

What do models get right about changes in evapotranspiration?

Attribution of mechanisms of ET change

Dr. Kirsten Findell
Geophysical Fluid Dynamics Laboratory

EPESC-LEADER JOINT SCIENCE MEETING

15-18 JULY 2025

*WITH HSIN HSU (STARTING AT NATIONAL TAIWAN UNI SOON!),
STEPHAN FUEGLISTALER, PAUL DIRMAYER, MIN-HUI LO, ANDREW FELDMAN,
SHA ZHOU, EUNKYO SEO, AND DIEGO MIRALLES*

Evapotranspiration: The Key Link

Evapotranspiration:

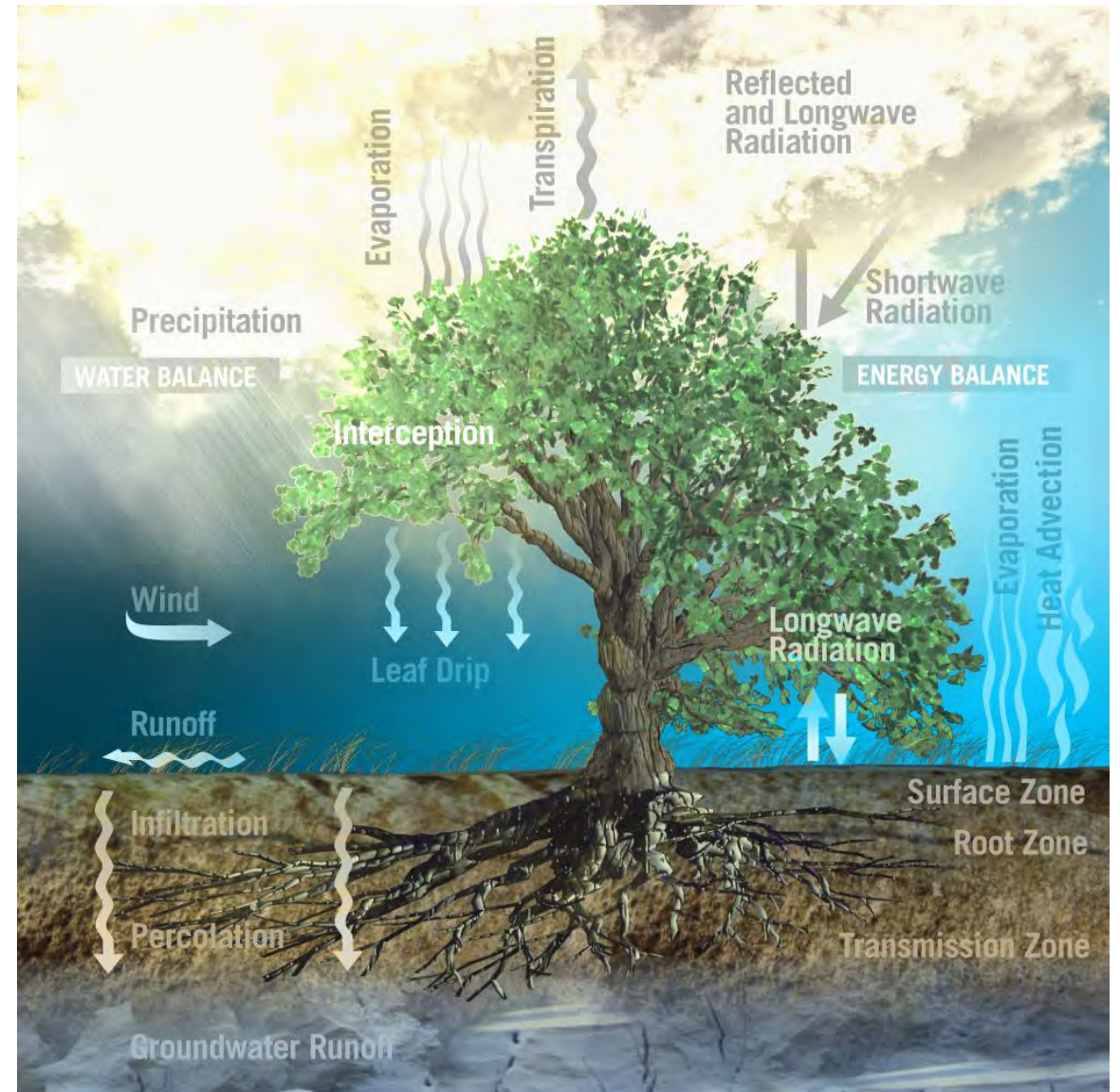
connects the land and the atmosphere,

connects the energy and water cycles

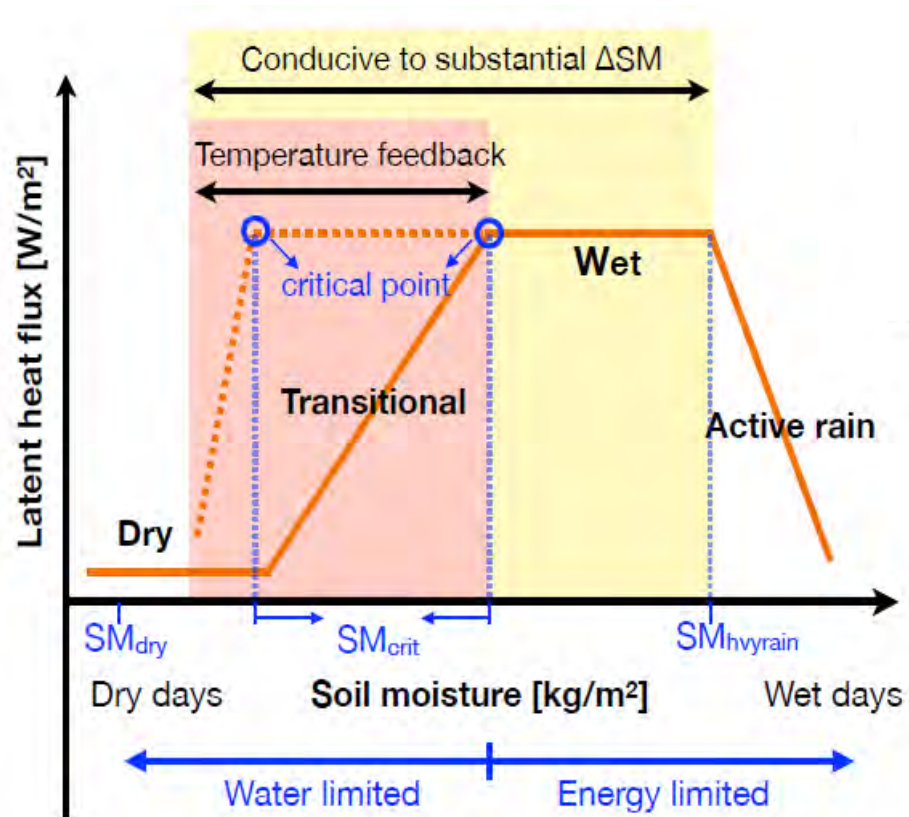
ET is sometimes limited by water availability,
sometimes limited by energy availability

Driven by sub-daily processes... high-temporal
frequency data are required (Findell et al., 2024)

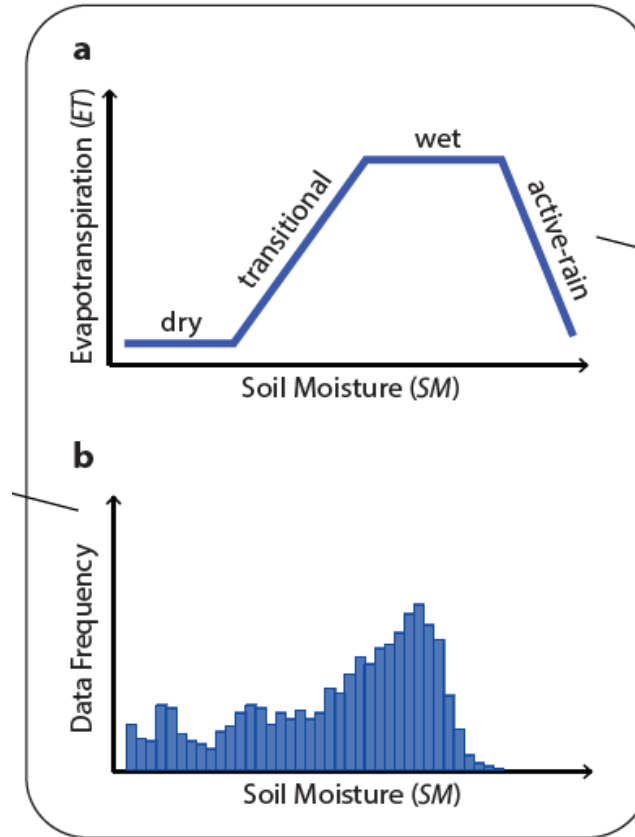
[ESA-AOES Medialab](#)



The non-linear relationship between Soil Moisture and ET



Duan, Findell and
Fueglistaler, 2023, GRL



Hsu, Findell, et al., subm.

