

Beating of the Amazon: The Diurnal Cycle of Amazonian Hydroclimate

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Abstract

Previous studies have estimated that 25% to 35% of Amazonian precipitation comes from evapotranspiration (ET) within the basin. However, due to simplifying assumptions of traditional models, these studies primarily focus on large spatial and temporal scales. In this work we use the Weather Research and Forecast (WRF) regional climate model with the added capability of water vapor tracers to track the moisture from Amazonian ET at the native WRF resolution. The tracers reveal that the well-mixed assumption of simpler models does not hold, as local ET is more efficiently rained out of the atmospheric column than remote sources of moisture, particularly in the eastern part of the basin. Recycled precipitation shows a strong annual and semi-annual signal, associated with the passage of the Inter-Tropical Convergence Zone. The tracers also reveal a strong diurnal cycle of Amazonian water vapor related to the diurnal cycle of ET, convective precipitation and advected moisture. ET increases from early morning into the afternoon, some of this moisture is rained out through convective storms in the early evening, while later in the night strong winds associated with the South American Low Level Jet advect moisture downwind. Visualizing the Amazonian water vapor highlights its diurnal beating pattern and suggests that the Amazon has “younger” water than other regions in the globe, with very efficient recycling of local moisture.

Amazonian Moisture Recycling Revisited Using WRF with Water Vapor Tracers

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Key Points:

- 30% of Amazonian precipitation comes from local evapotranspiration. Local moisture is more efficiently rained out than remote moisture.
- Recycled precipitation shows a strong annual and semi-annual signal, associated with the passage of the Inter-Tropical Convergence Zone.
- A diurnal cycle cycle of Amazonian water vapor reflects the variability of evapotranspiration, recycled precipitation and vapor flux.

Abstract

Previous studies have estimated that 25% to 35% of Amazonian precipitation comes from evapotranspiration (ET) within the basin. However, due to simplifying assumptions of traditional models, these studies primarily focus on large spatial and temporal scales. In this work we use the Weather Research and Forecast (WRF) regional climate model with the added capability of water vapor tracers to track the moisture from Amazonian ET at the native WRF resolution. The tracers reveal that the well-mixed assumption of simpler models does not hold, as local ET is more efficiently rained out of the atmospheric column than remote sources of moisture, particularly in the eastern part of the basin. Recycled precipitation shows a strong annual and semi-annual signal, associated with the passage of the Inter-Tropical Convergence Zone. The tracers also reveal a strong diurnal cycle of Amazonian water vapor related to the diurnal cycle of ET, convective precipitation and advected moisture. ET increases from early morning into the afternoon, some of this moisture is rained out through convective storms in the early evening, while later in the night strong winds associated with the South American Low Level Jet advect moisture downwind. Visualizing the Amazonian water vapor highlights its diurnal beating pattern and suggests that the Amazon has “younger” water than other regions in the globe, with very efficient recycling of local moisture.

Plain Language Summary

Evaporation from soil and transpiration from plants within the Amazon contribute to approximately one third of the precipitation that falls within the basin in a process known as “precipitation recycling”. This estimate represents an average over the basin and over many years. In this work we use numerical water tracers within an atmospheric model to quantify precipitation recycling at higher spatial and temporal resolution than previous studies. The tracers allow us to follow the water from the time it evaporates from the land until it falls as precipitation. Our work reveals cycles in water vapor and precipitation of Amazonian origin that had not been previously studied. In particular, the daily timescale shows how evaporation and transpiration increase from early morning into the afternoon and contribute to the accumulation of Amazonian water vapor in the atmosphere. Some of this moisture is rained out in the early evening in convective storms, while later in the night strong winds transport moisture away from the basin. Visualizing the Amazonian water vapor highlights its diurnal beating pattern.

1 Introduction

Early evidence of the importance Amazonian evapotranspiration (ET) for local precipitation came from observational analysis of oxygen-18 ($\delta^{18}\text{O}$) in precipitation, which was found to have an inland gradient of $\delta^{18}\text{O}$ much smaller than in other regions of the world (Salati et al., 1979). This observational result suggested that a significant part of the rainfall came from re-evaporated water, otherwise known as recycled precipitation. Subsequent analyses based on bulk numerical models estimated precipitation recycling (or the percent of precipitation that comes from local ET) ranging from 25%-40% (Brubaker et al., 1993; Eltahir & Bras, 1994; Burde, 2006). There is a strong gradient in recycled precipitation, increasing from east to southwest as the dominant flow enters from the Atlantic Ocean, traverses the basin and encounters the Andes mountains (Eltahir & Bras, 1994; Burde, 2006). It is important to keep in mind that these early bulk models assume that the atmospheric column is well mixed. In other words, they assume that, at the time scales of the model, turbulence effectively mixes the atmosphere so the proportion of precipitation from advected and recycled origin is the same as the proportion found in water vapor. However, early work of Lettau et al. (1979) argued that in a system such as the Amazon it is important to account for fast recycling as diurnal convection yields precipitation before all water vapor from the surface has enough time to mix with the ex-

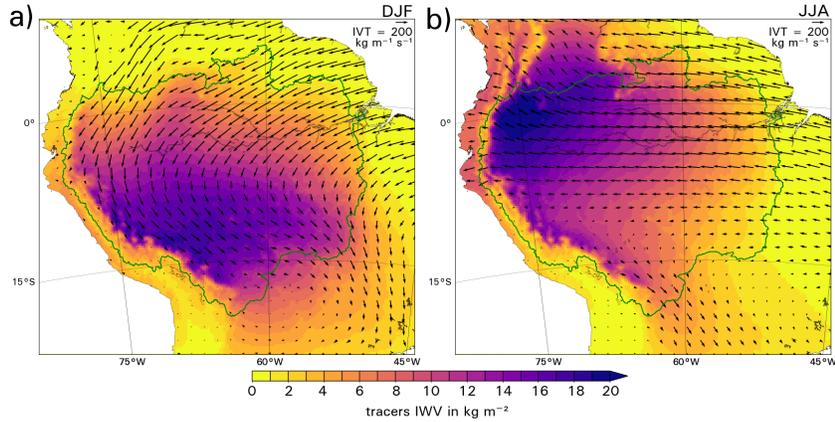


Figure 1. Shading indicates integrated water vapor IWV (kg m^{-2}) of Amazonian origin. Black vectors indicate total integrated vapor transport IVT ($\text{kg m}^{-1}\text{s}^{-1}$). Values are average for DJF (a) and JJA (b), for the period 2003-2013.

68 isting precipitable water in the column. In this sense, moisture from evapotranspiration
 69 would be more likely to rain out than advected moisture. When incomplete mixing is
 70 taken into account, the bulk model recycling estimates are closer to 45% (Lettau et al.,
 71 1979; Burde, 2006). When incorporating incomplete mixing, the spatial distribution of
 72 recycling remains similar but the recycling values increase towards the east (Burde, 2006).

