature in quantifying the evaporative response from canopy and soil to regional Amazonian deforestation, degredation and recovery.

5.3 Intersection of Seasonal Hydrology and Represented Plant Functional Types on Canopy Process

The two case studies showed that the structure of the vegetation canopy can influence the seasonal cycle of moisture storage and land-atmosphere moisture flux. At the Gran Chaco site, transpiration was greater in the *Potential* vegetation scenario during the wet season when the deeper roots and higher stomatal density of the open canopy forest could access available soil moisture. Alternatively, total evapotranspiration was greater in the *Actual* vegetation scenario at the onset of the dry-season, due mostly to the fact that the grasslands had more available water stored in the upper root-zone, recall Figure 010.

The natural vegetation at the Gran Chaco site is represented "in-model" with early successional broad-leaf evergreen plant functional types, with accompanying C4 grasses. While the demographic size structure, composition, and the openness of the canopy shows some similarity with dryforest structure, it must be realized that the wider range of water conservation strategies observed in nature could influence how the differences in surface to atmosphere energy fluxes play out at this site.

These findings suggest that the next generations of earth system models may benefit from improvements in representing plant diversity. The seasonal flux of surface to atmosphere water vapor is regulated by plants, and can potentially impact the hydrometeorological dynamic of the region. Total evapotranspiration during the transition from the late wet season to early dry season (April) at the Gran Chaco site was larger in the potential vegetation scenario. As shown in the hydrometeorological analysis in Section 4.4, this was a time in the seasonal cycle that exhibited relatively strong differences in the instability profiles in the atmosphere, albeit from competing local and regionaly driven mechanisms.

5.4 Drivers of Anomalous Convective Precipitation

The negative precipitation bias at the two focus areas in the *Actual* scenario were both accompanied by reductions in net radiation, decreased annual latent heat flux (evapotranspiration) and increased albedo. They also experienced boundary layers with lower mean equivalent potential temperature, which was then related to fewer cumulus events that lacked

sufficient convective available potential energy to generate deep clouds (diagnosed through convective logs). It is believed that differences in convective available potential en- 955 ergy underlies the convective precipitation anomaly.

Positive anomolies in sensible heat flux for the *Actual* scenario are concurrent with increased turbulent kinetic energy, which is thought to promote the circulations and boundary layer development that lifts air parcels to trigger convection 960 (Wang et al., 2009; Fisch et al., 2004). The case study at Pará did show increased boundary layer turbulent kinetic energy and sensible heat flux, and it is possible that increased boundary layer turbulence may had helped recoup some losses in precipitation. This is deductive reasoning however, following 965 that the Gran Chaco case study did not have increased turbulent kinetic energy associated with the *Actual* scenario and it did experience more failed convective events associated with the inability to overcome convective inhibition (where there was no such trend at Pará).

The two case studies offer evidence that the negative precipitation anomaly, can be mediated by by primarily local effects and also by a combination of local effects and regional circulation effects. The Pará case study suggests that the negative precipitation anomaly is driven primarily by changes 975 in the local surface energy flux. The dry season prevailing winds at Pará flowed up the moisture gradient (i.e. gaining moisture not losing moisture). And according to the flux vectors in Figure 013, despite the decreased precipitation experienced in the Actual scenario, the prevailing winds fluxed 980 more moisture into the domain when compared to the Potential case. In contrast, the Gran Chaco case study showed evidence that the precipitation anomaly was responding to a change in the regional circulation as well as changes in the local surface energy fluxes. Here, the prevailing winds came 985 out of the north and flowed down the moisture gradient (i.e. losing moisture). While the domain was a net source of moisture (divergent) the net flux into the domain via advection was positive. The Actual scenario experienced a relative decrease in advected moisture flux from the prevailing winds, 990 see Figure 016.

5.5 Uncertainty in Model Estimates

Model uncertainty exists in many sources, including the ini- 995 tial condition, the boundary conditions, within the scale limitations of the resolved processes, the mechanics of the model processes and the parameters that govern the processes. In limited area models, there is also variability in the lateral boundary conditions. For the current research, this variability can affect the differences detected between the two scenarios, potentially impacting results. Ideally, this variability space can be explored in depth, perhaps using ensembles over multiple decades, including differing starting years and perturbations. The range of the variability space that can be sam-1005 pled is limited by computational expense of the simulations. The four year simulations presented here took approximately

two months each using 96 parallel computational cores with high-speed interconnects, the computational time being an result of the highly memory intensive ecosystem model and atmospheric time-stepping that is relatively short (30 seconds) compared to general circulation models. However, we maintain that the pattern differences in precipitation and evaporation between the two scenarios showed consistency enough to merit commentary.

The simulations used a number of different external datasets, each of which contained information at different spatial scales. This includes the soils information (variable scales), the human disturbance transitions (1^o) , the scale at which the forest structure was "spun-up" from climate driver data (1^o) versus the scale at which that data was re-sampled (using nearest neighbor) in the coupled simulation (64 km). The uncertainty associated with the dynamics of the coupled simulations are subject to the scale of the information provided by these external datasets, as well as any biases that might be inherent in those datasets. This may be particularly true in the case of the lateral boundary condition data.

It is also necessary to acknowledge that there is uncertainty inherent in model process. For instance, in the absence of explicitly resolving bouyant updrafts (which is not possible in meso-scale simulations) the successful triggering of convection was based on the negative energy between the level of updraft and the level of free convection. There are alternative methods for estimating where updrafts start and how a parcel may or may not overcome inhibition to reach free bouyancy. We chose a straight forward parameterization that compares the negative buoyancy to a threshold. There are other trigger mechanisms that can be used, such as estimating the statistical distribution of vertical kinetic energy present in eddy motions at the level of updraft, and using that to estimating the likelihood that eddies will overcome negative buoyancy through force balance computations. This is simply an example of how the simulations were undoubtedly influenced by these choices, and this is something that must be considered when interpretting the output from any complex numerical model.

In the acknowledging uncertainty inherent in the simulations presented here, we have also tried to verify if these simulations can show agreement with observations. The Actual scenario's simulations were compared to a group of different observations with the intention of verifying the modeling system's ability to represent key processes. The model's regional demographic of vegetation biomass was compared to field inventory data, atmospheric thermodynamic profiles were compared to soundings, mean all-sky profiles of cloud water were compared to satellite estimates and the seasonality of precipitation, net-radiation, latent heat flux and sensible heat flux were compared with multi-data composite data products (see Appendices A2,A3,A4,A5 and A6). The comparisons with observations (which harbour their own uncertainty) suggest the model system is adequate to make a meaningful comparison between the two scenarios, yet not without room for improvement. In particular, this modeling study and of regional coupled simulations in general, would stand to benefit from improvements in how large-scale precipitation and canopy interception is handled.