It's not just the local
SM-ET relationship that
matters,

It's also the PDF of SM:
*how SM availability
aligns with that SM-ET
relationship*

Calculating ET change

Mean ET:

sum across all individual observations

$$\overline{ET} = \frac{1}{n_{obs}} \sum_{i=1}^{n_{obs}} ET_i$$

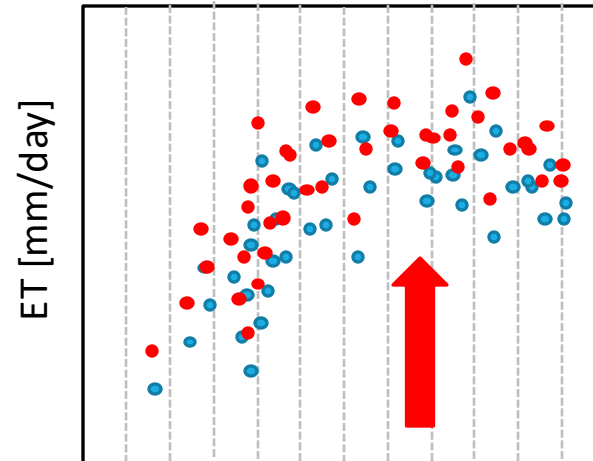
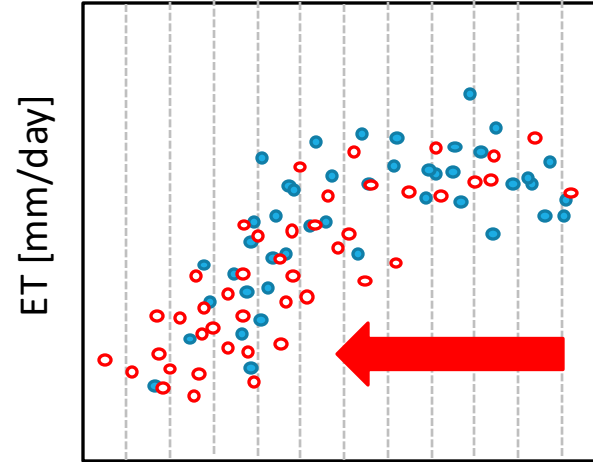
OR:

bin the data,

sum across (number in bin * mean ET in bin)

$$\overline{ET} = \frac{1}{n_{obs}} \sum_{i=1}^{k_{bins}} (n_i \times \overline{ET}_i)$$

Using a simple frequency-intensity decomposition



Change in ET between times 1 and 0:

$$\Delta \overline{ET} = \overline{ET}_1 - \overline{ET}_0$$

Could come from a change in number of obs in each bin

Change in Soil Moisture availability

OR

Could come from a change in mean ET in each bin

Change in the SM—ET relationship

Could reflect aerosol-radiative effects, LULCC, vegetation dynamics, etc.

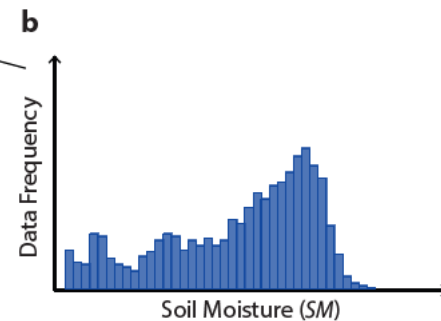
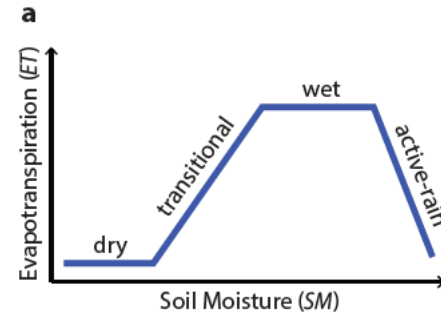
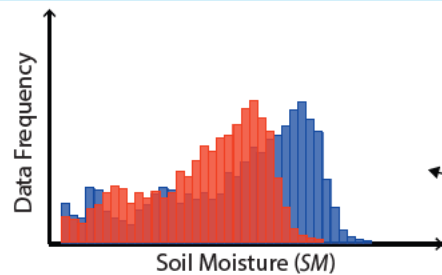
Pathways of ET change

$$\Delta \overline{ET} = ET_M + ET_N + ET_C$$

$$\Delta \overline{ET} = \overline{ET}_1 - \overline{ET}_0 = \frac{1}{n_{obs}} \sum_{i=1}^{k_{bins}} ((\Delta n_i \times \overline{ET}_{0i}) + (n_{0i} \times \Delta \overline{ET}_i) + (\Delta n_i \times \Delta \overline{ET}_i))$$

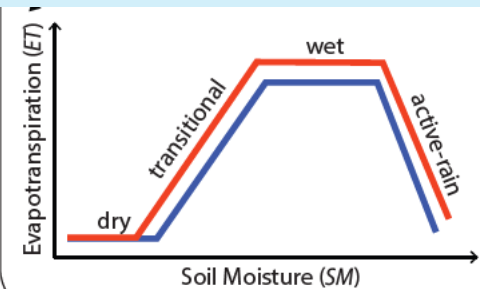
Moisture pathway

$$ET_M = \sum (\Delta n \times \overline{ET}_0)$$



Non-moisture pathway

$$ET_N = \sum (n_0 \times \Delta \overline{ET})$$



**Combined pathway
(Nonlinear term)**

$$ET_C = \sum (\Delta n \times \Delta \overline{ET})$$

Datasets considered

GLEAM4 as the observationally-based reference (Miralles et al., 2025)

- GLEAM4 provides data of the different components of evapotranspiration by maximizing the recovery of information on ET contained in current satellite observations of climatic and environmental variables
- Soil moisture constraints on evaporation; Detailed treatment of forest interception; Extensive use of microwave observations
- Long-term continuous records

CMIP6 historical simulations with **daily** soil moisture (SM) and evapotranspiration (ET) available

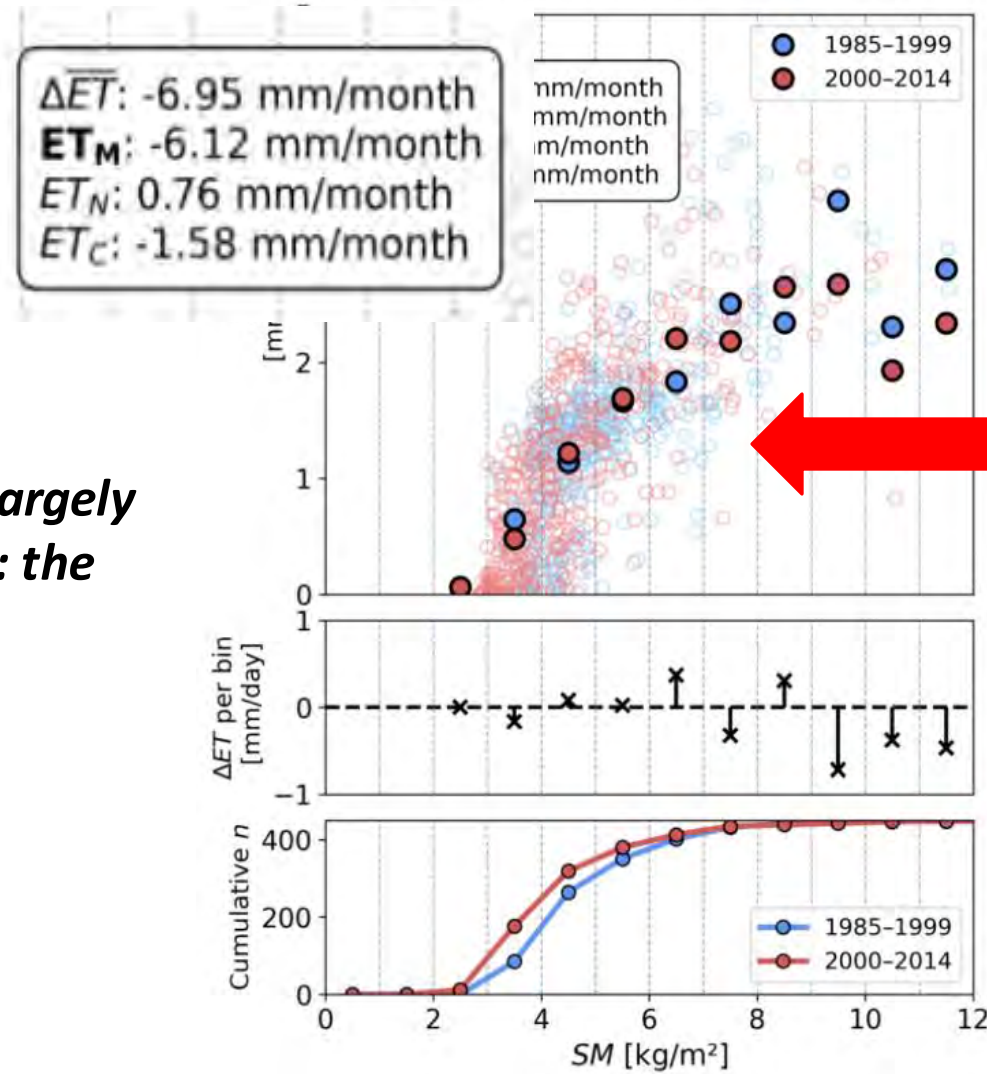
- 14 models

30 years of overlap: consider differences between two 15-year time blocks

- 2000-2014 compared to 1985-1999

Examples using GLEAM4 data

a Northwestern Mexico (27.05°N, 109.85°W, June)



