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## RESEARCH ARTICLE

# A regional view of the linkages between hydro-climatic changes and deforestation in the Southern Amazon

Sly Wongchuig<sup>1,2</sup>  | Jhan Carlo Espinoza<sup>1</sup>  | Thomas Condom<sup>1</sup>  |  
Hans Segura<sup>1</sup>  | Josyane Ronchail<sup>3</sup>  | Paola A. Arias<sup>4</sup>  |  
Clementine Junquas<sup>1</sup>  | Antoine Rabatel<sup>1</sup>  | Thierry Lebel<sup>1</sup> 

<sup>1</sup>Univ. Grenoble Alpes, IRD, CNRS, Grenoble INP, Institut des Géosciences de l'Environnement (IGE, UMR 5001), Grenoble, France

<sup>2</sup>Instituto de Pesquisas Hidráulicas IPH, Universidade Federal do Rio Grande do Sul UFRGS, Porto Alegre, Brazil

<sup>3</sup>Univ. Paris Diderot and Université de Paris, UMR LOCEAN (Sorbonne Universités-UPMC, CNRS, IRD, MNHN), Paris, France

<sup>4</sup>Grupo de Ingeniería y Gestión Ambiental (GIGA), Escuela Ambiental, Facultad de Ingeniería, Universidad de Antioquia, Medellín, Colombia

## Correspondence

Sly Wongchuig, Univ. Grenoble Alpes, IRD, CNRS, Grenoble INP, Institut des Géosciences de l'Environnement (IGE, UMR 5001), 38000 Grenoble, France.  
Email: sly.wongchuig-correa@univ-grenoble-alpes.fr

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## Abstract

In the last four decades, the Southern Amazon (south of 8°S) has shown changes in the spatial and temporal patterns of its hydro-climatic components, leading to drier conditions. Due to climate and land-use changes, this region is considered as a zone under biophysical transition processes. Previous studies have documented a complex interaction between climate and deforestation either on a large-scale or based on limited in situ data, typically covering the Brazilian Amazon. In this study, we analyse the relationships between hydro-climate, the surface water-energy partitioning and an index of regional forest cover change for the period 1981–2018. Additionally, we discretized three regions covering the Bolivian Amazon and the southern portions of the Peruvian and Brazilian Amazon due to their differences in the evolution of land use. In the Bolivian region, a high ratio of forest cover change, exceeding 40–50%, is related to a significant tendency to become water-limited. This change is associated with decreased rainfall, increased potential evapotranspiration and decreased actual evapotranspiration. Regardless of the region analysed, those that are characterized by a high ratio of forest cover change (>40–50%) show growing imbalance between increasing potential and decreasing actual evapotranspiration. However, in the Peruvian and Brazilian regions, hydro-climatic conditions remain energy-limited due to minor rainfall changes. The observed differences in surface water-energy partitioning behaviour evidence a complex dependence of both sub-regional (i.e., land cover changes) and large-scale (i.e., strengthening of the Walker and Hadley circulations) conditions. Our findings indicate a clear link between hydro-climatic changes and deforestation, providing a new perspective on their spatial variability on a sub-regional scale.

## KEYWORDS

forest loss, hydro-climatic variability, water-limited conditions

## 1 | INTRODUCTION

In tropical South America and particularly in the Amazon forest, which spans nine countries and more than 6 millions of square kilometres, climate and vegetation strongly interact, exchanging energy, moisture and momentum in a dynamic equilibrium (Salati and Vose, 1984; Nobre *et al.*, 1991; Salati and Nobre, 1991; Zeng and Neelin, 1999; Zemp *et al.*, 2017). While ongoing environmental changes (including climate change, deforestation and wildfires) are profoundly affecting the biodiversity of this unique ecosystem, a large debate remains about its possible tipping point into a new state with conditions similar to those of a savanna (Nobre and Borma, 2009; Hirota *et al.*, 2011; Davidson *et al.*, 2012; Nobre *et al.*, 2016). Based on model estimates, two thresholds leading to such a tipping have been identified, namely a global warming of 3–4°C or deforestation exceeding 40% (Lenton *et al.*, 2008; Nobre and Borma, 2009; Salazar and Nobre, 2010; Nobre *et al.*, 2016; Aragão *et al.*, 2018; Lovejoy and Nobre, 2018; Brando *et al.*, 2020). However, these results are based on large-scale approaches and hypothetical deforestation scenarios (Sampaio *et al.*, 2007; Lawrence and Vandecar, 2015), while the Amazon forest is a complex eco-climatic system where interactions between climate and vegetation occur over a wide range of scales (Cysneiros *et al.*, 2018). For instance, hydro-climatic features over the Amazonia are characterized by a complex spatio-temporal variability related to the large-scale atmospheric circulation (Figueroa and Nobre, 1990; Espinoza *et al.*, 2009a), interactions with the topography (Killeen *et al.*, 2007; Espinoza *et al.*, 2015) and interactions between land surface processes and the atmosphere, including moisture recycling by the forest (Zemp *et al.*, 2017; Staal *et al.*, 2018). Several authors have also found evidence that the hydrological seasonal cycle has changed in the past decades and that even these changes are different according to the regions, with more rainfall and discharge in the north but a longer dry season in the south (Espinoza *et al.*, 2009b; 2019a; García-García and Ummenhofer, 2015; Barichivich *et al.*, 2018; Carvalho Oliveira *et al.*, 2018; Liang *et al.*, 2020). Therefore, in order to better apprehend this complexity, it is necessary to conduct sub-regional scale studies that complement large-scale research and take into account natural and anthropogenic dynamics that may vary significantly from one area to another as well as across scales. Most notably, one factor that has received relatively little attention at sub-regional scale and in a spatially distributed manner is the modification – as the result of either climate change or vegetation changes (natural and anthropic) or both – of some key hydro-climatic variables that control the interactions between climate and vegetation dynamics.

In this respect, the Southern Amazon (south to 8°S) constitutes a peculiar region due to specific climatic characteristics such as its strong rainfall seasonality and the predominance of convective processes partly related to tropical-extratropical interactions (a brief section in the Supporting Information describes the different characteristics of the Southern Amazon with respect to the seasonal cycle of rainfall and the winds circulation, see Figure S1; see also Marengo, 1995; 2004; Espinoza *et al.*, 2009b; Fu *et al.*, 2013; Yin *et al.*, 2014). In addition, vegetation in the Southern Amazon is particularly sensitive to changes in the length of the dry season (Aragão *et al.*, 2008; Costa and Pires, 2010; Gutierrez-Cori *et al.*, 2021), and it has been widely impacted by anthropogenic forcing, especially along a curve that hugs the south-eastern edge of the forest that scientists and resource managers have come to call the ‘Arc of Deforestation’ (Costa and Pires, 2010; Davidson *et al.*, 2012; Brando *et al.*, 2014; Spracklen and Garcia-Carreras, 2015; Dias *et al.*, 2016; Alves *et al.*, 2017; Damien *et al.*, 2017; Le Page *et al.*, 2017). Large evidences support the coupling between forest lost in the Southern Amazonia and changes in different components of the hydrological cycle, such as evapotranspiration, rainfall amount (Sud *et al.*, 1996; Sampaio *et al.*, 2007; Bathiany *et al.*, 2010; Spracklen *et al.*, 2012; Lawrence and Vandecar, 2015; Leite-Filho *et al.*, 2021), duration of the dry-season and among others (Marengo *et al.*, 2011; Fu *et al.*, 2013; Arias *et al.*, 2015; Debortoli *et al.*, 2015; 2017; Khanna *et al.*, 2017; Wright *et al.*, 2017; Aragão *et al.*, 2018; Agudelo *et al.*, 2019; Leite-Filho *et al.*, 2019; 2021; Ruiz-Vásquez *et al.*, 2020). However, most of these studies are based either on limited in situ data covering only the Brazilian Amazon (Fu *et al.*, 2013; Debortoli *et al.*, 2015; 2017; Leite-Filho *et al.*, 2019) or on regional models of coarse resolution (>50 km) (Alves *et al.*, 2017; Ruiz-Vásquez *et al.*, 2020). Given the contrasting hydro-climatic changes that have recently been reported in the Southern Amazon due to regional (Marengo, 2004; Espinoza *et al.*, 2013; 2019a; Molina-Carpio *et al.*, 2017) and global processes such as anthropogenic warming (Song *et al.*, 2018), looking at the spatial patterns and temporal dynamics of key hydro-climatic variables in relation to land cover changes is a matter of particular interest.

This study aims to tackle this issue through two angles. The first consists in looking at the co-evolution of the annual precipitation ( $P$ ) and evapotranspiration (ET, potential and actual) with the surrounding percentage of non-forest vegetation – the idea being to diagnose whether areas of high non-forest vegetation display any significant evolution of  $P$  or ET – and, conversely, whether significant evolutions of  $P$  or ET may have occurred in areas of little non-forest vegetation. The non-forest vegetation is considered

here as a proxy for forest loss (see Section 2). Secondly, we investigate whether the decadal co-evolution (for a total of 38 years) of the proportion of non-forest vegetation areas has some implications for the water/energy partitioning by looking at changes in the components of Budyko's theoretical framework.

The procedures used to compute all these variables are detailed in Section 2 and a detailed discussion regarding the validity of these data sets for the Southern Amazon is provided in the Supporting Information.