A small selection of published research has conducted similar simulations to those reported here, with results thatoss offer limited comparison. Perhaps most similar, Bagley et al. compared different land-cover scenarios with some similarities to our *Potential* and *Actual* vegetation, finding a generally weaker patterning of precipitation anomaly. However, evidence of the strongest pattern differences also occured of south of the Amazon delta, which may be viewed as correlated with the depression in the precipitation dipole presented in this work. The simulations conducted by Bagley et al. also used a different modeling system (The Weather Research and Forecasting System WRF) and land-surface model (Noah 1075 LSM), suggesting that model intercomparison would be a useful endeavor in rectifying the regional hydrometeorological effects of South American land-cover change.

Other studies have used similar simulation approaches to the work described here, but targetted land-atmosphere ef-1080 fects under future conditions. Global simulations conducted by Medvigy et al. found that future business as usual (aggressive) deforestation would generate dipole precipitation differential with respect to conservative deforestation scenarios. The length scales of the patterning in the dipoles (100s₁₀₈₅ of kilometers) showed similarity with results presented here. However, the locations of differences they presented were different, as of course were the driving land-use scenarios as well. Walker et al. also evaluated the effects of future deforestation scenarios on the regional hydrometeorology, find-1090 ing that massive deforestation outside of protected areas will lead to basin-wide changes in rainfall, both positive and negative. In summary, there are some commonalities between the various regional simulations, albeit from differing landuse scenarios; impacts on hydrometeorological climate are 1095 expressed most strongly as changes in dry-season precipitation, and that patterning exists to varying degrees at large length scales (> 100 kilometers) (Walker et al., 2009; Bagley et al., 2014; Medvigy et al., 2011).

6 Conclusions

The simulations presented here produced negative anomolies in evaporatranspiration rates that showed pattern similarity₁₁₀₅ with areas where deforestation had driven noticable differences in aboveground biomass. The results showed mixed support that historical land conversion has had influence on the patterning of South American precipitation. Over four years of simulation most of the region showed little consis-1110 tency in annual anomolies. However patterns in the precipitation anomoly emerged at specific locations, where mean annual precipitation showed moderate yet consistent differences as evaluated through a standard score. One pattern

showed a dipole structure that occured near eastern Pará Brazil. The other pattern, showed a negative anomaly in the Gran Chaco region of central and southern Bolivia. In this area, their was positive precipitation anomaly at the Peru-Bolivia border, yet the general patterning may not-necessarily be considered a dipole.

The simulations suggest land conversion in South America has not had a large impact on mean precipitation in the region as a whole. Mean regional precipitation was lower in the *Actual* scenario in only three of the four simulation years. In contrast, differences in mean continental runoff (increased with human land-conversion) and total evapotranspiration (decreased with human land-conversion) were both consistent from year to year and showed greater differences compared to precipitation.

It was also observed that a large-scale shift in forest composition to early-successional tropical evergreens following deforestation and abandonement in central-eastern Amazon produced a noticeable increase in grid-average transpiration. This increase was striking not only because the pattern difference matched the pattern shift in composition, but particularly because the overall biomass in this area had decreased compared to the natural (*Potential*) condition, suggesting that the physiological differences of the vegetation had influenced regional fluxes and not necessarily the quantity.

Beyond characterizing mean and pattern differences in the regional water balance, it was identified that changes (be they moderate) in precipitation may have occured in locations where terrestrial vegetation actively regulates photosynthesis due to water limitations and that the changes in precipitation can be attributed to combinations of changes in local surface fluxes and changes in the regional atmospheric circulation. An assessment of the regional vegetation's response to moisture stress has indicated that both these locations show some susceptibility to changes in root-zone soil moisture (plant stomata were actively regulated to conserve water over a broad range depending on exact location, varying from 20 - 80% of the time), however the ecosystems of Gran Chaco are generally dryer and show greater susceptibility. In both focus cases, deforestation led to increases in total surface albedo, driving decreases in net-radiation, boundary layer moist static energy and ultimately convective precipitation. However, the differences in precipitation at Pará Brazil are more strongly connected with these localized differences in land-surface energy flux. The hydrometeorological analysis near Gran Chaco suggests that human land-conversion has had some impact on the strength of the South American moisture circulation in the southwestern portion of the Amazon, which claims partial responsibility along with differences in surface fluxes for an estimated decrease in annual precipitation in the Gran Chaco.

A1 Land-Atmosphere Model Coupling

The atmospheric model (BRAMS) and the land surface model (ED2) are loosely and asynchronously coupled. This assumes that the two models pass each other boundary conditions at a frequency that captures the natural variability of the flux, yet the fluxes between models are not dynamically changing as prognostic variables within the numerical integrator sub-stepping. The atmosphere (BRAMS) provides information at the beginning of the ED2 forward step, while ED2 assumes this information is constant over the duration of its forward step and makes a time average of the fluxes during the step to return.

Meso-scale atmospheric models at spatial resolutions of tens of kilometers typically perform integrations on the order of tens of seconds to maintain numerical stability and convergence. The simulations in this work used an atmospheric time-step of 30 seconds (for non-accoustic dynamics,1175 accoustics were 10 seconds). Ideally, the land surface and atmospheric models operate at time-steps that approach the infintessimaly small. The ED2 has a non-trivial computational overhead due to the large number of vegetation cohorts experiencing energy balancing within each grid cell. Ultimately,1180 the ED2.1 model used a 120 second time-step while coupled to BRAMS.

Atmospheric variables such as air temperature, humidity and wind-speed are provided to ED2 at a reference height of 70 meters. This is required because ED2 internally calculates turbulent transport of heat, moisture and momentum through the canopy and into the inertial sub-layer of the lower atmosphere. The turbulent transport of scalars through the sub-layer above the canopy were calculated following Beljaars and Holtslag. These calculations relied on gradients that were 190 baseed on the enthalpy, density and specific humidity of air at reference temperature and within the canopy interstitial air-space. A list of the variables required to drive ED2 is provided in Table 05.

The ED2 model passes boundary fluxes at the grid scale 195 to the atmospheric surface layer as an area-weighted average across patches. There are three groups of information the land surface must provide the atmosphere model: 1) the topography which governs the geometry of the atmosphere's coordinate system and drag, 2) a lower boundary condition 200 for turbulent closure, i.e., the vertical velocity perturbations of momentum, heat, moisture and carbon, and 3) a surface albedo for the atmospheric radiative transfer calculations. These variables are listed in Table 06. Because the ED2 model prognoses spatial variables at the patch sale (such as 205 canopy temperature, humidity, etc), spatial averaging is required for flux variables. For any generic variable S, at a grid cell with M patches of area A_{patch} , the area averaged quantity is straight forward.

$$S_{grid} = \sum_{j}^{M} S_{patch,j} A_{patch,j}$$
 (A1)

$$\sum_{j}^{M} A_{patch,j} = 1 \tag{A2}$$

More detailed explanation of how turbulent fluxes are calculated in the ED2 model can be found in Medvigy et al. (2009) and Knox (2012).