73 Currently, the most physically realistic way to numerically estimate precipitation
 74 recycling is using water vapor tracers embedded within atmospheric models, as this method
 75 requires the least number of assumptions. Water vapor tracers (WVT) do not make the
 76 assumption of a well-mixed atmospheric column used in most bulk models. Embedding
 77 tracers within the atmospheric model takes into account the changes in moisture con-
 78 tent due to turbulent transport in the planetary boundary layer, cloud microphysics, and
 79 convection (Dominguez et al., 2020). Using WVT within a global climate model at a 2°
 80 by 2.5° resolution, M. G. Bosilovich and Chern (2006) found a recycling ratio of about
 81 27% in the peak recycling season, a 50% higher value than the estimate using bulk mod-
 82 els. This again confirms the idea of “fast recycling” in the Amazon. Interestingly, they
 83 found that the inter-annual variability of recycling mostly related to variability in mois-
 84 ture transport, not evapotranspiration. WVT embedded within regional climate mod-
 85 els allow us to examine processes at a much higher spatial resolution. Using the WVT
 86 embedded within the Weather Research and Forecasting (WRF) model (WRF-WVT),
 87 Yang and Dominguez (2019) tracked the water that originated from the Amazon basin
 88 to understand how it contributes to precipitation throughout the continent. They found
 89 that around 30% of the total precipitation over the Amazon and about 16% of the pre-
 90 cipitation in the downwind region of the La Plata River basin originates from Amazo-
 91 nian ET.

92 All of the work to date has focused on quantifying precipitation recycling over the
 93 Amazon at the monthly or longer time scale. However, the dominant processes that af-
 94 fect recycling variability have several characteristic scales of variability. Terrestrial evap-
 95 otranspiration, which is the source of recycled water vapor, has a strong diurnal cycle.
 96 Precipitation in the Amazon is driven by small-scale and organized convection, and also
 97 shows a clear diurnal cycle (Tanaka et al., 2014). However, intra-daily time scales of re-
 98 cycled precipitation have not been previously studied in the Amazon because bulk mois-
 99 ture source methods cannot provide information at these time scales due to their under-

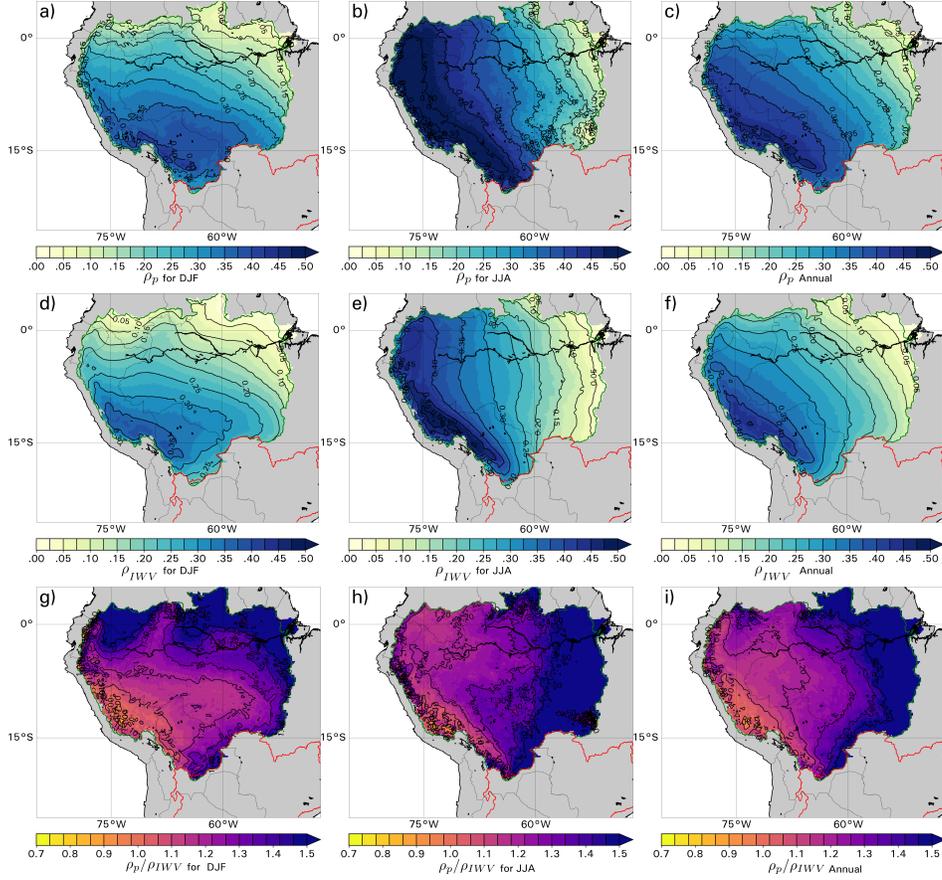


Figure 2. Row 1) Fraction of Amazonian precipitation to total precipitation ($\rho_P = \text{rain}_t / \text{rain}$) for DJF (a), JJA (b) and Annual means (c). Row 2) Fraction of precipitable water (IWV) of Amazonian origin to total precipitable water ($\rho_{IWV} = IWV_t / IWV$) for DJF (d), JJA (e) and Annual means (f). Row 3) “Precipitation efficiency” (ρ_P / ρ_{IWV}), or row 1 divided by row 2.

100 lying assumptions. Evidence of the terrestrial signature on the atmosphere at the global
 101 scale was highlighted in Tuinenburg and van der Ent (2019) as a daily cycle in the at-
 102 mospheric residence time of land evaporation which is different from that of water va-
 103 por of oceanic sources. A diurnal signature in water vapor would imply much shorter
 104 times than the traditional 8-10 day global estimates (van der Ent & Tuinenburg, 2017),
 105 or even the recent 4-5 days estimates (Laderach & Sodemann, 2016). WVT-WRF is an
 106 ideal tool to study fast turnover processes as it allows us to isolate the terres-
 107 tral contributions to atmospheric vapor without having to make assumptions about verti-
 108 cal mixing in the atmosphere. The fact that this is a regional model allows us to delve into the
 109 smaller spatial and temporal scales.

110 In this work we use WRF-WVT to characterize recycling of evapotranspiration over
 111 the Amazon. First, results from our analyses are compared to previous work using bulk
 112 models. Then, we analyze the characteristic spatial and temporal scales of moisture of
 113 Amazonian origin, with particular emphasis on the diurnal timescale.

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2 Methods

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A 10-year simulation (2003-2013) provided by the mesoscale model WRF (20 km \times 20 km) using the entire South American continent as domain of simulation is used here. Water vapor tracers are only available for one set of physics options. As such, the Kain-Fritsch (KF) (Kain & Fritsch, 1990; Kain, 2004) parametrization is used for subgrid-scale convection; the Yonsei University (YSU) (Hong & Pan, 1996) parametrization for turbulent mixing, and the single moment “6-class” (WSM6) (Hong & Lim, 2006) solved the microphysics in phase change and precipitation processes. ERA-Interim (Dee et al., 2011) provides boundary and initial conditions for the simulations.

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Two additional tools are used in the WRF simulation to improve the representation of land-atmosphere interactions, as well as provide the ability to track the fate of the Amazonian ET. We used the Noah-Multiparametrization land surface model (Noah-MP LSM) (Niu et al., 2011) with the MMF groundwater scheme developed by Fan et al. (2007) and Miguez-Macho et al. (2007). This scheme is used to take into account the interaction between the shallow aquifers and soil moisture, which affects Amazonian ET, particularly in water limited regions during dry periods (Miguez-Macho & Fan, 2012; Martinez et al., 2016b, 2016a).