## 2 | MATERIALS AND METHODS

### 2.1 | Study region and periods

We performed analyses for the Southern Amazon Basin (south to 8°S). Over 1992–2018, the Southern Amazon presents an average of ~14% of deforestation compared to a lower value (~3%) for the Northern Amazon (Figure 1b). These averages were computed for areas below 1,000 m above sea level (m.a.s.l.) and values between Southern and Northern Amazon are statistically significantly different at a confidence level of 99%, which provides an additional reason for focusing our study on the Southern Amazon. This large region is itself divided into three regions of analysis, delimited on the basis of their geopolitical, climate and deforestation characteristics (Eva *et al.*, 2005; Espinoza *et al.*, 2009a; 2019b; Hansen *et al.*, 2013). As shown in Figure 1a, the first region analysed belongs to Bolivian Amazon (hereafter called Bolivian region), the second region belongs to Peruvian Amazon (hereafter called Peruvian region) and finally the third and largest region belongs to Brazilian Amazon (hereafter called Brazilian region). We have taken into account the geopolitical division of these countries, as they have different policies on soil conservation management, so a study at the regional level will also help to generate discussions at this level. Additionally, the areas corresponding to altitudes above 1,000 m. a.s.l. were disregarded from our analyses to avoid the massive topography that affects convection, producing the highest rainfall amounts over the Amazon Basin in the so-called “rainfall hotspots” region (Eva *et al.*, 2005; Espinoza *et al.*, 2015; Chavez and Takahashi, 2017). Our study covers the period 1981–2018 for the hydro-climatic variables and 1992–2018 for the land cover maps. Although the temporal coverage of these data sets is not the same, analyses for years prior to 1992 (non-common period) can still be performed because the land cover maps also contain information on the cumulative changes of past years, that is, before 1992 (see Section 2.3).

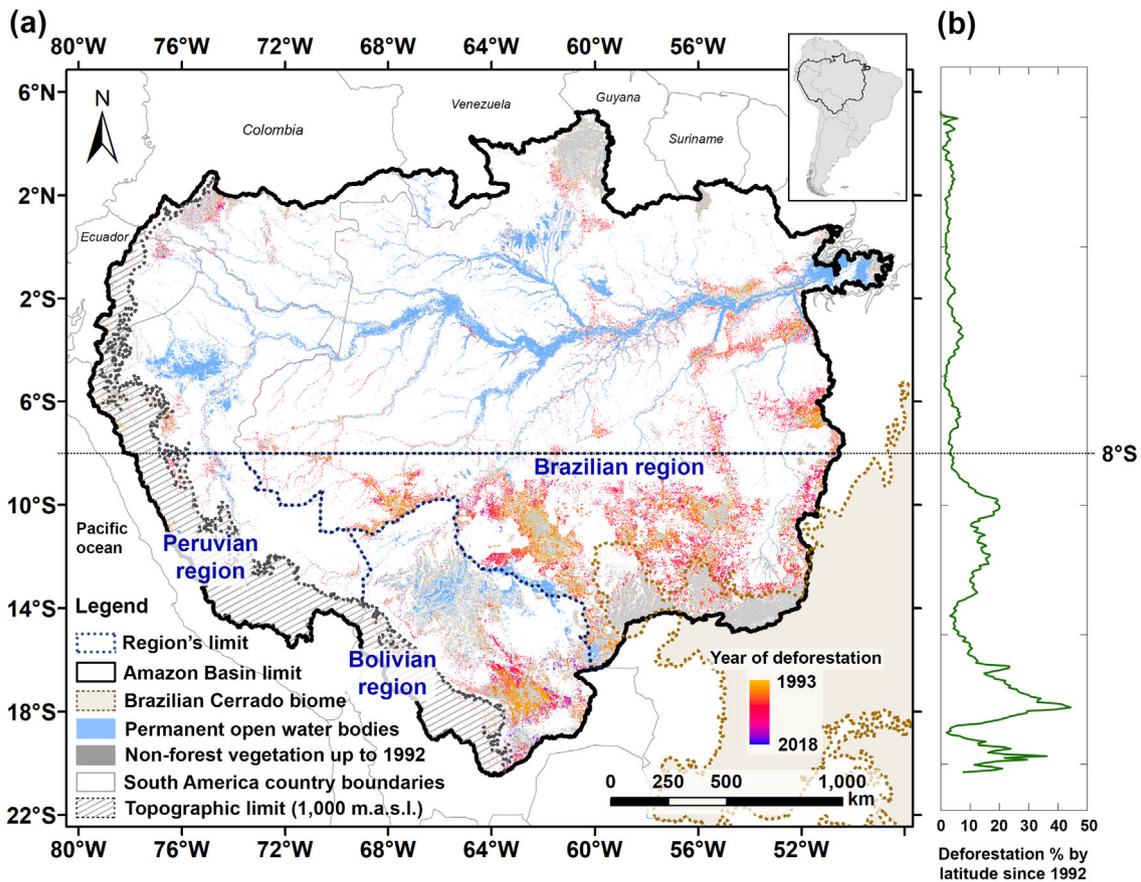
### 2.2 | Hydrometeorological data sets

For the hydro-climate analysis of  $P$  and  $ET$ , we use CHIRPS and GLEAM data sets, respectively. These products have shown an adequate representation of long-term and interannual variability in previous studies in South America (Beck *et al.*, 2017; Wongchuig *et al.*, 2017; Casagrande *et al.*, 2018; da Motta Paca *et al.*, 2019; Moreira *et al.*, 2019; Pan *et al.*, 2020; Espinoza *et al.*, 2019b; Haghtalab *et al.*, 2020).

The evolution of  $P$  over 1981–2018 is characterized in terms of amounts (annual totals) and occurrence – as measured by the dry day frequency (DDF) and the wet day frequency (WDF), based on the precipitation data set CHIRPS at 5 km of spatial resolution (Funk *et al.*, 2015). The evolution of  $ET$  is assessed by looking at both the potential and actual evapotranspiration (PET and AET, respectively). In order to consider the uncertainty of temporal trend of PET and AET, six land surface, global hydrological models and global model and satellite-based data set were analysed. These models use different configurations for the calculation of PET and AET (see Supporting Information). The data set is from ORCHIDEE, PCR-GLOBWB, W3RA, WaterGAP3, HBV-SIMREG and GLEAM. Since most of the temporal trend results of these global models coincide with the results of GLEAM-v3 (Martens *et al.*, 2017), the analyses in the main document are performed using only GLEAM due to: (a) has a longer period (1981–2018 vs. 1972–2012), (b) coincides with the period of the precipitation data set and (c) has a higher spatial resolution (25 km vs. 50 km). A more detailed discussion on the validation of these products is shown in the Supporting Information.

#### 2.2.1 | Climate Hazards Group InfraRed Precipitation with Station

We used the Climate Hazards Group InfraRed Precipitation with Station data (CHIRPS) that was developed by the United States Geological Survey (USGS) and the Earth Resources Observation and Science (EROS) (Funk *et al.*, 2015). This data set is based on merged precipitation models by interpolation of rainfall in situ data. The version 2 (CHIRPS v2) has a spatial resolution of 5 km and showed an adequate performance for the Amazon Basin when evaluated against ground-based rainfall stations and in situ discharge observations to represent interannual and multi-decadal variability and interannual trends (Wongchuig *et al.*, 2017; Haghtalab *et al.*, 2020; Fassoni-Andrade *et al.*, 2021).



**FIGURE 1** (a) Map of non-forest vegetation up to 1992 (grey) and deforested areas (yellow-red-blue bar colours) for the 1992–2018 period in the Amazon Basin. (b) Percentage of deforestation in relation to latitude and for regions below 1,000 m.a.s.l. [Colour figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]

## 2.2.2 | Global Land Evaporation Amsterdam Model

We used the Global Land Evaporation Amsterdam Model (GLEAM) (Martens *et al.*, 2017), a monthly global evaporation dataset which is based on the Priestley-Taylor equation. This data set-based model was designed to be driven by remote sensing observations only. These observations are primarily derived from microwave sensors, including soil moisture and vegetation optical depth used in GLEAM to constrain the potential evaporation rates. Four types of land covers are possible in GLEAMv3: bare soil, low vegetation, tall vegetation and open water. The vegetation cover is produced using dynamic land cover information based on the MEaSUREs Vegetation Continuous Fields dataset at yearly intervals from 1982 to 2016 itself based on sensors from NOAA-7, NOAA-9, NOAA-11, NOAA-14 and Terra. The variations in vegetation status is considered in the stress-factor (used to convert PET to AET) which is based on remote sensed Vegetation Optical Index from CCI-LPRM (Liu *et al.*, 2013) and soil

moisture both from remote sensing (ESA CCI SM v2.3; Liu *et al.*, 2012) and land-reanalysis (GLDAS Noah; Rodell *et al.*, 2004) products assimilated into the GLEAM multi-layer soil model. The present paper evaluates the last version (v3.3a) of the product, which updates most components and has a spatial resolution of 25 km.

## 2.2.3 | earth2Observe

The outputs from the earth2Observe first water resources reanalysis (WRR) multi-model ensemble, which is a global inter-comparison project produced by different institutions for the period 1979–2012 at 0.5° spatial resolution, were used in this research. We used the first version of the WRR based on the current modelling systems of each institution forced with a state-of-the-art atmospheric reanalysis. We only evaluated those models that have PET and AET available, these are ORCHIDEE, PCR-GLOBWB, W3RA, WaterGAP3 and HBV-SIMREG (see Supporting Information).

