



PT-JPL EVAPOTRANSPIRATION

REMOTE SENSING DEVELOPMENTS AT
NASA'S JET PROPULSION LABORATORY

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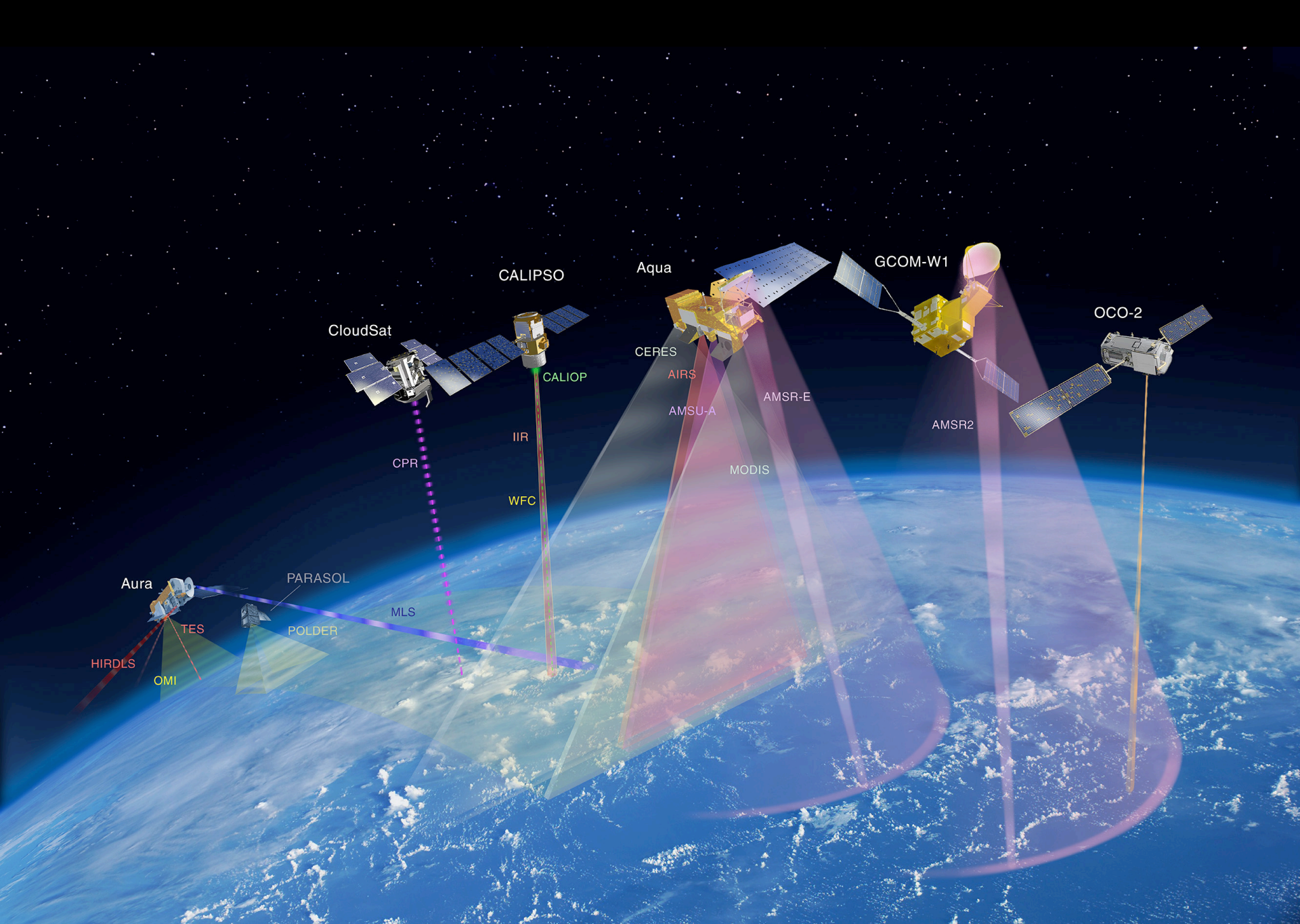
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Pasadena, California











JPL EVAPOTRANSPIRATION (PT-JPL)

Global estimates of the **land–atmosphere water flux** based on monthly AVHRR and ISLSCP-II data, validated at 16 FLUXNET sites

[JB Fisher](#), [KP Tu](#), [DD Baldocchi](#) - *Remote Sensing of Environment*, **2008** - Elsevier

Numerous models of evapotranspiration have been published that range in data-driven complexity, but global estimates require a model that does not depend on intensive field measurements. The Priestley–Taylor model is relatively simple, and has proven to be ...

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PT-JPLET

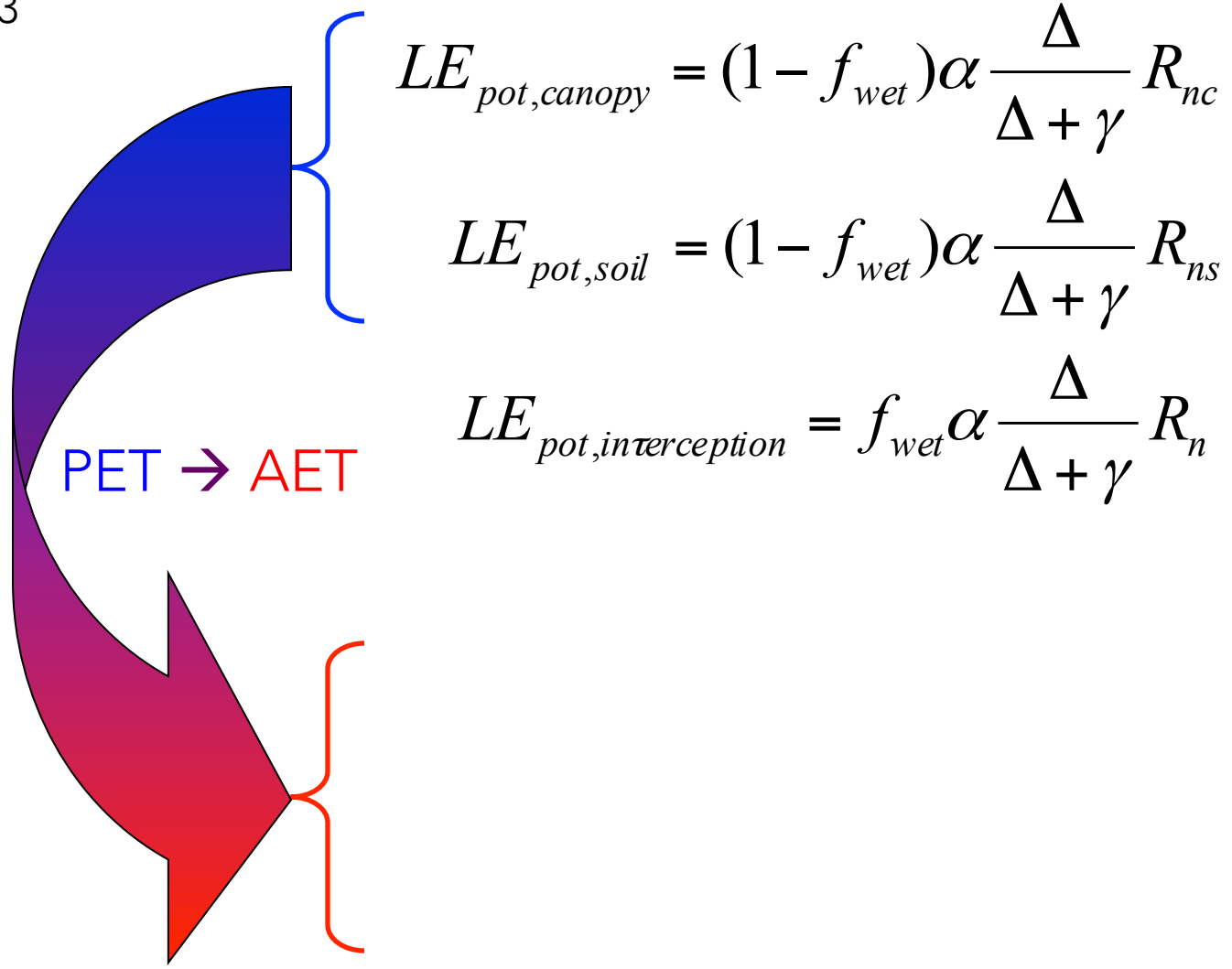
1/3

$$LE_{pot} = \alpha \frac{\Delta}{\Delta + \gamma} R_n$$

Priestley & Taylor (1972)

PT-JPL ET

2/3



PT-JPL ET

3/3

$$LE_c = (1 - f_{wet}) f_g f_T f_M \alpha \frac{\Delta}{\Delta + \gamma} R_{nc}$$

$$LE_s = (f_{wet} + f_{SM} (1 - f_{wet})) \alpha \frac{\Delta}{\Delta + \gamma} (R_{ns} - G)$$

1.0 7

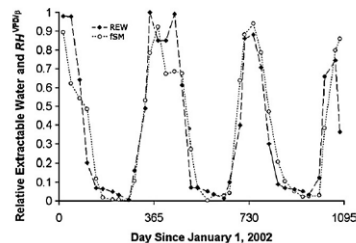
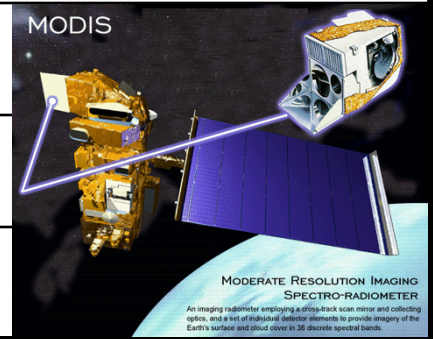


Fig. 1. Comparison of monthly f_{SM} to normalized volumetric water content (VWC), or relative extractable water – $REW = (VWC - VWC_{min}) / (VWC_{max} - VWC_{min})$ – at an oak-savanna site.

PT-JPL ET

$$LE = LE_s + LE_c + LE_i$$

<i>Net Radiation</i>	MODIS CERES SRB ISCCP
<i>Air Temperature</i>	MODIS AIRS
<i>Relative Humidity</i>	MODIS AIRS
<i>Vegetation fraction</i>	MODIS Landsat



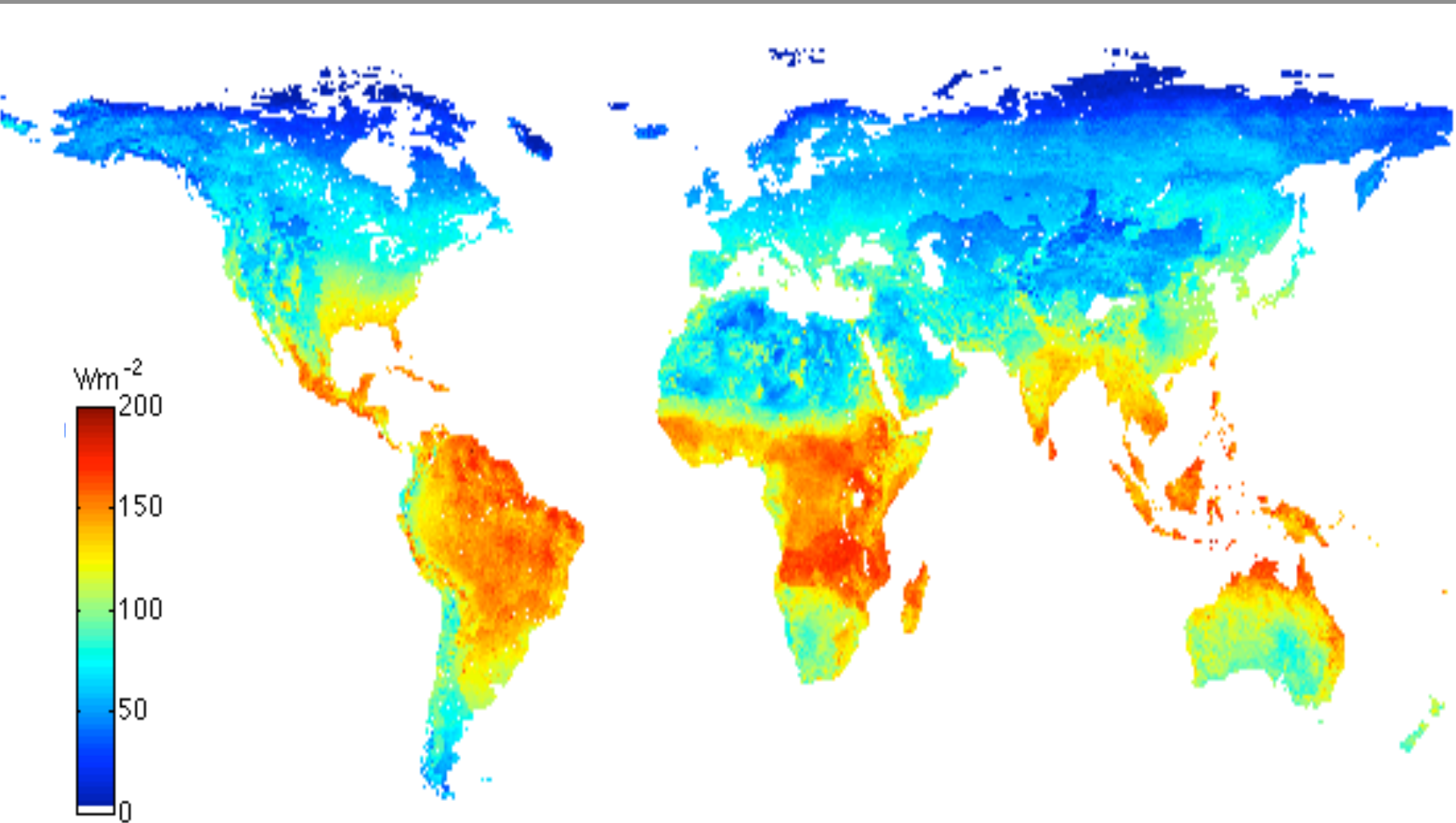
JPL MODIS R_n

$$R_n = (1 - \textit{albedo}) * SW_{dn} + LW_{dn} - LW_{up}$$

- We combined *11 variables* from *6 different MODIS products* daily over the MODIS era to estimate the components of R_n .

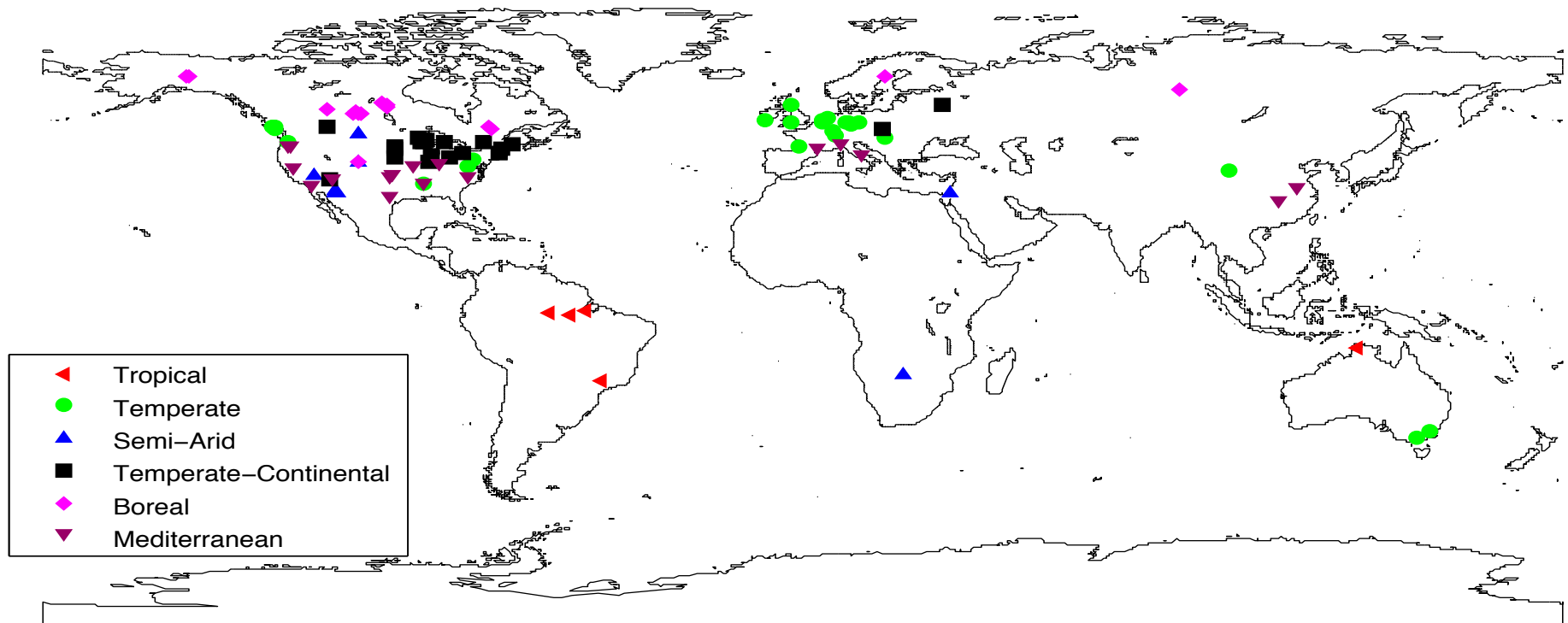
Component of R_n	MODIS products and method
Albedo	Albedo from MCD43 (500 m, 8-day), land cover from MCD12 (500 m, annual)
Incoming Shortwave (SW_{dn})	Cloud optical thickness, cloud top altitude, and solar zenith angle from MOD06 (5 km, daily); aerosol optical thickness at 550 nm from MOD04 (10 km, daily); Input MODIS data to a radiative transfer model (Kobayashi et al., 2008)
Incoming Longwave (LW_{dn})	Near surface air temperature and vapor pressure from MOD07 (5 km, daily); estimate emissivity from vapor pressure and temperature
Outgoing Longwave (LW_{up})	Land <i>surface temperature</i> and emissivity from MOD11 (1 km, daily); estimate broadband emissivity