A2 Regional Above Ground Biomass

The model estimated live above ground biomass (AGB) and basal area (BA) that represent the initial condition is compared with a collection of census measurements in (Baker et al., 2004a,b), see Figure 017. A map is provided showing the locations of the plot experiments, see Figure 018. The coordinates from the measurement stations were matched with ED2 nearest neighbor grid cells. Consistent with observations, only ED2 plants greater than 10cm in primary forests were included in the comparison. The published measurements were taken at different times over the previous decades, the lag between these sites and the time of the simulation initial condition (January 2008) varies and is reported. Tree diameters in ED2 is diagnosed allometrically from structural carbon, similar to allometric equations in (Chave et al., 2001) and (Baker et al., 2004b), with differences in parametrization to reflect functional groups.

Please note that the forest inventory data makes no assumption that aboveground biomass is in equilibrium. This comparison is only intended to evaluate the present day static representation of forest structure. As applied to the coupled simulation, the were treated as static. As stated earlier, the length of the simulation did not merit the need to incorporate dynamics.

The majority of sites show fair agreement with model estimated above ground biomass and basal area. The exception to this is the cluster of sites located in eastern Bolivia at Huanchaca Dos (HCC), Chore 1 (CHO), Los Fierros Bosque (LFB) and Cerro Pelao (CRP). There are several potential reasons for this discrepancy attributed to the model such as: variability in 1) climate forcing data (most notably precipitation and vapor pressure deficit as these are water limited growth conditions), 2) edaphic conditions and 3) plant functional parameters. These plots are close to the southern Amazonian transition between tropical rain-forests and Cerrado type open canopy forests where gradients in vegetation types are large and uncertainty is expected to be greater. Large spatial gradients in biomass are also reflected in the differences among the cluster of plots (HCC, CHO, LFB and CRP), (124.8 Mg ha⁻¹ above ground biomass at CHO and 260.0 Mg ha⁻¹ AGB at HCC). The sharp gradient in forest biomass suggests that the omission of sub-pixel variability in 1220

the modified DS314 climate forcing could explain a portion of this difference, along with any persistent biases.

A3 Thermodynamic Mean Profiles

The objectives of this experiment require that modeling system output match mean observations to such a degree that there is trust in the model's ability to represent physical processes. It is believed that the relative differences between the two simulations have validity if model processes repro-1265 duce mean observed tendencies in the land surface and atmosphere.

Mean monthly profiles of model estimated air temperature, specific humidity and moist static energy are compared with mean radiosonde data over Manaus Brazil (see Figures₁₂₇₀ 019 and 020). Comparisons are made at 00:00Z UTC (8pm local time) and 12:00Z UTC (8am local time). The model estimates a consistently warmer atmosphere, in the range of about two degrees both morning and evening. Model estimated specific humidity in the lower troposphere is lowen₂₇₅ than the radiosondes, see Figure 019. Moist static energy is slightly underestimated by the model in the lower troposphere and then overestimated in the mid to upper troposphere. This may suggest that the model is convecting relatively large quantities of warm moist air at the surface and entraining it to the upper atmosphere.

A4 Inter-seasonal Precipitation and Surface Radiation

Monthly precipitation and down-welling shortwave radiation 285 in the model is evaluated as spatial means within five separate zones of analysis. The boundaries of the zones of analysis are shown in Figure 018. Monthly mean model estimates are again compared to precipitation estimates from the Tropical Rainfall Measurement Mission (TRMM) 3B43 product²⁹⁰ and surface radiation from the Global Energy and Water Exchanges Project (GEWEX) Surface Radiation Budget (SRB) product version 2.5.3 There is generally acceptable agreement between the model and TRMM estimated precipitation. The seasonal variability in both datasets is greater than thein 295 differences, see Figure 021. The largest differences are reflected in the strength of of the wet-season peak precipitation in Zone 3 (central-eastern Amazon) and the severity of the dry season precipitation in Zone 5 (southern Brazil). The timing of peak and minimum rainfall show generally good:300 agreement, particularly in Zones 2-5. The lower estimate of mean precipitation in southern Brazil is consistent with decreased cloud albedo and increased down-welling short-wave radiation at the surface, see Figure 022. Surface shortwave radiation is modestly over-estimated compared with the SRB1305 estimates for most other cases.

A5 Inter-seasonal Latent and Sensible Heat Flux

Similar to the comparison in Appendix A4, monthly mean model estimated latent (see Figure 023) and sensible heat flux (see Figure 024) are compared with means from a benchmark. In this case, we compare output with the products of Jung et al.. These products are based off of surface observations, which are upscaled using gridded explanatory variables from various sources including the Climate Research Unit (CRU), Global Precipitation Climatology Centre (GPCC) as well as the ERA-Interim product used in this study. The inter-seasonal means also incorporate spatial means within the domains shown in Figure 018.

The purpose of this comparison is to give some benchmark of the ecosystem model's ability to partition energy flux at the land-surface. The patterned flux of energy is dependant on the atmospheric model and its lateral boundary conditions as well. Compared to the synthesis product, the model system estimated a stronger inter-seasonal variability of latent and sensible heat flux over southern Brazil and the South American Convergence zone. Both model and observations showed relatively lower seasonal variability over the Amazon. In these regions, the model estimated small biases in sensible heat flux.

The greatest discrepancies between benchmark and model output was a latent heat flux bias (higher in the EDBRAMS model) among the three zones covering the Amazon basin. There are several possible explanations for this, outside of uncertainty inherent in the benchmark product. Latent and sensible heat flux contribute a portion of the total energy flux balance through the land-surface, which also includes contributions from change in storage, diffusive ground heat flux, net radiation and the enthalpy contained in the mass flux of precipitation and runoff (enthalpy flux from density and pressure changes can be assumed near zero). Latent heat flux also contributes to a portion of the water mass balance at the landsurface, which also includes precipitation mass flux, change in storage and total runoff. During previous experiments with the ED2 model used in offline simulations, we found that the surface water balance was sensitive to the scale of the precipitation input. When driving the land-surface model with precipitation resolved at coarse scales (such as native NCEP and ECMWF products), the leaf evaporation rates were dispoportionately high. It was found that low but continuous precipitation rates from these products promoted a slow wetting of canopy leaves, and as a result the canopy leaves overflowed to the point of generating throughfall with less frequency and magnitude. Point scale precipitation rates from raingauges in the Amazon showed a much larger variability in precipitation intensity. After using downscaling routines based on Lammering and Dwyer to preserve the monthly volume of gridcell precipitation and creating point-scale precipitation intensity, these products (specifically the DS314 from UCAR) elicited a shift in canopy throughfall rates thereby decreasing latent heat flux and increasing surface runoff. Precipi-

³These data were obtained from the NASA Langley Research Center Atmospheric Sciences Data Center NASA/GEWEX SRB Project.

tation scale and how it affects the distribution of intensity, storm duration and the time-between storms is a challenge in couple model simulation, that cannot be overcome using 1360 the same techniques as offline simulations. The spatial and temporal resultion of the simulations used in this work (40 kilometers with 15 minute time-step between convective precipitation calls) are smaller than the reanalysis models (larger than 1 degree), yet they cannot generate point-scale precipitation. There are various approaches to ameliorating precipitation scale effects, such as using multi-atmosphere multiland (MAML) sub-grid methods and by employing creative ways at the land-surface to generate throughfall volumes that match observations even when driven with precipitation rates that cannot match those that are observed. Regional and meso-scale couple simulations such as the work presented 1370 here, could benefit greatly from advances in this area.