The change in ET is largely attributable to ET_M : the moisture pathway

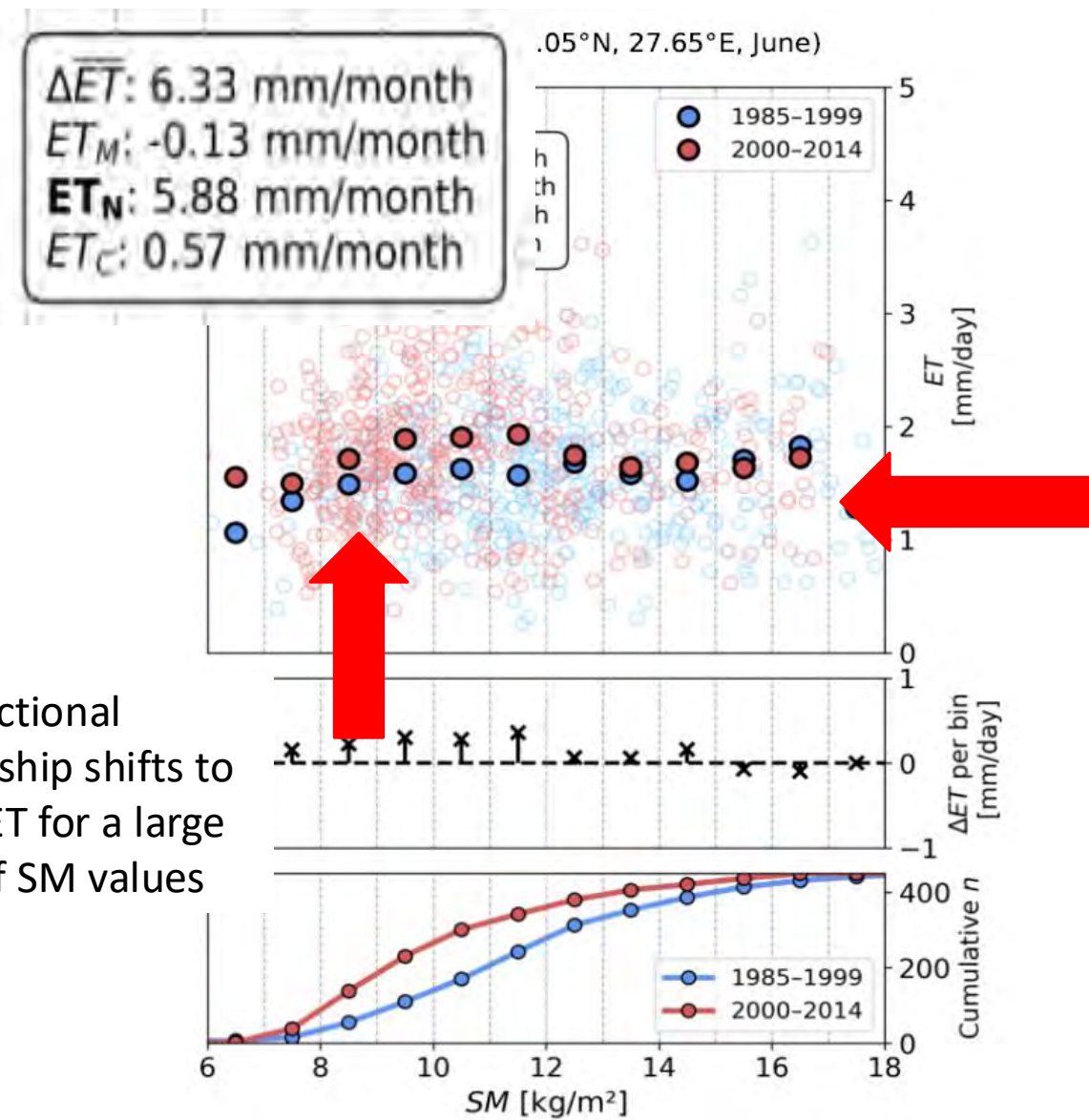
The PDF shifts left, towards drier conditions

In the transitional regime: strong sensitivity of ET to SM

Examples using GLEAM4 data

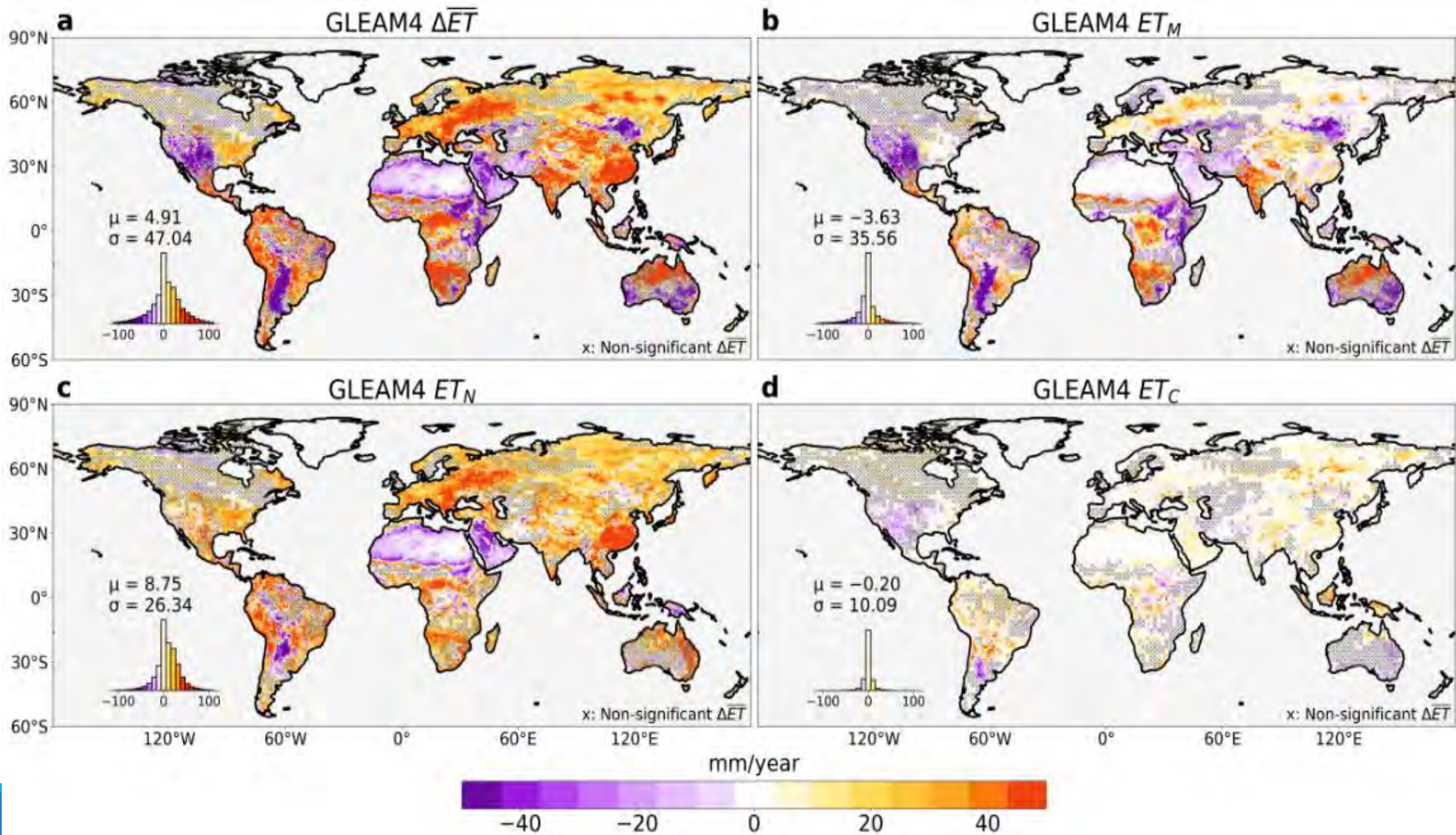
The change in ET is largely attributable to ET_N : the non-moisture pathway