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We also use the WRF Water Vapor Tracer Tool (WRF-VT) which allows the tracking of the moisture evapotranspired from the Amazon (Dominguez et al., 2016; Eiras-Barca et al., 2017; Insua-Costa & Miguez-Macho, 2018). This moisture can either leave the Amazonian domain or precipitate within the basin (as recycled precipitation). WRF-WVT includes additional output variables related to water vapor and precipitation of tracer origin, as well as other species (see Insua-Costa and Miguez-Macho (2018) for details). In this way, we can calculate the precipitation recycling ratio ($\rho_P = rain_t / rain$) where $rain$ is the total precipitation which includes convective and non-convective processes, while $rain_t$ is the precipitation from Amazonian ET. In a similar way, we can calculate the integrated water vapor (IWV) recycling ratio ($\rho_{IWV} = IWV_t / IWV$). Note that this calculation is done on a cell-by-cell basis. Also note that this method does not require the assumption of a the well-mixed atmosphere in the vertical column that is traditionally used in analytical models such as Brubaker et al. (1993) and Eltahir and Bras (1994). This configuration of WRF was recently used by Eiras-Barca et al. (2020) to an-

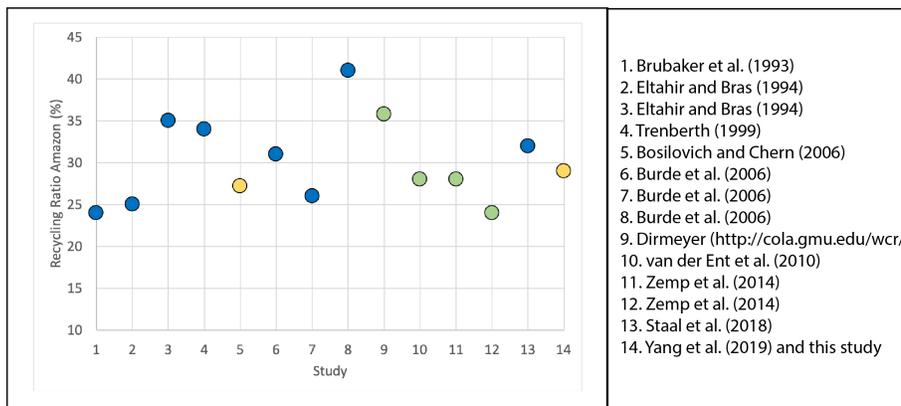


Figure 3. Amazonian climatological recycling ratio as estimated in previous studies which using bulk models (blue), offline models (green) and online models (yellow) to estimate the recycling ratio in the Amazon basin.

145 analyze the impacts of a realistic future deforestation scenario over the same domain, and
 146 a comprehensive validation of these simulations can also be found in Yang and Dominguez
 147 (2019). Along with the mean time series obtained with the WRF-VT we also plot the
 148 equivalent observations obtained with ERA5 and TRMM (3B42 subset, Huffman et al.
 149 (2010)). This helps evaluate WRF’s ability to represent the annual and diurnal cycles
 150 of the variables of interest.

151 To extract the dominant signals in the time series of total and tracer ET, precip-
 152 itation, IWV and IVT we use fast Fourier transform analysis (FFT). Using the 10-year
 153 time series, area-averaged over the Amazon basin, the FFT allows us to identify and quan-
 154 tify the strength of the dominant modes with daily to annual periodicity (Wilks, 2011).
 155 However, we cannot extract interannual signals with this short timeseries.

156 3 Results

157 The moisture transpired by the vegetation and evaporated from the soil of the Ama-
 158 zon basin is carried by the prevalent winds. These winds change dramatically depend-
 159 ing on the season. During the Austral summer (DJF), moisture laden winds enter from
 160 the tropical North Atlantic Ocean (Drumond et al., 2014), cross the Amazon and travel
 161 in a southwesterly direction until they encounter the Andes Mountains and are forced
 162 to veer south and east toward southeastern South America (Satyamurty et al., 2013; Se-
 163 gura, 2019). The moisture of Amazonian origin accumulates along the southwest of the
 164 Amazon basin (Fig. 1a). During the Austral winter, the trade winds traverse the Ama-
 165 zon in a much more zonal direction, and veer northwest toward Colombia and Venezuela.
 166 Moisture of Amazonian origin accumulates along the western and northwestern part of
 167 the basin (Fig. 1b). The autumn (March-May, MAM) and spring (September-November,
 168 SON) seasons have characteristics that reflect the transition between the Austral win-
 169 ter and summer, these can be found in supplementary material (Fig. S1).

170 The fraction of precipitation of Amazonian origin to total precipitation (recycling
 171 ratio) follows this same spatial pattern, increasing gradually from the Atlantic coast to
 172 around 40% in the southwest part of the basin during DJF (Fig. 2a). Interestingly, the
 173 recycling ratio is higher during the dry season (JJA) when more than 50% of the pre-
 174 cipitation along the eastern Andes is of Amazonian origin (Fig 2b). The corresponding
 175 plot for the transitional months (MAM and SON) can be found in Fig. S2. If we ana-
 176 lyze recycled to total precipitable water (IWV), the pattern is similar to that of the re-
 177 cycled precipitation, but smaller in magnitude (Fig. 2d-f). This implies that local pre-
 178 cipitable water is more efficiently rained out of the atmospheric column than advected
 179 moisture. This is likely due to the fact that the moisture for convection is sourced from
 180 lower levels, as argued by Lettau et al. (1979). The geographical pattern of efficiency,
 181 defined as ρ_P/ρ_{IWV} shows higher efficiencies in the eastern part of the basin, and effi-
 182 ciencies close to one in the west along the Andes. Efficiencies close to one indicate that
 183 the atmosphere is well mixed (Eltahir & Bras, 1994). Higher efficiencies in the east in-
 184 dicate that low-level moisture is rained out before it has fully mixed in the atmosphere.
 185 In these regions of poor mixing, lower-level humidity experiences an efficient ascent mech-
 186 anism that leads to precipitation. However, as stated before, the assumption of a well-
 187 mixed atmosphere is common among bulk recycling models (Brubaker et al., 1993; Eltahir
 188 & Bras, 1994; Dominguez et al., 2006). As seen in Fig. 2 g-i, this assumption is not valid
 189 throughout the domain and would result in an under-estimation of recycled precipita-
 190 tion, particularly in the eastern part of the region. However, recycling values in the east
 191 are significantly smaller than in the west, so the area-average differences are not as large.
 192 We find that area-average annual precipitation recycling is 29%. This compares well with
 193 the results from studies using bulk models, offline models and online models (Brubaker
 194 et al., 1993; Eltahir & Bras, 1994; Trenberth, 1999; M. Bosilovich & Chern, 2006; Burde,
 195 2006; van der Ent et al., 2010; Zemp et al., 2014; Staal et al., 2018; Yang & Dominguez,
 196 2019) as shown in Figure 3. In all but one study, Amazonian recycling ranges between

