

**Supplementary Material**

**Thermodynamic Bounds of Terrestrial Water-Energy Coupling and Resiliency in Global Ecosystems**

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**Table S1.** Signal to Noise ratio (SNR) blending period for the combined Copernicus Climate Change Service (C3S) soil moisture product.

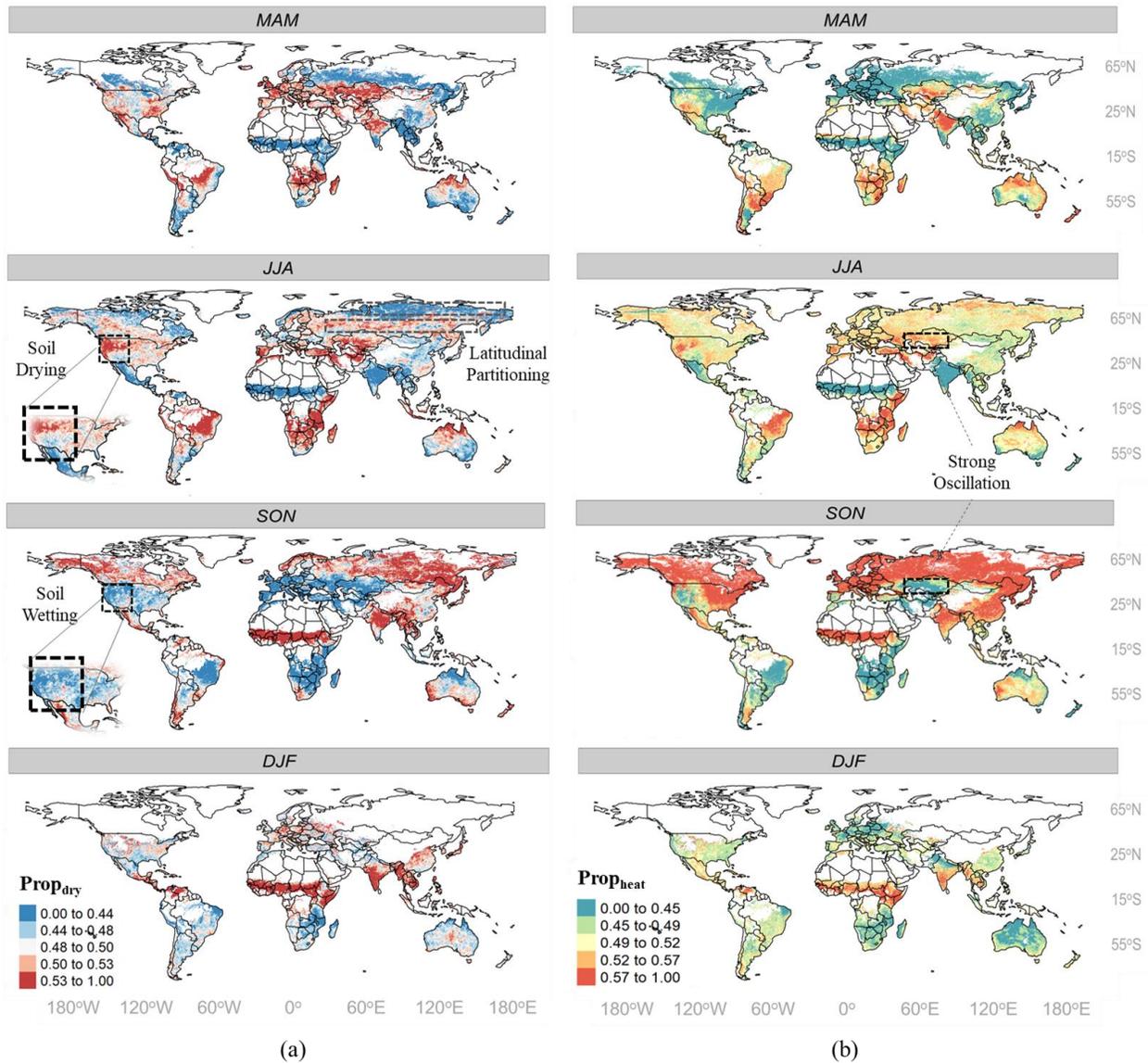
<b>Time Period</b>	<b>Passive Sensors</b>	<b>Active Sensors</b>
2010-01-15 to 2011-10-04	AMSR-E, WindSat, SMOS	ASCAT-A
2011-10-05 to 2012-06-30	WindSat, SMOS	ASCAT-A
2012-07-01 to 2015-03-30	SMOS, AMSR2	ASCAT-A
2015-03-31 to 2015-07-20	SMOS, AMSR2, SMAP	ASCAT-A
2015-07-21 to 2020-12-31	SMOS, AMSR2, SMAP	ASCAT-A, ASCAT-B

**Table S2.** Description of surface meteorological (atmospheric), soil and vegetation variables used for quantification of dominant drivers controlling SM-ET coupling and entropy production.