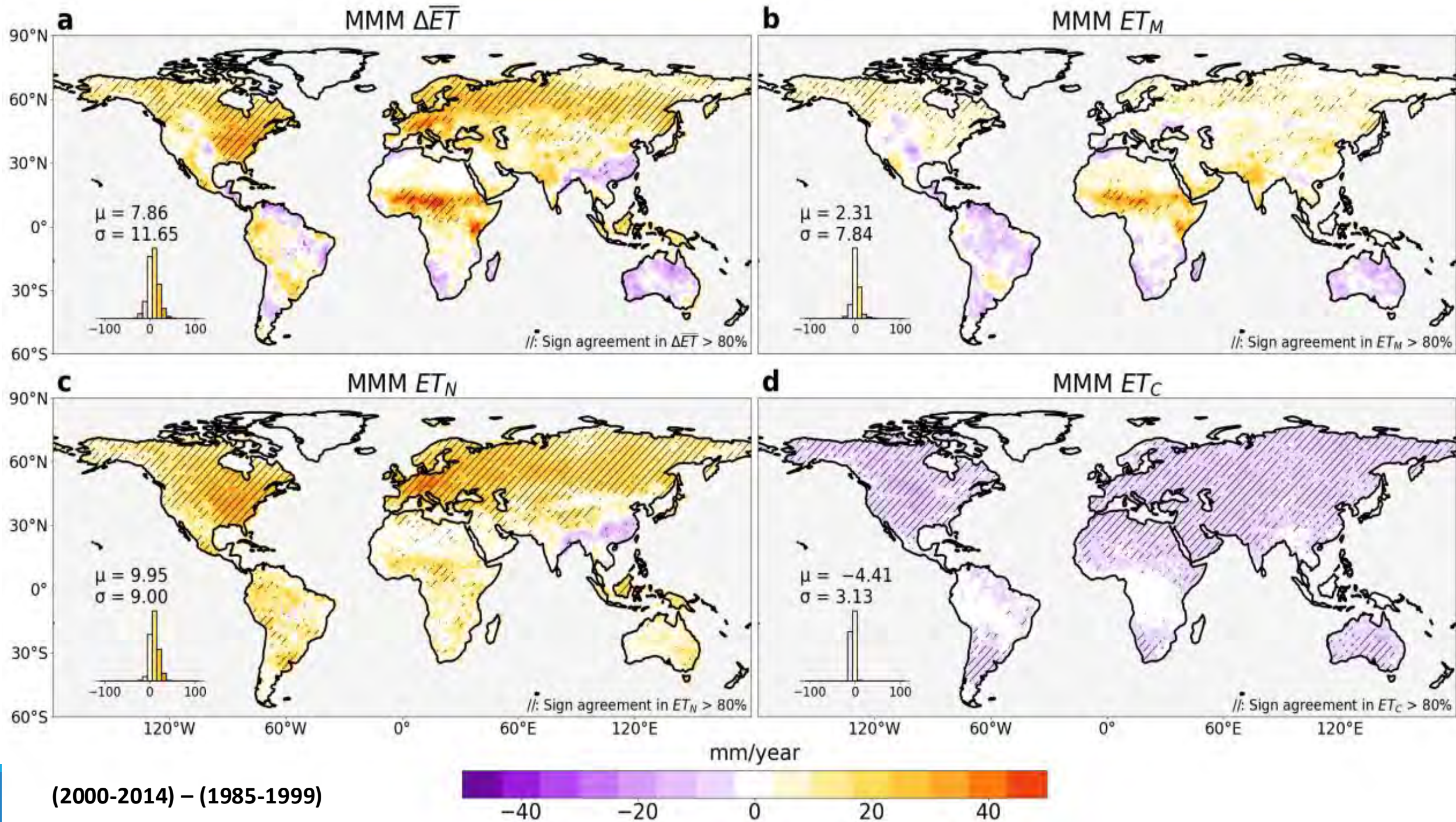
The functional relationship shifts to higher ET for a large range of SM values



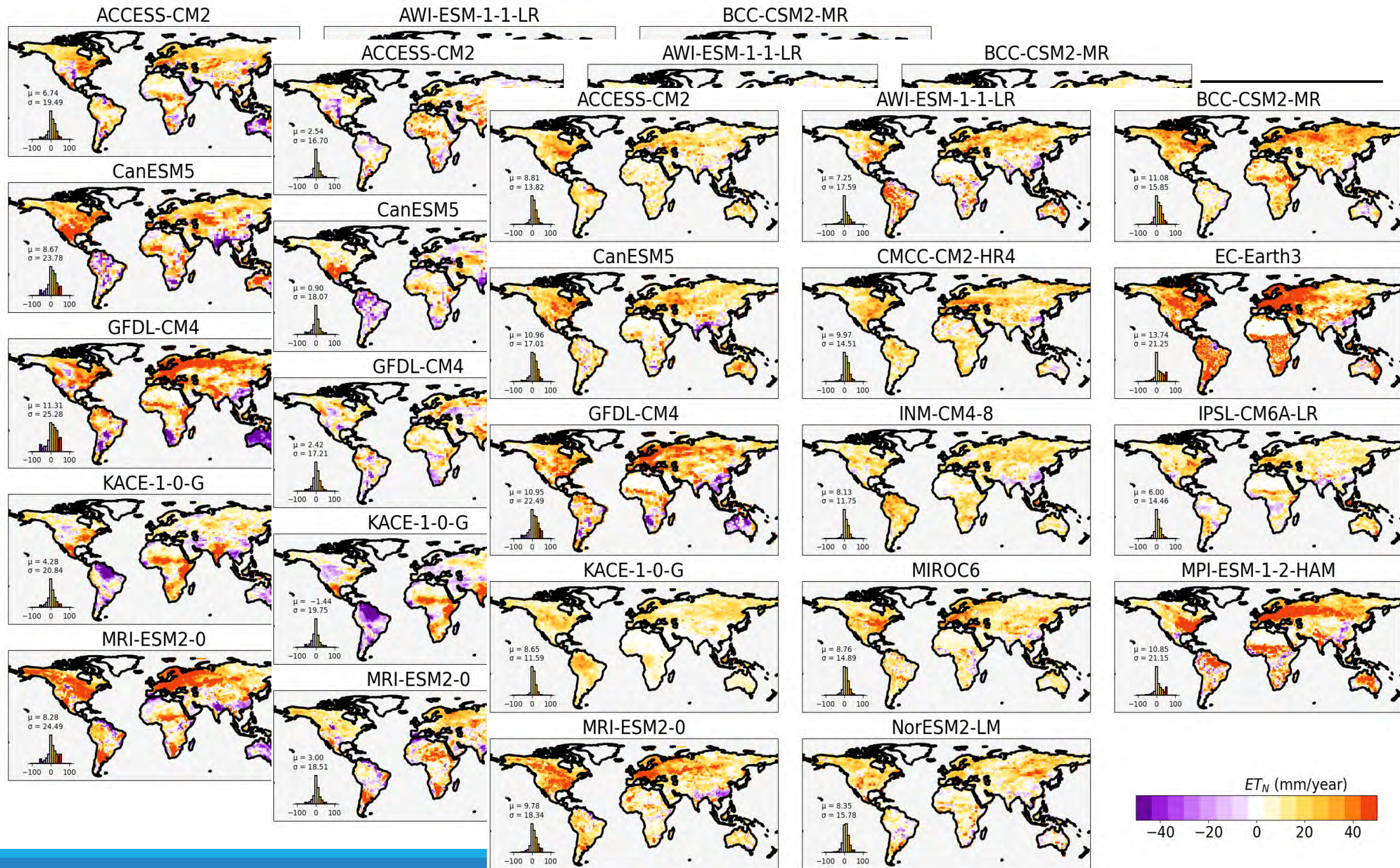
The PDF shifts left, towards drier conditions