JPL MODIS R_n

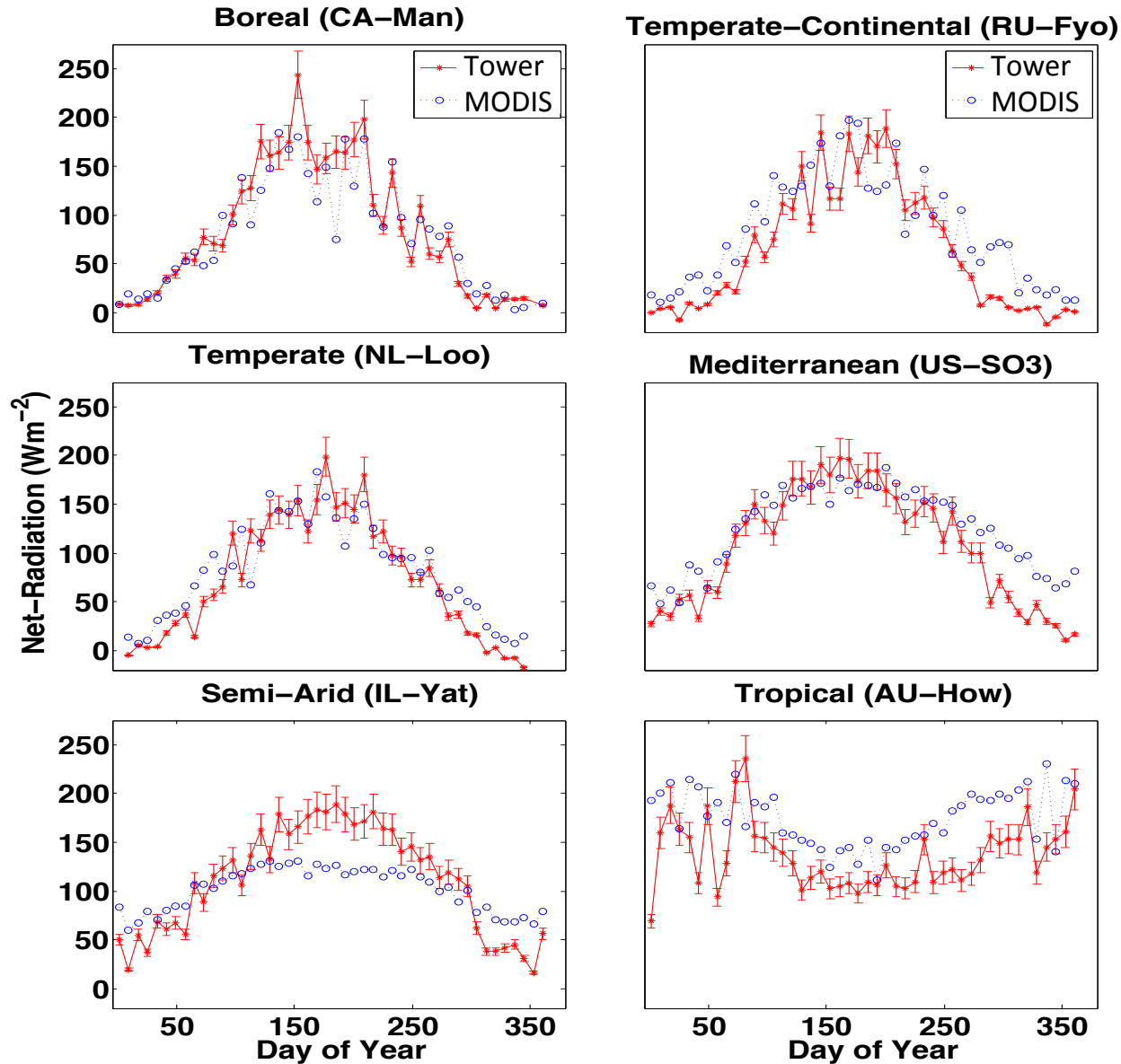


JPL MODIS R_n

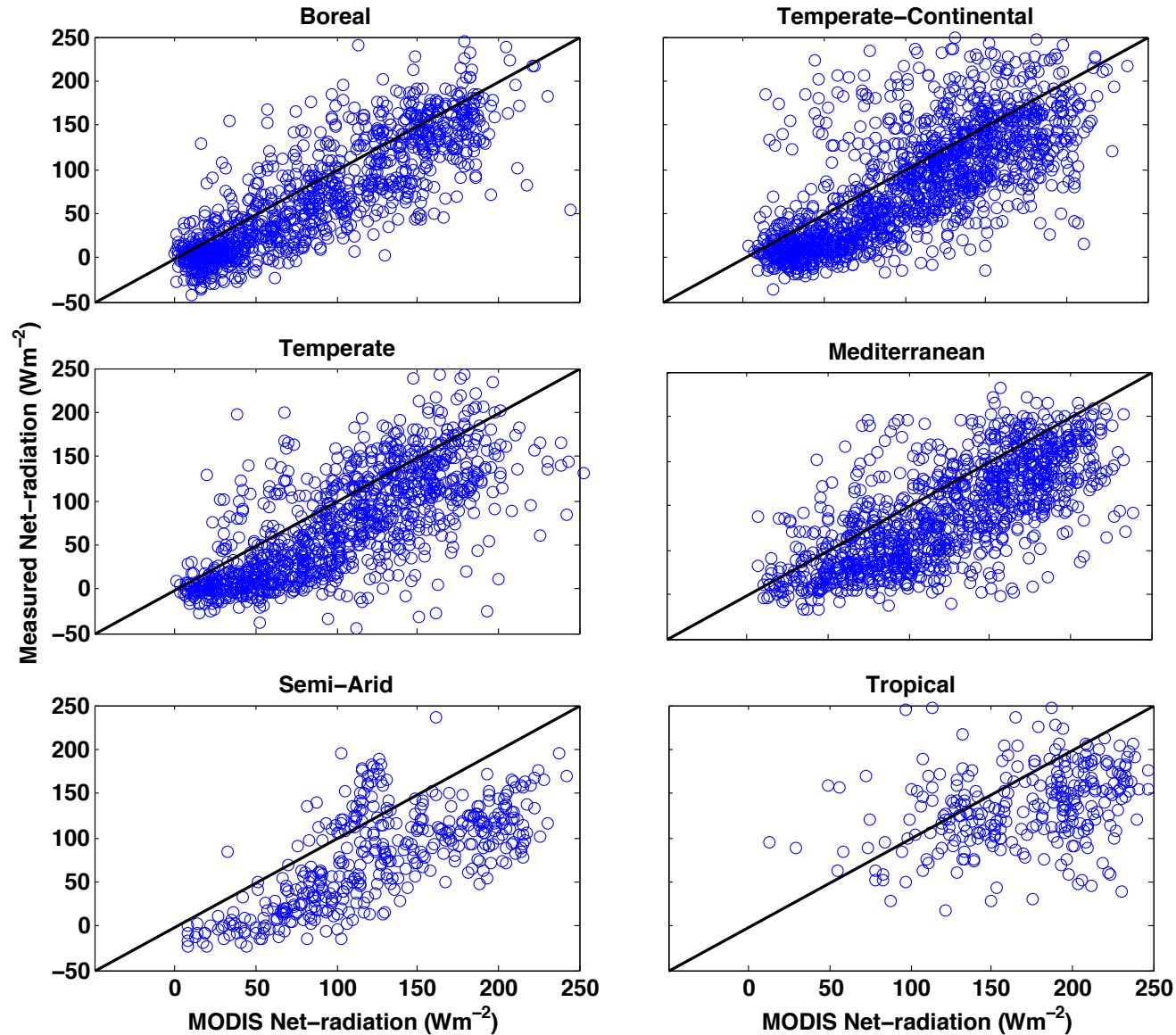
- Validation at 126 sites across FLUXNET and SURFRAD.

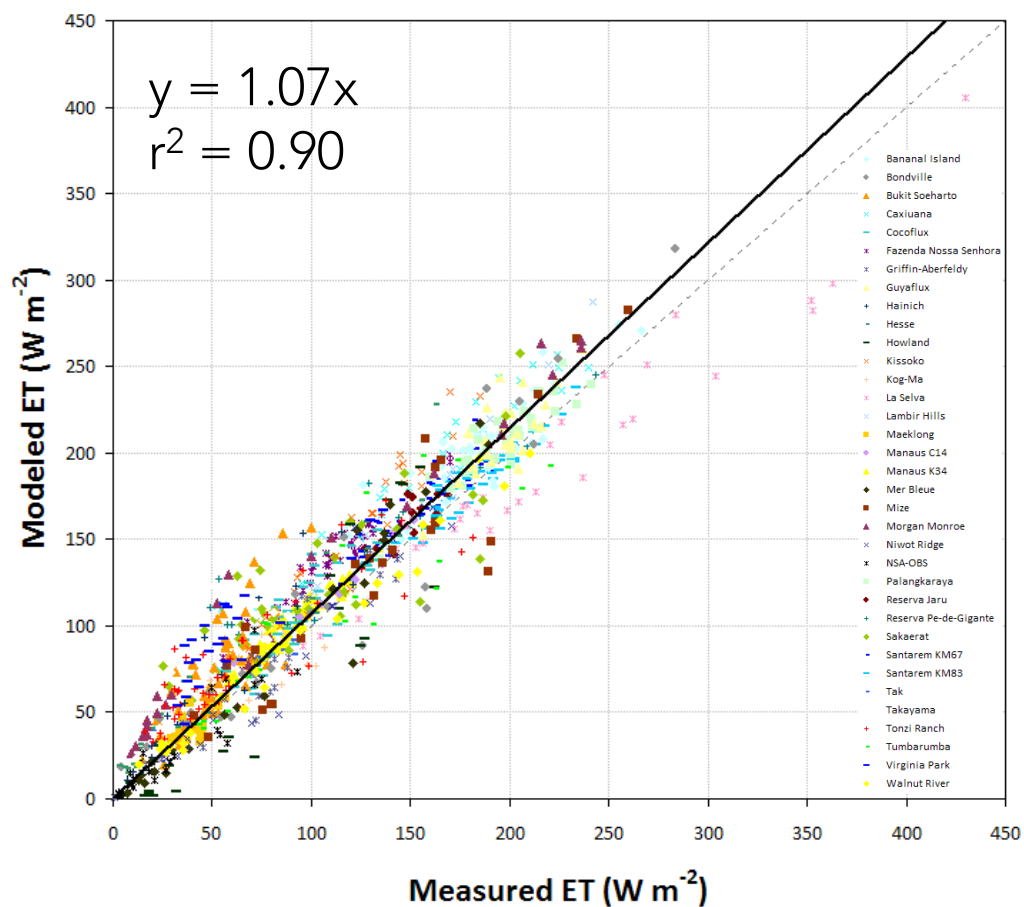


JPL MODIS R_n



JPL MODIS R_n

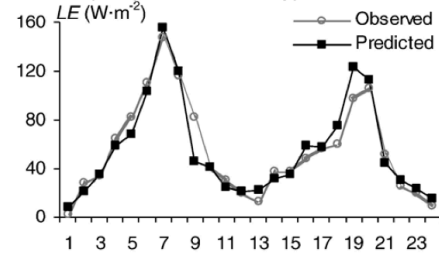




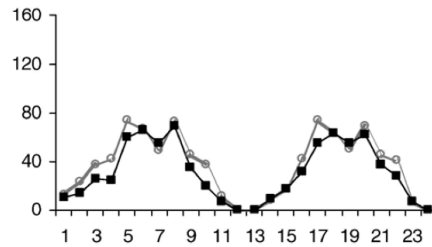
PT-JPL ET VALIDATION

PT-JPL ET VALIDATION

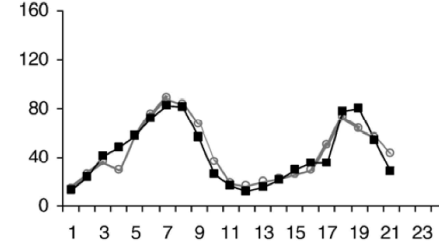
Bondville (Temperate C3/C4 Crop)



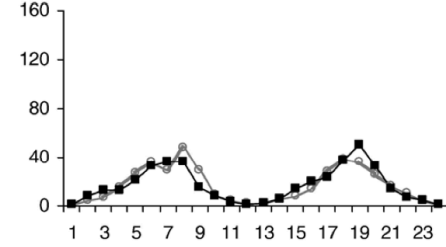
Griffin (Temperate Evergreen Needleleaf Forest)



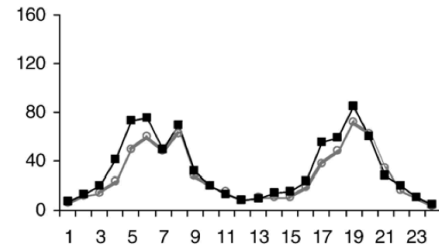
Niwot (Sub-Alpine Evergreen Needleleaf Forest)



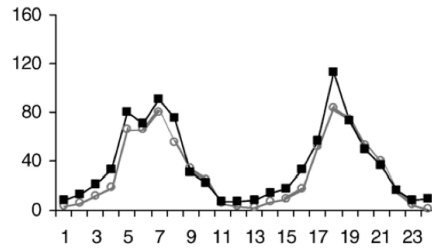
NSA-OBS (Boreal Evergreen Needleleaf Forest)



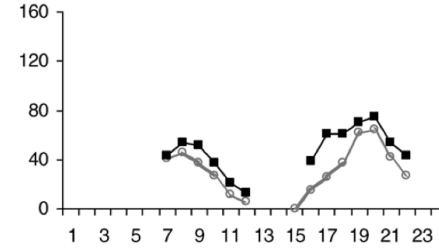
Hainich (Temperate Deciduous Broadleaf Forest)



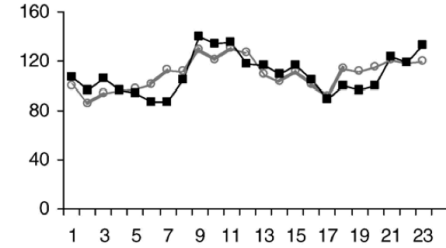
Hesse (Temperate Deciduous Broadleaf Forest)



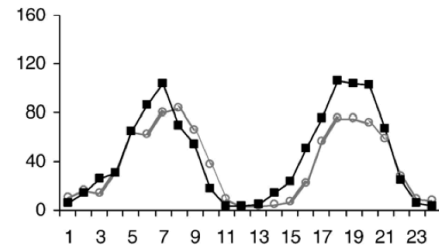
Takayama (Cold-Temperate Deciduous Broadleaf Forest)



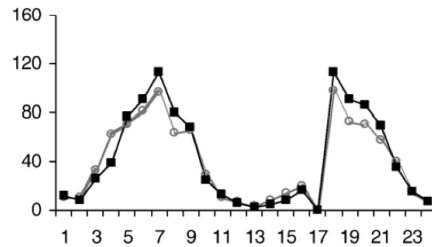
Tapajos (Tropical Evergreen Broadleaf Forest)



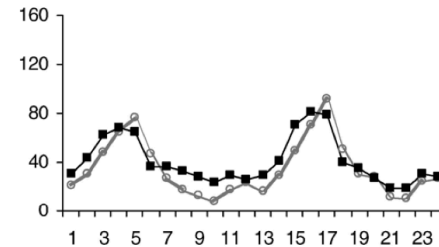
Howland (Cold-Temperate Evergreen Needleleaf Forest)



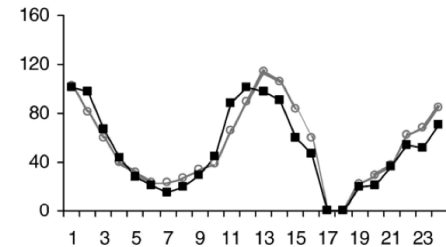
Mer Bleue (Boreal Wetland)



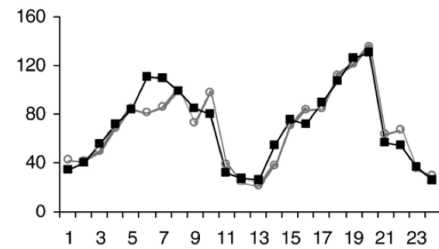
Tonzi (Mediterranean Savanna)



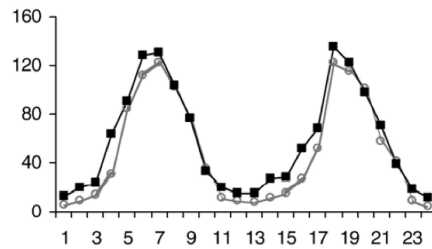
Tumbarumba (Temperate Evergreen Broadleaf Forest)



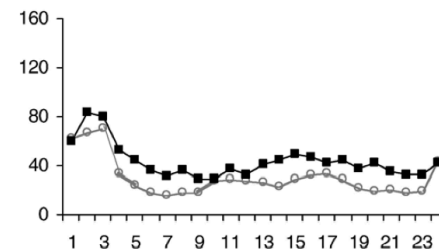
Mize (Subtropical Evergreen Needleleaf Forest)