A6 All-Sky Cloud Water Content Profiles

Cloud profile validation datasets were constructed from CloudSat Cloud Water content (2B-CWC-RO) and Cloud Classification (2B-CLDCLASS-LIDAR) datasets. 4 Overpasses during February 2007-2011 that intersected the geographic subset between 3^oN-12^oS and 70^oW-55^oW were collected and interpolated to a constant vertical datum above the surface. Overpasses typically occurred near 17hrs UTC.

Making a rigorous comparison of the model estimated cloud water and observation is challenging. Consider that,385 the simulation time frame does not overlap with the Cloud-Sat mission time-frame, so these comparisons are treated as proxies to climatology and not weather validation. Cloud-Sat measurements are known to have signal loss, attenuation and clutter during moderate to intense rainfall; events such 390 as these could not be filtered from the comparison. It must also be assumed that the cloud classification algorithm is not without error. Nonetheless, the purpose of the comparison was to get a sense of whether the simulations estimated rea-1395 sonable mean ranges of water contents and cloud fractions, and also if the phase transitions (liquid to ice) were occurring at reasonable elevations.

The all-sky cloud water content profiles for both cumulus and non-cumulus clouds are provided in Figure 025.1400 The peaks in model estimated mean cloud water content showed reasonable agreement across liquid and ice cloud types. The model estimated generally more water content in both phases, skewed towards higher altitudes and showed a uni-modal shape in the vertical distreribution. It is possible that CloudSat relative underestimation could be explained by the omission of precipitating clouds.

⁴Cloudsat data sets were provided on-line by CloudSat Data Processing Center, courtesy of NASA, Colorado State University and their partners.

Acknowledgements. This work was made possible through both the National Science Foundation Grant ATM-0449793 and National Aeronautics and Space Administration Grant NNG06GD63G. The authors would like to thank D. Entekhabi and E.A.B Eltahir for their generous discussion and council. Further, the peer reviewers provided us with thorough, challenging and fair recommendations. We believe their efforts have contributed to significant improvements to this manuscript and we thank them.

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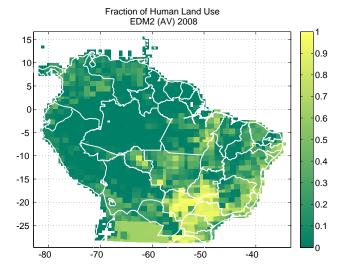


Fig. 01. Fraction of the land surface with human land-use. Output is taken from the ED2 the *Actual Vegetation* (AV) simulation, which was driven with Global Land Use (GLU) and SIMAMAZONIA-1 land-use transition data.

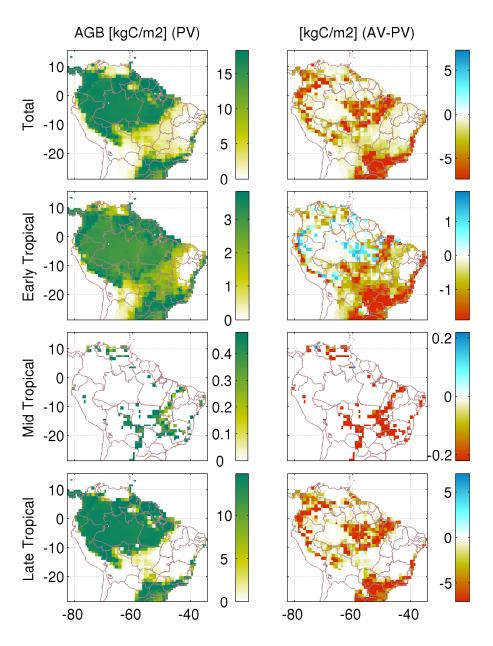


Fig. 02. Regional maps of total above ground biomass (AGB) $[kg\,m^{-2}]$ from the ED2 initial condition. The left column indicates results are from the *Potential* Vegetation condition, the right column is the relative differences between the *Actual* and *Potential* scenarios, $(AGB_{AV} - AGB_{PV})$. Each row represents the partitioning of the above ground biomass into respect plant functional types. "Early Tropical", "Mid Tropical" and "Late Tropical" refer to broad-leaf tropical evergreen plant functional types.

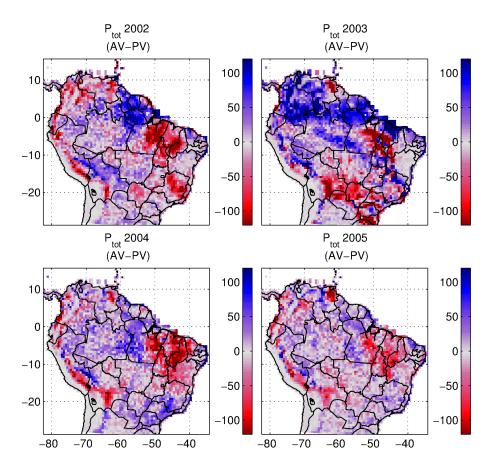


Fig. 03. Differences in total annual surface precipitation [mm], 2002-2005. The *Potential* Vegetation (PV) Condition is subtracted from the *Actual* Vegetation (AV) condition, (AV-PV).

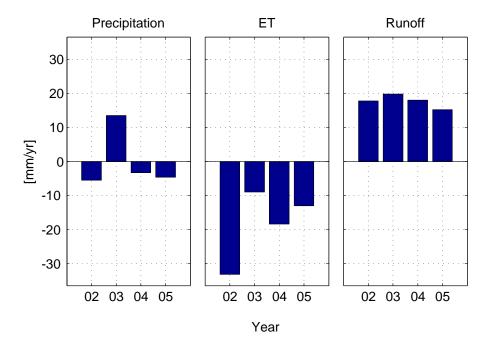


Fig. 04. Difference in mean continental precipitation, evaporation and total runoff, between the *Actual* Vegetation case and *Potential* Vegetation case (AV-PV).

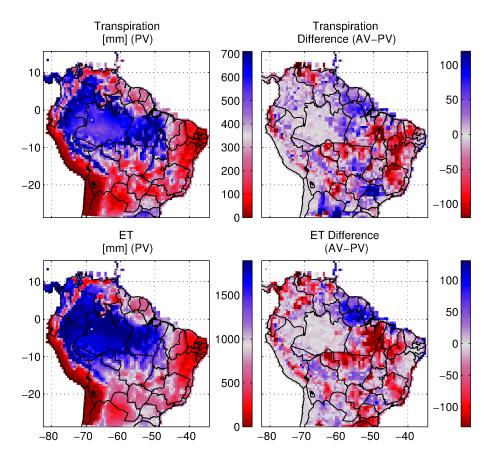


Fig. 05. Left Panels: Mean annual transpiration and evapotranspiration in the *Potential* vegetation (PV) scenario, from 2002-2005 [mm]. Right panels: difference in mean annual transpiration and evapotranspiration between the *Actual* Vegetation case and *Potential* Vegetation case (AV-PV).