<b>Variable</b>	<b>Unit</b>	<b>Description</b>
Near-surface air temperature ( $T_{\text{air}}$ )	K	The temperature of air at 2 meters above the surface of land.
Near-surface specific humidity ( $q$ )	kg kg <sup>-1</sup>	The amount of moisture in the air divided by the amount of air plus moisture at that location.
Near-surface wind speed ( $u_{\text{surf}}$ )	m s <sup>-1</sup>	The horizontal wind speed at a height of 10 meters above the surface of the Earth.
Surface air pressure ( $p_{\text{atm}}$ )	Pa	The pressure (force per unit area) of the atmosphere at the surface of land.
Land Surface Temperature	K	Surface temperature of 0-5 cm depth soil profile.

### Dry-Delayed vs Wet-delay Systems

These systems are supported by moisture influx from deeper horizons, necessitating the decoupling of surface moisture with subsurface moisture dynamics. Such coupling is most likely to prevail in mixed forests and native prairies. On the contrary, wet-delayed systems incorporate regions with delayed transfer of increased SM to an increase in ET. These regions are most likely energy limited and are typically found at higher latitudes such as the cold deserts of Siberia.



**Figure S1.** Global seasonal maps of (a) proportional duration spent by the pixel in SM drying (SM<sub>AIF</sub> < 0) or SM wetting (SM<sub>AIF</sub> > 0), and (b) proportional duration spent by the pixel in

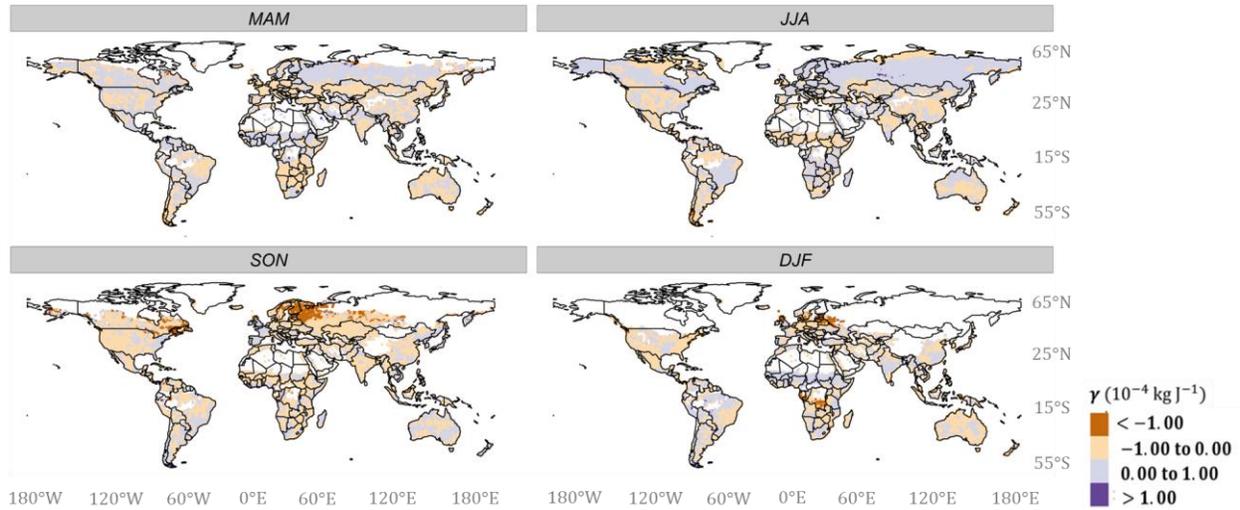
atmosphere heating ( $ET_{AIF} < 0$ ) or cooling ( $ET_{AIF} > 0$ ) for four seasons, namely, March through May—MAM, June through August—JJA, September through November—SON, and December through February—DJF. The color sequential follows a quantile division of data points. Missing/masked data are represented in white color.

**Table S3.** Statistical summary for seasonal Wasserstein distance, WD (-) across hydroclimates (IQR - Interquartile Range, SD - Standard Deviation, SE - Standard Error).