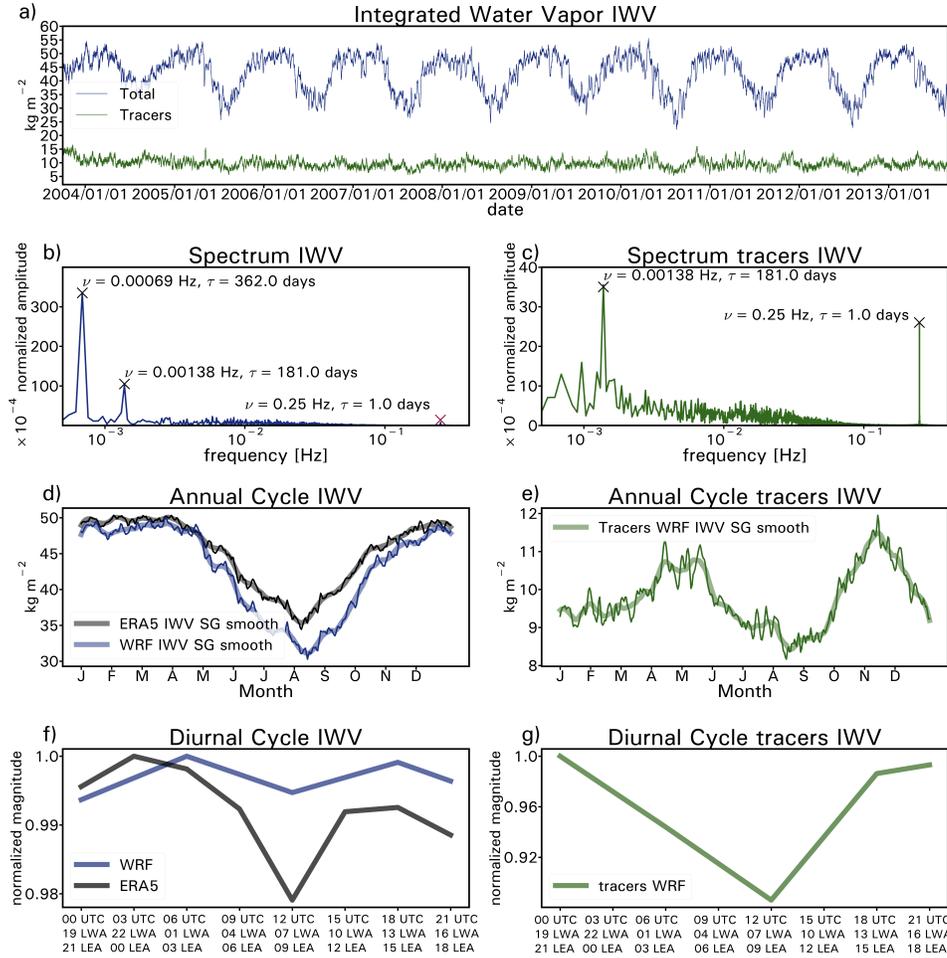


Figure 4. Fast-Fourier transform (FFT) analysis of IWV. a) Raw signals of total IWV (blue) and tracers IWV (green) for a 6h-time step time series 2003-2013. b and c) FFT frequencies spectrum for total IWV (blue) and tracers IWV (green). d) Mean annual cycle for IWV as obtained from the WRF simulations (blue) and ERA5 reanalysis (black). e) Mean annual cycle for tracers IWV as obtained from WRF. f) Mean diurnal cycle for IWV as obtained from the WRF simulations (blue) and ERA5 reanalysis (black). g) Mean diurnal cycle for tracers IWV as obtained from WRF. Note that a Savitzky-Golay (SG) smooth filtered signal is also plotted along with the raw data to ease the visualization when necessary.

197 25-35%. This result is rather surprising given the large differences in the methods and
 198 data sources used. The results also agree with previous analyses in terms of spatial pat-
 199 tern. In particular, those of Eltahir and Bras (1994) using a very simplified two-dimensional
 200 model that assumes a well-mixed atmosphere, and atmospheric data at a 2.5° resolution
 201 (compared to the three dimensional 20 km resolution of our analyses).

3.1 Temporal Analysis

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 203 Using FFT analysis, we find that the characteristic timescale of total IWV has a
 204 very clear annual cycle, with a maximum during the Austral summer and a minimum
 205 during the winter (blue line Fig. 4a and 4b). The annual signal of WRF is very simi-

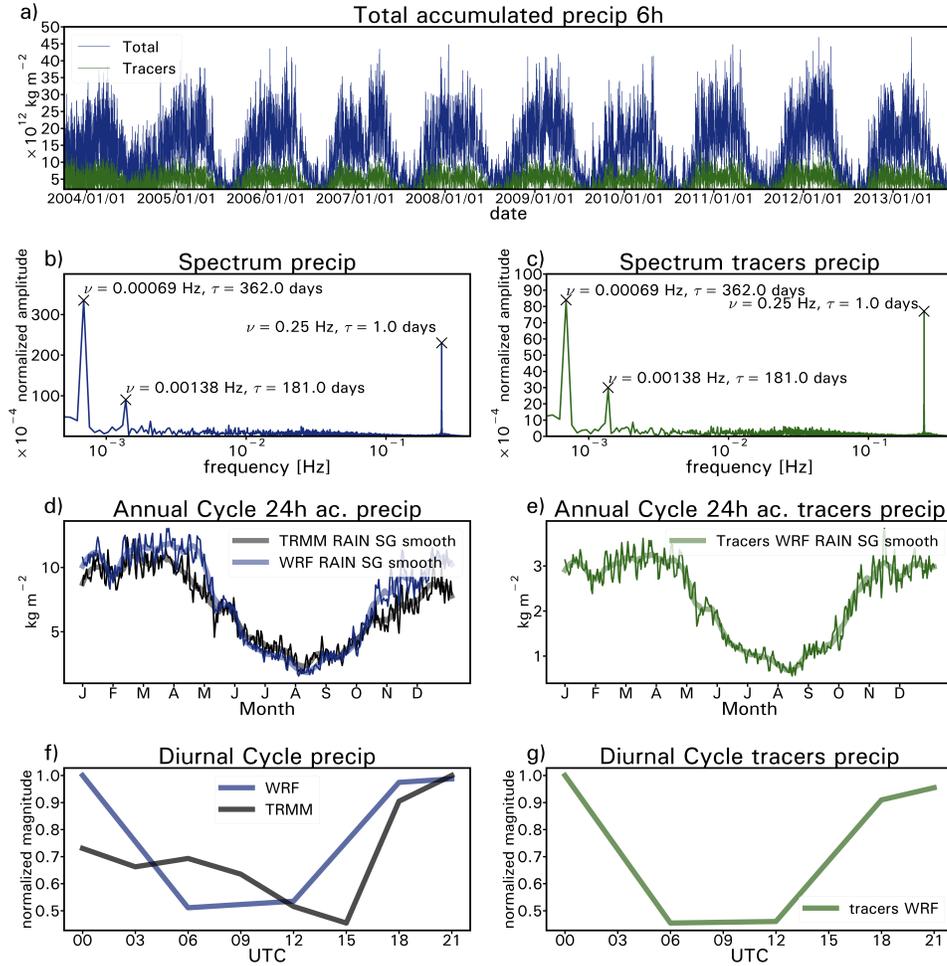


Figure 5. Fast-Fourier transform (FFT) analysis of Precipitation (P) a) Raw signals of total P (blue) and tracers P (green) for a 6h-time step time series 2003-2013. b and c) FFT frequency spectrum for total P (blue) and tracers P (green). d) Mean annual cycle for P as obtained from the WRF simulations (blue) and ERA5 reanalysis (black). e) Mean annual cycle for tracers P as obtained from WRF. f) Mean diurnal cycle for P as obtained from the WRF simulations (blue) and ERA5 reanalysis (black). g) Mean diurnal cycle for tracers P as obtained from WRF.

206 lar to that of ERA5 (Fig. 4d) . We see another signal with a periodicity of around 6 months
 207 (181 days). This signal is due to the migration of the ITCZ, which reaches its southern-
 208 most extent in the Austral summer and its northern-most extent during the Boreal sum-
 209 mer. The central Amazon basin experiences a peak in IWV as the ITCZ migrates north,
 210 and another peak as the ITCZ migrates south. WRF shows a very weak diurnal cycle
 211 in total IWV (Fig. 4f), while ERA5 has a slightly stronger diurnal cycle with minimum
 212 IWV around 12Z (8am local) and increasing values as the day progresses and a maxi-
 213 mum around 3Z (11pm local).