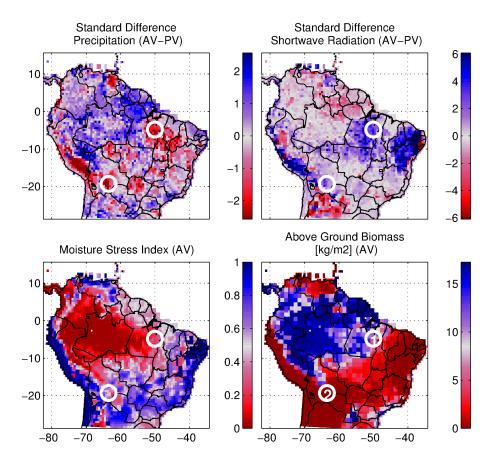


Fig. 06. Combined assessment of the regional significance in differences between precipitation and radiation, and the susceptibility of the ecosystems. Upper panels show standard scores for consistency of differences *Actual* and *Potential* (AV-PV) surface precipitation and surface down-welling shortwave radiation. The lower left panel shows the moisture stress index for the *Actual* (AV) scenario, see Equation (3). For reference, *Actual* (AV) scenario Above Ground Biomass is provide in the bottom right panel.

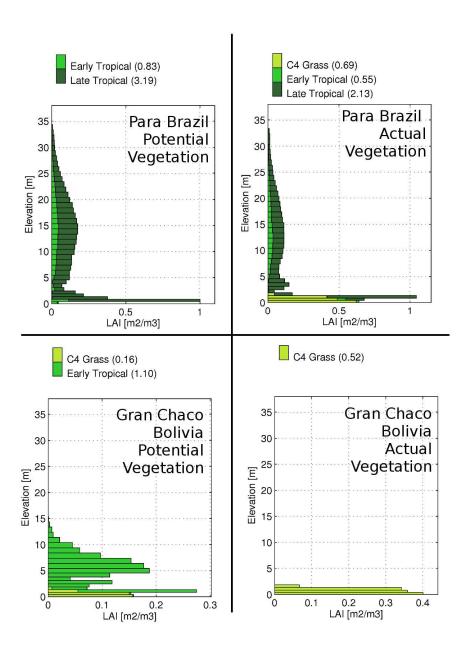


Fig. 07. Mean vertical leaf area index profiles $[m^2m^{-3}]$ estimated by the Ecosystem Demography Model 2 at the two focus areas. Vertically integrated leaf area index $[m^2m^{-2}]$ per each represented functional type of plant is shown in the key above each plot.

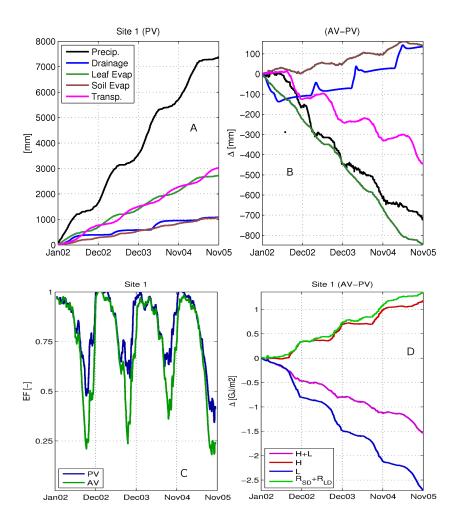


Fig. 08. Time series analysis of the surface water and energy balance at the Pará focus site, 2002-2005. Upper left panel (A): Accumulated water flux for the *Potential* scenario. Upper right panel (B): Accumulated differential water flux between *Actual* (AV) and *Potential* (PV) scenarios. Lower left panel (C): mean evaporative fraction, latent heat flux (L) divided by the sum of latent and sensible (H) heat flux (L/(L+H)). Lower right panel (D): accumulated differential energy flux in GigaJoules per square meter. R_{SD} is down-welling short-wave radiation incident on the surface, R_{LD} is down-welling long-wave radiation incident on the surface.

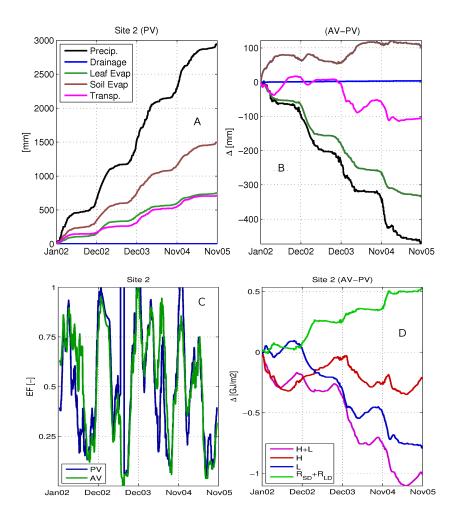


Fig. 09. Time series analysis of the surface water and energy balance at the Gran Chaco site, 2002-2005. Upper left panel (A): Accumulated water flux for the *Potential* scenario. Upper right panel (B): Accumulated differential water flux between *Actual* (AV) and *Potential* (PV) scenarios. Lower left panel (C): mean evaporative fraction, latent heat flux (L) divided by the sum of latent and sensible (H) heat flux (L/(L+H)). Lower right panel (D): accumulated differential energy flux in GigaJoules per square meter. R_{SD} is down-welling short-wave radiation incident on the surface, R_{LD} is down-welling long-wave radiation incident on the surface.

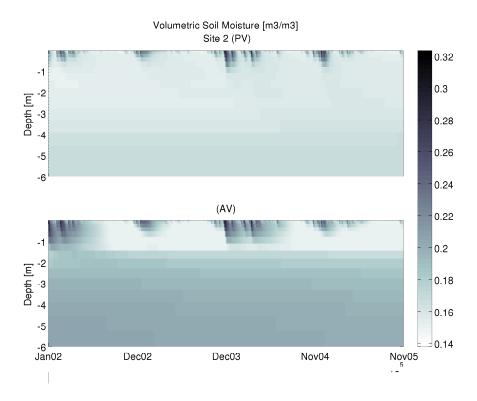


Fig. 010. Time series profile of volumetric soil water at the Gran Chaco focus area. Both scenarios, *Potential* (PV) and *Actual* (AV) are shown separately.