<b>Climate</b>	<b>WD Median</b>	<b>WD IQR</b>	<b>WD Mean</b>	<b>WD SD</b>	<b>WD SE</b>
Super Humid	0.806	0.172	0.834	0.147	0.002
Humid	0.993	0.351	1.017	0.229	0.002
Temperate	1.112	0.370	1.111	0.253	0.001
Arid	0.983	0.427	1.025	0.281	0.002
Hyper Arid	0.766	0.201	0.812	0.183	0.001

### Seasonal Slope Factor ( $\gamma$ )

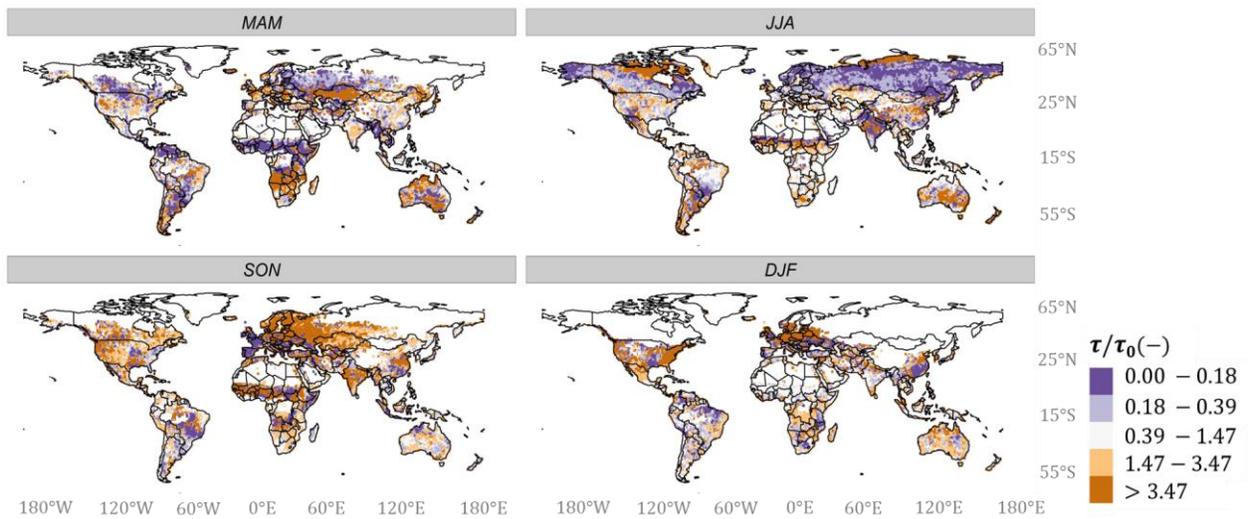
To compute the seasonal slope factor ( $\gamma$ ), we compared two different methods – (a) quantile regression, (b) piecewise linear regression. Quantile regression divided the time series of a location based on four quantiles (i.e., 25<sup>th</sup>, 50<sup>th</sup>, 75<sup>th</sup>, 100<sup>th</sup> percentile) while piecewise regression discretized the data based into chunks of 3 months. However, the resulting raster’s from both the methods did not have much difference, hence we selected piecewise linear regression for representing  $\gamma$  due to its conceptual proximity with the definition of “seasonality” (i.e., MAM, JJA, SON, DJF) used in the study.



**Figure S2.** Global maps of seasonal slope factor,  $\gamma$  expressed in  $\text{kg J}^{-1}$  for four seasons, namely, March through May—MAM, June through August—JJA, September through November—SON, and December through February—DJF, computed using piecewise linear regression. Missing/masked data are represented in white color.

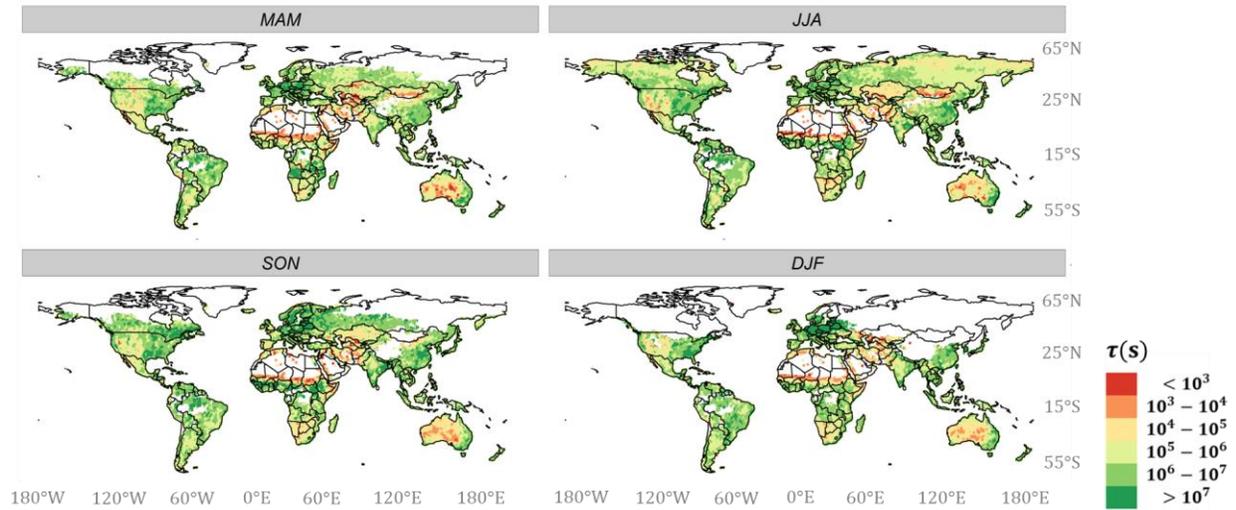
### Relative time ( $\tau/\tau_0$ )