In the wet regime: little sensitivity of ET to SM





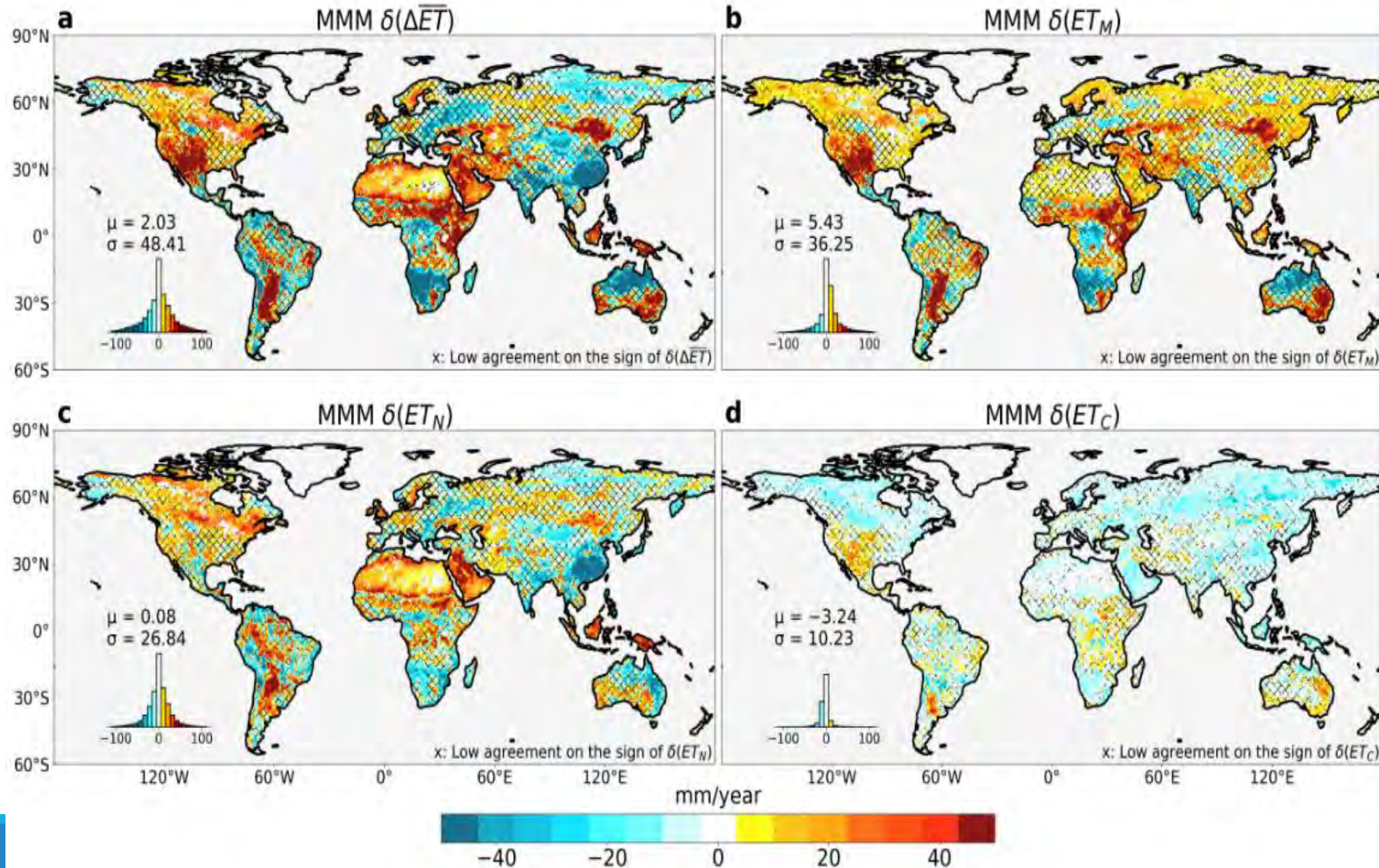
ΔET ,
 ET_M ,
 ET_N in
different
models



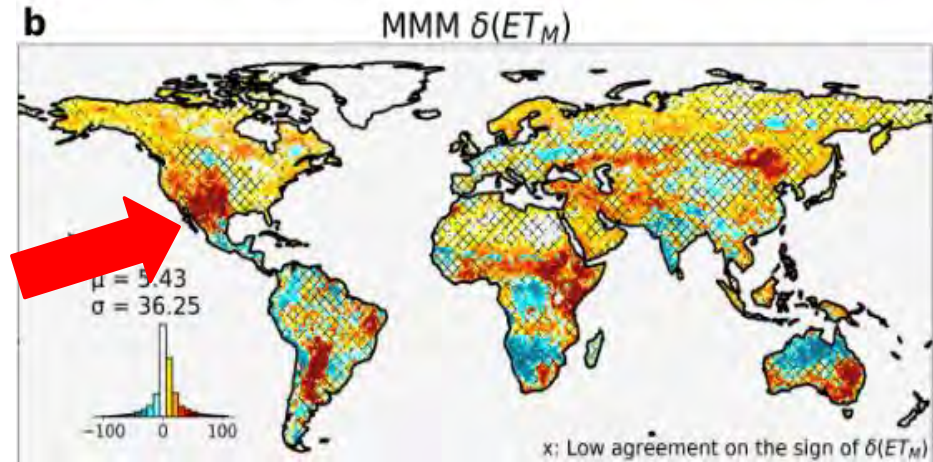
δ : individual model minus GLEAM

Deviation from GLEAM.

Where $MMM=0$, sign of δ will be opposite of GLEAM signal.



δ : individual model minus GLEAM



US SW: Model deviations from GLEAM attributable to the moisture pathway ET_M

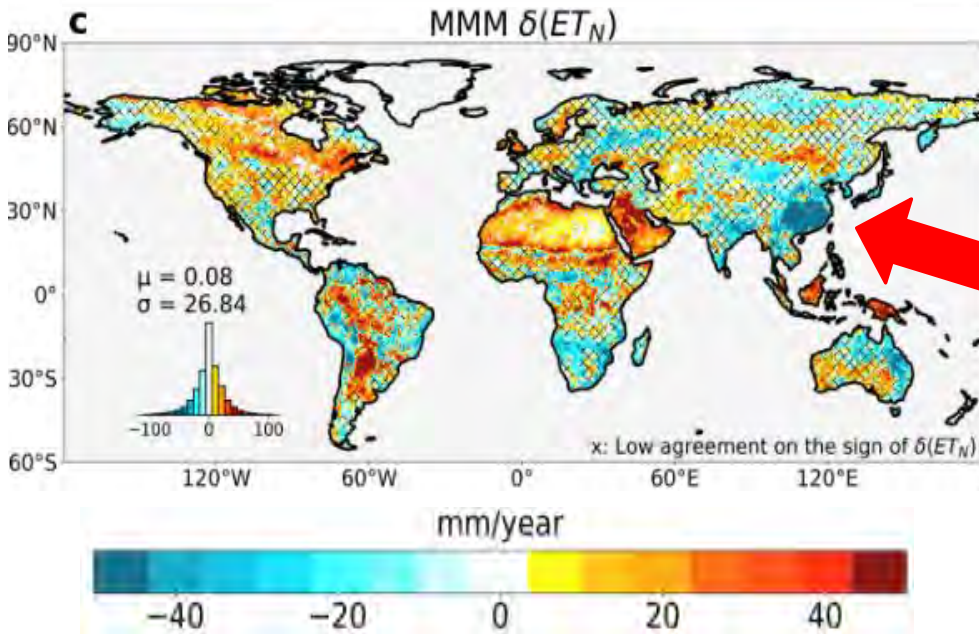
- GLEAM shows substantial ET decline, consistent with Simpson et al (2024) and others
- ET in most models changes very little, *largely because the SM PDF does not dry out enough*

SE China: Model deviations from GLEAM attributable to the non-moisture pathway ET_N

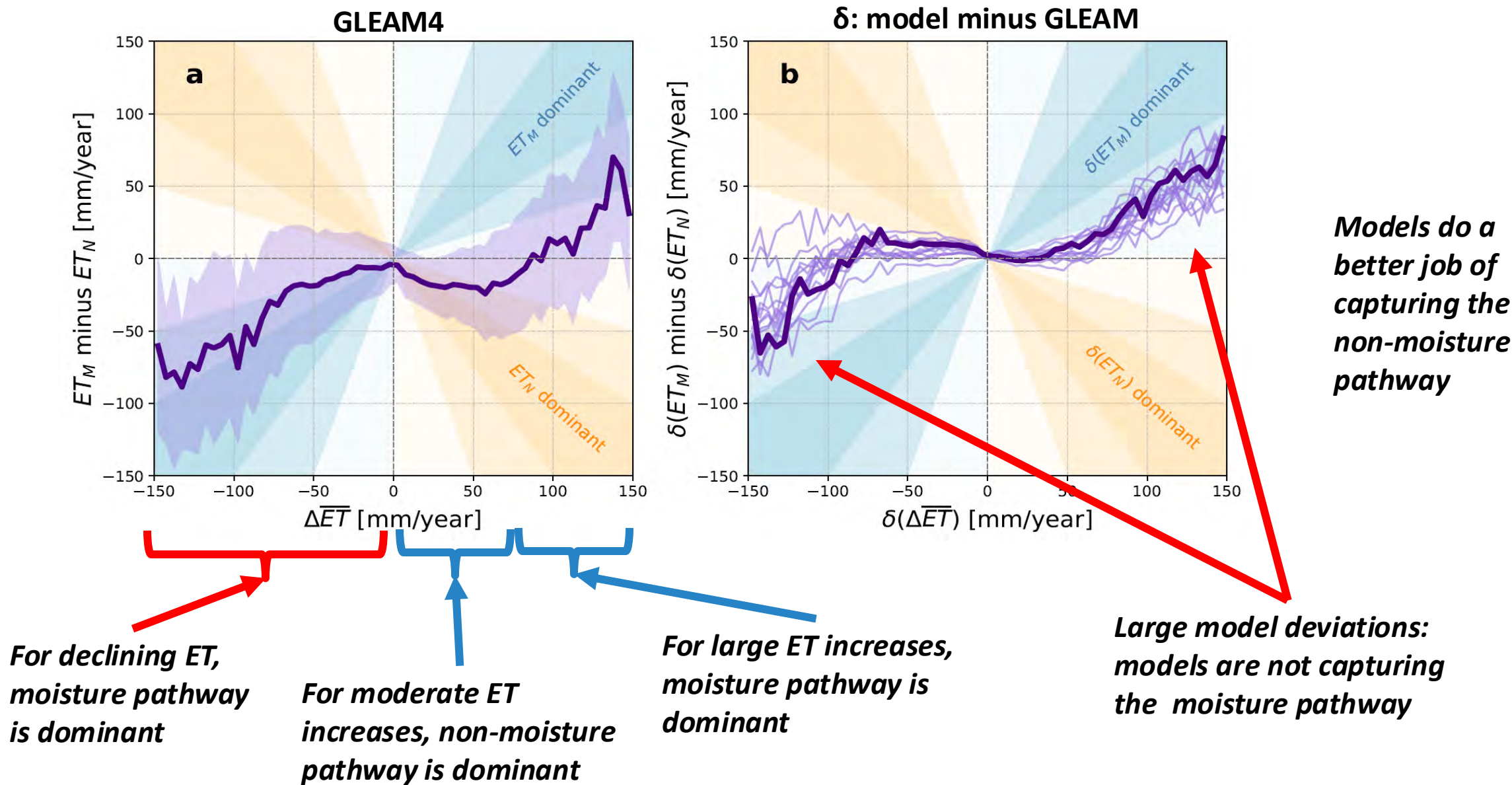
- GLEAM shows ET increases in this region
- While the models capture most ET_N -forced signals, this area is entirely missed
- Wang et al (2021): “*Incorrect Asian aerosols affecting the attribution and projection of regional climate change in CMIP6 models*”

What do the models fail to capture?