214 Unlike total IWV, tracer IWV (or water vapor of Amazonian origin) does not show
 215 the annual signal, however, it does have the 6-month signal associated to the ITCZ pas-
 216 sage (green line in Fig. 4a and c). The six-month signal has a peak from late April to
 217 early June, and another peak from October to early December (Fig. 4e). In addition,
 218 there is a very clear diurnal signal (Fig. 4c). Figure S3 in the supplementary material

219 shows the 50th and 80th percentile contours of daily mean tracers IWV corresponding
 220 to 00, 06, 12 and 18 UTC. Despite the fact that the static representation provided by
 221 the figure does not allow a full appreciation of the diurnal cycle, the position of these
 222 percentiles varies significantly throughout the day. The tracer IWV signal generates a
 223 pattern that can be clearly seen when visualizing the data as a movie: [https://www.youtube](https://www.youtube.com/watch?v=sVP9B.85jfw)
 224 [.com/watch?v=sVP9B.85jfw](https://www.youtube.com/watch?v=sVP9B.85jfw). We can think of this signal as “the heartbeat of the Ama-
 225 zon”. Tracer water vapor is minimum in the early morning and maximum in the late af-
 226 ternoon and evening (Fig. 4g). The physical processes that give rise to characteristic sig-
 227 nals in integrated water vapor are due to the sources and sinks of atmospheric moisture:
 228 precipitation, evapotranspiration and moisture advection. In the analysis that follows
 229 we will focus on each of these sources and sinks.

230 Total precipitation and recycled precipitation show the same characteristic time
 231 scales of variability: annual, 6-month and diurnal (Fig. 5). This implies that the same
 232 physical processes that give rise to total precipitation affect recycled precipitation. To-
 233 tal precipitation peaks during the warmer months (Nov-April), with a slight lull in Febru-
 234 ary as the ITCZ is located south and the northern Amazon has a decrease in precipi-
 235 tation (Fig. 5d and Eiras-Barca et al. (2020) Fig. 4). The WRF annual cycle of precipi-
 236 tation coincides with that of TRMM estimates. Tracer precipitation shows a strong an-
 237 nual cycle that was not clear for IVW (compare Figure 5c and 4c). Note that the an-
 238 nual and semi-annual cycle in precipitation has also been detected in the tropical And-
 239 es mountains (Segura, 2019), in particular, the authors find a strong semi-annual cy-
 240 cle along the transition zone between the southern and equatorial Andes.

241 Total precipitation shows a much stronger diurnal cycle than IWV (compare Fig
 242 5b and 4b). The diurnal cycle of precipitation in the Amazon is well known. Hourly sta-
 243 tion data from the Manaus area shows that precipitation frequency peaks in the after-
 244 noon between 14 and 17 local time, and this agrees with remote sensing estimates (Tanaka
 245 et al., 2014). It is important to highlight, however, that the diurnal peak depends on the
 246 type of precipitation. Late afternoon and early evening peaks are associated with non-
 247 mesoscale convective system (MCS) precipitation, while MCS precipitation tends to peak
 248 in the early morning (Wu & Lee, 2019). When looking at the overall convective precipi-
 249 tation in the Amazon, regardless of type, we see a peak in the late afternoon and evening,
 250 as shown in the TRMM estimates (5f). Precipitation derived from TRMM shows a strong
 251 peak between 18-21Z (14-17 local). WRF total and tracer precipitation shows a simi-
 252 lar afternoon peak, but sustains the precipitation until around 0z (8pm) which is not seen
 253 in the remote sensing estimates.

254 IVT combines the effect of winds and water vapor (Fig. 6). We see that total IVT
 255 is dominated by the annual cycle, as warm season IVT is larger than during the cooler
 256 months. This coincides with previous results of Satyamurty et al. (2013) who found that
 257 moisture convergence in the basin accounts for most of the rainy season precipitation.
 258 High IVT also coincides with the intensification of the South American low-level jet (SALLJ),
 259 which transports moisture from the tropics into higher latitudes (Salio et al., 2002; Ar-
 260 raut et al., 2012). Interestingly total IVT shows a stronger diurnal cycle than IWV (Com-
 261 pare Figs 6b and 4b). This highlights the strong diurnal cycle of winds, and its effect
 262 on IVT variability. The diurnal cycle of IVT is also closely related to the diurnal cycle
 263 of the SALLJ, as has been shown in observations (Vera et al., 2006). Tracer IVT is nois-
 264 ier, with several peaks in addition to the diurnal peak (Fig. 6c). Interestingly, tracer IVT
 265 shows intensification during the summer and winter months, indicating that despite lower
 266 tracer IWV in the winter, the winds play an important role in moisture transport (Fig.
 267 6d). The diurnal cycle shows weaker IVT between 18Z and 0Z (14-20 local time) as bound-
 268 ary layer turbulence weakens horizontal winds. Higher IVT during the night and early
 269 morning would correspond to a stable boundary layer and higher wind speed. Tracer IVT
 270 also shows a similar diurnal cycle.