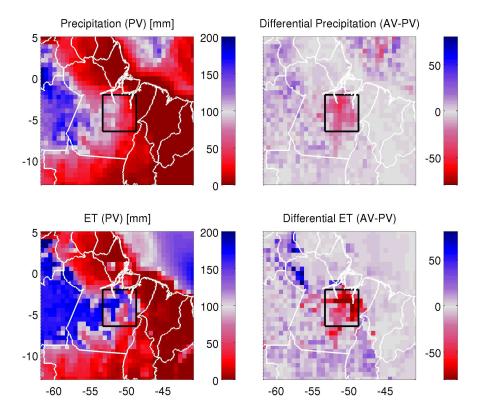


Fig. 011. Monthly integrated surface water fluxes over the Pará focus region, September 2003. Upper left panel: map of integrated monthly precipitation for the *Potential* vegetation simulation (PV). Upper right panel: map of the integrated difference in monthly precipitation, *Actual* vegetation case minus the *Potential* vegetation case (AV-PV). Lower left panel: map of integrated monthly evapotranspiration for the *Potential* vegetation simulation. Lower right panel: map of the integrated difference in monthly evapotranspiration, *Actual* vegetation case minus the *Potential* vegetation case.

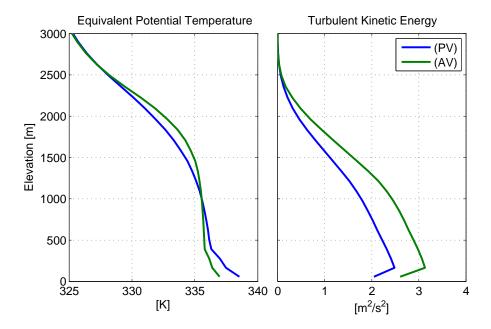


Fig. 012. Mean profiles of Equivalent Potential Temperature and Turbulent Kinetic Energy at 15Z within the bounded domain at Pará, September 2003.

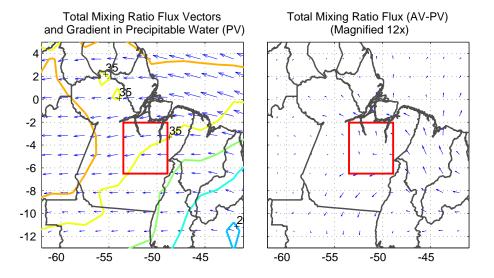


Fig. 013. Left Panel: Map of vertically integrated total water advective flux vectors (quivers) and vertically integrated precipitable water (contours) for the *Potential* vegetation case (PV), region near the Pará site, September 2003. Quivers are scaled and convey only directionality and relative magnitude. Contours of low precipitable water are shown by cool colors (blues) and high precipitable water with warm colors (reds). Right Panel: The differential in vertically integrated advection of total precipitable water, *Actual* vegetation minus *Potential* vegetation (AV-PV). Quivers are scaled to 12 times relative to the left panel. In both panels the sub-domain bounding the Pará focus region is shown with a red-box.

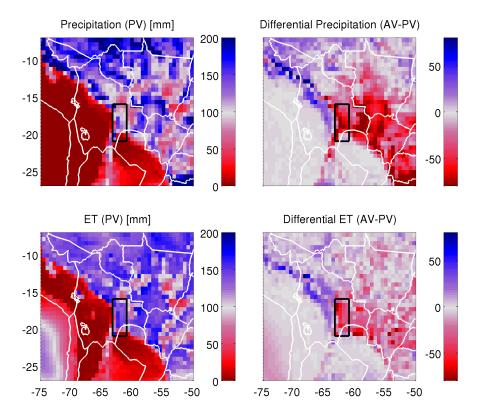


Fig. 014. Monthly integrated surface water fluxes over the Gran Chaco focus region, April 2003. Upper left panel: map of integrated monthly precipitation for the *Potential* vegetation simulation (PV). Upper right panel: map of the integrated difference in monthly precipitation, *Actual* vegetation case minus the *Potential* vegetation case (AV-PV). Lower left panel: map of integrated monthly evapotranspiration for the *Potential* vegetation simulation. Lower right panel: map of the integrated difference in monthly evapotranspiration, *Actual* vegetation case minus the *Potential* vegetation case.

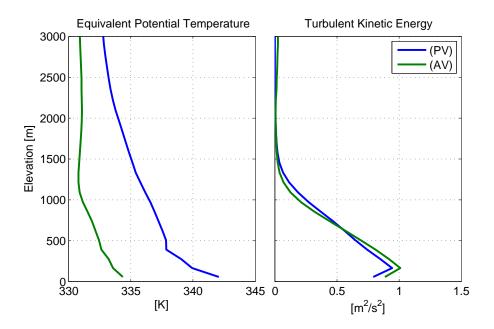


Fig. 015. Mean profiles of Equivalent Potential Temperature and Turbulent Kinetic Energy at 15Z within the bounded domain at Gran Chaco, April 2003.

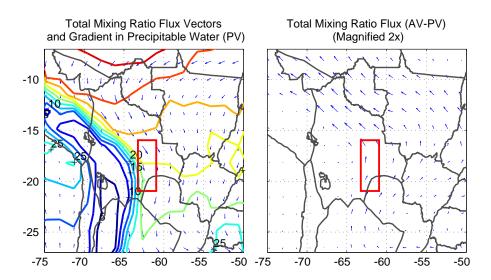


Fig. 016. Left Panel: Map of vertically integrated total water advective flux vectors (quivers) and vertically integrated precipitable water (contours) for the *Potential* vegetation case (PV), region near the Gran Chaco site, April 2003. Quivers are scaled and convey only directionality and relative magnitude. Contours of low precipitable water are shown by cool colors (blues) and high precipitable water with warm colors (reds). Right Panel: The differential in vertically integrated advection of total precipitable water, *Actual* vegetation minus *Potential* vegetation (AV-PV). Quivers are scaled to 2 times relative to the left panel. In both panels the sub-domain bounding the Gran Chaco focus region is shown with a red-box.

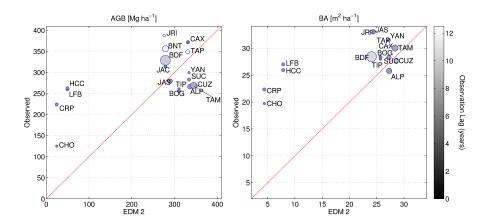


Fig. 017. Comparison of model estimated mean Above Ground Biomass (AGB) and Basal Area (BA) with measurements presented in (Baker et al., 2004a) and (Baker et al., 2004b). Circle size shows relative approximation of the number of census sites used in the field measurements reported in Baker (maximum=11 separate plots at BDF). Darker circles indicate that measurements were taken more recently and therefore have less time-lag in the comparison with the ED2 initial condition (January 2008). In accord with methods of (Baker et al., 2004a) and (Baker et al., 2004b), model estimates were filtered to include only primary forests and ignored vegetation less than 10 centimeters diameter. Coarse woody debris was excluded from comparison, only live stems were accounted for.

Table 01. Simulation constraints describing the spin-up process creating the initial boundary conditions. ¹ Modified DS314 data is derived from (Sheffield et al., 2006), precipitation down-scaling and radiation interpolation is applied, see footnote for data availability.

