# Amazonian Moisture Recycling Revisited Using WRF with Water Vapor Tracers

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November 16, 2022

#### 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.







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#### Key Points:

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12	•	30% of Amazonian precipitation comes from local evapotranspiration. Local mois-
13		ture is more efficiently rained out than remote moisture.
14	•	Recycled precipitation shows a strong annual and semi-annual signal, associated
15		with the passage of the Inter-Tropical Convergence Zone.
16	•	A diurnal cycle cycle of Amazonian water vapor reflects the variability of evap-
17		otranspiration, recycled precipitation and vapor flux.

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#### 18 Abstract

Previous studies have estimated that 25% to 35% of Amazonian precipitation comes from 19 evapotranspiration (ET) within the basin. However, due to simplifying assumptions of 20 traditional models, these studies primarily focus on large spatial and temporal scales. 21 In this work we use the Weather Research and Forecast (WRF) regional climate model 22 with the added capability of water vapor tracers to track the moisture from Amazonian 23 ET at the native WRF resolution. The tracers reveal that the well-mixed assumption 24 of simpler models does not hold, as local ET is more efficiently rained out of the atmo-25 spheric column than remote sources of moisture, particularly in the eastern part of the 26 basin. Recycled precipitation shows a strong annual and semi-annual signal, associated 27 with the passage of the Inter-Tropical Convergence Zone. The tracers also reveal a strong 28 diurnal cycle of Amazonian water vapor related to the diurnal cycle of ET, convective 20 precipitation and advected moisture. ET increases from early morning into the afternoon, 30 some of this moisture is rained out through convective storms in the early evening, while 31 later in the night strong winds associated with the South American Low Level Jet ad-32 vect moisture downwind. Visualizing the Amazonian water vapor highlights its diurnal 33 beating pattern and suggests that the Amazon has "younger" water than other regions 34 in the globe, with very efficient recycling of local moisture. 35

#### <sup>36</sup> Plain Language Summary

Evaporation from soil and transpiration from plants within the Amazon contribute 37 to approximately one third of the precipitation that falls within the basin in a process 38 known as "precipitation recycling". This estimate represents an average over the basin 39 and over many years. In this work we use numerical water tracers within an atmospheric 40 model to quantify precipitation recycling at higher spatial and temporal resolution than 41 previous studies. The tracers allow us to follow the water from the time it evaporates 42 from the land until it falls as precipitation. Our work reveals cycles in water vapor and 43 precipitation of Amazonian origin that had not been previously studied. In particular, 44 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 46 the atmosphere. Some of this moisture is rained out in the early evening in convective 47 storms, while later in the night strong winds transport moisture away from the basin. 48 Visualizing the Amazonian water vapor highlights its diurnal beating pattern. 49

#### 50 1 Introduction

Early evidence of the importance Amazonian evapotranspiration (ET) for local pre-51 cipitation came from observational analysis of oxygen-18 ( $\delta^{18}$ O) in precipitation, which 52 was found to have an inland gradient of  $\delta^{18}$ O much smaller than in other regions of the 53 world (Salati et al., 1979). This observational result suggested that a significant part of 54 the rainfall came from re-evaporated water, otherwise known as recycled precipitation. 55 Subsequent analyses based on bulk numerical models estimated precipitation recycling 56 (or the percent of precipitation that comes from local ET) ranging from 25%-40% (Brubaker 57 et al., 1993; Eltahir & Bras, 1994; Burde, 2006). There is a strong gradient in recycled 58 precipitation, increasing from east to southwest as the dominant flow enters from the At-59 lantic Ocean, traverses the basin and encounters the Andes mountains (Eltahir & Bras, 60 1994; Burde, 2006). It is important to keep in mind that these early bulk models assume 61 that the atmospheric column is well mixed. In other words, they assume that, at the time 62 scales of the model, turbulence effectively mixes the atmosphere so the proportion of pre-63 cipitation 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 65 the Amazon it is important to account for fast recycling as diurnal convection yields pre-66 cipitation before all water vapor from the surface has enough time to mix with the ex-67



Figure 1. Shading indicates integrated water vapor IWV ( kg m<sup>-2</sup>) of Amazonian origin. Black vectors indicate total integrated vapor transport IVT ( kg m<sup>-1</sup>s<sup>-1</sup>). Values are average for DJF (a) and JJA (b), for the period 2003-2013.