Due to discretization of  $\gamma$ , the resulting dimensionless quantity ( $\tau/\tau_0$ ) might have few outliers which needs to be removed from further analysis. For example, if  $\gamma \rightarrow 0$ ,  $\tau/\tau_0 \rightarrow \infty$ ; such large outliers are impractical and were discarded from further analysis. Hence, values beyond 95th percentile were flagged out (i.e., values  $> 10000$  were set equal to *NA*). Furthermore, it is important to note that as  $\tau/\tau_0$  approaches zero (i.e.,  $\tau/\tau_0 \approx 0$  or  $\tau_0 \gg \tau$ ), it physically represents the case where effective conductivity ( $K_{eff}$ ) is very low or effective resistance ( $r_{eff}$ ) is very high. This is visible (Fig. S3) in energy limited regions of Northern Eurasia and North America in the season of JJA, while regions with highest perturbations in atmospheric conditions, reflect the most variations in  $\tau/\tau_0$ .



**Figure S3.** Global maps of relative time ( $\tau/\tau_0$ ) for four seasons, namely, March through May—MAM, June through August—JJA, September through November—SON, and December through February—DJF. The color sequential follows a quantile division of data points. Missing/masked data are represented in white color.

## Memory timescale ( $\tau$ )



**Figure S4.** Global maps of memory timescale ( $\tau$ ), expressed in secs, for four seasons, namely, March through May—MAM, June through August—JJA, September through November—SON, and December through February—DJF. Missing/masked data are represented in white color.

**Table S4.** Statistical summary for seasonal entropy production,  $\Sigma^*$  ( $10^{-6}$  J m<sup>-2</sup> K<sup>-1</sup>) across hydroclimates (IQR - Interquartile Range, SD - Standard Deviation, SE - Standard Error).

<b>Climate</b>	<b><math>\Sigma^*</math> Median</b>	<b><math>\Sigma^*</math> IQR</b>	<b><math>\Sigma^*</math> Mean</b>	<b><math>\Sigma^*</math> SD</b>	<b><math>\Sigma^*</math> SE</b>
Super Humid	24.200	6.670	24.300	5.250	0.0793
Humid	20.800	6.490	21.100	4.880	0.0422
Temperate	18.900	6.110	19.000	4.770	0.0212
Arid	14.000	8.530	13.800	5.680	0.0256
Hyper Arid	3.340	2.780	4.140	2.840	0.0186

**Table S5.** Statistical summary for seasonal entropy production,  $\Sigma^*$  ( $10^{-6} \text{ J m}^{-2} \text{ K}^{-1}$ ) across terrestrial ecosystems (IQR - Interquartile Range, SD - Standard Deviation, SE - Standard Error).

<b>IGBP</b>	<b><math>\Sigma^*</math> Median</b>	<b><math>\Sigma^*</math> IQR</b>	<b><math>\Sigma^*</math> Mean</b>	<b><math>\Sigma^*</math> SD</b>	<b><math>\Sigma^*</math> SE</b>
<b>Forests</b>	21.100	5.980	21.400	4.600	0.029
<b>Savannas</b>	18.600	7.280	18.300	6.440	0.031
<b>Croplands</b>	14.900	8.210	14.500	5.640	0.040
<b>Shrublands</b>	14.100	15.000	12.400	7.950	0.047
<b>Grasslands</b>	10.100	9.630	11.100	6.230	0.029
<b>Barren</b>	4.240	13.000	8.4500	7.750	0.195

**Table S6.** Dissipative energy barriers ( $\Delta e$ ,  $10^{-6} \text{ J m}^{-2} \text{ K}^{-1}$ ) between terrestrial ecosystems computed as the difference between median entropy thresholds ( $\Sigma^*$  Median from Table 5).

	<b>Forests</b>	<b>Savannas</b>	<b>Croplands</b>	<b>Shrublands</b>	<b>Grasslands</b>	<b>Barren</b>
<b>Forests</b>	0.000					
<b>Savannas</b>	2.400	0.000				
<b>Croplands</b>	6.200	3.700	0.000			
<b>Shrublands</b>	7.000	4.500	0.800	0.000		
<b>Grasslands</b>	11.000	8.500	4.800	4.000	0.000	
<b>Barren</b>	16.860	14.360	10.660	9.860	5.860	0.000