## 2.3 | Deforestation and non-forest vegetation buffer approach

Estimates of changes in forest cover were based on the methodology used by (Hansen *et al.*, 2013), where the forest loss is defined as the stand-replacement disturbance of tree cover canopy (complete removal of tree cover canopy at each pixel). Therefore, the forest loss value is assigned for each pixel relative to the first year of the land cover map. To make these estimates, we use the yearly ESA CCI Land Use maps, which cover the period 1992–2018 and were produced by the European Space Agency (ESA) Climate Change Initiative (CCI) (ESA, 2017; Lamarche *et al.*, 2017).

We used the already provided data set classification of each pixel to determine whether it corresponds to forest or non-forest cover annually, and which is accumulated as deforestation has occurred until the desired analysis period. The ratio of the regional deforested area is computed based on the concept of buffer zones, similar to (Debortoli *et al.*, 2015). The buffer zones approach corresponds to the surrounding neighbourhood around the 300 m pixels of the ESA CCI Land Use map given by a circular area from each hydro-climatic pixel centroid. Several radii of influence from 25 to 100 km were assessed in order to find a representative one at regional scale to finally choose the radius of 50 km for this study (Figure 2a). Non-forested areas (a temporal evolution concept of vegetation cover at time  $t$ ) is quantified as a ratio referred as ratio of non-forest vegetation (RNF), therefore the RNF considers the area within the buffer zone (Figure 2b) and refers to the non-forest vegetation cover until each period. The criteria to compute it was considering two periods, before and after 1992, which is when we started to have observations (hereafter called ESA-CCI LC). For periods before 1992, non-forest vegetation cover was considered as grass, shrubs, sparse vegetation and crops (low vegetation types) represented here in the 1992 map. These areas are represented as grey pixels in Figure 1a. Besides, for periods after 1992, changes from forest to these other low vegetation types were considered (Figure 2c), which are represented as pixels in the orange, red and violet colour palette in Figure 1a. These areas do not overlap with the non-forest vegetation cover before 1992 because, by definition, these areas should have been covered by forest in 1992.

In addition, the RNF is divided in classes of 10% for this investigation (see Section 3.1). Here we highlight that for the analyses made in this manuscript, the 40% threshold is used for defining low- or high-RNF.

## 2.4 | DDF and WDF

The annual DDF is defined as the frequency of days with daily rainfall ( $P$ ) of less than 1 mm, while the annual WDF is defined as the frequency of day with  $P$  greater than 10 mm (Espinoza *et al.*, 2016; 2019a).

## 2.5 | Kendall trend test

To compute temporal trends of hydro-climatic variables over the last four decades, we assess the rank-based non-parametric Kendall test (Kendall, 1975). We use the Kendall's  $\tau$ , which is based on counting the number of  $(i, j)$  pairs, for  $i < j$ , that are concordant that is, for which  $X_{a,i} - X_{a,j}$  and  $Y_{b,i} - Y_{b,j}$  have the same sign. The equation for Kendall's  $\tau$  includes an adjustment for ties in the normalizing constant and is often referred to as  $\tau$ -b.

For column  $X_a$  in matrix  $X$  and column  $Y_b$  in matrix  $Y$ , Kendall's  $\tau$  coefficient is defined as:

$$\tau = \frac{2K}{n(n-1)}$$

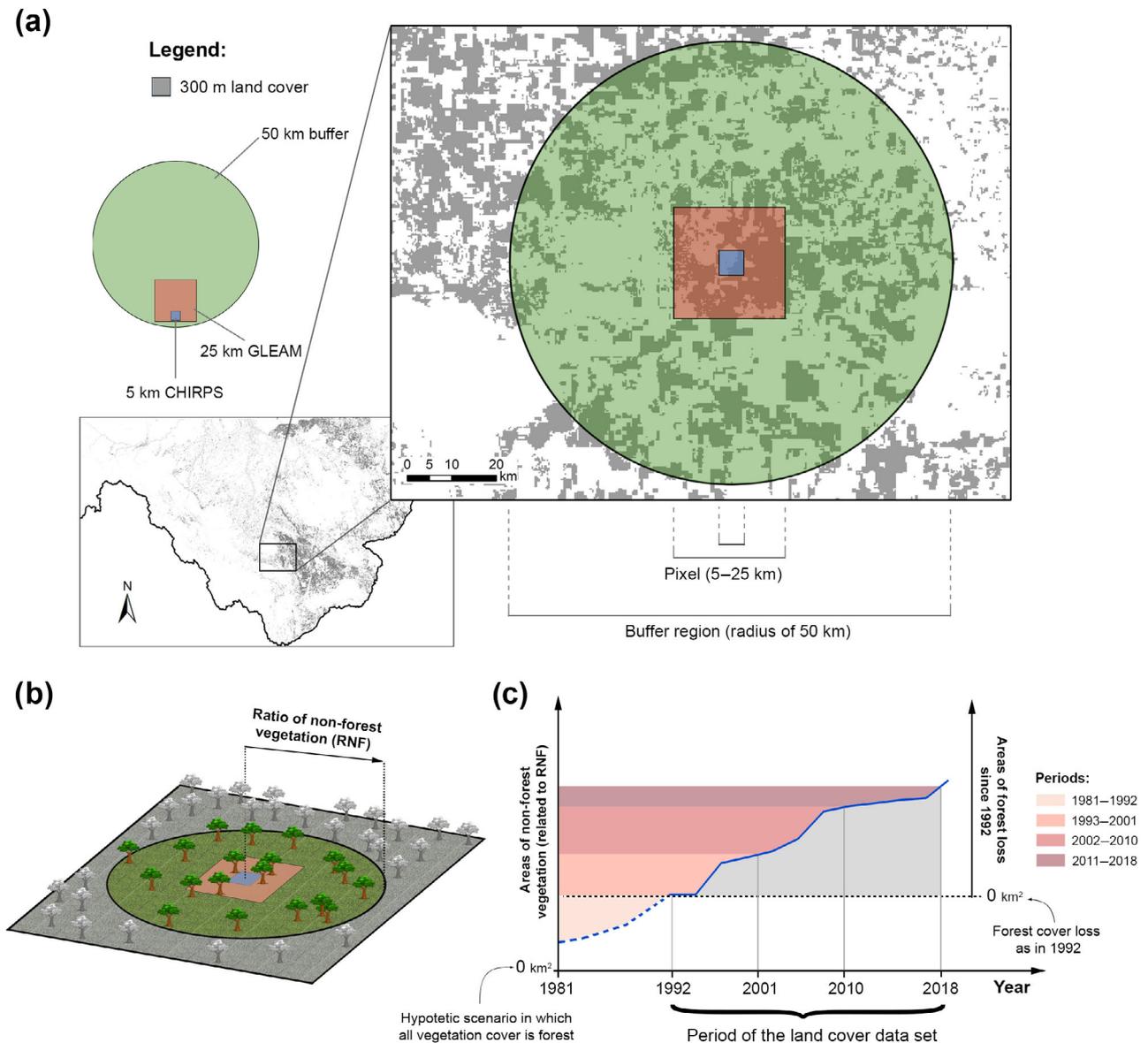
$$K = \sum_{i=1}^{n-1} \sum_{j=i+1}^n F(X_{a,i}, X_{a,j}, Y_{b,i}, Y_{b,j})$$

$$F(X_{a,i}, X_{a,j}, Y_{b,i}, Y_{b,j}) = \begin{cases} 1, & (X_{a,i} - X_{a,j})(Y_{b,i} - Y_{b,j}) > 0 \\ 0, & (X_{a,i} - X_{a,j})(Y_{b,i} - Y_{b,j}) = 0 \\ -1, & (X_{a,i} - X_{a,j})(Y_{b,i} - Y_{b,j}) < 0 \end{cases}$$

Values of the correlation coefficient can range from  $-1$  to  $+1$ . A value of  $-1$  indicates that one column ranking is the reverse of the other, while a value of  $+1$  indicates that the two rankings are the same. A value of  $0$  indicates no relationship between the columns. For this study, the significance of statistical tests is evaluated at 95% ( $p < .05$ ).

## 2.6 | Surface water and energy partitioning

Following the buffer-based approach (Debortoli *et al.*, 2015) of land use changes and their relations with the climate, the Budyko-like theoretical framework was applied in this study. The closure in this pixel-based approach of surface water and energy partitioning is not constrained by the watershed limits like usual applications; as a consequence, that is, AET can be greater than



**FIGURE 2** (a) Illustration of the applied methodology showing an example of the buffer zone of 50 km radio surrounding a single hydro-climatic pixel of rainfall (CHIRPS-5km) and actual-potential evapotranspiration (GLEAM-25km). The smaller pixels in grey represent the land cover change at 300 of spatial resolution. (b) Illustration of the covered area (in colours) for the ratio of non-forest vegetation (RNF). (c) Schematic representation of temporal area considerations for vegetation non-forest and forest loss since 1992. The blue line represents conceptually the value of land cover change. The periods up to which regional non-forest vegetation were set, are in red scale colours [Colour figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]

$P$  at a pixel scale, which has been aggregated in this analysis to 25 km, which is the coarsest resolution between the GLEAM and CHIRPS data.