Specification	Value		
climate data	modified DS314 ¹		
soils data	(Quesada et al., 2011) + IGBP-DIS		
plant types	late succession tropical evergreens		
	mid succession tropical evergreens		
	early succession tropical evergreens		
	subtropical grasses		
	C4 grasses		
simulation period	508 years		
spatial resolution	gridded 1°		
bounding domain $30^{\circ}S-15^{\circ}N, 85^{\circ}W-30^{\circ}W$			
tree allometry			
(DBH,height)	(Chave et al., 2001; Baker et al., 2004b)		
(crown properties)	(Poorter et al., 2006; Dietze et al., 2008)		
turbulent transport	(Beljaars and Holtslag, 1991) atmospheric boundary		
	(Massman, 1997) within canopy		
photosynthesis	(Collatz et al., 1991)		
& leaf conductance	(Collatz et al., 1992)		
	(Leuning, 1995)		
canopy radiation scattering	(Zhao and Qualls, 2005, 2006)		
soil hydrology	(Walko et al., 2000; Tremback and Kessler, 1985)		
	(Medvigy et al., 2009)		

Table 02. Run time parameters and specifications in the ED2-BRAMS coupled simulations. ¹ LFC = Level of Free Convection.

Specification	Value
simulation period	January 2002 - December 2005
grid projection	polar stereographic
grid dimensions	98 (E-W), 86 (N-S), 56 (vertical)
horizontal grid resolution	64 km
vertical grid resolution	110 m (lowest) stretching to 1500 m at 7%
atmospheric time-step	30 s
atmospheric accoustic time-step	10 s
land-surface model time-step	120 s
method of calculating	level of maximum sum of mean
updraft base	and variance of vertical velocity
number of prototype cloud scales	2
mean radius of cloud 1	20,000 m
minimum depth of cloud 1	4000 m
mean radius of cloud 2	800 m
minimum depth of cloud 2	80 m
cumulus convective scheme	(Grell and Dévényi, 2002)
cumulus convective trigger	pressure differential between
	updraft base and $LFC^1 < 100$ hpa
cumulus dynamic control	(Kain and Fritsch, 1990; Kain, 2004)
condensate to precipitation	
conversion efficiency	3%
cloud # concentrations	(Medvigy et al., 2010)
and distribution parameters	
turbulent closure	(Nakanishi and Niino, 2006)
short-wave radiation scattering	(Harrington and Olsson, 2001)
long-wave radiation scattering	(Chen and Cotton, 1983)
advection	monotonic, (Walcek and Aleksic, 1998)
	& (Freitas et al., 2012)
cumulus feedback on radiation?	Yes

Table 03. Hydrologic monthly means within the bounded area above the Pará case study, September 2003. Total change in column precipitable water for the month per square meter ΔM_{pw} , evapotranspiration ET, precipitation P and resolved moisture convergence Mc, 55 meter air temperature T, mixing ration (55m) r, equivalent potential temperature θ_e , surface albedo to short-wave radiation α , down-welling short-wave radiation R_{SD} , down-welling long-wave radiation R_{LD} , up-welling long-wave radiation R_{LU} , net surface radiation R_{net} , sensible heat flux SHF and latent heat flux LHF.

Case	ΔM_{pw}	ET	P	Mc	T	r
Units	${\rm kgm^{-2}}$	${\rm kgm^{-2}}$	${\rm kgm^{-2}}$	${\rm kgm^{-2}}$	$^{\circ}\mathrm{C}$	$\rm gkg^{-1}$
AV	-3.457	63.1	29.8	-37.14	32.83	12.18
PV	-3.515	94.7	47.3	-51.32	32.35	12.93

Case	θ_e	α	R_{SD}	R_{LD}	R_{LU}	R_{net}	SHF	LHF
Units	K	-	$ m Wm^{-2}$					
AV	336.8	0.262	300.2	443.3	513.2	180.9	139.25	70.97
PV	338.2	0.257	285.6	443.9	498.0	187.8	114.50	106.45

Table 04. Hydrologic monthly means within the bounded area above the Gran Chaco case study, April 2003. Total change in column precipitable water for the month per square meter ΔM_{pw} , evapotranspiration ET, precipitation P and resolved moisture convergence Mc, 55 meter air temperature T, mixing ration (55m) r, equivalent potential temperature θ_e , surface albedo to short-wave radiation α , downwelling short-wave radiation R_{SD} , down-welling long-wave radiation R_{LD} , up-welling long-wave radiation R_{LU} , net surface radiation R_{net} , sensible heat flux SHF and latent heat flux LHF.

Case	ΔM_{pw}	ET	P	Mc	T	r
Units	$ m kgm^{-2}$	${\rm kgm}^{-2}$	${\rm kgm}^{-2}$	${\rm kgm}^{-2}$	$^{\circ}\mathrm{C}$	$\rm gkg^{-1}$
AV	-11.42	82.95	41.89	-52.49	25.98	12.73
PV	-11.02	111.89	85.91	-36.99	27.36	15.15

Case	θ_e	α	R_{SD}	R_{LD}	R_{LU}	R_{net}	SHF	LHF
Units	K	-	${ m Wm^{-2}}$					
AV	334.4	0.330	252.6	400.2	466.9	111.74	38.54	91.0
PV	342.0	0.297	218.7	424.9	462.6	130.2	28.15	122.5

Table 05. Atmospheric boundary conditions provided by BRAMS, that drive the ED2 model.

Symbol	Units	Description	
u_x	[m/s]	Zonal wind speed	
u_y	[m/s]	Meridional wind speed	
T_a	[K]	Air temperature	
q_a	[kg/kg]	Air specific humidity	
\dot{m}_{pcp}	[kg/s]	Precipitation mass rate	
z_{ref}	[m]	Height of the reference point	
R_{ld}	$[w/m^2]$	Downward longwave radiation	
R_{vb}	$[w/m^2]$	Downward shortwave radiation, visible beam	
R_{vd}	$[w/m^2]$	Downward shortwave radiation, visible diffuse	
R_{nb}	$[w/m^2]$	Downward shortwave radiation, near infrared beam	
R_{nd}	$[w/m^2]$	Downward shortwave radiation, near infrared diffuse	

Table 06. ED2 flux variables providing the lower boundary condition for the BRAMS atmospheric model.