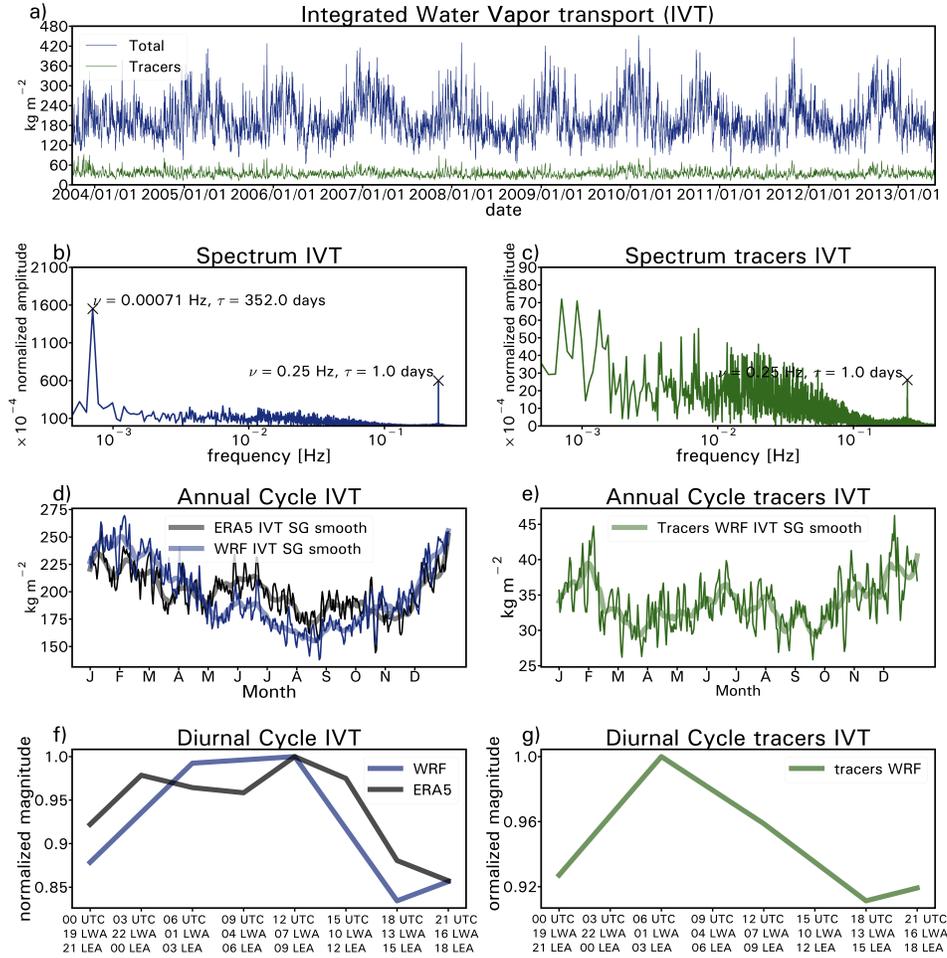


Figure 6. Fast-Fourier transform (FFT) analysis of IVT a) Raw signals of total IVT (blue) and tracers IVT (green) for a 6h-time step time series 2003-2013. b and c) FFT frequencies spectrum for total IVT (blue) and tracers IVT (green). d) Mean annual cycle for IVT as obtained from the WRF simulations (blue) and ERA5 reanalysis (black). e) Mean annual cycle for tracers IVT as obtained from WRF. f) Mean diurnal cycle for IVT as obtained from the WRF simulations (blue) and ERA5 reanalysis (black). g) Mean diurnal cycle for tracers IVT as obtained from WRF.

271 The noisy tracer IVT signal stands in sharp contrast to the clear diurnal signal of
 272 the ET timeseries, highlighting the noisiness of the lower level winds. ET diurnal cycle
 273 is stronger than the annual cycle (Fig. 7 a and b). ET peaks during the warm season
 274 (November through March). It is important to highlight that the annual cycle in ET is
 275 dominated by the signal in the southern Amazon, as the northern Amazon has weak annual
 276 variability (Fig. 4) [eiras2020changes]. In fact ET in the northern Amazon is sus-
 277 tained during the winter months with greener forests during the dry season (Huete et
 278 al., 2006). The diurnal cycle of ET is strong, with high ET values between 18Z and 0Z
 279 (14-20 local), associated with enhanced plant photosynthetic activity and higher atmo-
 280 spheric evaporative demand, and negligible nocturnal evapotranspiration. Note that we
 281 do not show tracer ET because the tracer flux at the surface is equal to ET, so the signal
 282 is the same.

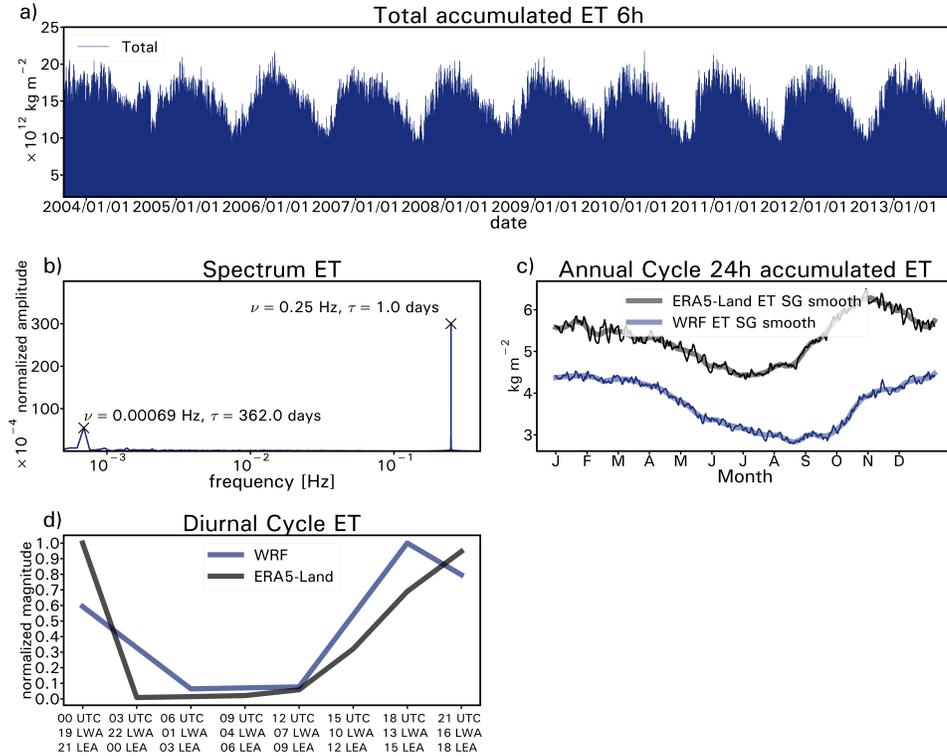


Figure 7. Fast-Fourier transform (FFT) analysis of IVT a) Raw signals of total ET for a 6h-time step time series 2003-2013. b) FFT frequencies spectrum for ET. c) Mean annual cycle for ET as obtained from the WRF simulations (blue) and ERA5-Land reanalysis (black). d) Mean diurnal cycle for IVT as obtained from the WRF simulations (blue) and ERA5-Land reanalysis (black).

4 Discussion and Conclusions

We use the WRF regional climate model with the added capability of water vapor tracers to track the moisture that evapotranspires from the Amazon basin and follow it in space and time until it rains out of the atmospheric column. This method allows us to quantify the amount of precipitable water, precipitation and integrated vapor transport that originates from Amazonian ET, without having to rely on simplifying assumptions of previous estimates. WRF-WVT allows us to analyze local moisture sources at shorter temporal and higher spatial scales than any other existing method. In addition to the annual and semi-annual cycle, our analysis revealed a clear diurnal cycle in all of the moisture budget variables. Our results show that:

- Precipitation of Amazonian origin (or recycled precipitation) has a strong zonal gradient with values gradually increasing towards the western part of the basin, following the prevailing wind patterns. Approximately half of the precipitation near the foothills of the Andes mountains is of Amazonian origin. Interestingly, this region of highest recycling in the Andean foothills is also one of the rainiest regions in the world due to exposure to easterly winds and orographic ascent mechanisms (Espinoza et al., 2015).
- Local evapotranspiration is more efficiently rained out of the column than remote moisture, particularly in the eastern part of the domain. This indicates that the

302 well mixed assumption used in most analytical recycling models, does not hold through-
 303 out the region. It also suggests an effective ascent mechanism in the eastern side
 304 of the domain for low-level moisture to contribute to precipitation.

- 305 • Climatologically, we find that nearly 30% of Amazonian precipitation comes from
 306 Amazonian evapotranspiration. This agrees with previous studies using a wide va-
 307 riety of models. The agreement is rather surprising given the simplifying assump-
 308 tions in previous estimates, and suggests a robust result: the recycling ratio in the
 309 Amazon basin is between 25-35%.
- 310 • Recycled precipitation shows a strong annual and 6-month signal. This is related
 311 to the ITCZ which passes once a year over the southern Amazon, the region that
 312 shows the strongest annual cycle. However, the ITCZ passes twice over the north-
 313 ern Amazon each year, once on its way south around October-November, and once
 314 on its way north around April-May (see Eiras-Barca et al. (2020), their Figure 4).
 315 ET, on the other hand, only shows a very weak annual cycle. This is related to
 316 the ability of the Amazonian forest, particularly in the north, to sustain ET dur-
 317 ing the dry season (da Rocha et al., 2009). Interestingly, tracer precipitable wa-
 318 ter shows a clear 6-month signal, unlike total IWV which has both an annual and
 319 6-month signal. This indicates that the high total IWV in December-March is pri-
 320 marily of oceanic origin, not of local origin, this can also be seen with the lower
 321 IWV recycling ratios during December-February than in other seasons.
- 322 • At the diurnal timescale, water vapor of Amazonian origin increases from early
 323 morning into the afternoon as evapotranspired moisture from the plants and soil
 324 accumulates in the atmospheric column, then some of this water is rained out through
 325 convective precipitation in the early evening. This agrees with observations of IWV
 326 and convection in Manaus that reveal that convective events are characterized by
 327 an 8 hour period of weak convergence, followed by a 4 hour period of intense con-
 328 vergence followed by a transition from shallow to deep convection (Adams et al.,
 329 2013). Later in the night, the water vapor is swept away by nocturnal winds as-
 330 sociated with the South American Low Level Jet. When visualized, the diurnal
 331 pattern of Amazonian IWV appears as a beating signal [https://www.youtube](https://www.youtube.com/watch?v=sVP9B_85jfw)
 332 [.com/watch?v=sVP9B_85jfw](https://www.youtube.com/watch?v=sVP9B_85jfw).