According to Budyko's schematic diagram, the vertical water budget is represented in  $y$  axis [evaporative index (EI)] and its changes represent the influence of climate, land use and land cover changes. Besides, drier or warmer conditions are shown in  $x$  axis [dryness index (DI)] and their changes are in principle only influenced by climatology (Wang *et al.*, 2015; Wang and Zhou, 2016).

As a summary of the methodology,  $P$ , WDF and DDF variables corresponding to the CHIRPS data were analysed at  $\sim 5$  km and their corresponding RNF within the 50 km radius buffer. For the PET and AET variables corresponding to the GLEAM data, these were analysed at  $\sim 25$  km and their corresponding 50 km radius buffer. The RNF analysis for this first part of this research considers the entire period of non-forest vegetation up to 2018. This means, we calculate the ratio of areas that were already grass, shrub, sparse vegetation and crops

until 1992 plus the areas that changed from forest in 1992 to these low vegetation covers until 2018.

Besides, for the analysis of the decadal evolution (the total period 1981–2018, divided into four periods) of surface water and energy partitioning, the hydro-climatic variables were aggregated at the coarsest resolution, that is, at 25 km. That is, the DI and the EI were analysed at 25 km with their corresponding buffer of radius 50 km and whose RNF values vary with each decade analysed. This means, for the first period (1981–1992) we assumed the representative RNF as the ratio of areas that were already grass, shrub, sparse vegetation and crops until 1992. For the subsequent periods, the RNF has been accumulated as deforestation has occurred until the end of each decade.

### 3 | RESULTS

#### 3.1 | Regional hydro-climatic trends and associated deforestation ratio

In this subsection, the results are displayed with different panels in Figures 3 and 4: panels 3a, 4a and 4c show the spatial distribution of the Kendall's  $\tau$  value of  $P$ , PET and AET, respectively for the statistically significant pixels. The palette colours indicate that when the trends are on the blue scale, there is a moisture addition in the water balance equation, on the contrary on the red scale is that there is an outflow of moisture from the water balance. The remaining panels show the scatter plot of the Kendall's  $\tau$  values of each pixel corresponding to each region associated with the RNF in their respective 50 km buffer. The boxplots corresponding to each 10% RNF interval have been superimposed.

##### 3.1.1 | Precipitation amount, wet and dry day frequencies

For the 1981–2018 period, the Bolivian region is characterized by a significant decrease in  $P$  in most of its southern area (Figures 3a and S6a), whose Kendall's  $\tau$  values are mainly in the range of  $-0.6$  to  $-0.2$  and which is consistent with a reduction in river flows as documented in previous studies (Molina-Carpio *et al.*, 2017; Wongchuig *et al.*, 2017; Espinoza *et al.*, 2019b). This significant reduction in  $P$  is systematically observed in areas with RNF values above 40–50% (Figure 3b,c, left column), meaning that areas with a RNF of at least 40–50% are always associated with a reduction in  $P$ , which translates into a systematic diminution in WDF and increase in DDF (Figure 3c,d, left column). Additionally, in the northern part of the Bolivian region, near the border with Brazil,

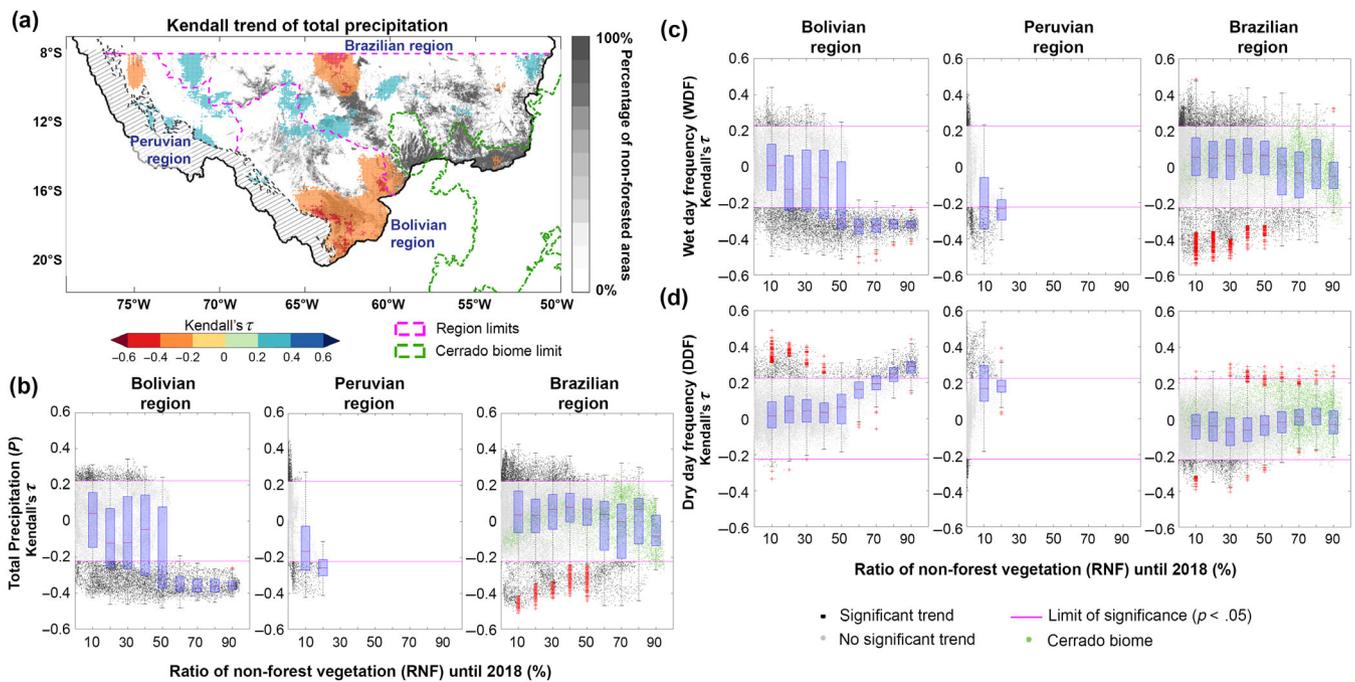
there are regions with a significant increase in  $P$  but only associated with low RNF values.

On the other hand, regions of the Peruvian and Brazilian Amazon show zones with significant decrease or increase in rainfall indices ( $P$ , WDF and DDF) with no systematic relationship with RNF values, contrary to what is observed in the Bolivian region. A marked decrease in  $P$  and WDF is observed in the north-central part of the Brazilian region (Figure 3a), as already reported in previous studies using rainfall stations (Debortoli *et al.*, 2017; Leite-Filho *et al.*, 2019), but not necessarily associated with high values of RNF. On the opposite, the occidental margins of areas with high RNF values, in the Brazilian region in Rondônia state (see Figure 7 for political divisions of states and cities) and close to the Bolivian region, exhibit significant increase in  $P$ . This could be explained by enhanced convection and an intensified atmospheric moisture transport from the nearby still forested regions, that could be reinforced by the elongated form of this region, as suggested by previous studies (Lawrence and Vandecar, 2015; Khanna *et al.*, 2017; Leite-Filho *et al.*, 2021).

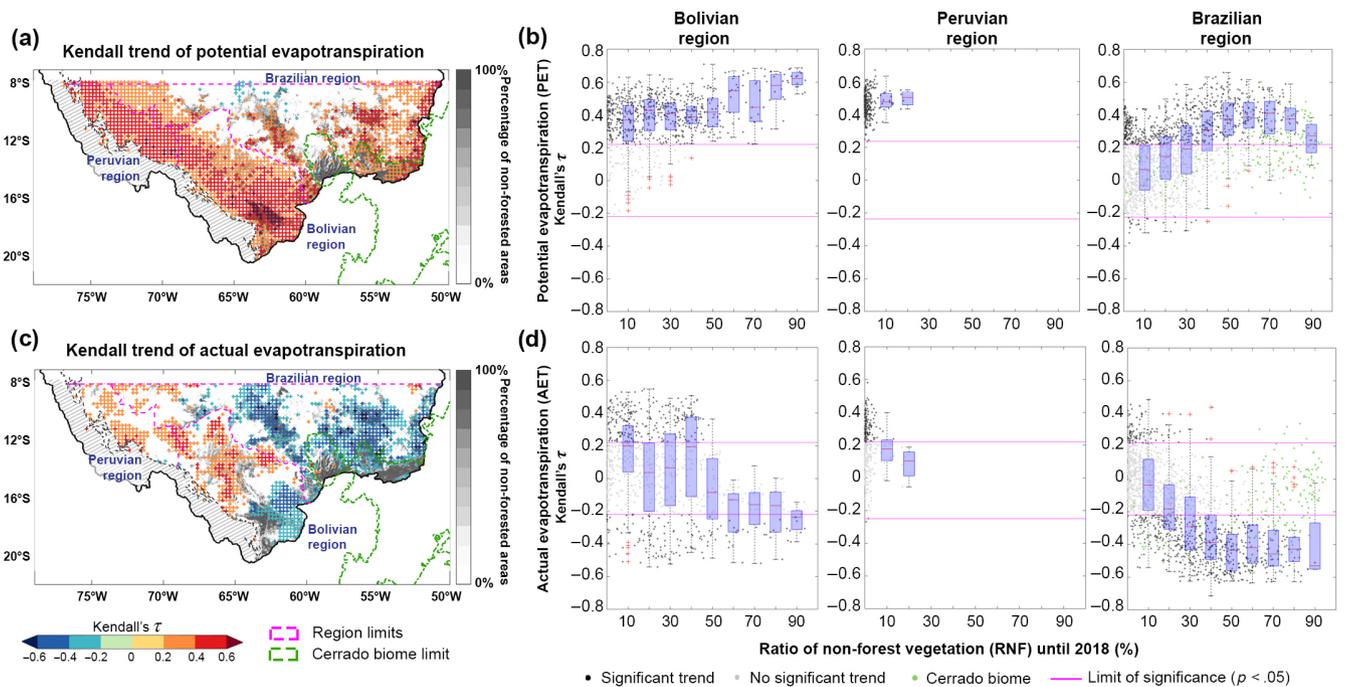
In addition, a part of the Brazilian region corresponds to the Cerrado region, which is described as a Brazilian biome with savanna vegetation (Ratter *et al.*, 1997). In general, the pixels that belong to the Cerrado biome (green dots in Figure 3b–d, right column) do not show significant increases or decreases in precipitation indices. Indeed, pixels showing statistically significant changes in the Cerrado biome represent less than 0.4% of the total Brazilian region for  $P$ , WDF and DDF.