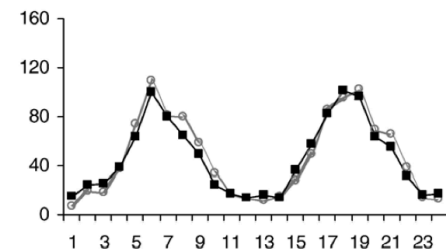
Morgan Monroe (Temperate Deciduous Broadleaf Forest)



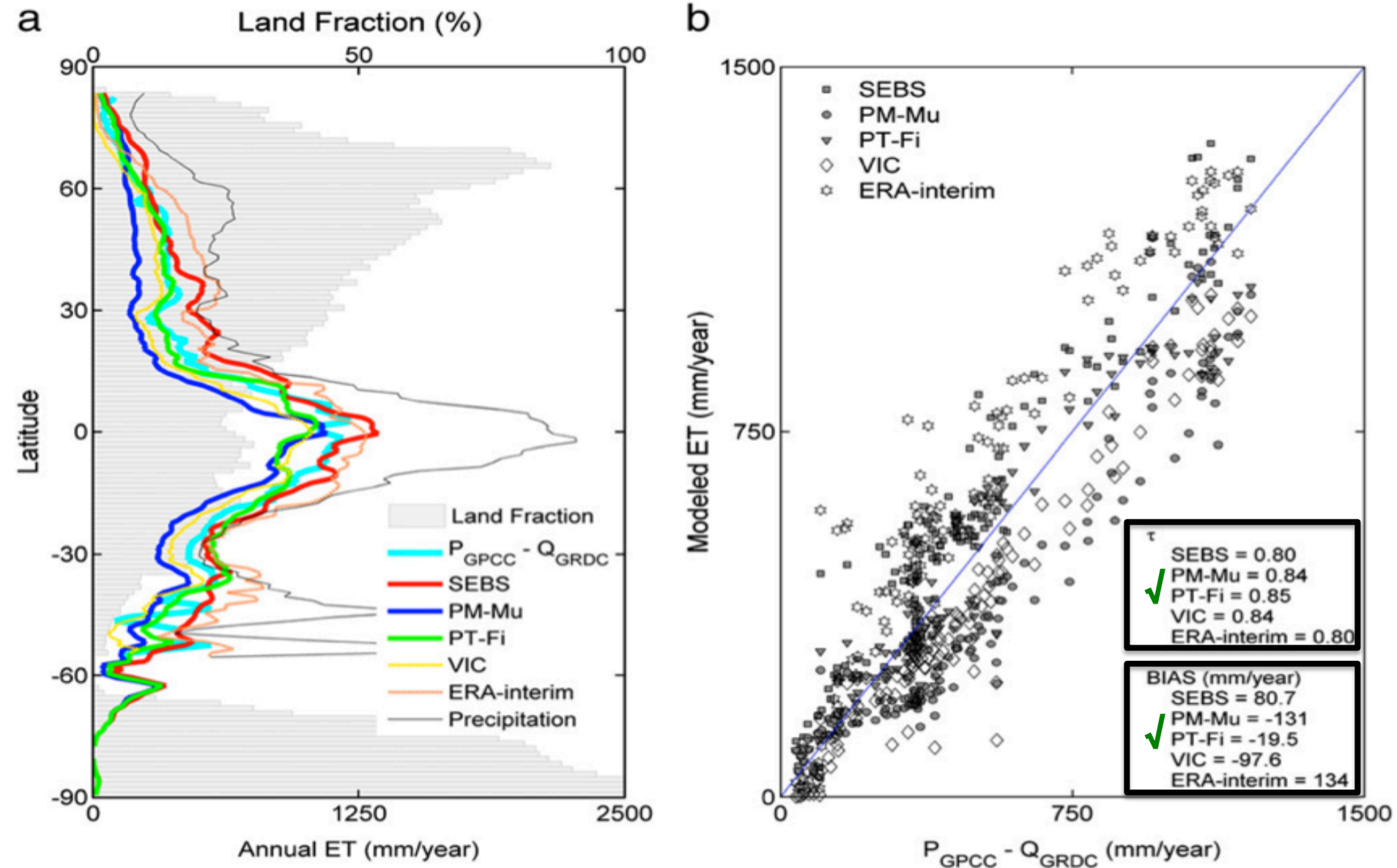
Virginia Park (Woody Savanna)



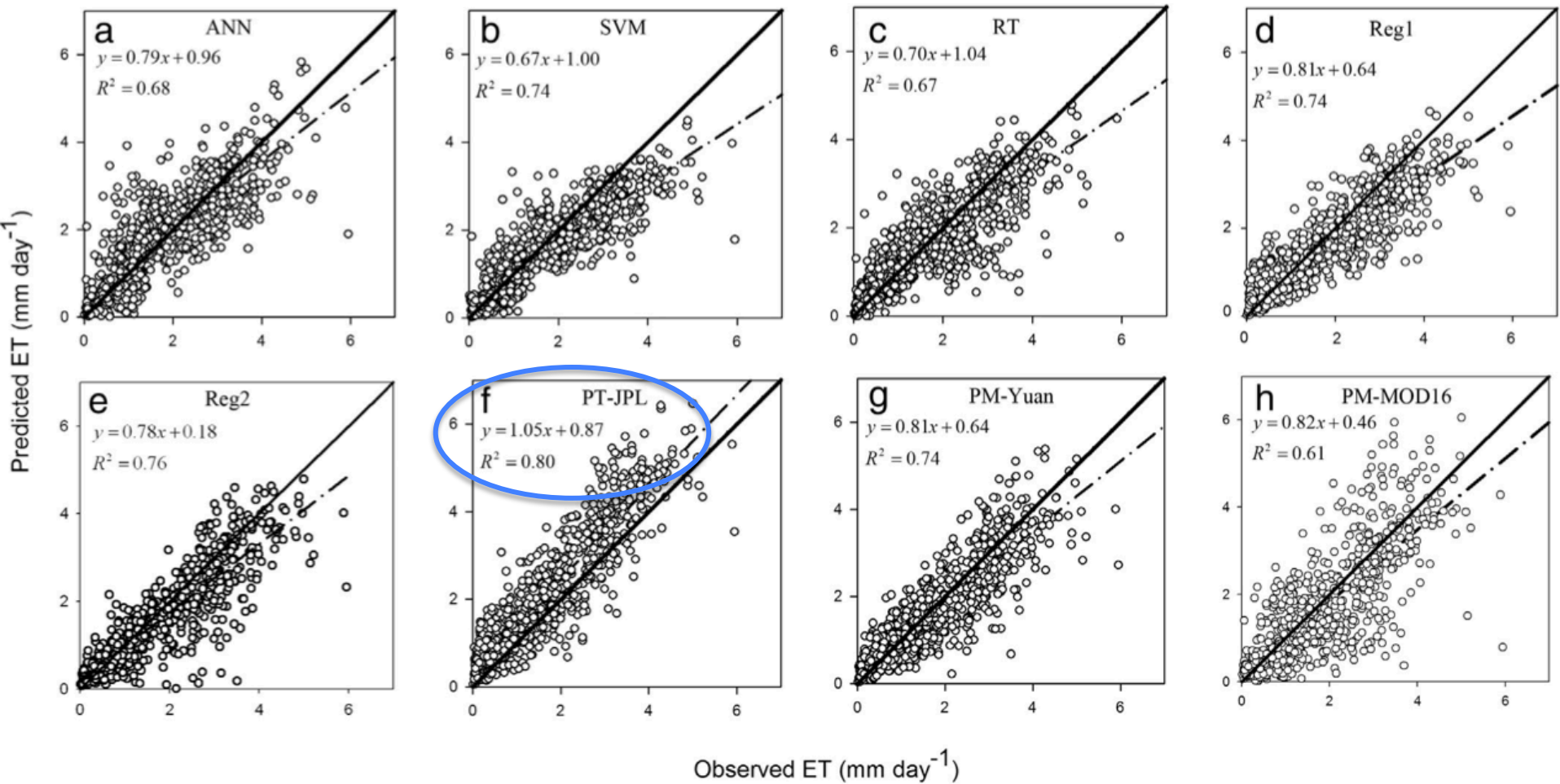
Walnut River (Temperate C3/C4 Grassland)



INDEPENDENT EVALUATION (GEWEX/PRINCETON)



INDEPENDENT EVALUATION (BEIJING)



INDEPENDENT EVALUATION (GEWEX/AUSTRALIA)

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Multi-site evaluation of terrestrial evaporation models using FLUXNET data

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ABSTRACT

We evaluated the performance of four commonly applied land surface evaporation models using a high-quality dataset of selected FLUXNET towers. The models that were examined include an energy balance approach (Surface Energy Balance System; SEBS), a combination-type technique (single-source Penman–Monteith; PM), a complementary method (advection–aridity; AA) and a radiation based approach (modified Priestley–Taylor; PT-JPL). Twenty FLUXNET towers were selected based upon satisfying stringent forcing data requirements and representing a wide range of biomes. These towers encompassed a number of grassland, cropland, shrubland, evergreen needleleaf forest and deciduous broadleaf forest sites. Based on the mean value of the Nash–Sutcliffe efficiency (NSE) and the root mean squared difference (RMSD), the order of overall performance of the models from best to worst were: ensemble mean of models (0.61, 64), PT-JPL (0.59, 66), SEBS (0.42, 84), PM (0.26, 105) and AA (0.18, 105) [statistics stated as (NSE, RMSD in W m^{-2})]. Although PT-JPL uses a relatively simple and largely empirical formulation of the evaporative process, the technique showed improved performance compared to PM, possibly due to its partitioning of total evaporation (canopy transpiration, soil evaporation, wet canopy evaporation) and lower uncertainties in the required forcing data. The SEBS model showed low performance over tall and heterogeneous canopies, which was likely a consequence of the effects of the roughness sub-layer parameterization employed in this scheme. However, SEBS performed well overall. Relative to PT-JPL and SEBS, the PM and AA showed low performance over the majority of sites, due to their sensitivity to the parameterization of resistances. Importantly, it should be noted that no single model was consistently best across all biomes. Indeed, this outcome highlights the need for further evaluation of each model's structure and parameterizations to identify sensitivities and their appropriate application to different surface types and conditions. It is expected that the results of this study can be used to inform decisions regarding model choice for water resources and agricultural management, as well as providing insight into model selection for global flux monitoring efforts.

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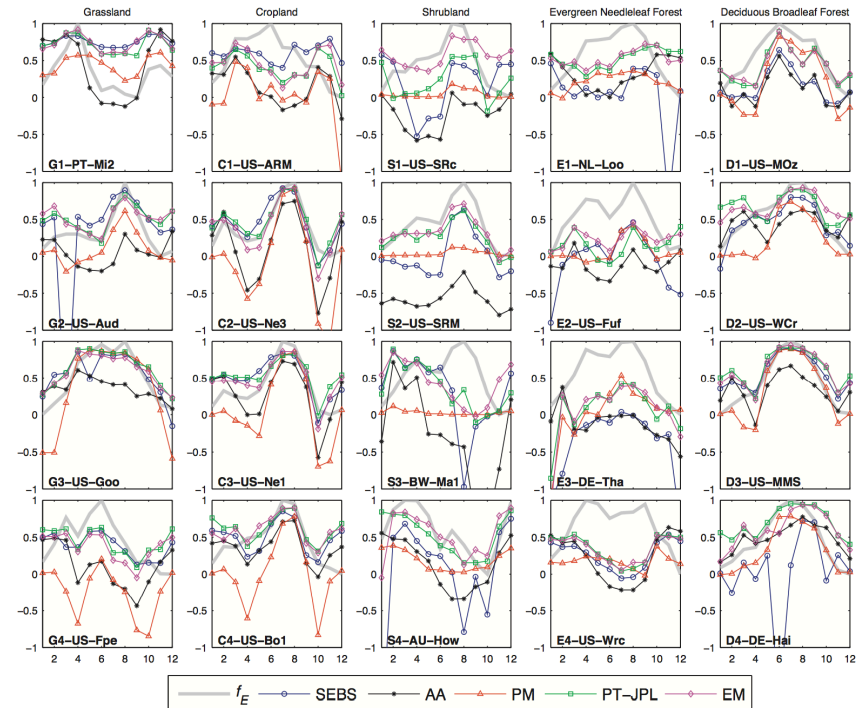
1. Introduction

Reliable estimates of evaporation (E) are required for the accurate representation of mass and energy exchanges at the land surface. In hydrological and water resource studies, an evaporation model is required to characterize the exchange of moisture between the surface and the overlying atmosphere. Not surprisingly, the choice of model can have considerable impact on water

resource planning and decision support across a range of temporal and spatial scales. Improved understanding of the influence of model choice on flux estimation is required in order to better characterize the fidelity of these simulations, particularly in light of an increasing number of regional and global scale efforts to produce land surface heat flux data products (Jiménez et al., 2011; Mueller et al., 2013).

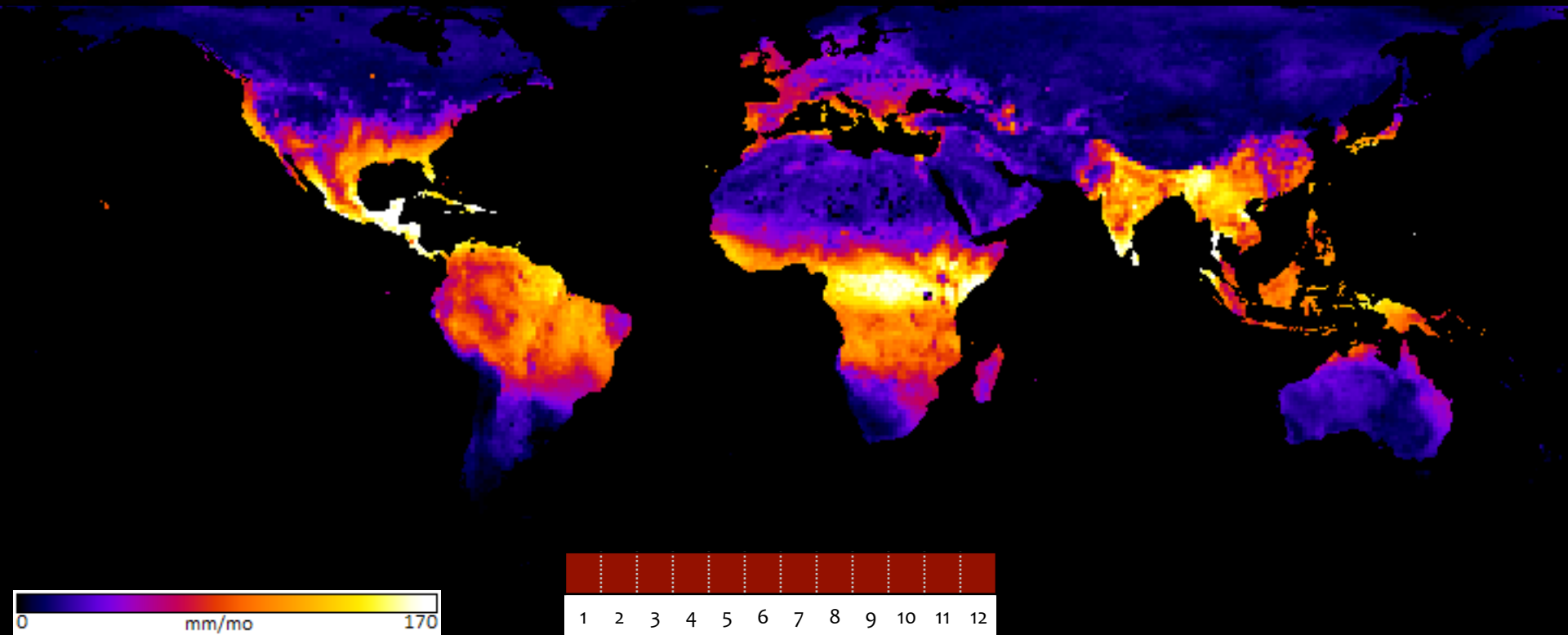
A number of models have been developed for the estimation of either the reference, potential or actual values of evaporation (see reviews of Kalma et al., 2008 and Wang and Dickinson, 2012). The reference evaporation is defined as the evaporation from a hypothetical, well-watered ‘reference’ crop (Allen, 2000), while potential evaporation is the maximum evaporation for a given surface if moisture is not limiting (Penman, 1948; Irmak and Haman, 2003). Estimation of the reference and potential evaporation is

PT-JPL was the best performing ET retrieval algorithm.