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

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

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



**Figure 2.** Row 1) Fraction of Amazonian precipitation to total precipitation ( $\rho_P$  =  $rain_t/rain$ ) 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" ( $\rho_P/\rho_{IWV}$ ), or row 1 divided by row 2.

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

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

#### 114 2 Methods

A 10-year simulation (2003-2013) provided by the mesoscale model WRF (20 km)115  $\times$  20 km) using the entire South American continent as domain of simulation is used here. 116 Water vapor tracers are only available for one set of physics options. As such, the Kain-117 Fritsch (KF) (Kain & Fritsch, 1990; Kain, 2004) parametrization is used for subgrid-scale 118 convection; the Yonsei University (YSU) (Hong & Pan, 1996) parametrization for tur-119 bulent mixing, and the single moment "6-class" (WSM6) (Hong & Lim, 2006) solved the 120 microphysics in phase change and precipitation processes. ERA-Interim (Dee et al., 2011) 121 122 provides boundary and initial conditions for the simulations.

Two additional tools are used in the WRF simulation to improve the representa-123 tion of land-atmosphere interactions, as well as provide the ability to track the fate of 124 the Amazonian ET. We used the Noah-Multiparametrization land surface model (Noah-125 MP LSM) (Niu et al., 2011) with the MMF groundwater scheme developed by Fan et 126 al. (2007) and Miguez-Macho et al. (2007). This scheme is used to take into account the 127 interaction between the shallow aquifers and soil moisture, which affects Amazonian ET, 128 particularly in water limited regions during dry periods (Miguez-Macho & Fan, 2012; Mar-129 tinez et al., 2016b, 2016a). 130

We also use the WRF Water Vapor Tracer Tool (WRF-VT) which allows the track-131 ing of the moisture evapotranspired from the Amazon (Dominguez et al., 2016; Eiras-132 Barca et al., 2017; Insua-Costa & Miguez-Macho, 2018). This moisture can either leave 133 the Amazonian domain or precipitate within the basin (as recycled precipitation). WRF-134 WVT includes additional output variables related to water vapor and precipitation of 135 tracer origin, as well as other species (see Insua-Costa and Miguez-Macho (2018) for de-136 tails). In this way, we can calculate the precipitation recycling ratio ( $\rho_P = rain_t/rain$ ) 137 where *rain* is the total precipitation which includes convective and non-convective pro-138 cesses, while  $rain_t$  is the precipitation from Amazonian ET. In a similar way, we can cal-139 culate the integrated water vapor (IWV) recycling ratio ( $\rho_{IWV} = IWV_t/IWV$ ). Note 140 that this calculation is done on a cell-by-cell basis. Also note that this method does not 141 require the assumption of a the well-mixed atmosphere in the vertical column that is tra-142 ditionally used in analytical models such as Brubaker et al. (1993) and Eltahir and Bras 143 (1994). This configuration of WRF was recently used by Eiras-Barca et al. (2020) to an-144



**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.

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

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

#### 156 **3 Results**

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

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

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**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 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.

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

#### 3.1 Temporal Analysis

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Using FFT analysis, we find that the characteristic timescale of total IWV has a very clear annual cycle, with a maximum during the Austral summer and a minimum 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.

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

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

shows the 50th and 80th percentile contours of daily mean tracers IWV corresponding 219 to 00, 06, 12 and 18 UTC. Despite the fact that the static representation provided by 220 the figure does not allow a full appreciation of the diurnal cycle, the position of these 221 percentiles varies significantly throughout the day. The tracer IWV signal generates a pattern that can be clearly seen when visualizing the data as a movie: https://www.youtube 223 .com/watch?v=sVP9B\_85jfw. We can think of this signal as "the heartbeat of the Ama-224 zon". Tracer water vapor is minimum in the early morning and maximum in the late af-225 ternoon and evening (Fig. 4g). The physical processes that give rise to characteristic sig-226 nals in integrated water vapor are due to the sources and sinks of atmospheric moisture: 227 precipitation, evapotranspiration and moisture advection. In the analysis that follows 228 we will focus on each of these sources and sinks. 229

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

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

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



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.