##### 3.1.2 | PET and AET

A significant PET increase is observed over 1981–2018 in most of the Southern Amazon (Figure 4a), particularly in the Bolivian and Peruvian regions and in the eastern Brazilian region. It can be noted that relatively high values of Kendall's  $\tau$  (0.4–0.6) occur particularly on the eastern flank of the Andes. This is coherent with the documented increase in temperature over this region as a regional climatic feature (Lavado Casimiro *et al.*, 2013; Seiler *et al.*, 2013). In the Bolivian and Brazilian regions, the highest PET increases are mostly recorded in the highest RNF areas (>40%), with relatively high Kendall's  $\tau$  values (0.4–0.6) but even higher (>0.6) in the highly deforested region of the Southern Bolivian region (Figure 4a,b). Indeed, the land surface temperature increment over tropical deforested areas is related to reduced latent heat flux, due to changes in land cover (Culf *et al.*, 1996; Fisch *et al.*, 2004; Bathiany *et al.*, 2010; Alkama and Cescatti, 2016; Prevedello *et al.*, 2019) and/or



**FIGURE 3** (a) Spatial distribution of the 1981–2018 Kendall coefficient ( $\tau$ ) showing the significant annual rainfall decrease (red) and increase (blue). The grey scale background indicates the percentage of total non-forested areas as for 2018. Relationships between the ratio of non-forest vegetation (RNF) in percentage up to 2018 and temporal trends [Kendall coefficient ( $\tau$ ) of the 1981–2018 period] of (b)  $P$ , (c) WDF and (d) DDF, respectively, in Bolivian (left), Peruvian (middle) and Brazilian (right) regions. The points belonging to the Cerrado biome in Brazilian region are plotted in green. Boxplots for 10% RNF intervals are superimposed on panels (b–d) [Colour figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]



**FIGURE 4** Spatial distribution of the 1981–2018 Kendall coefficient ( $\tau$ ) from the trend analysis of (a) potential and (c) actual evapotranspiration. Significant increase and decrease are indicated in red and blue, respectively. The grey scale background indicates the percentage of total non-forested areas as for 2018. Panels (b) and (d) indicate the relationship between the ratio of non-forest vegetation (RNF) in percentage over up to 2018 and the temporal trend [Kendall coefficient ( $\tau$ ) of the 1981–2018 period] of PET and AET, respectively, in Bolivian (left), Peruvian (middle) and Brazilian (right) regions. The points belonging to the Cerrado biome in Brazilian region are plotted in green. Boxplots for 10% RNF intervals are superimposed on panels (b) and (d) [Colour figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]

increase in radiation due to changes in the patterns of cloudiness (Wang *et al.*, 2009). Note that PET and forest loss data are computed using independent data sources (see Section 2). Remarkably, these results are also valid when five land surface and global hydrological models were analysed (see Figure S3).

On the other hand, spots with significant AET diminution are located in highly and extensively deforested areas (RNF > 40%) (Figure 4c,d), mostly in the Brazilian region and, to a lesser extent, in the Southern Bolivian region. In the Brazilian region, it can be noted that relatively high values of Kendall's  $\tau$  (−0.6 to −0.4) are associated with high values of deforestation, including some inland areas with lower values (less than −0.6). It is also interesting to note that this does not occur in the Cerrado region where there are also high values of RNF, which will be discussed later. These findings compare well with those of other studies (Gash and Nobre, 1997; von Randow *et al.*, 2004; 2012; Lee *et al.*, 2005; Baker and Spracklen, 2019; Prevedello *et al.*, 2019; Xu *et al.*, 2020) showing that in deforested areas AET decreases. While, as shown in Figure 4a,b, the capacity of the atmosphere to hold humidity increased in most of the Southern Amazon (PET increase), AET increased in non-deforested areas (Western Amazon) and systematically decreased in large deforested areas (Brazilian and Southern Bolivian regions) (Figure 4c). This is likely explained by the fact that tree canopy in the Amazon has larger access to soil moisture from a deeper soil layer (up to 8 m) than grasses, shrubland and sparse vegetation (hereafter called 'low-tree cover', see Supporting Information) and agricultural crops (Nepstad *et al.*, 1994; Kim *et al.*, 2012). This characteristic allows for the transfer of large quantities of water to the atmosphere via transpiration (Lee *et al.*, 2005). Therefore, deforestation favours the energy partitioning towards an increase in the sensible heat flux due to inaccessibility to soil water; moreover, a decrease in  $P$  could amplify this effect even in areas with lower RNF. Here again, pixels belonging to the Cerrado biome in the Brazilian region (green dots in Figure 4b,d, right column) show a wide spread of increasing and decreasing PET/AET pixels, indicating that there is no significant global trend over this specific biome. However, it is important to note that pixels corresponding to the Cerrado biome region with high values of RNF but no significant AET decrease could be explained by the fact that this region has a high rate of agricultural production (Reis *et al.*, 2017; Pires, 2020). Therefore, the expansion and intensification of irrigated agriculture that was identified in the Cerrado region is likely to influence the AET values (Arantes *et al.*, 2016; Spera *et al.*, 2016). This is also evidenced by the vegetation cover maps used in this research (ESA-CCI LC) where the changes from forest

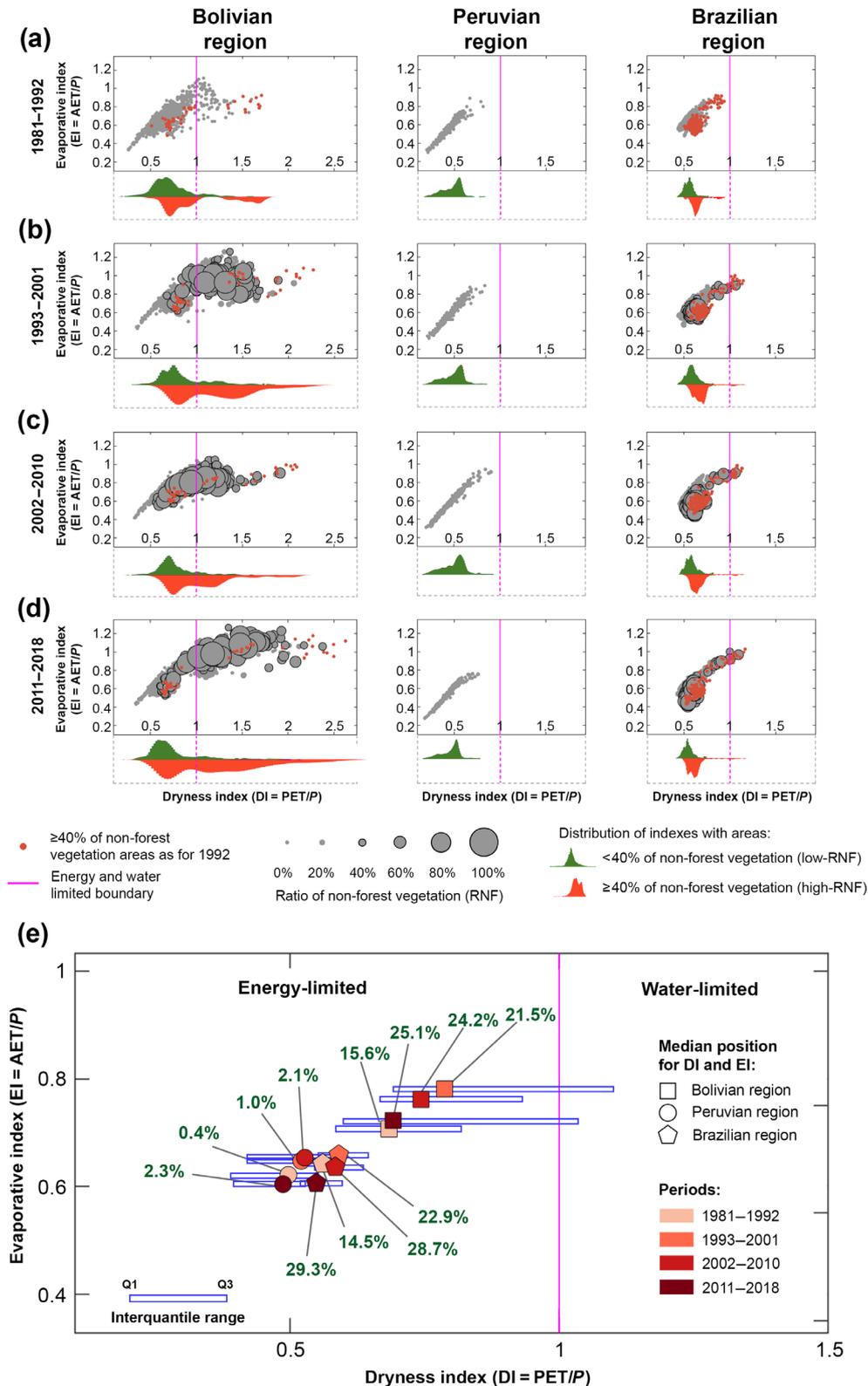
and mainly from low-tree to cropland have been significant in the Cerrado region (see Figure S5f).