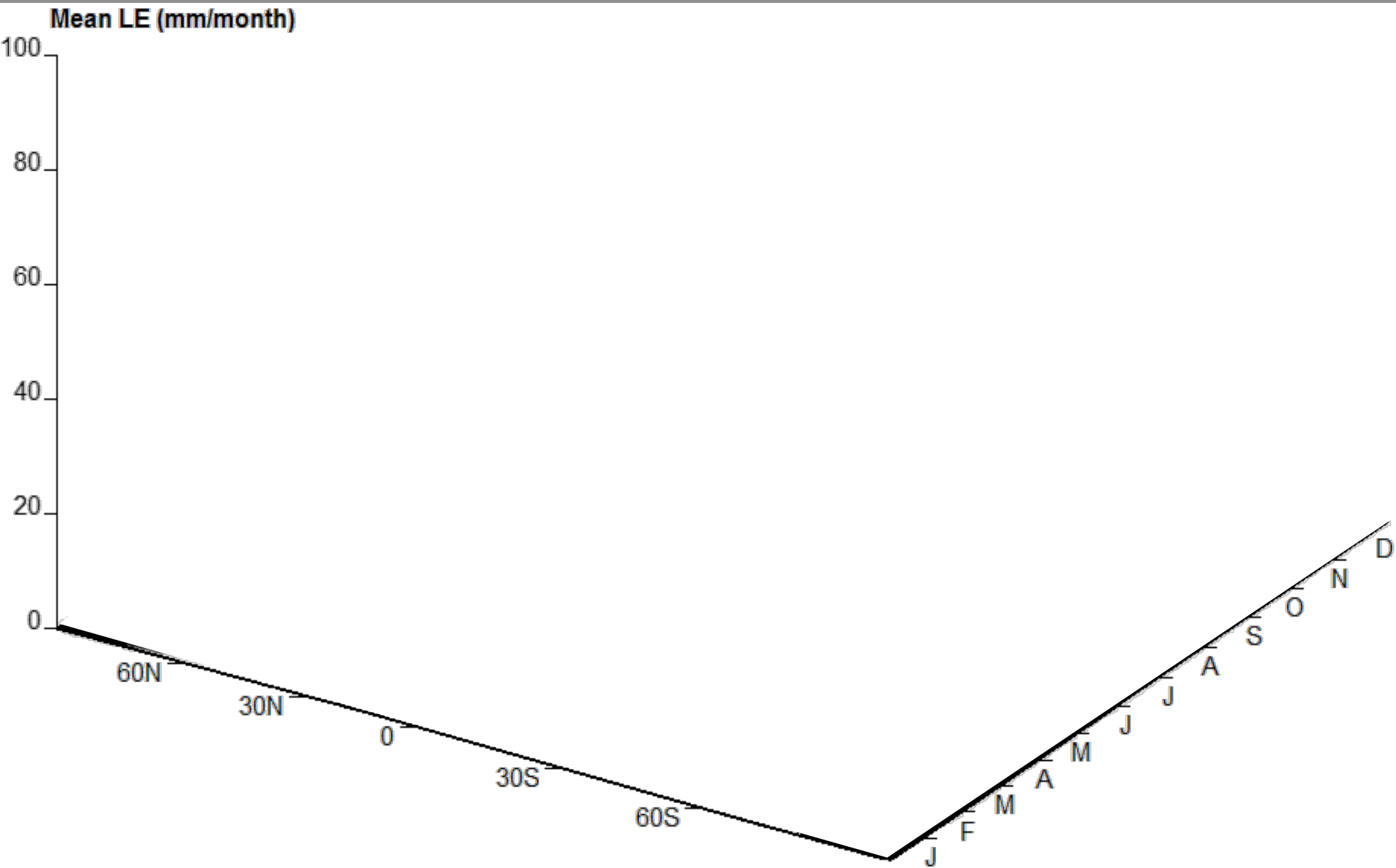
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PT-JPL GLOBAL ET

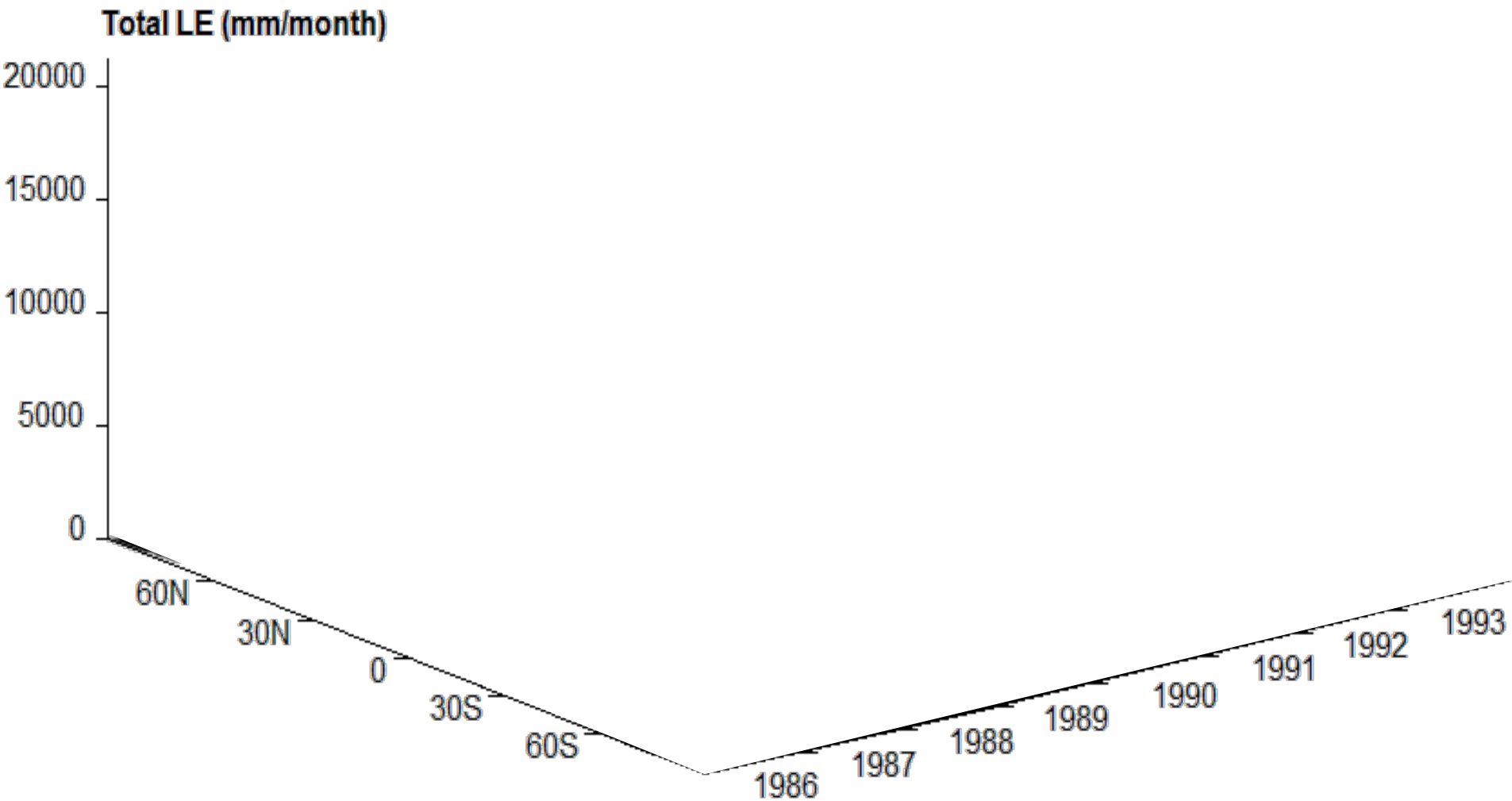


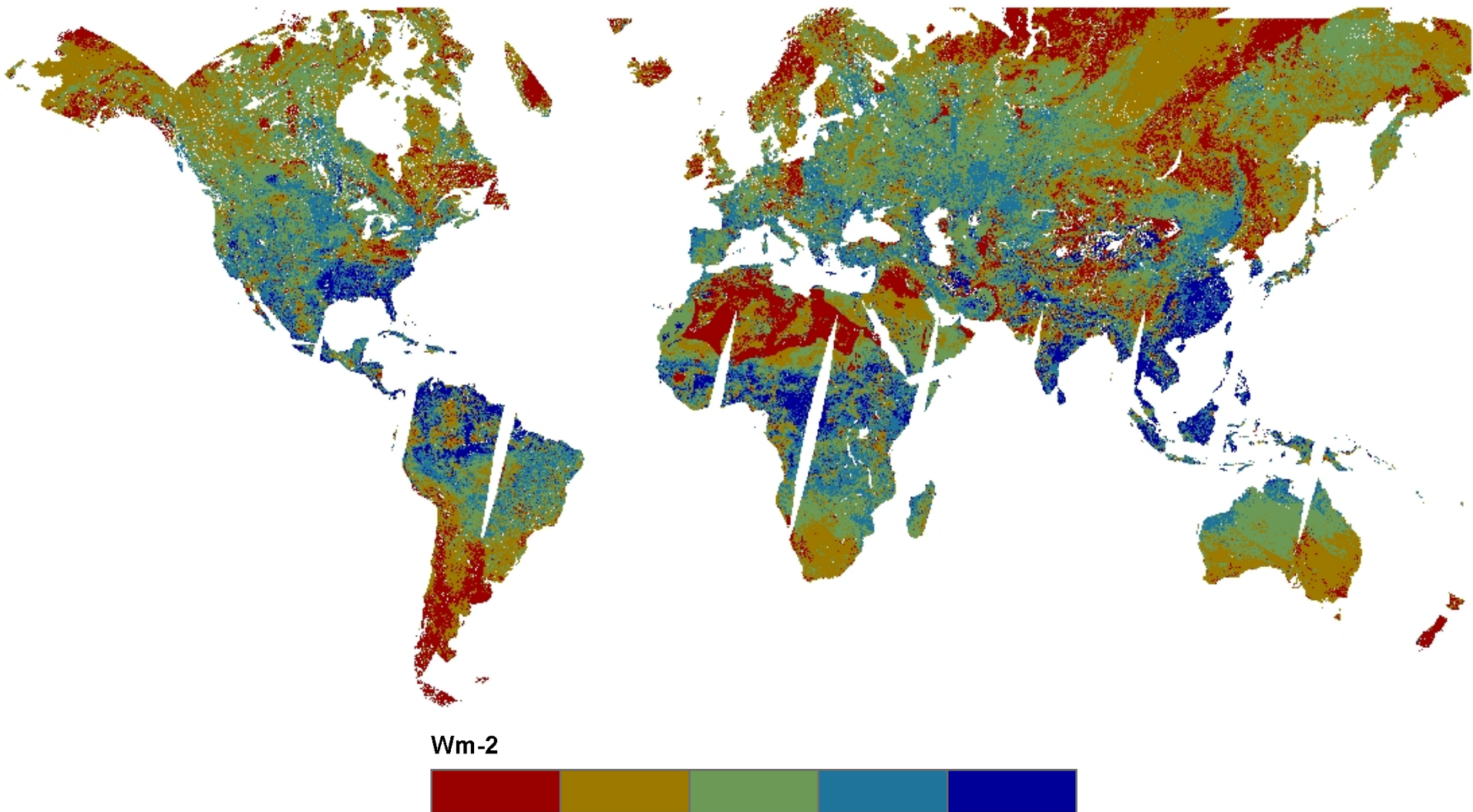
MONTHLY, 0.5 DEGREE

PT-JPL GLOBAL ET



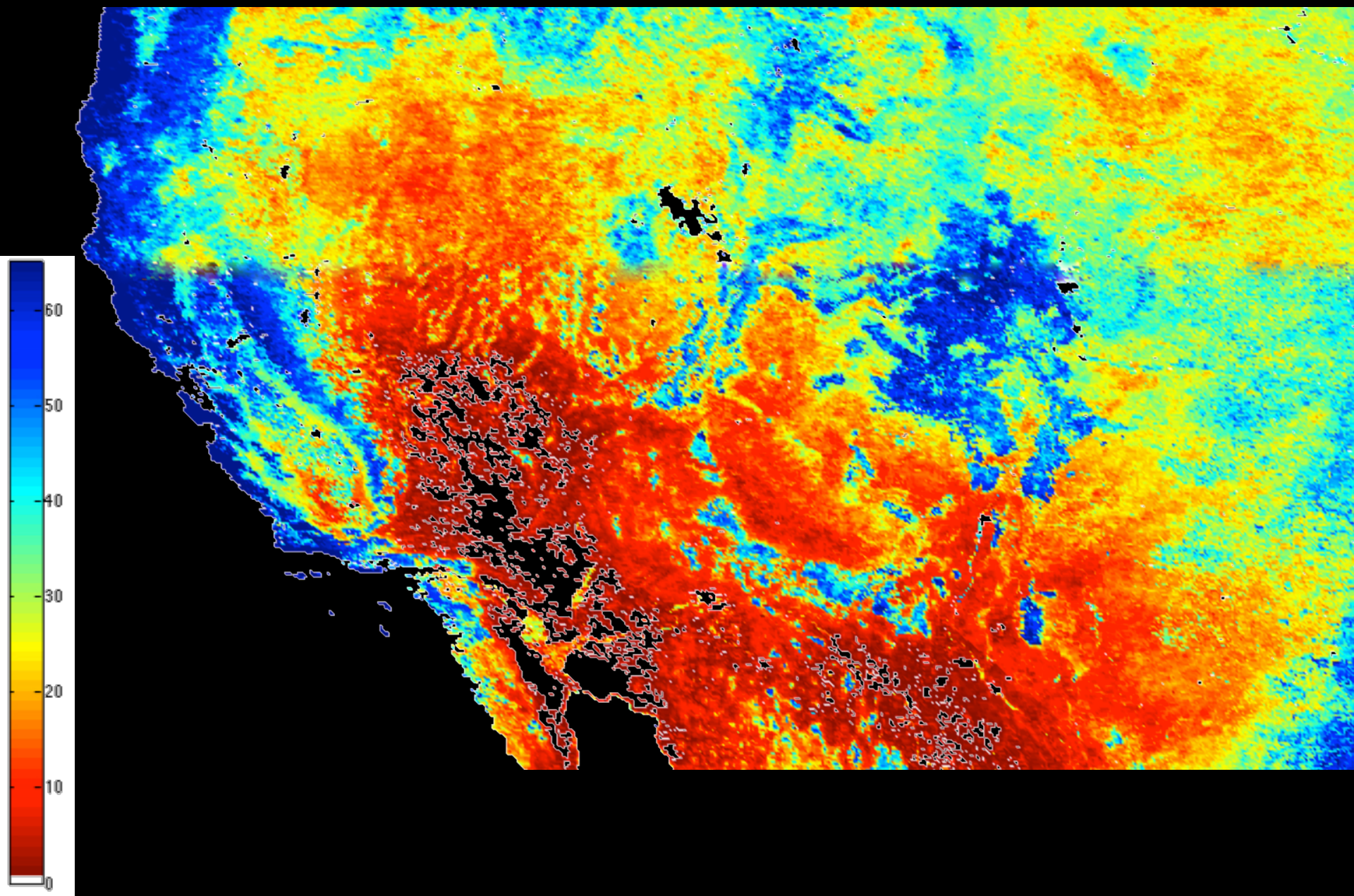
PT-JPL GLOBAL ET



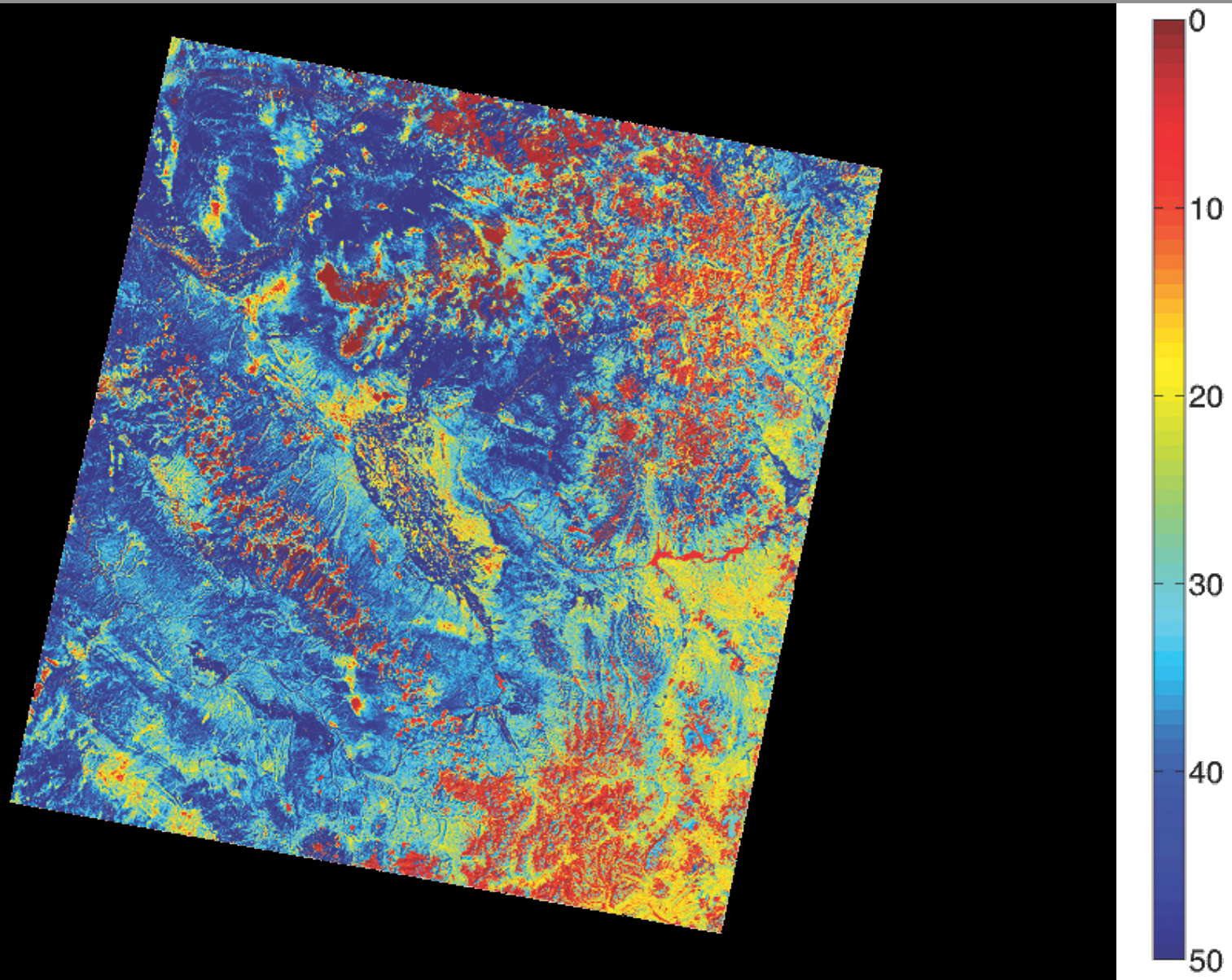


Work in progress:

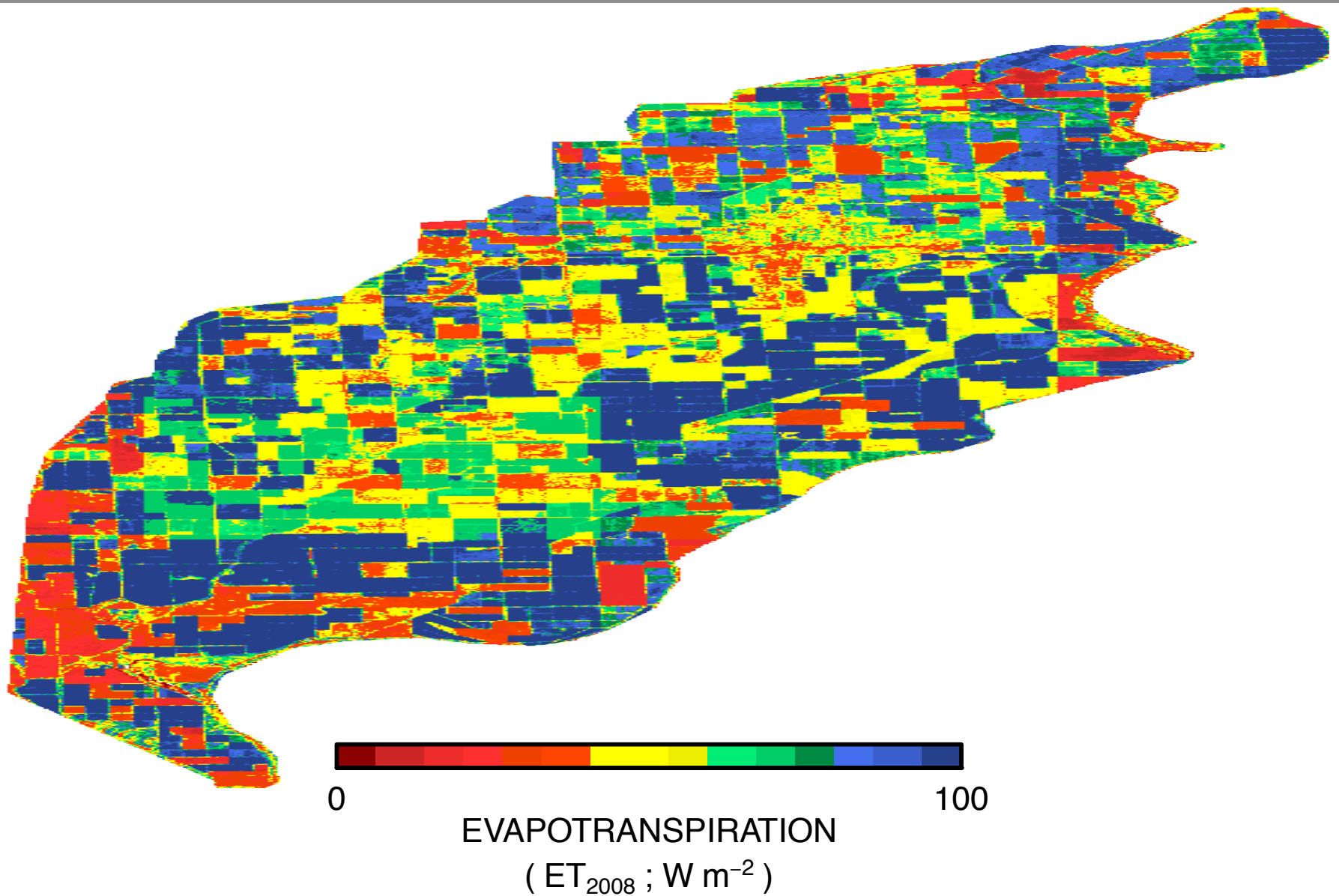
- Global, 1 km, daily, MODIS-era (10+ years)
- PT-JPL, PM-MOD16, SEBS, PMBL



PT-JPL ET Landsat



PT-JPL ET Landsat





Actual evapotranspiration in drylands derived from in-situ and satellite data: Assessing biophysical constraints

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Water-limited ecosystems

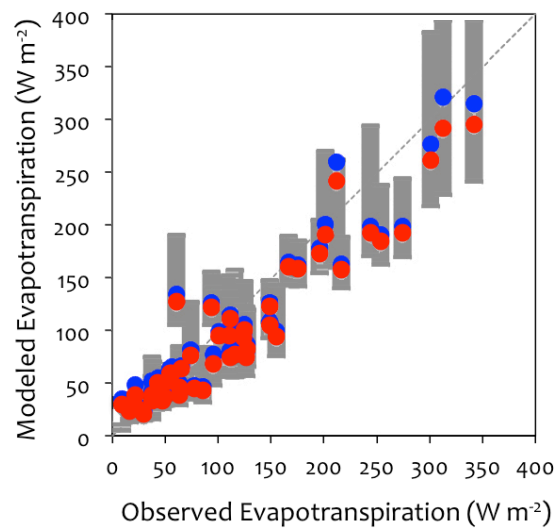
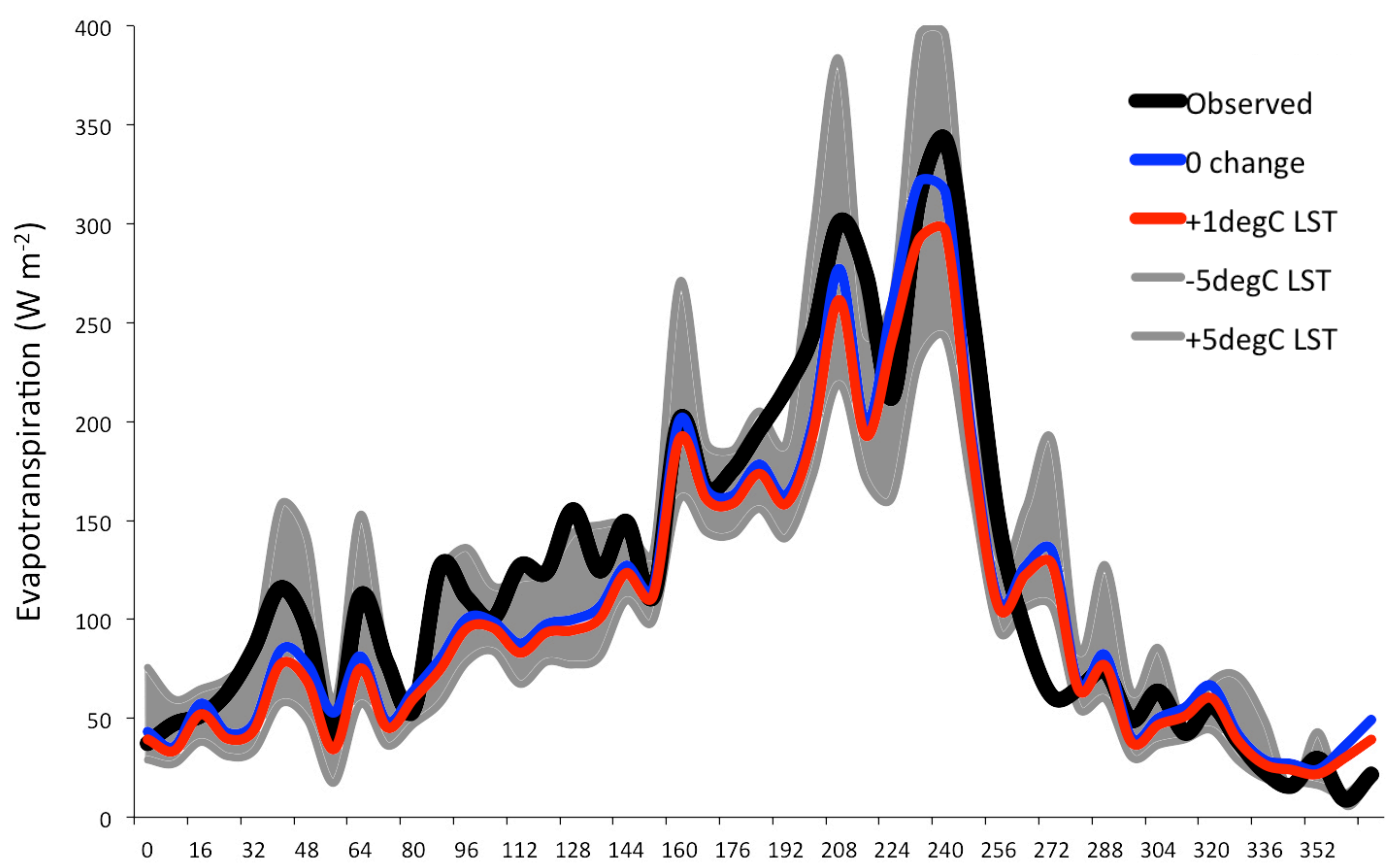
MODIS

ABSTRACT

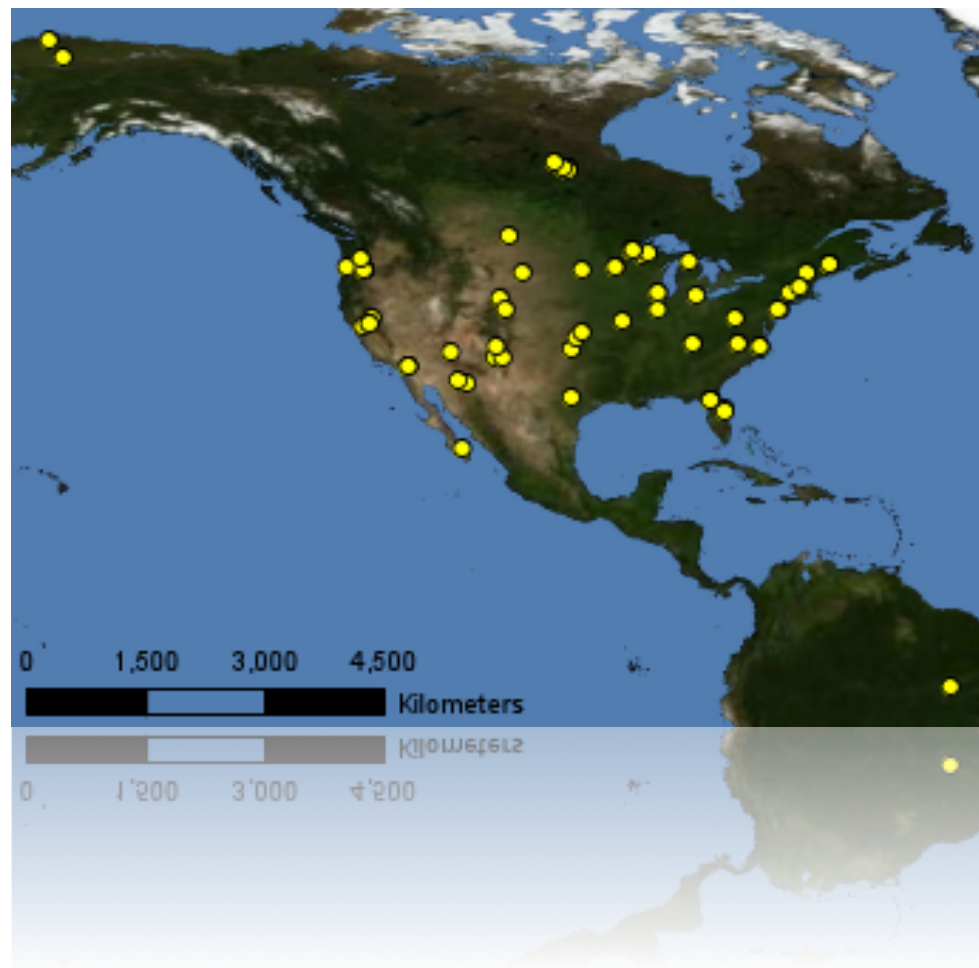
Improving regional estimates of actual evapotranspiration (λE) in water-limited regions located at climatic transition zones is critical. This study assesses an λE model (PT-JPL model) based on downscaling potential evapotranspiration according to multiple stresses at daily time-scale in two of these regions using MSG–SEVIRI (surface temperature and albedo) and MODIS products (NDVI, LAI and f_{PAR}). An open woody savanna in the Sahel (Mali) and a Mediterranean grassland (Spain) were selected as test sites with Eddy Covariance data used for evaluation. The PT-JPL model was modified to run at a daily time step and the outputs from eight algorithms differing in the input variables and also in the formulation of the biophysical constraints (stresses) were compared with the λE from the Eddy Covariance. Model outputs were also compared with other modeling studies at similar global dryland ecosystems.

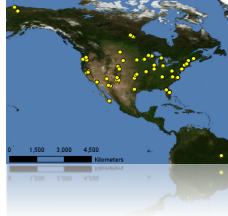
The novelty of this paper is the computation of a key model parameter, the soil moisture constraint, relying on the concept of apparent thermal inertia (f_{SM-ATI}) computed with surface temperature and albedo observations. Our results showed that f_{SM-ATI} from both in-situ and satellite data produced satisfactory results for λE at the Sahelian savanna, comparable to parameterizations using field-measured Soil Water Content (SWC) with r^2 greater than 0.80. In the Mediterranean grasslands however, with much lower daily λE values, model results were not as good as in the Sahel ($r^2 = 0.57–0.31$) but still better than reported values from more complex models applied at the site such as the Two Source Model (TSM) or the Penman–Monteith Leuning model (PML).

PT-JPL-daily model with a soil moisture constraint based on apparent thermal inertia, f_{SM-ATI} offers great potential for regionalization as no field-calibrations are required and water vapor deficit estimates, required in the original

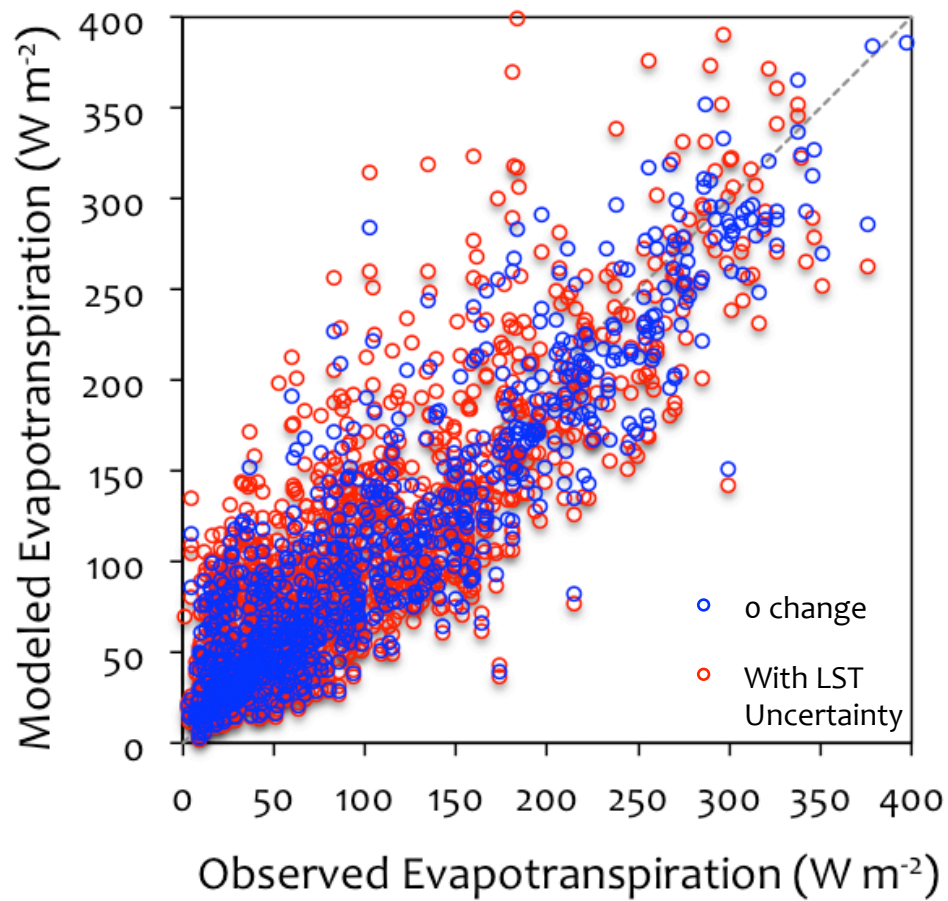


	0Δ	$+1^{\circ}\text{C}$	$\pm 5^{\circ}\text{C}$
m			
r^2			
RMSE			

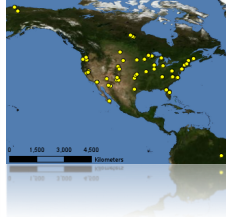




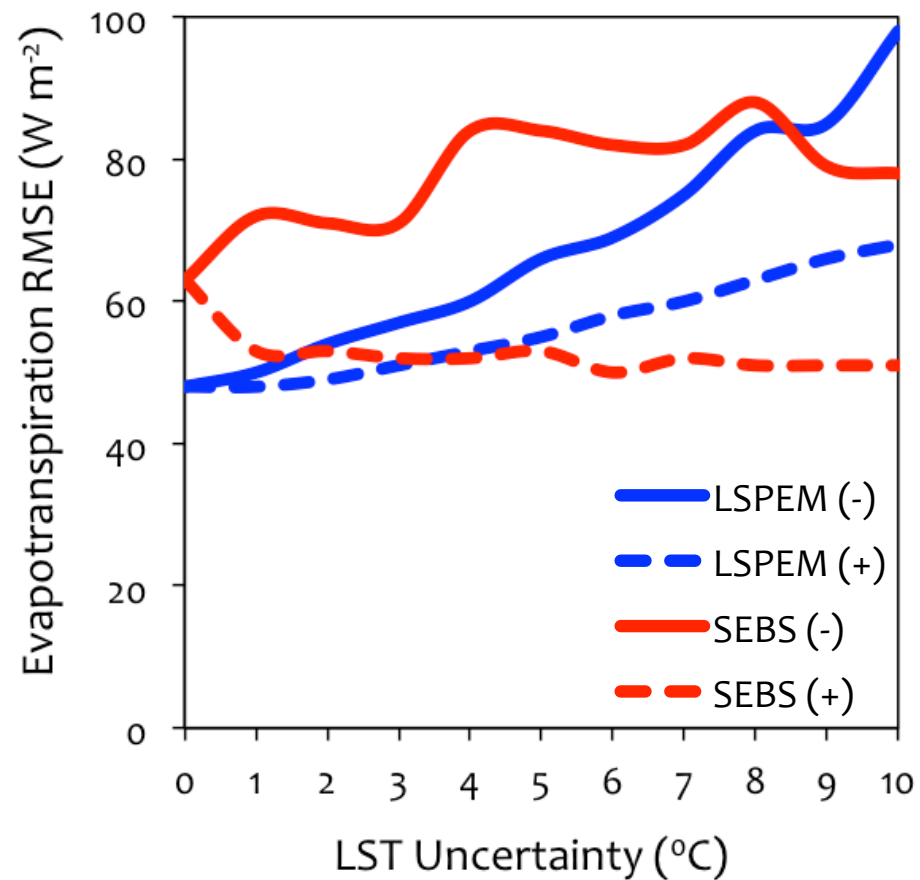
site-level ET uncertainty from LST uncertainty