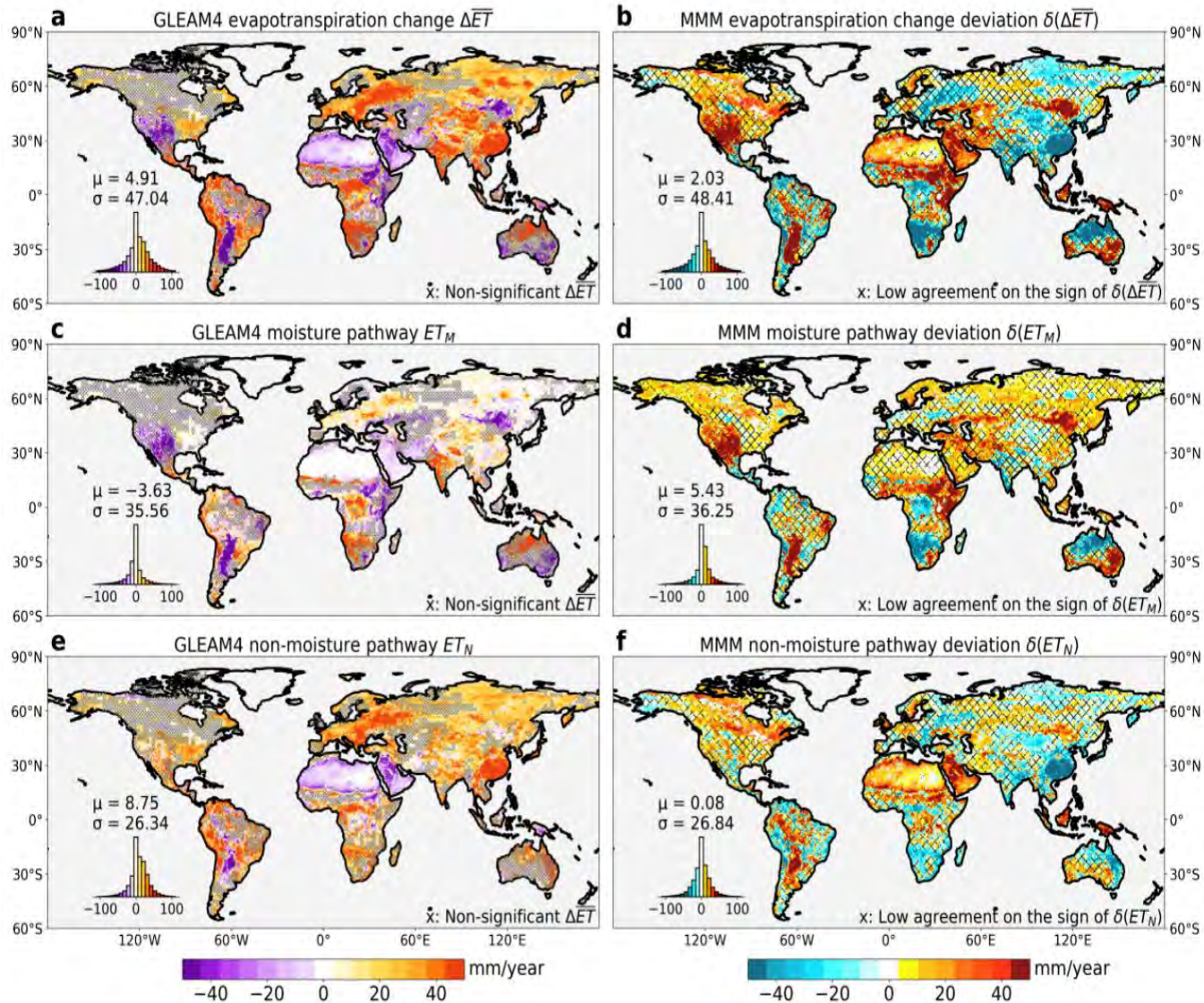
- Vegetation representation? Soil physics of arid regions?
- Aerosol forcing? SST biases?
- *Analyses of LESFMIP simulations could help sort this out*



Attributable responses



Summary



- ❖ An empirical approach separates ET changes into pathways related to
 - (1) **Soil Moisture availability** and
 - (2) **Other factors** (e.g., radiation, aerosols, vegetation dynamics, etc.)
- ❖ Observationally-based data show complicated regional changes in ET
- ❖ The biggest changes are attributable to Soil Moisture availability
→ *models fail to capture this pathway*
- ❖ Models do a better job capturing moderate increases in ET attributable to the non-moisture pathway

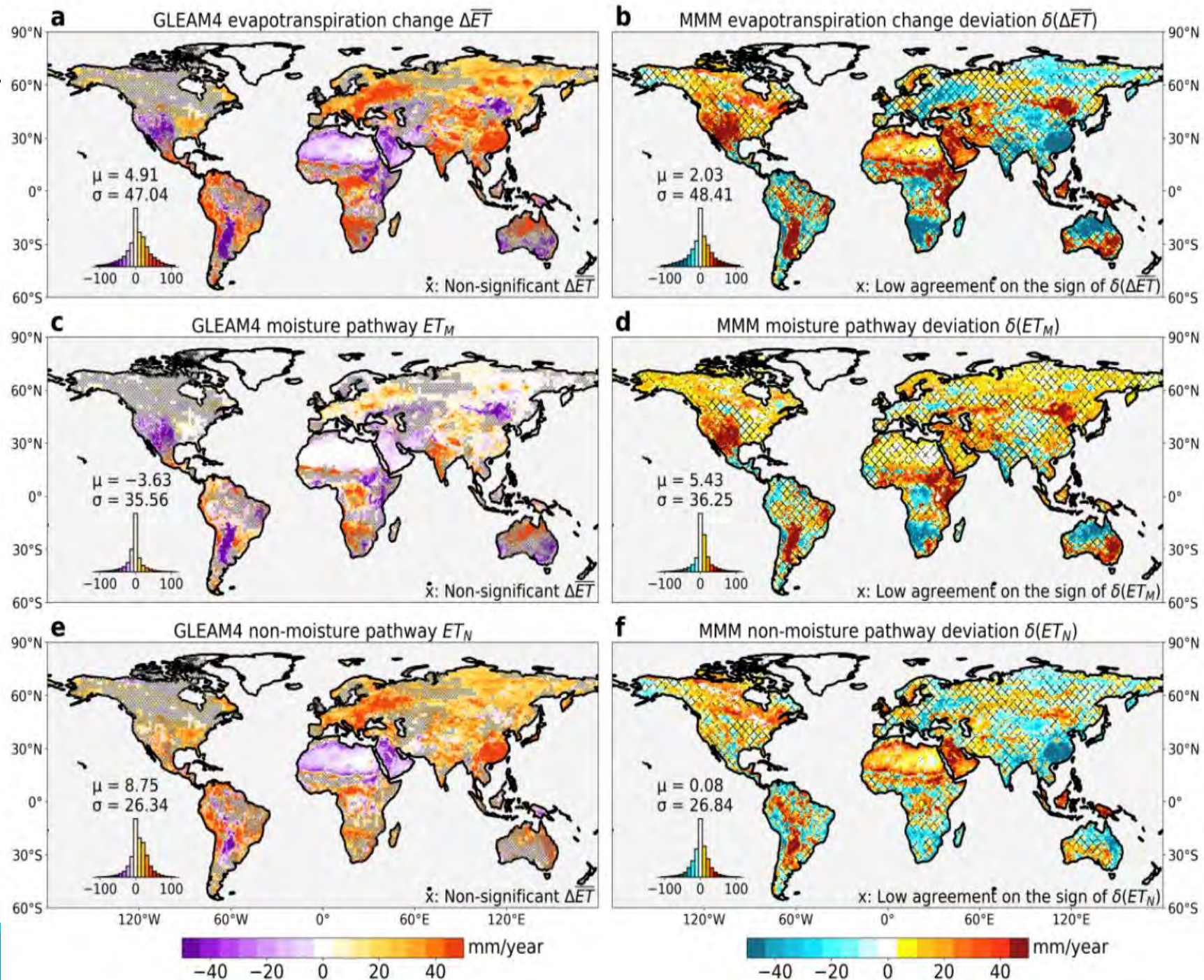
Thank you!