### 3.2 | Temporal evolution of surface water and energy partitioning for decadal periods

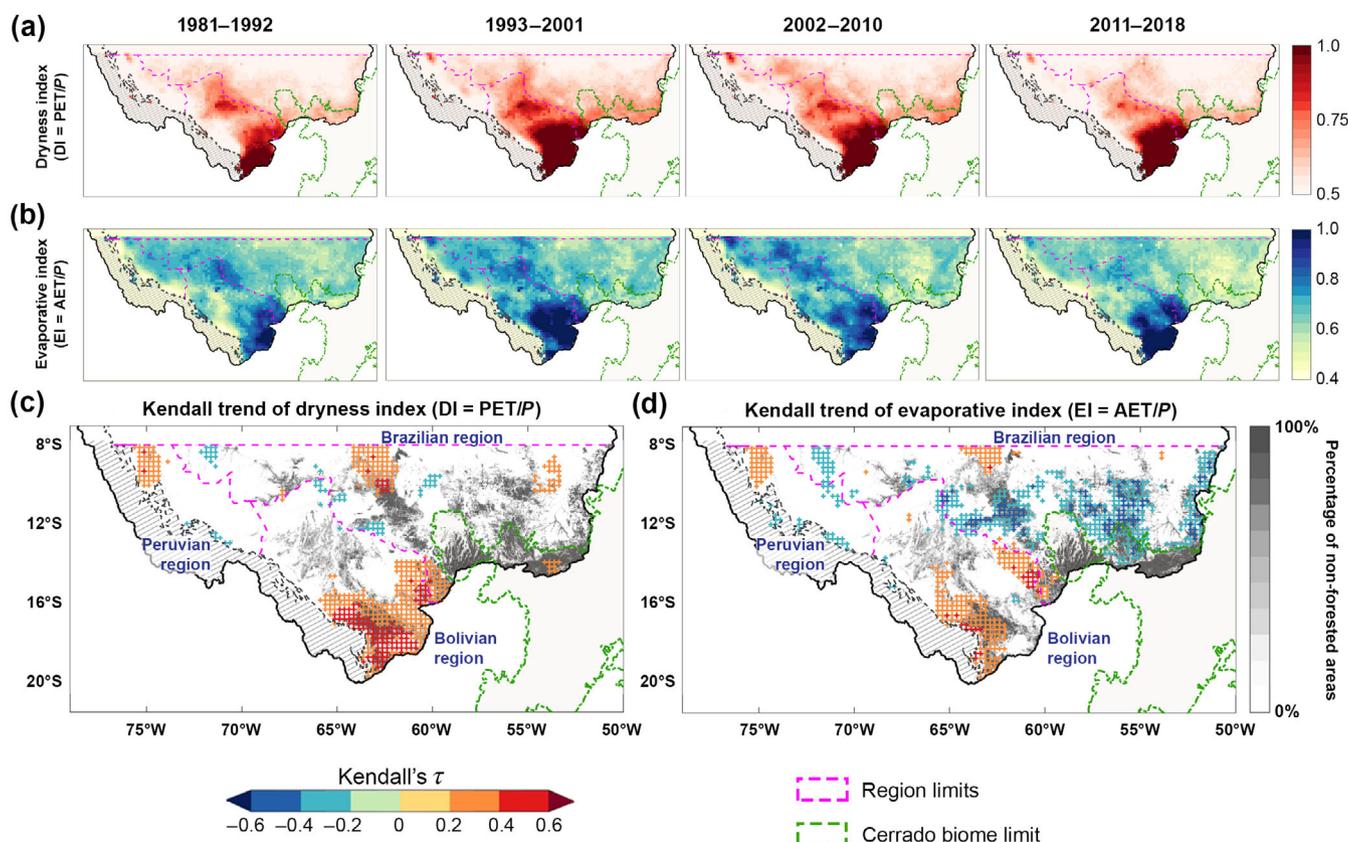
Here, the relationship between the surface water and energy partitioning and RNF are analysed using a pixel-based Budyko-like theoretical framework approach (see Section 2). In the Budyko framework, the surface water and energy partitioning are characterized by computing an evaporative index ( $EI = AET/P$ ) and a dryness index ( $DI = PET/P$ ). This approach allows for discriminating energy-limited ( $DI < 1$ ) regions from water-limited ( $DI > 1$ ) regions. Arguably, a long-term evolution from a neat energy-limited to a neat water-limited regime over large areas might lead to a biome transition. We further discriminate regions with RNF values above (hereafter called 'high-RNF') and below (hereafter called 'low-RNF') 40% (Figure 5). Note that RNF values evolve every ~10 years in this analysis. A high-RNF area in 2018 could be due to deforestation (during the land cover data set period, 1992–2018) or pre-existing non-forest vegetation cover. On the other hand, in areas characterized by a strong dynamics of forest loss, passing the RNF threshold of 40% has been hypothesized as a possible tipping point in terms of bioclimatic feedback (Sampaio *et al.*, 2007; Nobre and Borma, 2009; Nobre *et al.*, 2016). In this perspective, it is worth noting that areas with high-RNF values in the Bolivian region display a systematic decrease in  $P$  (Figure 3). This is why RNF is used here as both a spatial pattern descriptor for a given year or period of 10 years and a dynamic descriptor by looking at its evolution over successive 10-year periods.

#### 3.2.1 | Bolivian region

In the Bolivian region, most of the area for the 1981–1992 period is energy-limited ( $DI$  in the range of ~0.50–1.0; Figure 5a left column). However, after 1992, ~19% of the region becomes water-limited ( $DI$  in the range of ~1.0–1.5), mainly in high-RNF areas (grey circles with black borders), as can be seen from the statistical distribution plotted in red below the  $x$  axis of Figure 5a–d. The transition from an energy-limited to a water-limited condition in high-RNF areas is caused by both a PET increase and a  $P$  decrease over this region, mainly in southern areas (south of 15°S) (Zhang *et al.*, 2016; Espinoza *et al.*, 2019b). While  $P$  and PET evolutions are likely driven by large-scale climatic conditions, the PET



**FIGURE 5** (a–d) Dryness versus evaporative index for the three regions and the four periods. The circle size is proportional to the value of the ratio of non-forest vegetation (RNF) averaged at the pixel scale over each of the four ~10-year period. The probability distribution of the dryness index is shown for areas with less (greater) than 40% of RNF in green (red) as violin plots. The areas with a RNF above 40% as for 1992 are represented by red dots. (e) Distribution of the median values of DI and EI for each region (geometric shapes) and period (red scale colours); the corresponding percentage of non-forest vegetation value for each region and period is indicated in green. The interquartile range for DI is plotted as horizontal blue bars [Colour figure can be viewed at wileyonlinelibrary.com]



**FIGURE 6** Spatial distribution of the (a) dryness index (red range) and (b) evaporative index (green range) for the four decadal periods. Spatial distribution of the 1981–2018 Kendall coefficient ( $\tau$ ) of (c) dryness index and (d) evaporative index. Significant increase and decrease are indicated in red and blue, respectively. The grey scale background indicates the percentage of total non-forest vegetation areas as for 2018 [Colour figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]

increase could also be related to an increased sensible heat flux due to sub-regional changes in vegetation cover from forest to low-tree or crops. Considering the areas with low-RNF, the DI median value remains within the range  $\sim 0.5$  to  $\sim 1.0$  for all periods as can be observed from the frequency distribution plotted in green below the  $x$  axis of Figure 5a–d, left column.

Over this region, the EI median shifted from 0.8–1.0 in the 1981–1992 period to 1.0–1.2 afterwards for high-RNF areas; in fact,  $\sim 22\%$  of this region became EI values greater than 1.0 after 1992. This is clearly a consequence of a stronger decrease in  $P$  than in AET, with AET being altered by forest loss and  $P$  mainly by large-scale climate variability. In fact, previous studies have documented a rainfall diminution over the Bolivian Amazon related to an enhanced atmospheric subsidence over southern Amazonia, as a consequence of the intensification of the continental Hadley circulation (Arias *et al.*, 2015; Molina-Carpio *et al.*, 2017; Leite-Filho *et al.*, 2019; Espinoza *et al.*, 2019a; 2019b). Additionally, in low-RNF areas the EI median remained around 0.8–1.0 after 1992. Note that EI values may exceed 1.0 since a pixel-based Budyko-like

approach is used, which does not guarantee a closed vertical water budget (see Section 2).