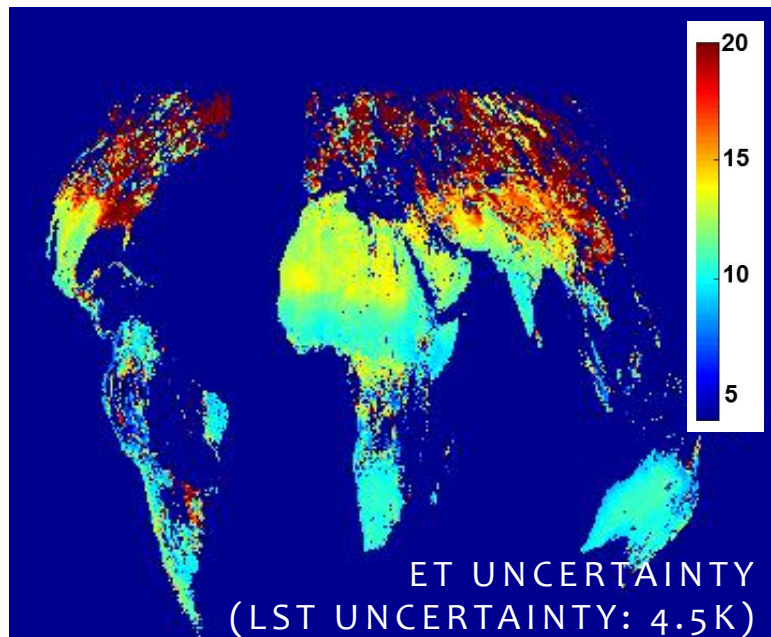
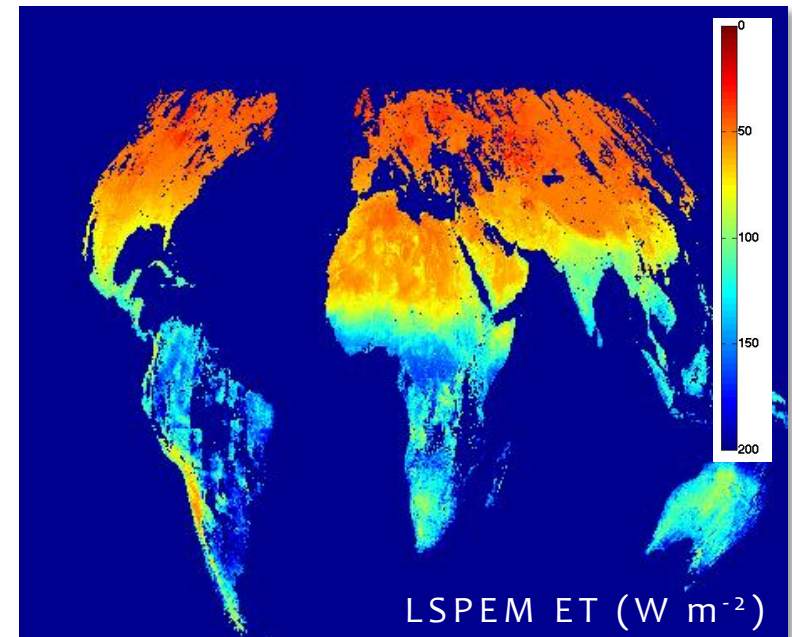
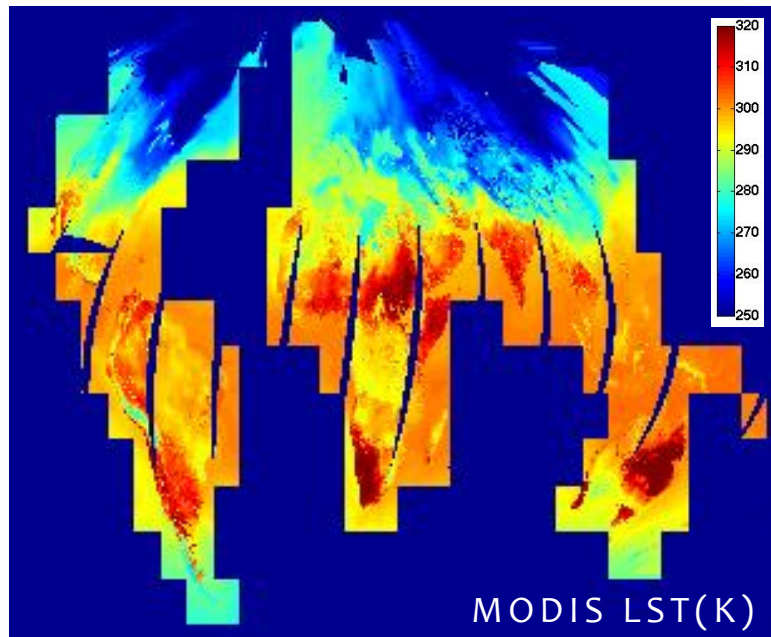


	$\sigma\Delta$	$\Delta^{\circ}\text{C}$
m	0.99	
r^2	0.82	
RMSE	37.7	



site-level ET uncertainty sensitivity from LST uncertainty





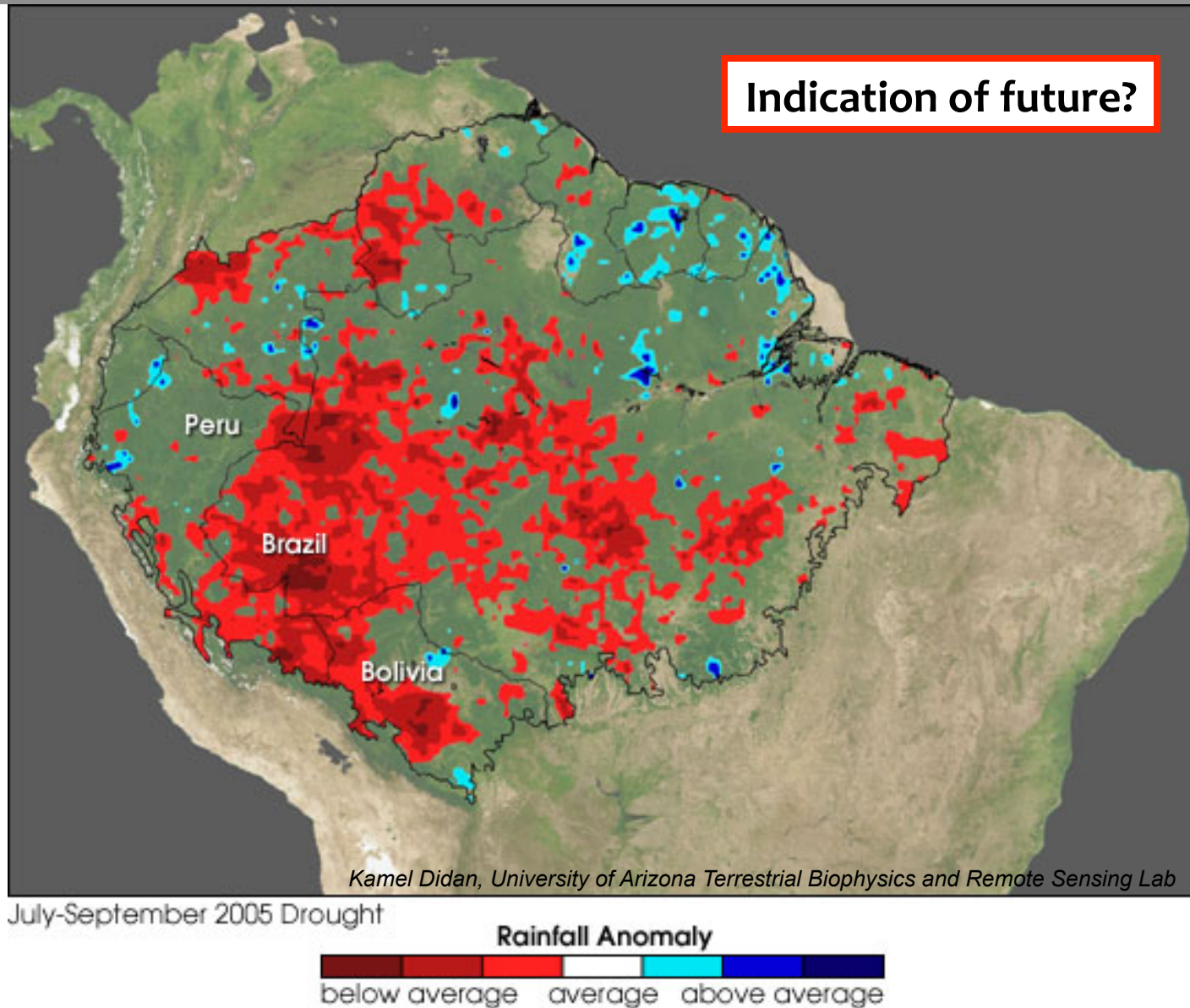
- **ET UNCERTAINTY**

$$\begin{aligned}
 &= \text{abs}(\text{ET}^{\text{LST}0} - \text{ET}^{\text{LST}+}) + \text{abs}(\text{ET}^{\text{LST}0} - \text{ET}^{\text{LST}-}) \\
 &= \text{abs}(50 - 51 \text{ W m}^{-2}) + \text{abs}(50 - 58 \text{ W m}^{-2}) \\
 &= \mathbf{9 \text{ W m}^{-2} \text{ (global mean)}}
 \end{aligned}$$

- **ET SENSITIVITY**

$$\begin{aligned}
 &= \text{abs}(\text{ET}^{\text{LST}+} - \text{ET}^{\text{LST}-}) / \text{ET}^{\text{LST}0} \\
 &= \text{abs}(51 - 58 \text{ W m}^{-2}) / 50 \text{ W m}^{-2} \\
 &= \mathbf{14\% \text{ (global mean)}}
 \end{aligned}$$

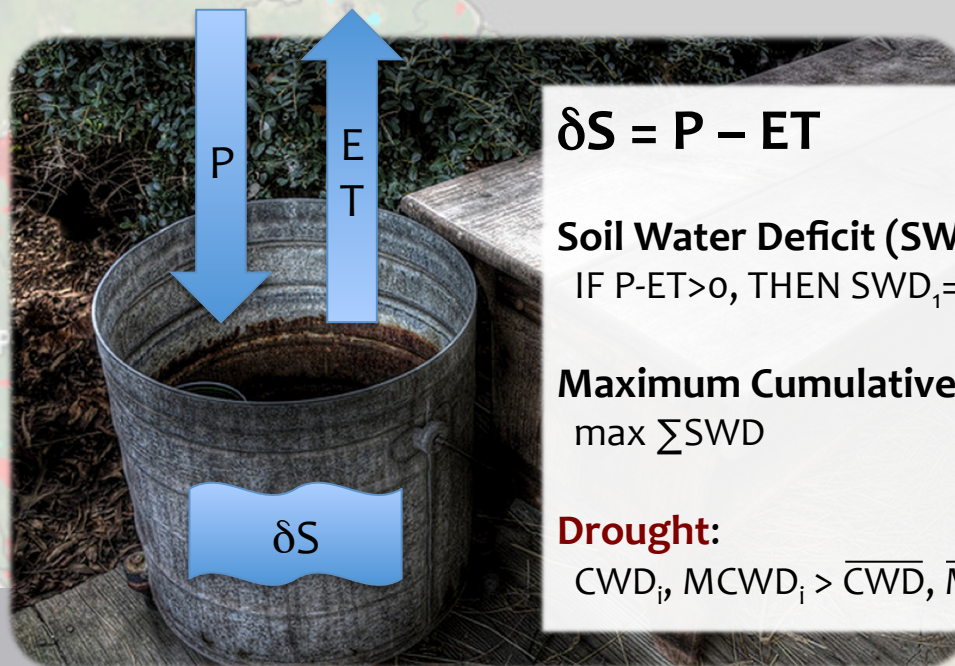
AMAZON DROUGHT 2005



DROUGHT?

Drought: meteorological (or, climatological, atmospheric), agricultural, hydrologic, socio-economic; exceedance of threshold.

Vegetation drought: physical drying of soil such that the overlying vegetation experiences physiological water stress manifested in a reduction of productivity, loss of leaves/needles, and, ultimately, mortality.



$$\delta S = P - ET$$

Soil Water Deficit (SWD):

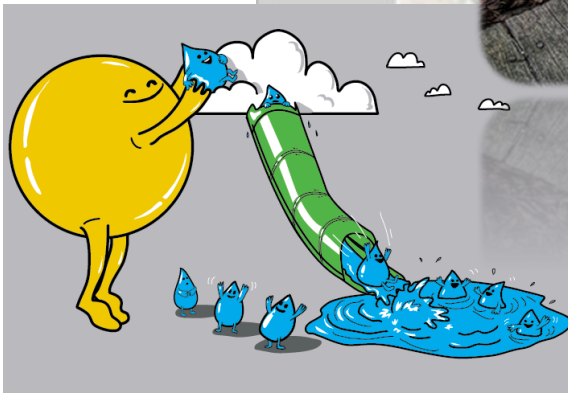
IF $P - ET > 0$, THEN $SWD_1 = 0$, ELSE $SWD_1 = P - ET + SWD_0$

Maximum Cumulative Water Deficit (MCWD):

$\max \sum SWD$

Drought:

$CWD_i, MCWD_i > \overline{CWD}, \overline{MCWD}$



Drought

Rainfall Anomaly

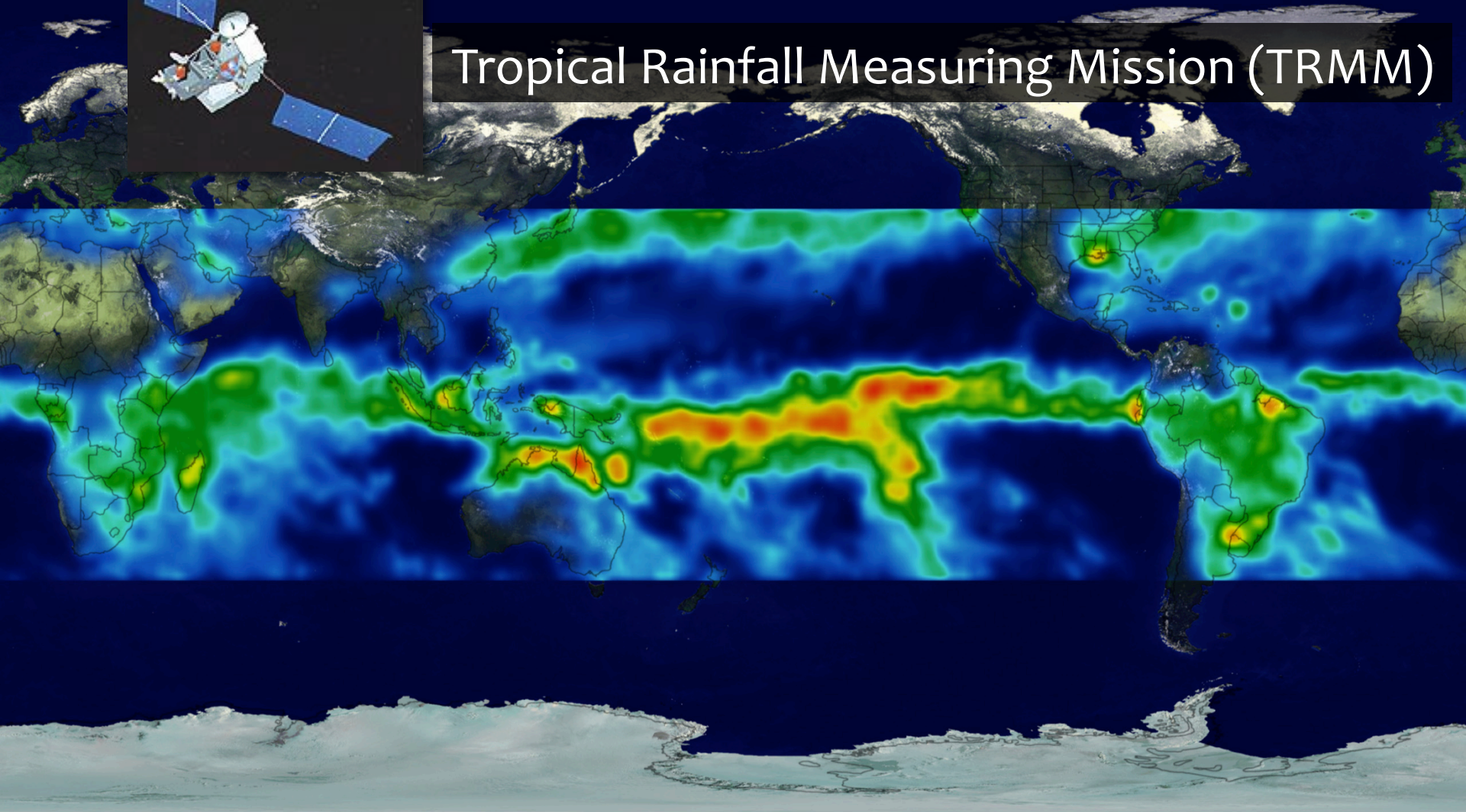


PRECIPITATION

January 1998



Tropical Rainfall Measuring Mission (TRMM)



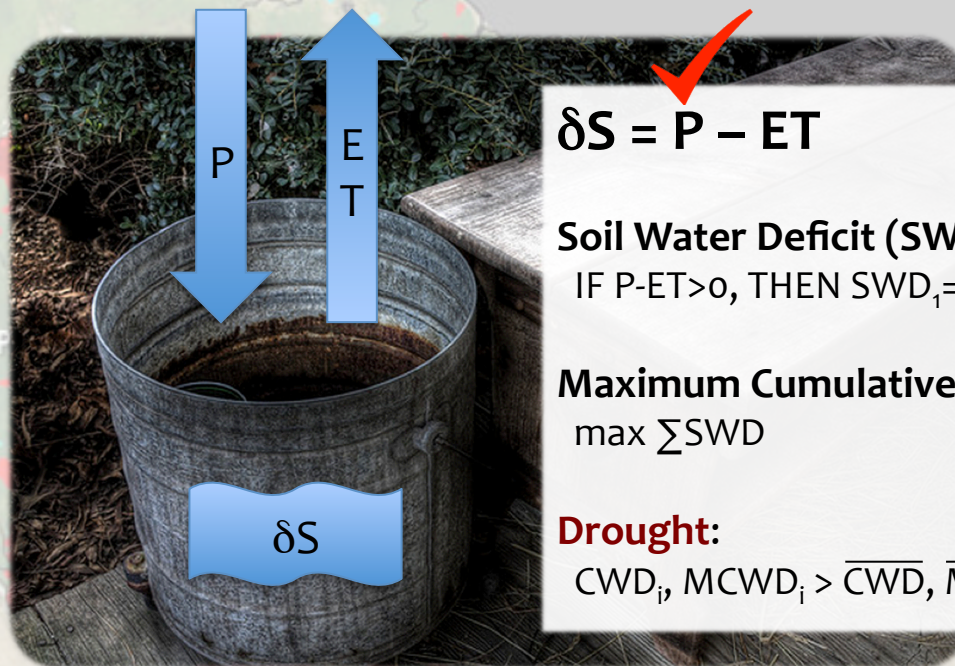
Monthly Rainfall (mm/day)

0 10 >50

DROUGHT?

Drought: meteorological (or, climatological, atmospheric), agricultural, hydrologic, socio-economic; exceedance of threshold.

Vegetation drought: physical drying of soil such that the overlying vegetation experiences physiological water stress manifested in a reduction of productivity, loss of leaves/needles, and, ultimately, mortality.



$$\delta S = P - ET$$

Soil Water Deficit (SWD):

IF $P-ET > 0$, THEN $SWD_1 = 0$, ELSE $SWD_1 = P-ET+SWD_0$

Maximum Cumulative Water Deficit (MCWD):

$\max \sum SWD$

Drought:

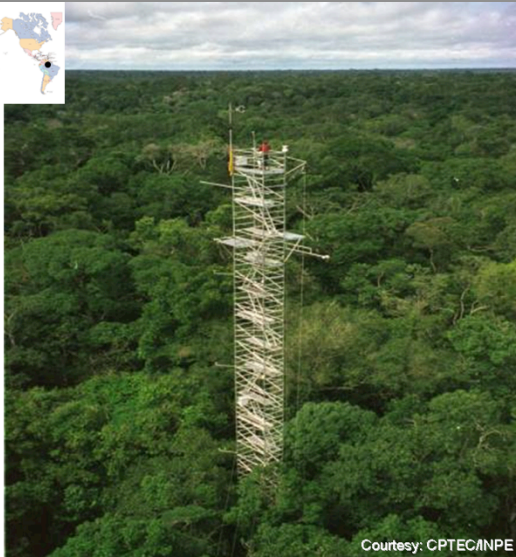
$CWD_i, MCWD_i > \overline{CWD}, \overline{MCWD}$

July-September 2005 Drought

Rainfall Anomaly



EVAPOTRANSPIRATION



Manaus

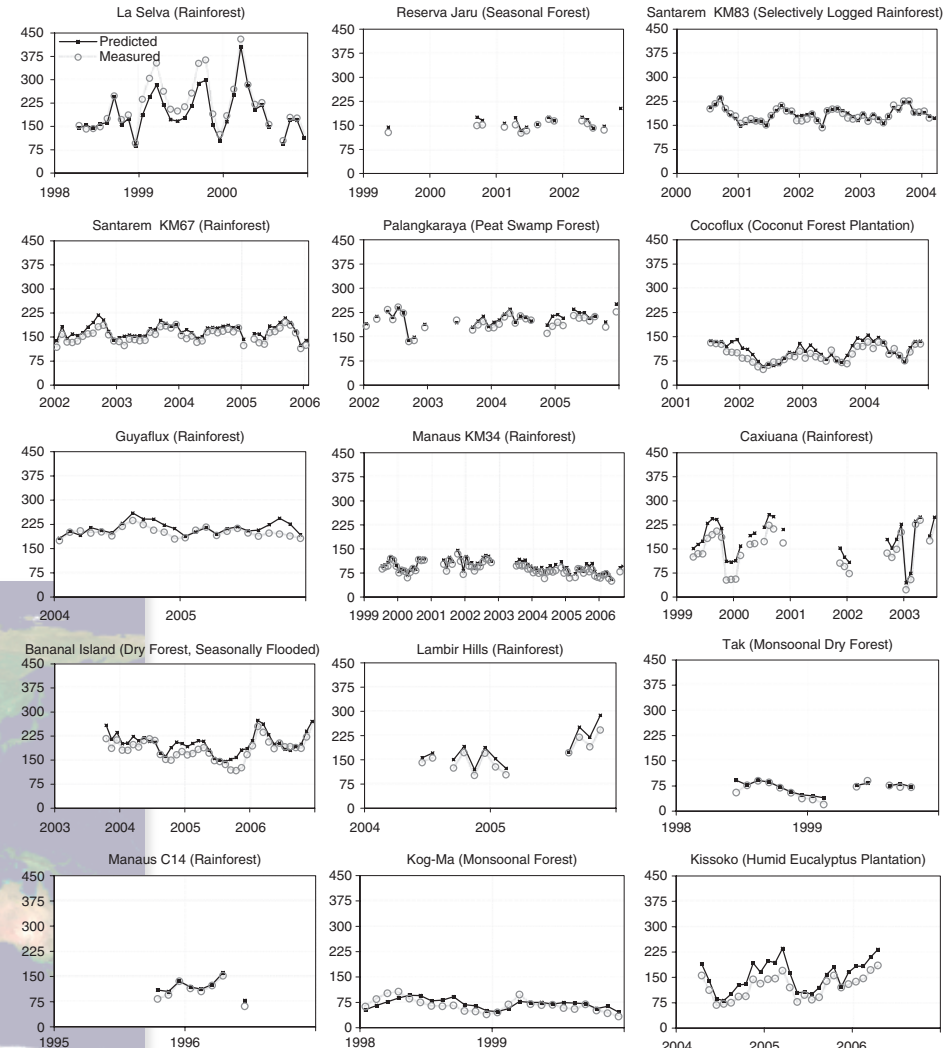
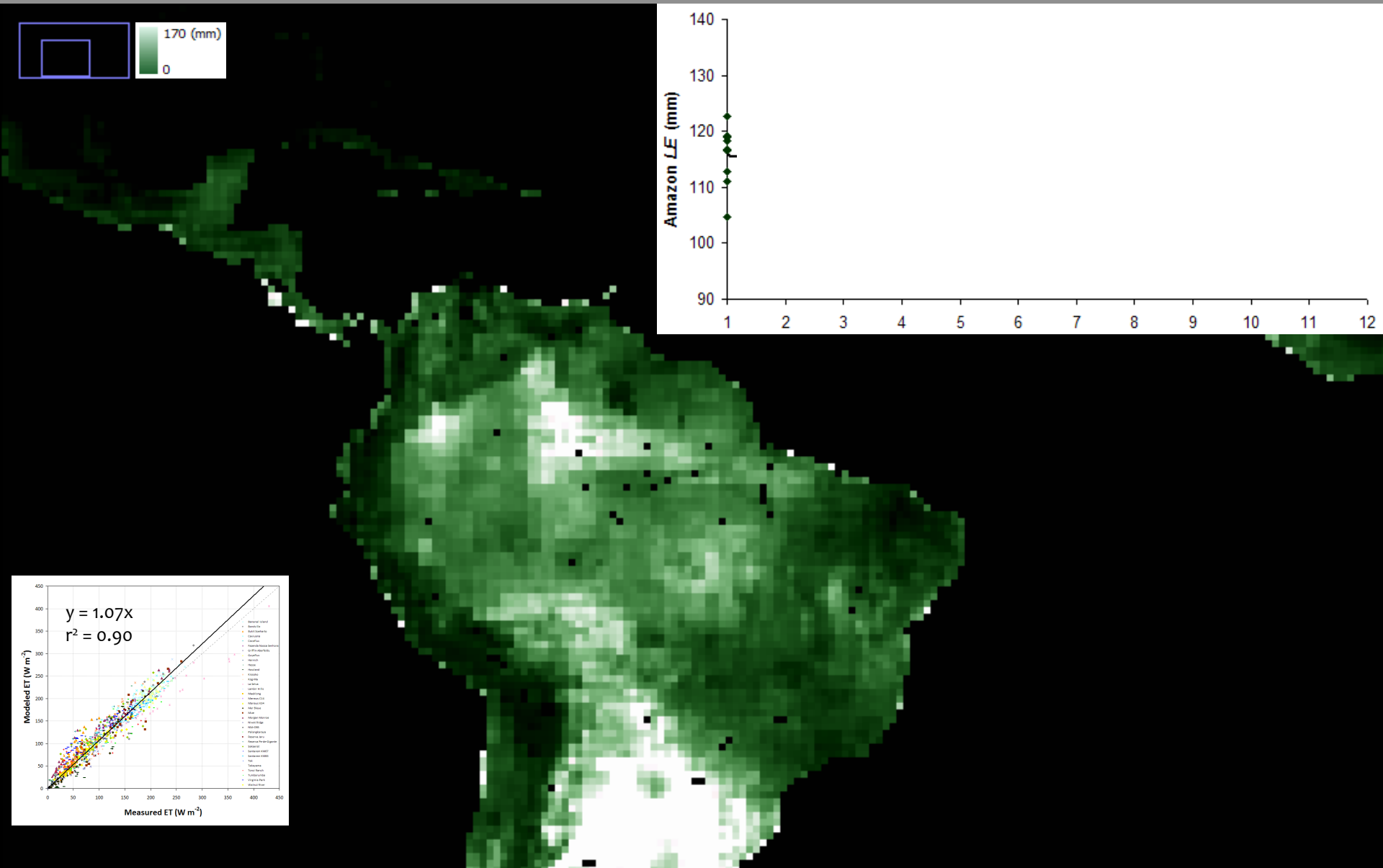


Fig. 6 Time series of tower measurements and model predictions at each site sorted in descending order by evaporative fraction. The y-axis is latent heat of evaporation (LE) (W m^{-2}) and x-axis is year. Open circles are observed and closed squares are predicted LE based on the FC model.

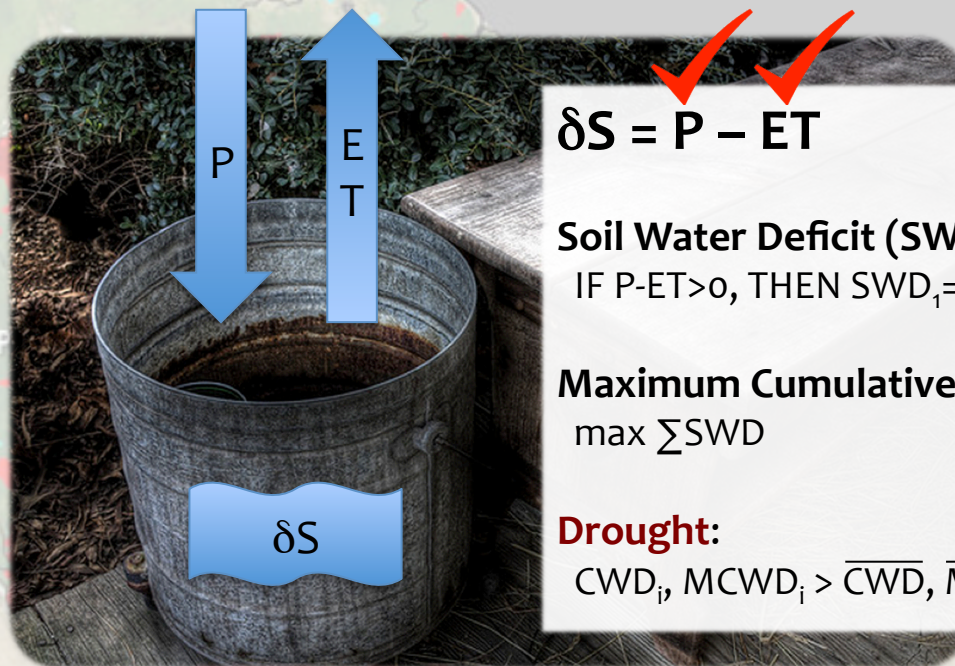
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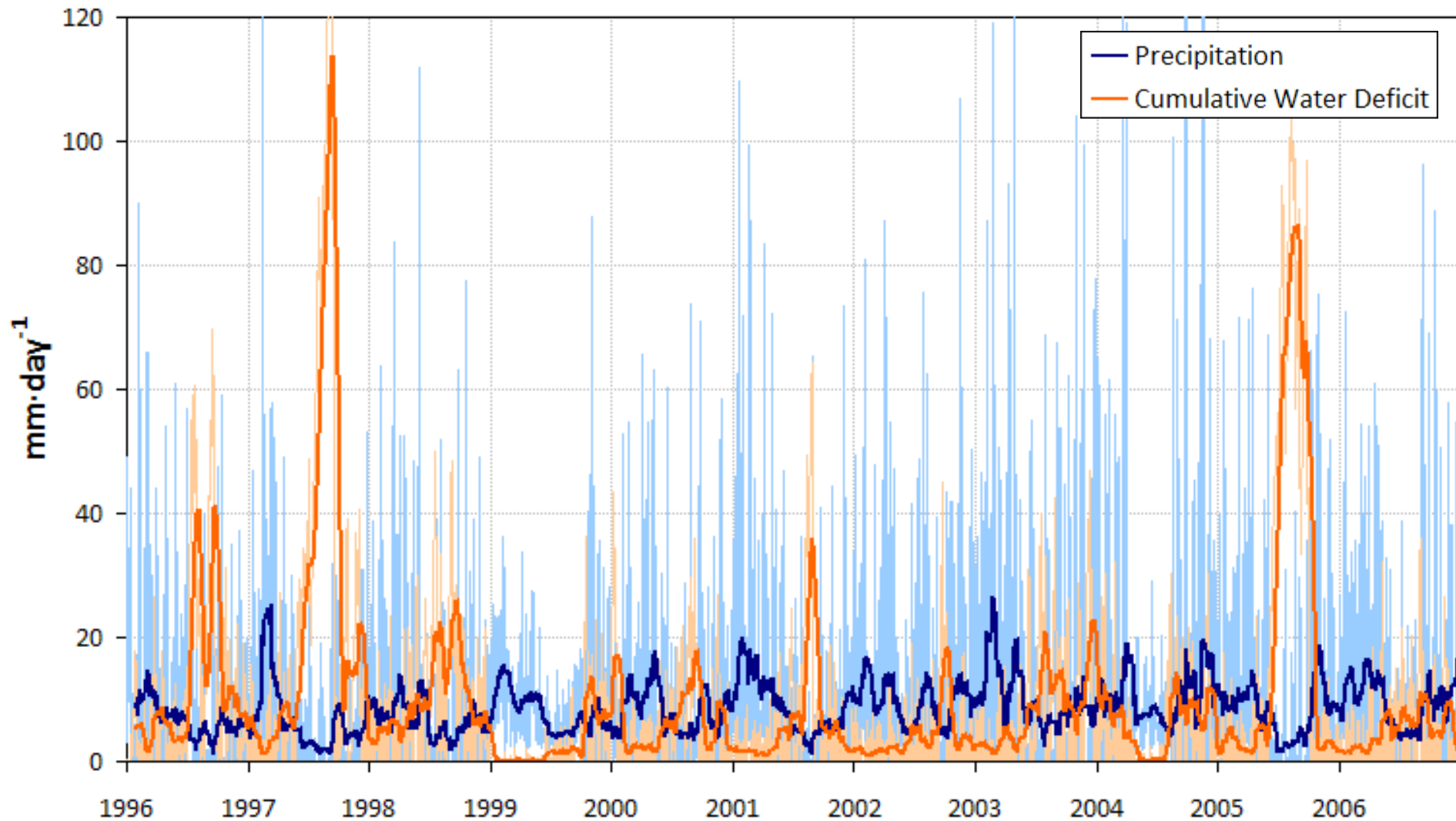
$CWD_i, MCWD_i > \overline{CWD}, \overline{MCWD}$