333 Our results imply that, compared to the traditional 8-10 day lifetime of water va-
 334 por as calculated by global analysis (van der Ent & Tuinenburg, 2017), or even the 4-
 335 5 day estimates of (Läderach & Sodemann, 2016) the lifetime of water vapor over the
 336 Amazon forest is much shorter. In fact, in global analyses the Amazon basin stands out
 337 as having some of the shortest lifetimes and length scales of terrestrial moisture on Earth
 338 (van der Ent et al., 2014), with average distance between transpiration and precipita-
 339 tion of about 600km (Staal et al., 2018). So the Amazon has younger atmospheric wa-
 340 ter than other places in the globe. The efficiency of recycling, short lifetime and length
 341 scale of Amazonian moisture also implies that ET of Amazonian origin is less likely to
 342 contribute to downwind precipitation than originally thought. In fact, once the air masses
 343 leave the Amazonian forest, studies have found an exponential decrease of precipitation
 344 with distance (Molina et al., 2019). Future studies should focus on observational vali-
 345 dation of these results using observations of water isotopes, with in-situ or remote sens-
 346 ing observations.

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362 References

- 363 Adams, D. K., Gutman, S. I., Holub, K. L., & Pereira, D. S. (2013, June). GNSS
 364 observations of deep convective time scales in the Amazon. *Geophys. Res.*
 365 *Lett.*, *40*(11), 2818–2823.
- 366 Arraut, J. M., Nobre, C., Barbosa, H. M., Obregon, G., & Marengo, J. (2012).
 367 Aerial rivers and lakes: looking at large-scale moisture transport and its re-
 368 lation to amazonia and to subtropical rainfall in south america. *Journal of*
 369 *Climate*, *25*(2), 543–556.
- 370 Bosilovich, M., & Chern, J. (2006). Simulation of water sources and precipitation
 371 recycling for the MacKenzie, Mississippi, and Amazon River basins. *Journal of*
 372 *Hydrometeorology*, *7*(3), 312–329.
- 373 Bosilovich, M. G., & Chern, J.-D. (2006). Simulation of water sources and precipita-
 374 tion recycling for the mackenzie, mississippi, and amazon river basins. *Journal*
 375 *of Hydrometeorology*, *7*(3), 312–329.
- 376 Brubaker, K. L., Entekhabi, D., & Eagleson, P. (1993). Estimation of continental
 377 precipitation recycling. *Journal of Climate*, *6*(6), 1077–1089.
- 378 Burde, G. I. (2006). Bulk recycling models with incomplete vertical mixing. part i:
 379 Conceptual framework and models. *Journal of climate*, *19*(8), 1461–1472.
- 380 da Rocha, H. R., Manzi, A. O., Cabral, O. M., Miller, S. D., Goulden, M. L.,
 381 Saleska, S. R., ... Maia, J. F. (2009, January). Patterns of water and heat
 382 flux across a biome gradient from tropical forest to savanna in Brazil. *J. Geo-*
 383 *phys. Res.*, *114*, G00B12.
- 384 Dee, D. P., Uppala, S. M., Simmons, A., Berrisford, P., Poli, P., Kobayashi, S., ...
 385 others (2011). The era-interim reanalysis: Configuration and performance of
 386 the data assimilation system. *Quarterly Journal of the royal meteorological*
 387 *society*, *137*(656), 553–597.
- 388 Dominguez, F., Hu, H., & Martinez, J. (2020). Two-layer dynamic recycling model
 389 (2l-drm): Learning from moisture tracking models of different complexity.
 390 *Journal of Hydrometeorology*, *21*(1), 3–16.
- 391 Dominguez, F., Kumar, P., Liang, X.-Z., & Ting, M. (2006). Impact of atmospheric
 392 moisture storage on precipitation recycling. *Journal of climate*, *19*(8), 1513–
 393 1530.
- 394 Dominguez, F., Miguez-Macho, G., & Hu, H. (2016). Wrf with water vapor tracers:
 395 A study of moisture sources for the north american monsoon. *Journal of Hy-*
 396 *drometeorology*, *17*(7), 1915–1927.
- 397 Drumond, A., Marengo, J., Ambrizzi, T., Nieto, R., Moreira, L., & Gimeno, L.
 398 (2014). The role of the Amazon Basin moisture in the atmospheric branch of
 399 the hydrological cycle: a Lagrangian analysis. *Hydrol Earth Syst Sc*, *18*(7),
 400 2577–2598.
- 401 Eiras-Barca, J., Dominguez, F., Hu, H., Garaboa-Paz, D., & Miguez-Macho, G.
 402 (2017). Evaluation of the moisture sources in two extreme landfalling atmo-
 403 spheric river events using an eulerian wrf tracers tool. *Earth Syst. Dyn*, *8*(4),
 404 1247–1261.
- 405 Eiras-Barca, J., Dominguez, F., Yang, Z., Chug, D., Nieto, R., Gimeno, L., &
 406 Miguez-Macho, G. (2020). Changes in south american hydroclimate under