Figure 5e synthesizes on a single graph the distributions shown in Figures 5a–d, by plotting the median of DI and EI for each region and each period. For the Bolivian region, surface water and energy partitioning conditions indicate a trend to shift from energy-limited to water-limited conditions (vertical line in magenta) mainly for areas located in the southern Bolivian region during the analysed period (Figure 6). This could be interpreted as this hydro-climatic shift having created the conditions for a forest dieback in this region. It is also worth noting an apparent trend to return to energy-limited conditions for the last two periods (2002–2010 and 2011–2018), with great uncertainty however (DI equals to  $0.74 \pm 0.3$  and  $0.69 \pm 0.4$ , respectively). Two factors may explain this somewhat uncertain and temporary trend to return to an energy-limited regime from 1993–2001 to the 2002–2010 and 2011–2018 periods: (a) PET decreased due to changes in surface temperature, humidity or wind speed; and (b) the increase in  $P$  amount in northern Bolivian Amazon, mainly for the period 2002–2010 (Espinoza *et al.*, 2019b). In fact, results suggest that

$P$  variation plays a more important role, because despite the large interquartile intervals of DI, there is a general trend towards drying out mainly for areas in the southern areas (see Figure 6a,c) but with significant decadal fluctuations.

### 3.2.2 | Peruvian and Brazilian regions

For the Peruvian and Brazilian regions, the regime mainly remains energy-limited over the entire 1981–2018 period (Figure 5a–d, middle and right columns), except for a few areas in the Brazilian region. However, slightly higher values of the DI median are observed for high-RNF areas ( $\sim 0.65$ ) compared to low-RNF areas ( $\sim 0.6$ ) in the Brazilian region. Besides, in the Peruvian region, there are only areas with RNF lower than 40% (low-RNF) and all of them remain in the energy-limited region during all periods. The EI values for the Peruvian region show a higher dispersion than for the Brazilian region but with the median remaining around  $\sim 0.7$  with a slight decline in the last period. Besides, over the high-RNF areas in the Brazilian region, EI also decreases slightly from  $\sim 0.6$  to  $\sim 0.5$  until the end of the period. This reflects the fact that, under relative stable rainfall conditions (Figure 3a), low-tree cover areas or crops (high-RNF) display a larger decrease in AET than forested areas (Figure 4c), inducing this slight decrease of EI.

In the Brazilian region, we observe (Figure 5e) that the greatest RNF increase is recorded from the first (1981–1992) to the second (1993–2001) period ( $14.5$  to  $22.9\% \pm 0.5\%$ ); however, the transition trend to water-limited conditions occurs mainly the Bolivian region, which also has a large and similar increase in RNF for the same period ( $15.6$  to  $21.5\% \pm 0.9\%$ ), exemplifying the fact that the dependence of surface water and energy partitioning to RNF is controlled by both sub-regional and large-scale conditions.

In order to provide a spatial overview of the observed changes in the DI and the EI, we analysed the spatial distribution by the four decadal periods (Figure 6a,b) and the temporal trend for the total period (1981–2018) (Figure 6c,d). Evidence suggests that zones showing decreased precipitation tend towards a water-limited state. Most of the significant trends occur in zones that shows both scenarios, decreased precipitation and forest loss greater than 40%, which are characterized by having crossed from an energy-limited to a water-limited behaviour over the study period, as it is observed in the Southern Bolivian region.

A significant increase in the trends of the DI and EI is also observed in the north-west of Peruvian region

and in the north-central part in the Brazilian region. For the Peruvian region, this is mainly a consequence of a reduction in  $P$ , while for the Brazilian region it may be due to a greater decrease in  $P$  than in AET (see Figures 3b and 4c). It is interesting to note that DI and EI (Figure 6c,d, respectively) have the same spatial trend pattern than  $P$  (see Figure 3a), which suggests that decrease in  $P$  plays a major influence in the surface water and energy partitioning in Southern Amazonia.

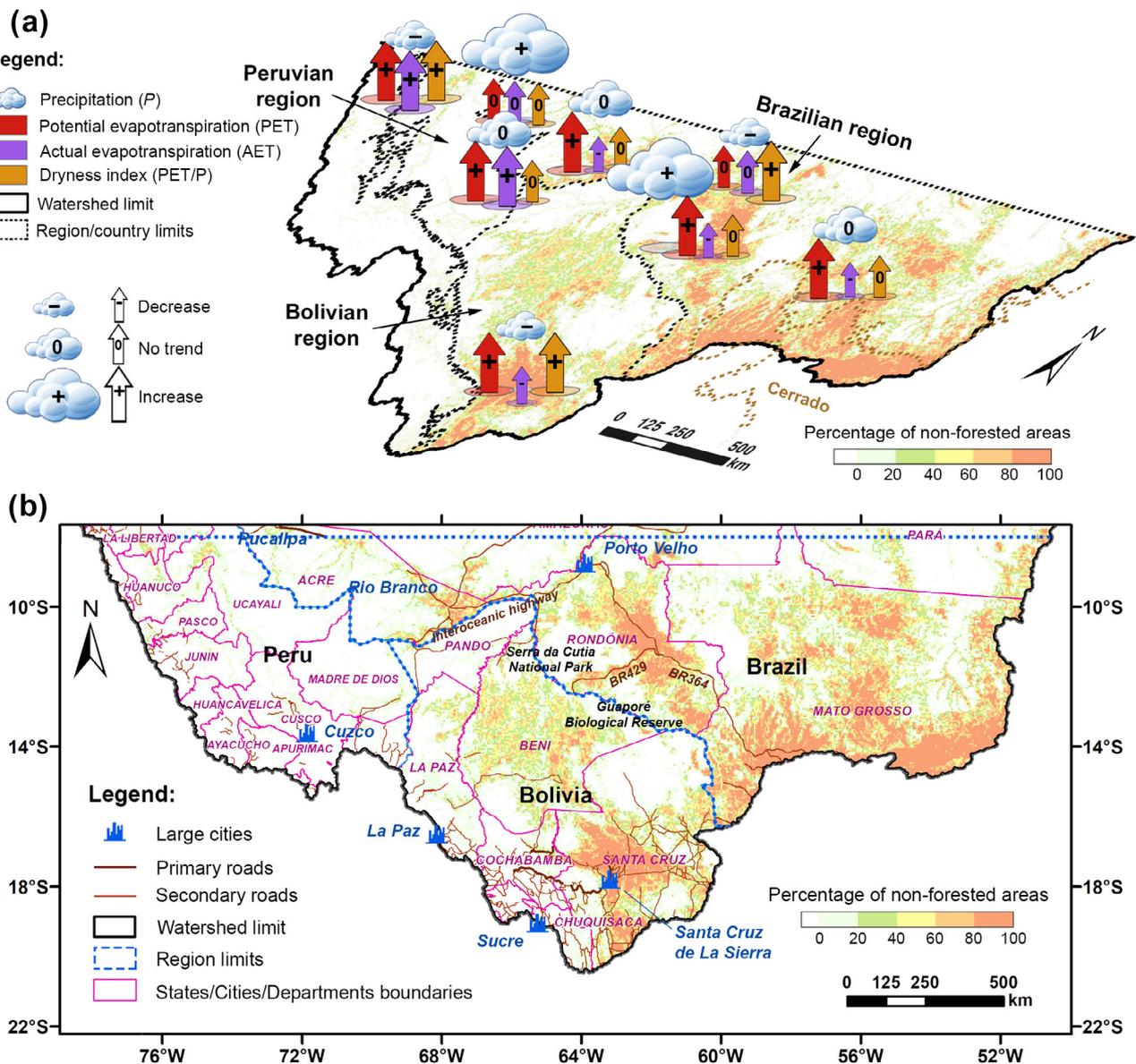
## 4 | DISCUSSIONS

### 4.1 | Potential factors related to land use changes and large-scale circulation

Beyond the sole value of RNF evolution over decades, the different results obtained for the three regions are likely also related to their different proportion of forest converted into cropland versus low-tree cover, as shown in Figure S5. For instance, the changes in land cover from forest to low-tree cover (interpreted as the natural process of forest death), in relation with the total deforested areas, are significantly higher in the Bolivian region ( $\sim 28\%$ ) as compared to the Peruvian and Brazilian regions (3% and 11%, respectively). Among the factors that imply the loss of forest is the increase in fires, which are primarily a man-induced phenomenon (Cochrane, 2003). Additionally, this can also have an opposite meaning since according to Xu *et al.* (2020), the fraction of burned area increase in the Southern Amazon is also largely dependent on the ratio of forest loss. Fires can cause a rapid transition from tropical forest to savanna-like cover, however existing databases are much more recent than our analysis period (e.g., 2000's of MODIS with the Fire Information for Resource Management System database or the Global Fire Emissions Database) and therefore not fully relevant for this analysis.