July-September 2005 Drought

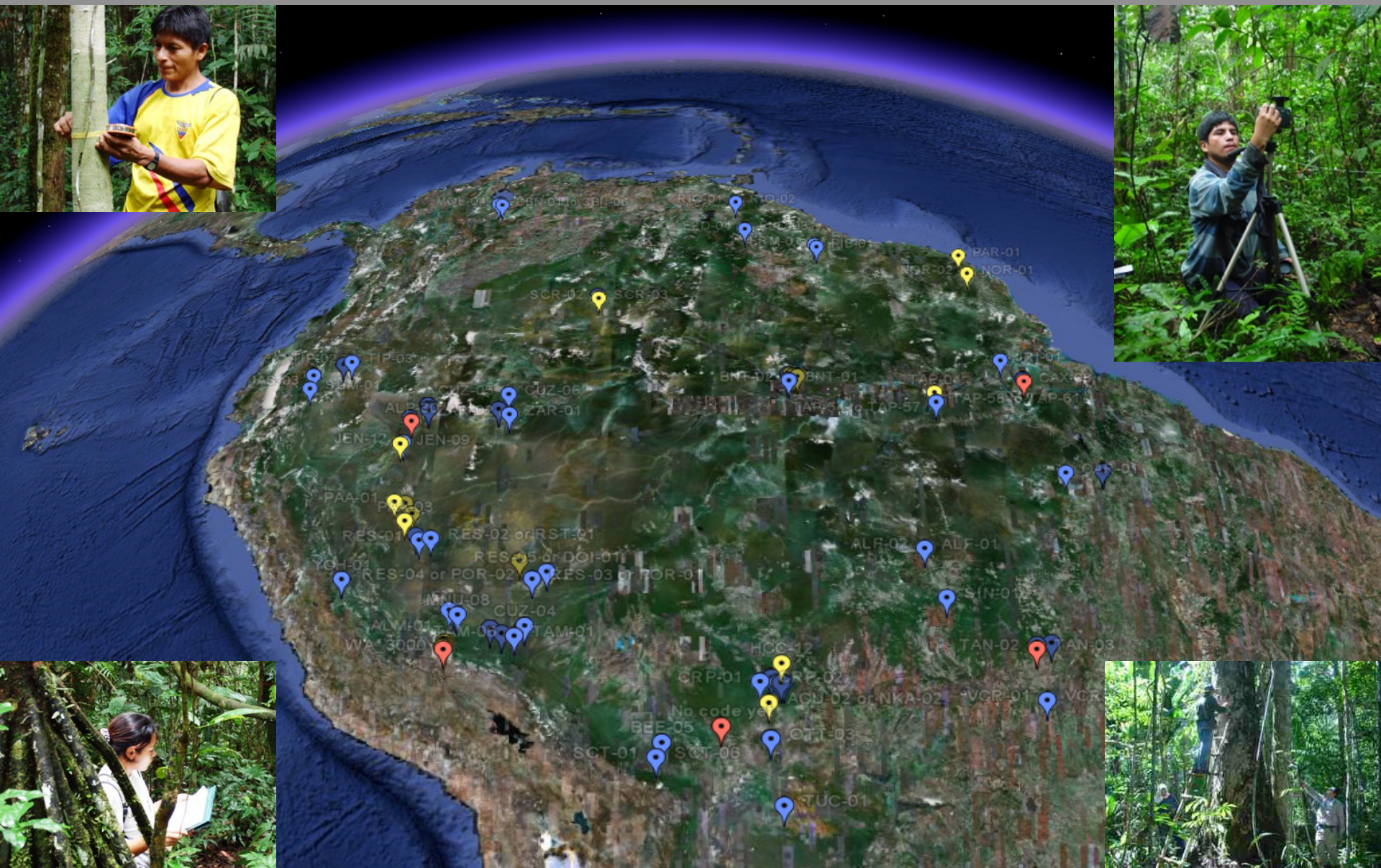
Rainfall Anomaly

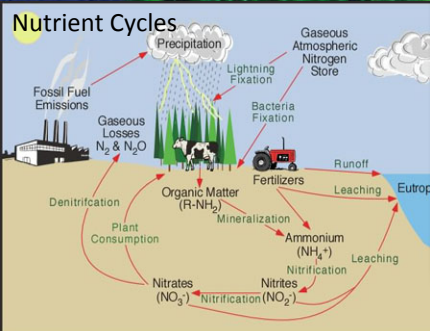
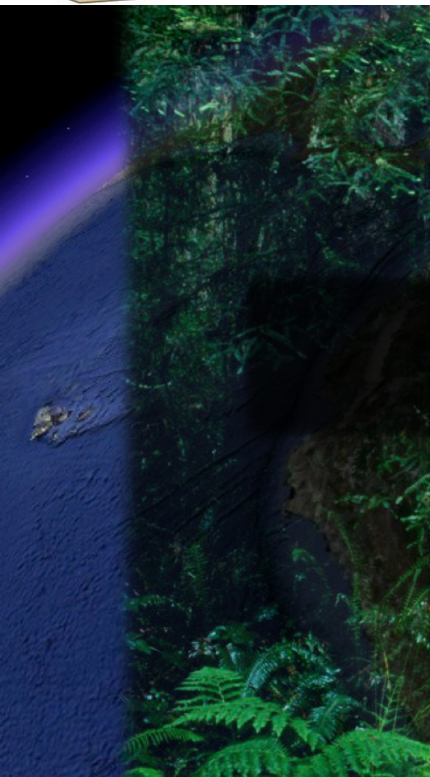
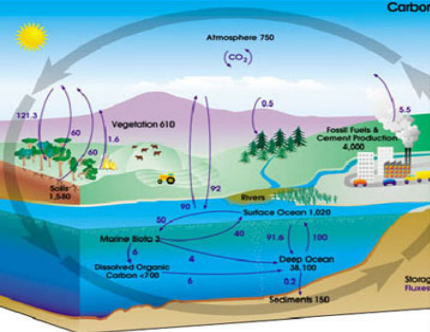


DROUGHT?



MORTALITY?





BREVIA

The 2010 Amazon Drought

Simon L. Lewis,^{1,†} Paulo M. Brando,^{2,3} Oliver L. Phillips,¹
Geertje M. F. van der Heijden,⁴ Daniel Nepstad²

Several global circulation models (GCMs) project an increase in the frequency and severity of drought events affecting the Amazon region as a consequence of anthropogenic greenhouse gas emissions (1). The proximate cause is twofold, increasing Pacific sea surface temperatures (SSTs), which may intensify El Niño Southern Oscillation events and associated periodic Amazon droughts, and an increase in the frequency of historically rarer droughts associated with high Atlantic SSTs and northwest displacement of the intertropical convergence zone (1, 2). Such droughts may lead to a loss of some Amazon forests, which would accelerate climate change (3). In 2005, a major Atlantic SST-associated drought occurred, identified as a 1-in-100-year event (2). Here, we report on a second drought in 2010, when Atlantic SSTs were again high.

We calculated standardized anomalies from a decade of satellite-derived dry-season rainfall data (Tropical Rainfall Measuring Mission, 0.25° resolution) across 5.3 million km² of Amazonia for 2010 and 2005 (4). We used identical reference periods to allow a strict comparison of both drought events (4). On the basis of this index, the 2010 drought was more spatially extensive than the 2005 drought (rainfall anomalies ≤ -1 SD over 3.0 million km² and 1.9 million km² in 2010 and 2005, respectively; Fig. 1 and fig. S1). Because dry-season

anomalies do not necessarily correlate with water stress for forest trees, we also calculated the maximum climatological water deficit (MCWD) for each year as the most negative cumulative value of water input minus estimated forest evapotranspiration (5). This measure of drought intensity correlates with Amazon forest tree mortality (6). In 2010, the difference in MCWD from the decadal mean that significantly increases tree mortality (≤ -25 mm) spanned 3.2 million km², compared with 2.5 million km² in 2005. The 2010 drought had three identifiable epicenters in southwestern Amazonia, north-central Bolivia, and Brazil's Mato Grosso state. In 2005 only a single southwestern Amazonia epicenter was detectable (fig. S1).

The relationship between the change in MCWD and changes in aboveground carbon storage derived from forest inventory plots affected by the 2005 drought (6) provides a first approximation of the biomass carbon impact of the 2010 event. Summing the change in carbon storage predicted by the 2010 MCWD difference across Amazonia gives a total impact of 2.2 Pg C [95% confidence intervals (CI) 1.2 and 3.4], compared with 1.6 Pg C for the 2005 event (CI 0.8, 2.6). These values are relative to the predrought carbon uptake and represent the sum of (1) the temporary cessation of biomass increases over the 2-year drought measurement interval (-0.8 Pg C) and (2) biomass loss

via tree mortality, a committed carbon flux from decomposition over several years (~ 1.4 Pg C after the 2010 drought). In most years, these forests are a carbon sink; drought reverses this sink.

Considerable uncertainty remains, related to the soil characteristics within the epicenters of the 2010 drought, which could moderate or exacerbate climatic drying, whether a second drought will kill more trees (i.e., those damaged by the initial drought) or fewer (i.e., if most drought-susceptible trees are already dead), and whether drought slows soil respiration (temporarily offsetting the biomass carbon source). New field measurements will be required to refine our initial estimates.

The two recent Amazon droughts demonstrate a mechanism by which remaining intact tropical forests of South America can shift from buffering the increase in atmospheric carbon dioxide to accelerating it. Indeed, two major droughts in a decade may largely offset the net gains of -0.4 Pg C year⁻¹ in intact Amazon forest aboveground biomass in nondrought years. Thus, repeated droughts may have important decadal-scale impacts on the global carbon cycle.

Droughts co-occur with peaks of fire activity (5). Such interactions among climatic changes, human actions, and forest responses represent potential positive feedbacks that could lead to widespread Amazon forest degradation or loss (7). The significance of these processes will depend on the growth response of tropical trees to increases in atmospheric carbon dioxide concentration, fire management, and deforestation trends (3, 7). Nevertheless, any shift to drier conditions would favor drought-adapted species, and drier forests store less carbon (8). If drought events continue, the era of intact Amazon forests buffering the increase in atmospheric carbon dioxide may have passed.

References and Notes

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Supporting Online Material

www.sciencemag.org/cgi/content/full/331/6017/554/DC1
Materials and Methods

Fig. S1
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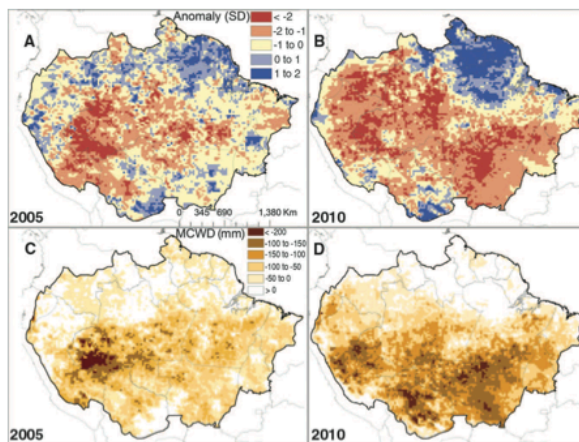
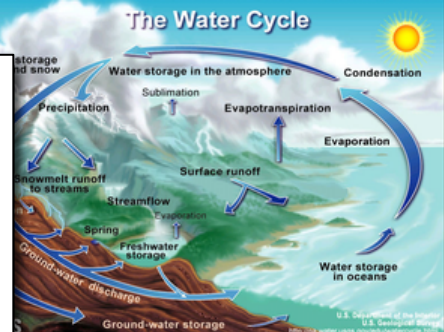


Fig. 1. (A and B) Satellite-derived standardized anomalies for dry-season rainfall for the two most extensive droughts of the 21st century in Amazonia. (C and D) The difference in the 12-month (October to September) MCWD from the decadal mean (excluding 2005 and 2010), a measure of drought intensity that correlates with tree mortality. (A) and (C) show the 2005 drought; (B) and (D) show the 2010 drought.



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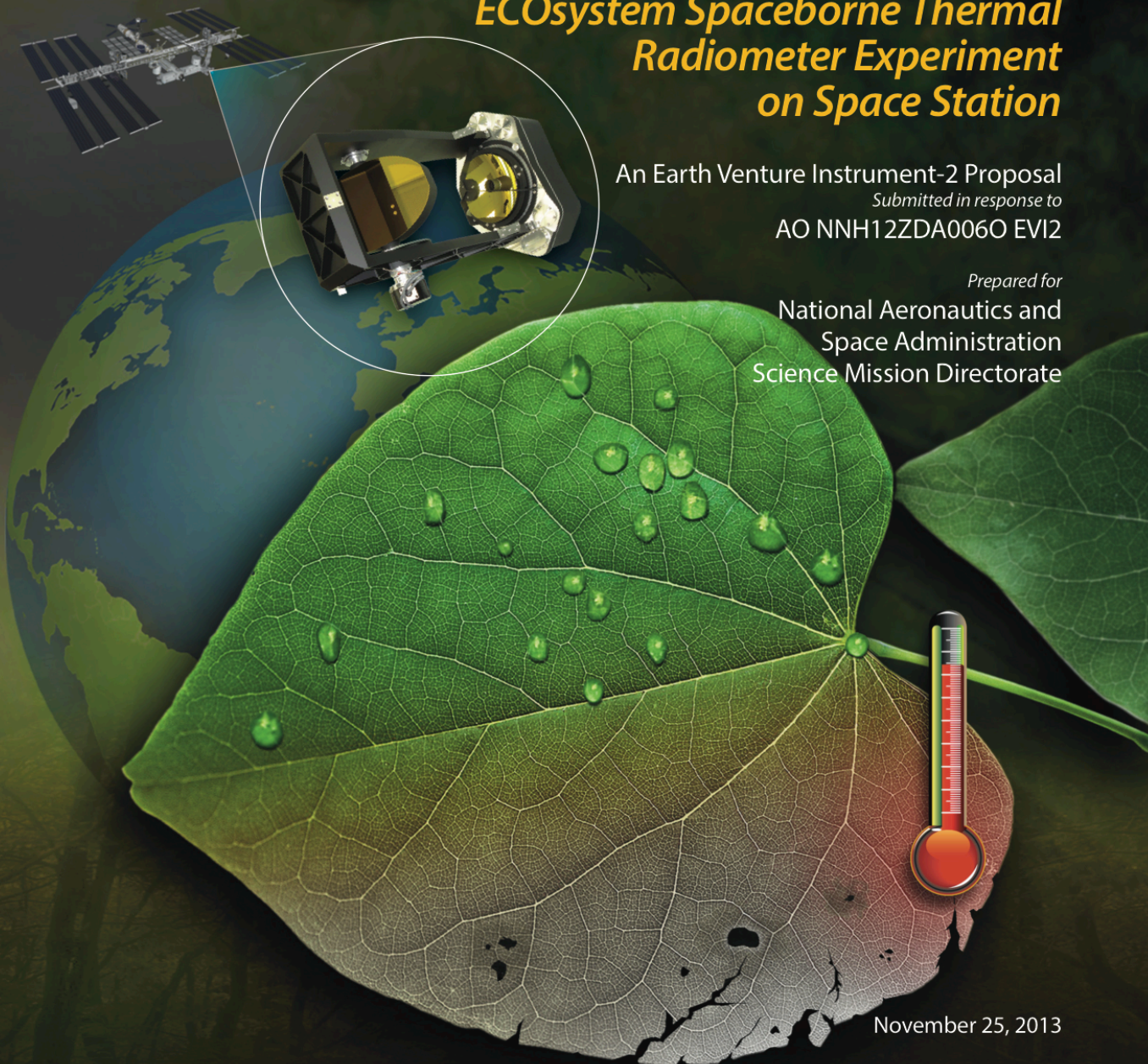


ECOSTRESS

ECOsystem Spaceborne Thermal Radiometer Experiment on Space Station

An Earth Venture Instrument-2 Proposal
Submitted in response to
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Science Mission Directorate



November 25, 2013