- 407 projected amazonian deforestation. *Annals of the New York Academy of Sci-*
 408 *ences*.
- 409 Eltahir, E. A., & Bras, R. L. (1994). Precipitation recycling in the amazon basin.
 410 *Quarterly Journal of the Royal Meteorological Society*, *120*(518), 861–880.
- 411 Espinoza, J. C., Chavez, S., Ronchail, J., Junquas, C., Takahashi, K., & Lavado,
 412 W. (2015, May). Rainfall hotspots over the southern tropical Andes: Spa-
 413 tial distribution, rainfall intensity, and relations with large-scale atmospheric
 414 circulation. *Water Resour Res*, *51*(5), 3459–3475.
- 415 Fan, Y., Miguez-Macho, G., Weaver, C. P., Walko, R., & Robock, A. (2007). Incor-
 416 porating water table dynamics in climate modeling: 1. water table observations
 417 and equilibrium water table simulations. *Journal of Geophysical Research:*
 418 *Atmospheres*, *112*(D10).
- 419 Hong, S.-Y., & Lim, J.-O. J. (2006). The wrf single-moment 6-class microphysics
 420 scheme (wsm6). *Asia-Pacific Journal of Atmospheric Sciences*, *42*(2), 129–
 421 151.
- 422 Hong, S.-Y., & Pan, H.-L. (1996). Nonlocal boundary layer vertical diffusion in a
 423 medium-range forecast model. *Monthly weather review*, *124*(10), 2322–2339.
- 424 Huete, A. R., Didan, K., Shimabukuro, Y. E., Ratana, P., Saleska, S. R., Hutyrá,
 425 L. R., . . . Myneni, R. (2006). Amazon rainforests green-up with sunlight in
 426 dry season. *Geophys. Res. Lett.*, *33*(6), L06405–4.
- 427 Huffman, G. J., Adler, R. F., Bolvin, D. T., & Nelkin, E. J. (2010). The trmm
 428 multi-satellite precipitation analysis (tmpa). In *Satellite rainfall applications*
 429 *for surface hydrology* (pp. 3–22). Springer.
- 430 Insua-Costa, D., & Miguez-Macho, G. (2018). A new moisture tagging capability
 431 in the weather research and forecasting model: formulation, validation and
 432 application to the 2014 great lake-effect snowstorm. *Earth System Dynamics*,
 433 *9*(1), 167.
- 434 Kain, J. S. (2004). The kain–fritsch convective parameterization: an update. *Journal*
 435 *of applied meteorology*, *43*(1), 170–181.
- 436 Kain, J. S., & Fritsch, J. M. (1990). A one-dimensional entraining/detraining plume
 437 model and its application in convective parameterization. *Journal of the Atmo-*
 438 *spheric Sciences*, *47*(23), 2784–2802.
- 439 Laderach, A., & Sodemann, H. (2016). A revised picture of the atmospheric mois-
 440 ture residence time. *Geophys. Res. Lett.*, *43*, 924–933.
- 441 Läderach, A., & Sodemann, H. (2016, February). A revised picture of the atmo-
 442 spheric moisture residence time. , 1–10.
- 443 Lettau, H., Lettau, K., & Molion, L. C. B. (1979). Amazonia’s hydrologic cycle
 444 and the role of atmospheric recycling in assessing deforestation effects. *Monthly*
 445 *Weather Review*, *107*(3), 227–238.
- 446 Martínez, J. A., Dominguez, F., & Miguez-Macho, G. (2016a). Effects of a ground-
 447 water scheme on the simulation of soil moisture and evapotranspiration over
 448 southern south america. *Journal of Hydrometeorology*, *17*(11), 2941–2957.
- 449 Martínez, J. A., Dominguez, F., & Miguez-Macho, G. (2016b). Impacts of a ground-
 450 water scheme on hydroclimatological conditions over southern south america.
 451 *Journal of Hydrometeorology*, *17*(11), 2959–2978.
- 452 Miguez-Macho, G., & Fan, Y. (2012). The role of groundwater in the amazon wa-
 453 ter cycle: 1. influence on seasonal streamflow, flooding and wetlands. *Journal*
 454 *of Geophysical Research: Atmospheres*, *117*(D15).
- 455 Miguez-Macho, G., Fan, Y., Weaver, C. P., Walko, R., & Robock, A. (2007). Incor-
 456 porating water table dynamics in climate modeling: 2. formulation, validation,
 457 and soil moisture simulation. *Journal of Geophysical Research: Atmospheres*,
 458 *112*(D13).
- 459 Molina, R. D., Salazar, J. F., Martínez, J. A., Villegas, J. C., & Arias, P. A. (2019,
 460 March). Forest-Induced Exponential Growth of Precipitation Along Climato-
 461 logical Wind Streamlines Over the Amazon. *J Geophys Res-Atmos*, *124*(5),

- 2589–2599.
- 462 Niu, G.-Y., Yang, Z.-L., Mitchell, K. E., Chen, F., Ek, M. B., Barlage, M., . . . oth-
 463 ers (2011). The community noah land surface model with multiparameteriza-
 464 tion options (noah-mp): 1. model description and evaluation with local-scale
 465 measurements. *Journal of Geophysical Research: Atmospheres*, *116*(D12).
- 466 Salati, E., Dall’Olio, A., Matsui, E., & Gat, J. R. (1979). Recycling of water in the
 467 amazon basin: an isotopic study. *Water resources research*, *15*(5), 1250–1258.
- 468 Salio, P., Nicolini, M., & Saulo, A. (2002). Chaco low-level jet events characteriza-
 469 tion during the austral summer season. *Journal of Geophysical Research: At-
 470 mospheres*, *107*(D24), ACL–32.
- 471 Satyamurty, P., da Costa, C. P. W., & Manzi, A. O. (2013). Moisture source for the
 472 amazon basin: a study of contrasting years. *Theoretical and Applied Climatol-
 473 ogy*, *111*(1-2), 195–209.
- 474 Segura, H. (2019, June). New insights into the rainfall variability in the tropical An-
 475 des on seasonal and interannual time scales. *Clim Dynam*, *53*(1), 405–426.
- 476 Staal, A., Tuinenburg, O. A., Bosmans, J. H. C., Holmgren, M., van Nes, E. H.,
 477 Scheffer, M., . . . Dekker, S. C. (2018, May). Forest-rainfall cascades buffer
 478 against drought across the Amazon. *Nature Publishing Group*, 1–8.
- 479 Tanaka, L. d. S., Satyamurty, P., & Machado, L. A. T. (2014). Diurnal variation
 480 of precipitation in central a mazon b asin. *International journal of climatology*,
 481 *34*(13), 3574–3584.
- 482 Trenberth, K. E. (1999, May). Atmospheric moisture recycling: Role of advection
 483 and local evaporation. *J. Climate*, *12*(5), 1368–1381.
- 484 Tuinenburg, O., & van der Ent, R. (2019). Land surface processes create patterns in
 485 atmospheric residence time of water. *Journal of Geophysical Research: Atmo-
 486 spheres*, *124*(2), 583–600.
- 487 van der Ent, R. J., Savenije, H. H. G., Schaeffli, B., & Steele-Dunne, S. C. (2010).
 488 Origin and fate of atmospheric moisture over continents. *Water Resour Res*,
 489 *46*, W09525.
- 490 van der Ent, R. J., & Tuinenburg, O. A. (2017). The residence time of water in the
 491 atmosphere revisited. *Hydrol. Earth Syst. Sci.*, *21*(2), 779–790.
- 492 van der Ent, R. J., Wang-Erlandsson, L., Keys, P. W., & Savenije, H. H. G. (2014).
 493 Contrasting roles of interception and transpiration in the hydrological cycle –
 494 Part 2: Moisture recycling. *Earth Syst. Dynam.*, *5*(2), 471–489.
- 495 Vera, C., Baez, J., Douglas, M., Emmanuel, C., Marengo, J., Meitin, J., . . . others
 496 (2006). The south american low-level jet experiment. *Bulletin of the American
 497 Meteorological Society*, *87*(1), 63–78.
- 498 Wilks, D. S. (2011). *Statistical methods in the atmospheric sciences* (Vol. 100). Aca-
 499 demic press.
- 500 Wu, M., & Lee, J. E. (2019, August). Thresholds for Atmospheric Convection in
 501 Amazonian Rainforests. *Geophys. Res. Lett.*, *46*(16), 10024–10033.
- 502 Yang, Z., & Dominguez, F. (2019). Investigating land surface effects on the mois-
 503 ture transport over south america with a moisture tagging model. *Journal of
 504 Climate*, *32*(19), 6627–6644.
- 505 Zemp, D. C., Schleussner, C. F., Barbosa, H. M. J., van der Ent, R. J., Donges,
 506 J. F., Heinke, J., . . . Rammig, A. (2014). On the importance of cascading
 507 moisture recycling in South America. *Atmos. Chem. Phys.*, *14*(23), 13337–
 508 13359.
- 509