Evidence suggests that decrease in  $P$  in Southern Amazonia is also likely driven by large-scale processes related to a strengthening of the Walker and Hadley circulations. This strengthening enhances wind subsidence over the  $10^{\circ}\text{S}$ – $20^{\circ}\text{S}$  which suggests a reduction in deep convection and cloudiness (Arias *et al.*, 2011; Espinoza *et al.*, 2019a). Authors suggest that warm conditions in the north tropical Atlantic are associated with the enhanced wind subsidence over southern Amazonia and dry conditions over this region (Arias *et al.*, 2020).

In addition to this, it is worth recalling that deforestation is also related to other variables influencing convective processes, such as vertical wind (see Supporting Information), which is statistically significantly correlated with RNF. Results indicate significant influence of land use



**FIGURE 7** (a) Schematic distribution of increased (large icon with '+'), decreased (small icon with '-') and non-trend (medium icon with '0') for precipitation (cloud), potential evapotranspiration (red arrow), actual evapotranspiration (purple arrow) and dryness index (gold arrow), for areas displaying the most significant variations, over 1981–2018. The percentage of non-forest vegetation areas as for 2018 is represented by the green-red colour bar. (b) Map of political boundaries, roads, and main departments, cities and states for each region (Peruvian, Bolivian and Brazilian) [Colour figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]

change (from forest to low-tree cover) on vertical wind, since the Southern Bolivian region (south of 15°S where high-RNF are mainly placed) and the Brazilian regions have the highest rate of deforestation ( $16.95\% \pm 0.9\%$  and  $14.7\% \pm 0.5\%$ , respectively) as compared to the Peruvian region ( $1.7\% \pm 0.5\%$ ) (Figure S7). This means that in areas with high-RNF, there is a significant decrease in vertical wind. In the Southern Amazon, convective processes are particularly active (Vera *et al.*, 2006); therefore, a decrease in vertical wind could indicate a decrease in convective activity and rainfall. To gain a complete understanding of what is at play when relating the evolution of forest cover

with rainfall changes, local processes should therefore be accounted for, through the use of tools such as numerical models (Eiras-Barca *et al.*, 2020; Ruiz-Vásquez *et al.*, 2020; Sierra *et al.*, 2021). For instance, in a recent high-resolution modelling work, Sierra *et al.* (2021) found that a deforestation scenario of 45% of the Amazonian forest produce decrease in evapotranspiration and vertically integrated moisture convergence in central and southern Amazon, which also implies a reduction in the vertical velocity. These regional changes due to Amazon deforestation are associated with a rise in the atmospheric pressure at

lower levels. This result suggests a complex interaction between changes in large-scale atmospheric circulation (e.g., Walker and Hadley circulations) and regional changes due to forest loss (Sierra *et al.*, 2021).

In this study, we have identified spatially distributed linkages between hydro-climatic conditions and forest cover change in the Southern Amazon based on observations (semiological study). While the scope of this study was not to establish a cause-effect diagnosis of the complex physical interactions between climate and changes in vegetation cover, our observational analysis can be useful to complement modelling studies.

## 4.2 | Geographical implications

This section briefly synthesizes and classifies the detected hydro-climatic trends in a geographical perspective (for localization of regions and cities, see Figure 7b). We would like to highlight that this geographical perspective could be useful from a governance point of view for decision makers regarding future threats posed by changes in forest cover and associated changes in the main hydro-climatic variables.

A decrease in  $P$  is observed in the Southern Bolivian Amazon, around Santa Cruz de La Sierra city (Bolivian region), as well as in small spots in the Peruvian Amazon (in Huánuco and Ucayali Regions, close to the Pucallpa city) in the Peruvian region and in the centre of the Rondônia State in the Brazilian region. Increase in  $P$  occurs on the border between Peru and Brazil (Department of Ucayali and the Brazilian State of Acre, respectively) in the Peruvian region, but also, in the south of Rondônia state in the Brazilian region, where there is a 'tongue' of deforestation between the national park Serra da Cutia and the Guaporé Biological Reserve, along the BR429 and near to the BR364 highways (Figure 7b).

Increase in PET and decrease in AET mainly occurs in places with high forest loss such as around Santa Cruz de La Sierra city (Bolivian region). This is also witnessed in the Brazilian Arc of Deforestation (close to the border between the Brazilian Acre State and the Bolivian Pando Department), especially along the inter-oceanic highway and near Rio Branco city, the capital of the Acre State (east of the Peruvian region). The same applies to the Southern Rondônia and the Central Mato Grosso states (Brazilian region). Increase in PET and AET together only occur in most of the Peruvian Amazon (westernmost part of the Peruvian region), where the level of deforestation is still low compared to the Southern Brazilian and Bolivian Amazon. In areas where the ~40–50% RNF threshold is exceeded, a

significant decrease in  $P$  and increase in AET are practically only observed in the Bolivian and Brazilian regions and according to our evidence, this threshold is still not exceeded in Peru.

Observed increase in the DI occurs in the same regions that are dominated by  $P$  decreases, pointing out a hydro-climatic condition (water-limited) favourable to develop a low-tree cover, potentially threatening the maintenance of the remaining tropical forest. This occurs mostly in the Bolivian region (the Bolivian city of Santa Cruz de La Sierra being a hot spot in this respect), and in the north-westernmost parts of the Peruvian region (Ucayali Region in Peru) and Brazilian region (area of the Porto Velho city in the Northern Rondônia State in Brazil).

## 5 | CONCLUSION AND FINAL REMARKS

The relationships existing at sub-regional scales between forest loss and the evolution of water and energy partitioning in the Southern Amazon basin vary largely depending on the location considered (e.g., Southern Amazon forest as a whole, Southern Bolivian Amazon, Southwestern Amazon or just the Arc of Deforestation zone), as a result of different dynamics in climate and land use changes. Most important is the fact that the decrease in precipitation induced either by large-scale climate modifications or by changing local conditions may lead to a transition from energy- to water-limited conditions, in case of high deforestation ratios (e.g., above ~40%). This situation is only observed in the Southern Bolivian Amazon at the moment, which highlights the necessity of working at local scales if one is to apprehend the conditions driving to a forest dieback.

In many areas of large forest loss, such as the Southern Bolivian and Brazilian Amazon, a growing imbalance is observed between increasing potential and decreasing AET, showing that extensively forested lands are crucial to prevent critical changes in the sub-regional water and energy partitioning. The central Peruvian Amazon is a particular case where the forest loss is not high while a decrease in precipitation affects the water balance, but without any obvious land cover transition. Additionally, the land use transition from forest to a low-tree cover is already three times higher in percentage for the Southern Bolivian region than for the Peruvian and Brazilian regions. This makes this region prone to large-scale shifts especially if areas that are not still highly deforested become water-limited due to changing climate conditions.

Recent studies show that hydro-climate changes in tropic regions are globally related to modifications of the large-scale atmospheric circulation associated with

anthropogenic global warming. Coherent with this, it is noticeable that changes in the regional Hadley circulation over tropical South America have been related to a rainfall diminution in Southern Amazon and an increased dry-season length. Thus, large-scale climate changes may combine with regional vegetation changes to produce a positive feedback leading to an increased forest loss (e.g., in Southern Bolivian Amazon). This picture is somewhat different from the one observed in the Southern Brazilian Amazon, where deforestation is occurring at a high rate, due to anthropogenic changes (cultivation of formerly forested areas), but where no systematic change in rainfall is observed.

We would like to point out that the methodology presented in this research can be used in any region of the Amazon or in any part of the world, since the information we used has a global coverage. However, due to the sub-regional scale of our analyses and due to the different conditions and results of interactions between land use change and hydro-climatic variables shown in this research, our conclusions cannot be extrapolated to other regions of the Amazon basin (i.e., north of 8°S) that have not been the object of this study.

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#### CONFLICT OF INTEREST

The authors declare no conflicts of interest.

#### AUTHOR CONTRIBUTIONS

**Sly Wongchuig:** Conceptualization; formal analysis; data curation; visualization; methodology; writing – review and editing. **Jhan Carlo Espinoza:** Conceptualization; formal analysis; funding acquisition; methodology; writing – review and editing. **Thomas Condom:** Conceptualization; formal analysis; methodology; writing – review and editing. **Hans Segura:** Writing – review and editing. **Josyane Ronchail:** Formal analysis; writing

– review and editing. **Paola Arias:** Formal analysis; writing – review and editing. **Clementine Junquas:** Writing – review and editing. **Antoine Rabatel:** Formal analysis; writing – review and editing. **Thierry Lebel:** Formal analysis; investigation; supervision; writing – review and editing.

#### DATA AVAILABILITY STATEMENT

The data that support the findings of this study are available upon reasonable request from the authors.

#### ORCID

**Sly Wongchuig**  <https://orcid.org/0000-0002-1116-0742>  
**Jhan Carlo Espinoza**  <https://orcid.org/0000-0001-7732-8504>  
**Thomas Condom**  <https://orcid.org/0000-0002-4408-8580>  
**Hans Segura**  <https://orcid.org/0000-0002-6720-5882>  
**Josyane Ronchail**  <https://orcid.org/0000-0002-8824-6771>  
**Paola A. Arias**  <https://orcid.org/0000-0003-1726-6000>  
**Clementine Junquas**  <https://orcid.org/0000-0003-1542-5602>  
**Antoine Rabatel**  <https://orcid.org/0000-0002-5304-1055>  
**Thierry Lebel**  <https://orcid.org/0000-0002-1297-6751>

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