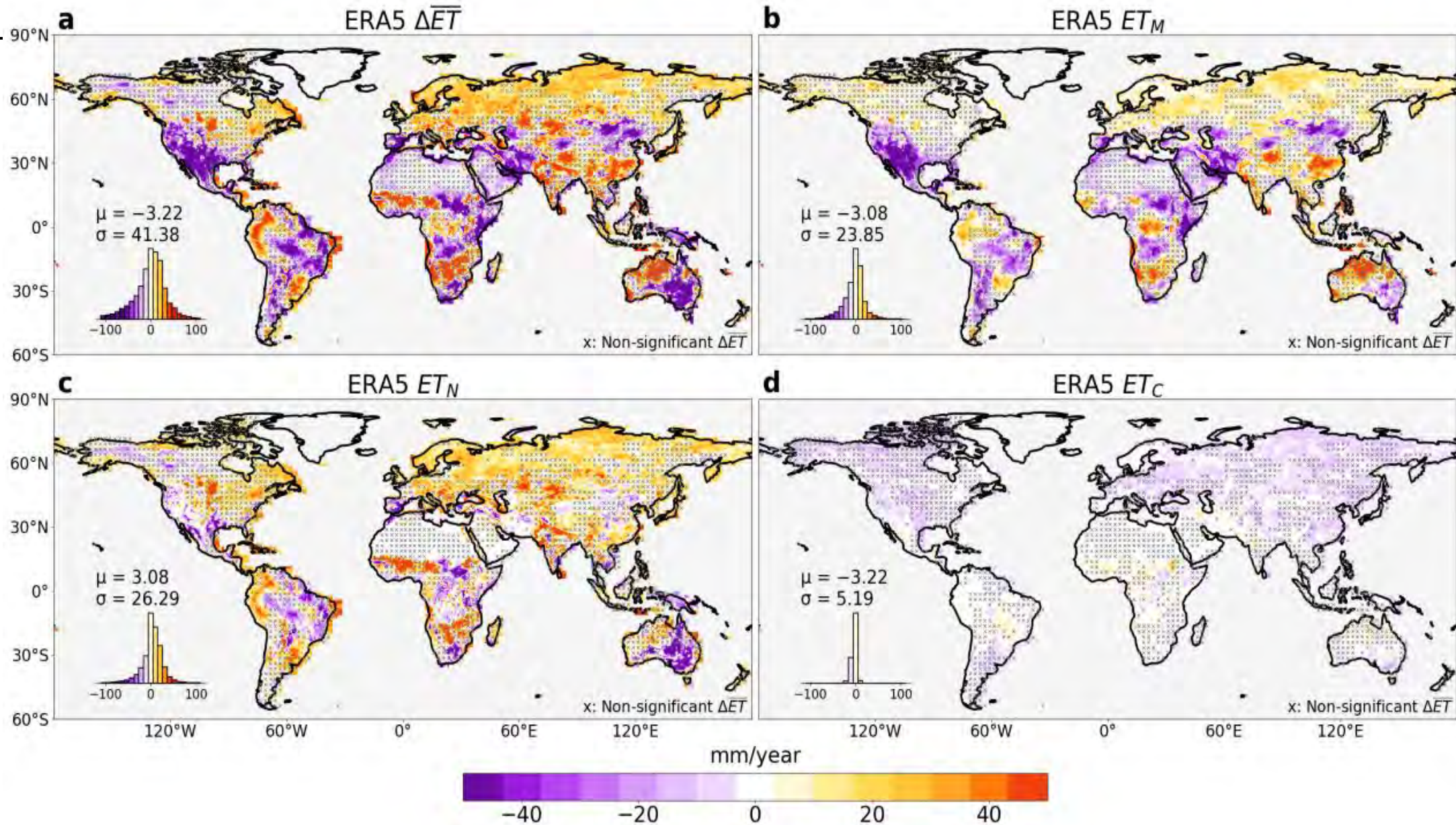


Dr. Hsin Hsu
Faculty position at NTU
beginning next month

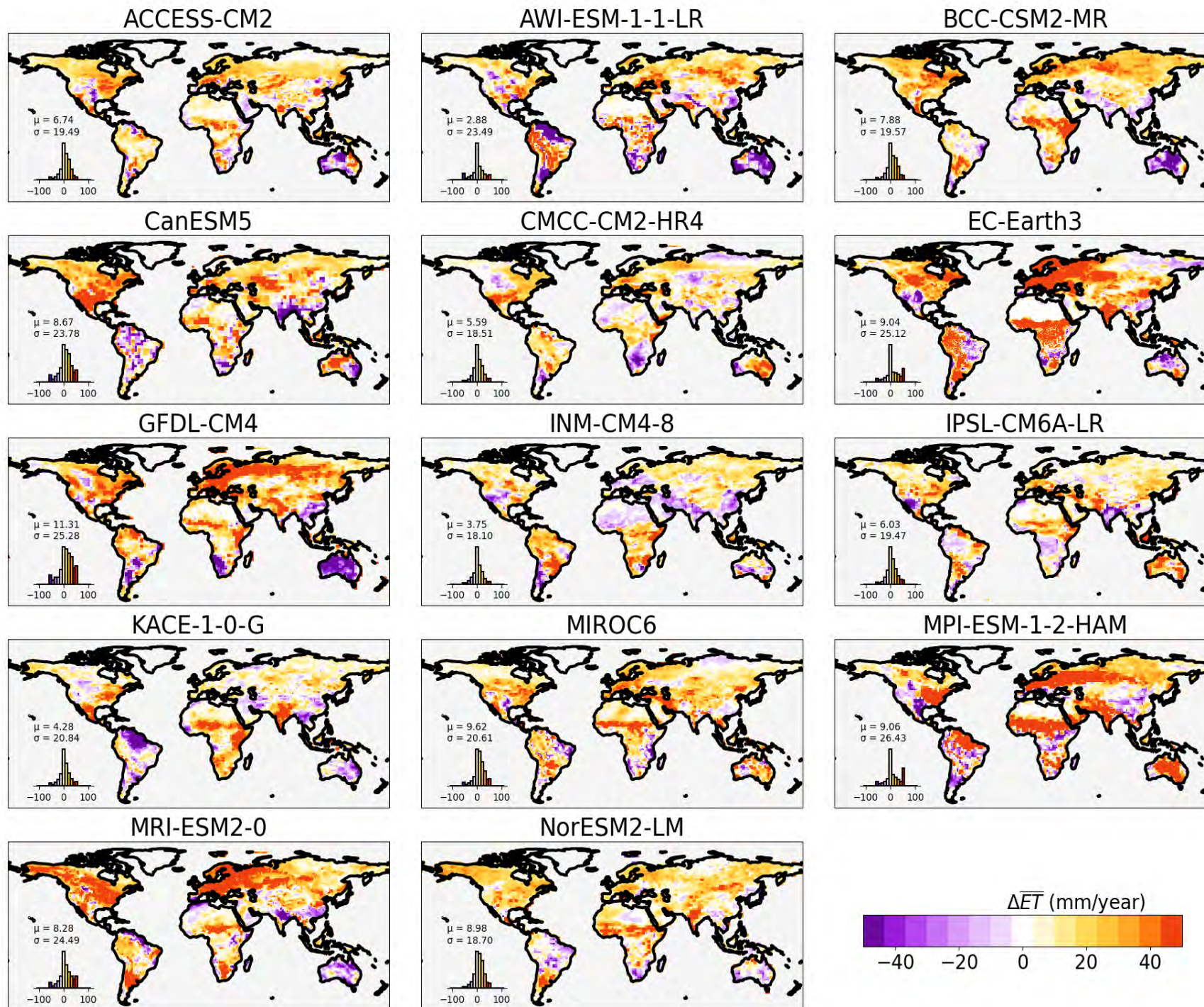
KIRSTEN: KIRSTEN.FINDELL@NOAA.GOV

HSIN: HH9736@PRINCETON.EDU

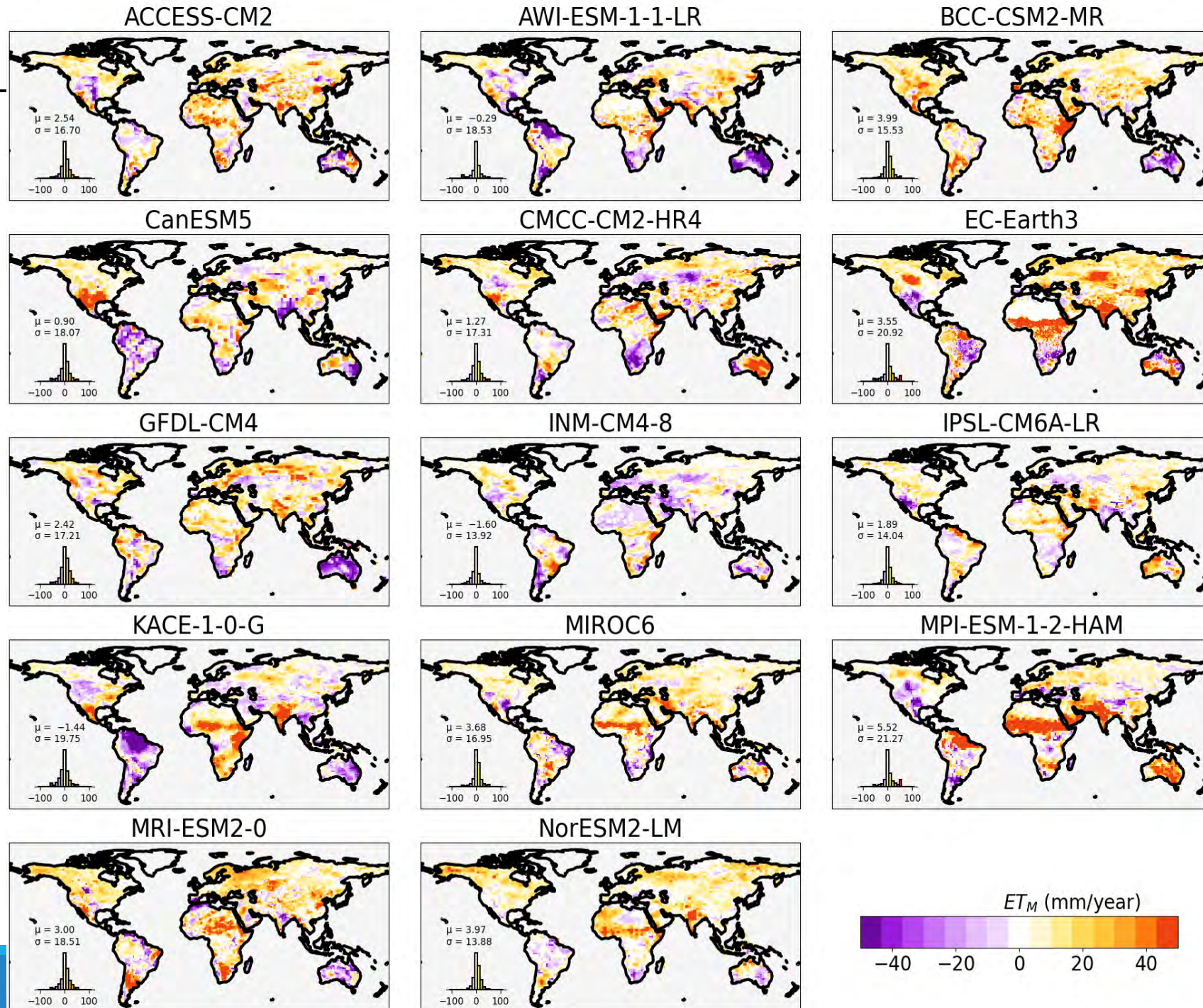




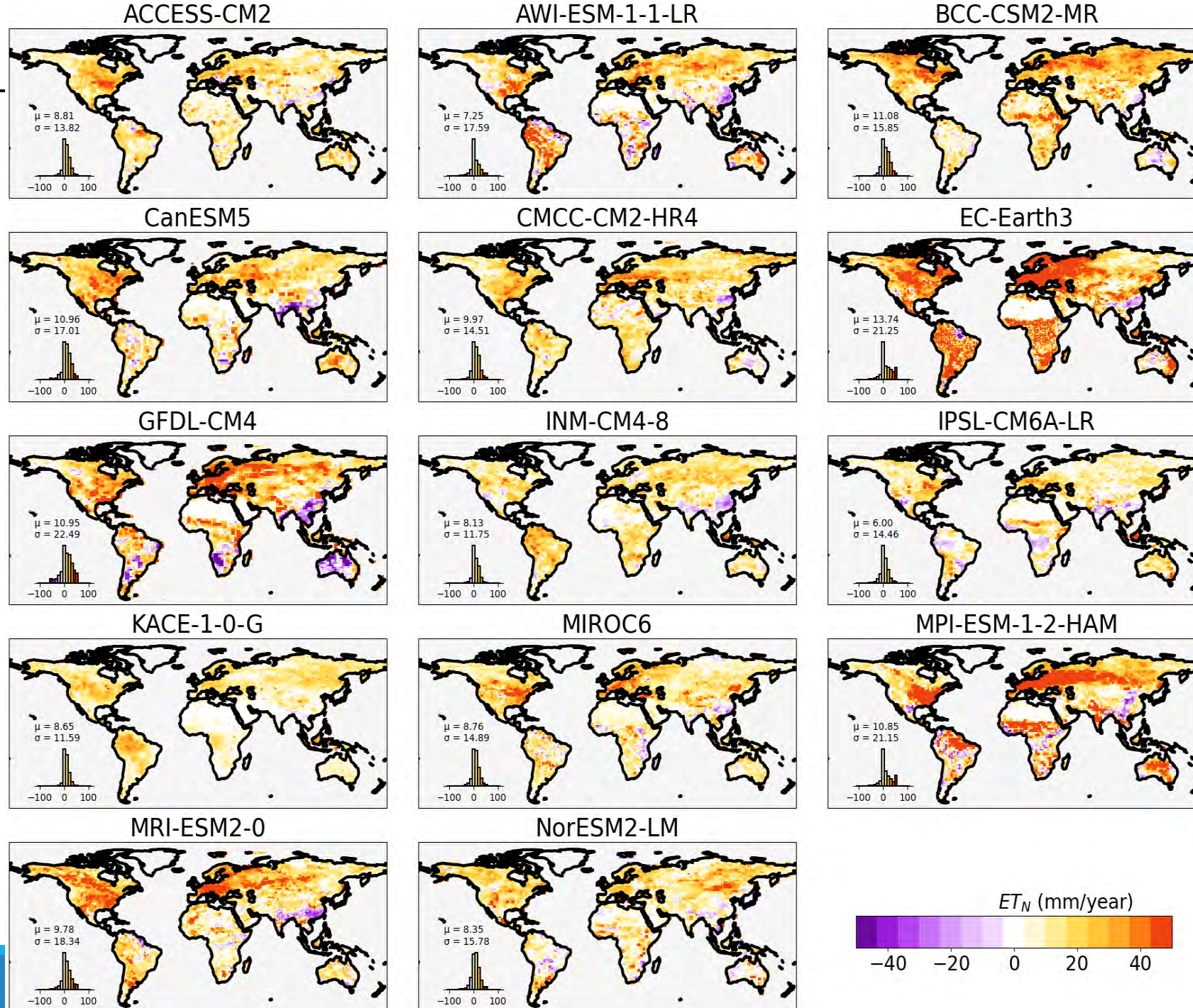
ΔET in different models



ET_M in different models



ET_N in different models



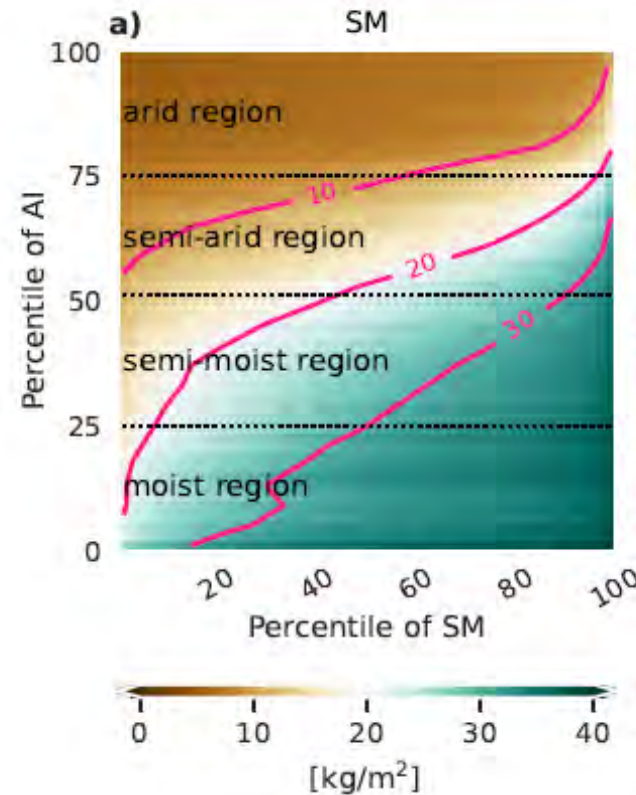
The Soil Moisture - Aridity Index (SM-AI) Percentile phase space