Symbol	Units	Description
$\overline{(u'w')}$	$[m^2/s^2]$	Average vertical flux of horizontal wind velocity perturbations
$\overline{(w'w')}$	$[m^2/s^2]$	Average vertical flux of vertical wind velocity perturbations
$\overline{(t'w')}$	$[mK/s^2]$	Average vertical flux of temperature perturbations
$\overline{(q'w')}$	$[kg/m^2/s^2]$	Average vertical flux of moisture perturbations
$\overline{(c'w')}$	$[\mu mol/m^2/s^2]$	Average vertical flux of carbon perturbations
χ_s	[-]	Average total shortwave albedo
χ_l	[-]	Average total longwave albedo
R_{lu}	$[w/m^2]$	Average up-welling longwave radiation

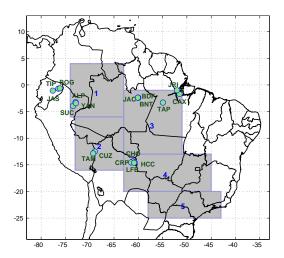


Fig. 018. Locations of zones and sites of analysis. Zones are numbered 1 through 5, and reflect geographic areas where model and observation spatial means are compared for validation (see Appendix). Forest census plot sites from (Baker et al., 2004a) and (Baker et al., 2004b) are referenced with green markers and their station code. Station codes designate the following site names: Allpahuayo (ALP), BDFFP (BDF), Bionte (BNT), Bogi (BOG), Caxiuana (CAX), Chore (CHO), Cerro Pelao (CRP), Cuzco Amazonico (CUZ), Huanchaca Dos (HCC), Jacaranda (JAC), Jatun Sacha (JAS), Jari (JRI), Los Fierros Bosque (LFB), Sucusari (SUC), Tambopata (TAM), Tapajos (TAP), Tiputini (TIP) and Yanamono (YAM).

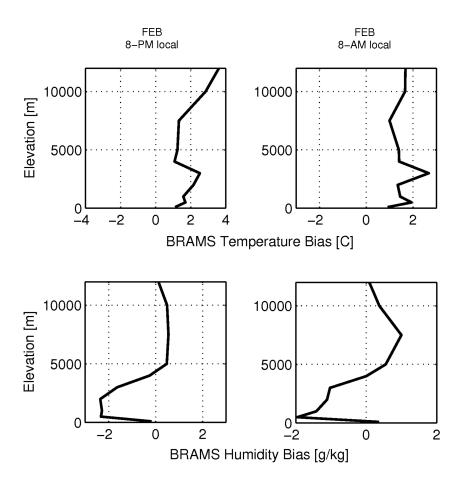


Fig. 019. Comparison of model estimates with radiosonde data, differences in mean air temperature and specific humidity. Manaus, February 2003.

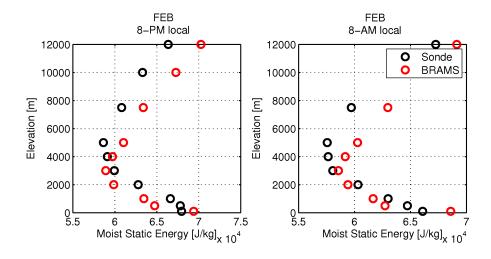


Fig. 020. Comparison of model estimated mean moist static energy with rawinsonde measurements. Manaus, February 2003.

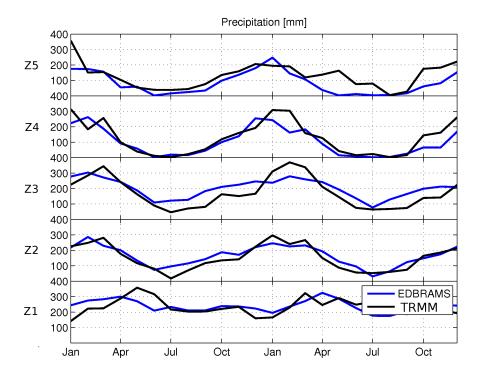


Fig. 021. Mean monthly precipitation from ED2-BRAMS and the Tropical Rainfall Measurement Mission (TRMM) 3B43 product, years 2002-2003. Spatial means are taken within zones according to Figure 018.

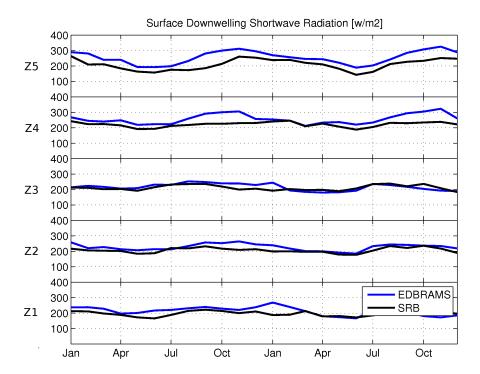


Fig. 022. Mean monthly surface radiation from ED2-BRAMS and the Surface Radiation Budget (SRB) product version 2.5, years 2002-2003. Spatial means are taken within zones according to Figure 018.

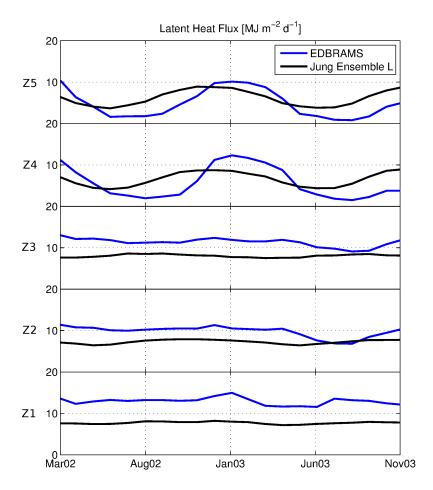


Fig. 023. Mean monthly surface to atmosphere latent heat flux from ED2-BRAMS and the synthesis product from (Jung et al., 2011, 2009). Spatial means are calculated over the zones shown in Figure 018.

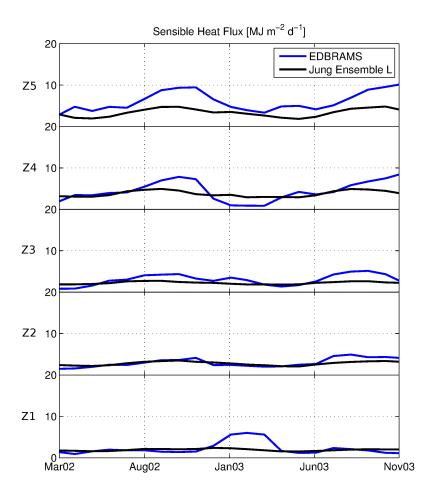


Fig. 024. Mean monthly surface to atmosphere sensible heat flux from ED2-BRAMS and the synthesis product from (Jung et al., 2011, 2009). Spatial means are calculated over the zones shown in Figure 018.

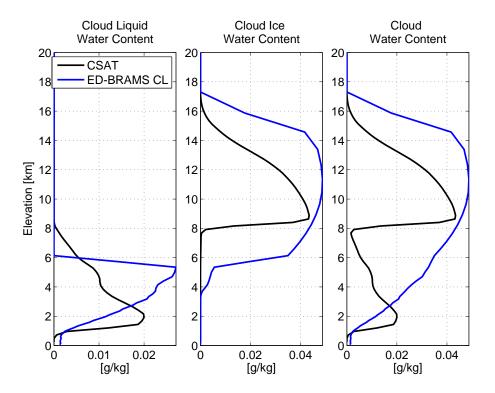


Fig. 025. CloudSat climatological water content profiles and model estimated water content profiles for February 2003, 17 UTC, $3^oN - 12^oS$ and $70^oW - 55^oW$.