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



**Figure 7.** Fast-Fourier transform (FFT) analysis of IVT a) Raw signals of total ET for a 6htime 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).

#### <sup>283</sup> 4 Discussion and Conclusions

We use the WRF regional climate model with the added capability of water vapor 284 tracers to track the moisture that evapotranspires from the Amazon basin and follow it 285 in space and time until it rains out of the atmospheric column. This method allows us 286 to quantify the amount of precipitable water, precipitation and integrated vapor trans-287 port that originates from Amazonian ET, without having to rely on simplifying assump-288 tions of previous estimates. WRF-WVT allows us to analyze local moisture sources at 289 shorter temporal and higher spatial scales than any other existing method. In addition 290 to the annual and semi-annual cycle, our analysis revealed a clear diurnal cycle in all of 291 the moisture budget variables. Our results show that: 292

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

- Climatologically, we find that nearly 30% of Amazonian precipitation comes from Amazonian evapotranspiration. This agrees with previous studies using a wide variety of models. The agreement is rather surprising given the simplifying assumptions in previous estimates, and suggests a robust result: the recycling ratio in the Amazon basin is between 25-35%.
- Recycled precipitation shows a strong annual and 6-month signal. This is related 310 to the ITCZ which passes once a year over the southern Amazon, the region that 311 shows the strongest annual cycle. However, the ITCZ passes twice over the north-312 ern Amazon each year, once on its way south around October-November, and once 313 on its way north around April-May (see Eiras-Barca et al. (2020), their Figure 4) 314 ET, on the other hand, only shows a very weak annual cycle. This is related to 315 the ability of the Amazonian forest, particularly in the north, to sustain ET dur-316 ing the dry season (da Rocha et al., 2009). Interestingly, tracer precipitable wa-317 ter shows a clear 6-month signal, unlike total IWV which has both an annual and 318 6-month signal. This indicates that the high total IWV in December-March is pri-319 marily of oceanic origin, not of local origin, this can also be seen with the lower 320 IWV recycling ratios during December-February than in other seasons. 321
- At the diurnal timescale, water vapor of Amazonian origin increases from early 322 morning into the afternoon as evapotranspired moisture from the plants and soil 323 accumulates in the atmospheric column, then some of this water is rained out through 324 convective precipitation in the early evening. This agrees with observations of IWV 325 and convection in Manaus that reveal that convective events are characterized by 326 an 8 hour period of weak convergence, followed by a 4 hour period of intense con-327 vergence followed by a transition from shallow to deep convection (Adams et al., 328 2013). Later in the night, the water vapor is swept away by nocturnal winds as-329 sociated with the South American Low Level Jet. When visualized, the diurnal 330 pattern of Amazonian IWV appears as a beating signal https://www.youtube 331 .com/watch?v=sVP9B\_85jfw. 332

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

#### 347 Acknowledgments

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This research was supported by the National Science Foundation (NSF) CAREER Award
AGS 1454089. Visualization by David Bock, Lead Visualization Programmer, National
Center for Supercomputing Applications, University of Illinois, Champaign Urbana, IL.
Visualizations supported by XSEDE award ATM170030. L.G., R.N. and J.E.B were funded
by the Spanish government within the LAGRIMA (RTI2018-095772-B-I00) project, funded
by Ministerio de Ciencia, Innovacion y Universidades, Spain, which are also funded by

FEDER (European Regional Development Fund, ERDF). J.E.B was also supported by

the Xunta de Galicia (Galician Regional Government) under grant and by the Fulbright

Program (US Department of State). Z. Y was supported by the Office of Science of the

U.S. Department of Energy (DOE) as part of the Atmospheric System Research (ASR)

Program via Grant KP1701000/57131. All results from model simulations are available

as: Dominguez, Francina (2021): WRF model output with water vapor tracers over Ama-

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<sup>361</sup> B2IDB-8790412\_V1.

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