X-axis: daily SM percentiles

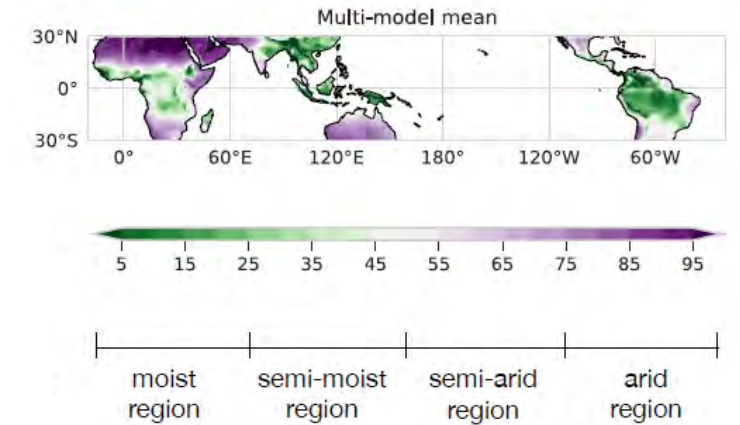
- Provides a visual of the PDF of soil moisture
- Can be linked to percent of time a given location is in different SM-ET behavioral regimes

Y-axis: Climatological Aridity Index (AI) percentiles

- An indication of the broad context of each grid point



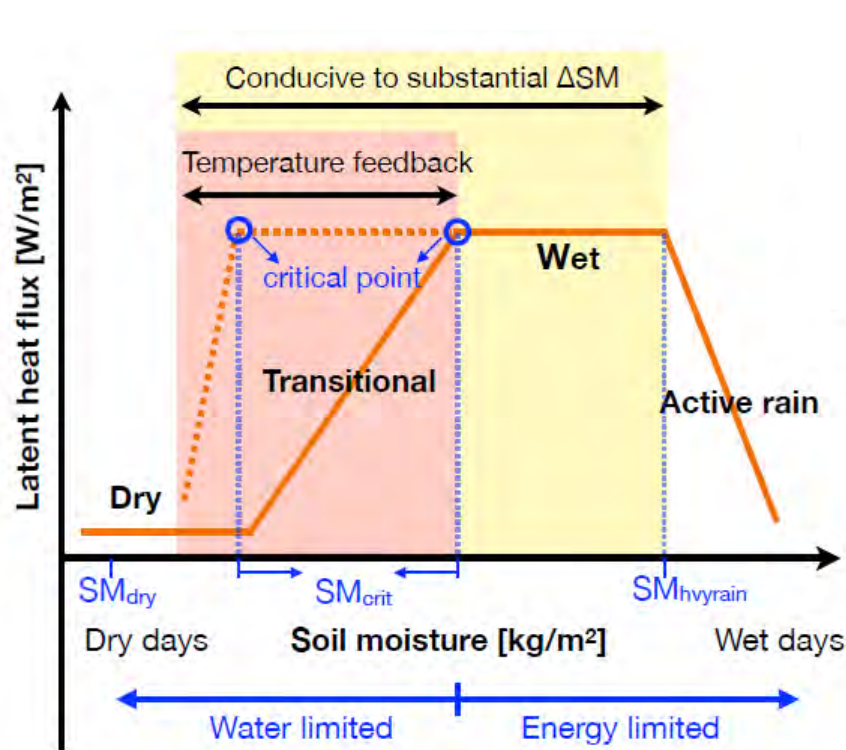
b) Percentile of climatological warm-season Aridity Index



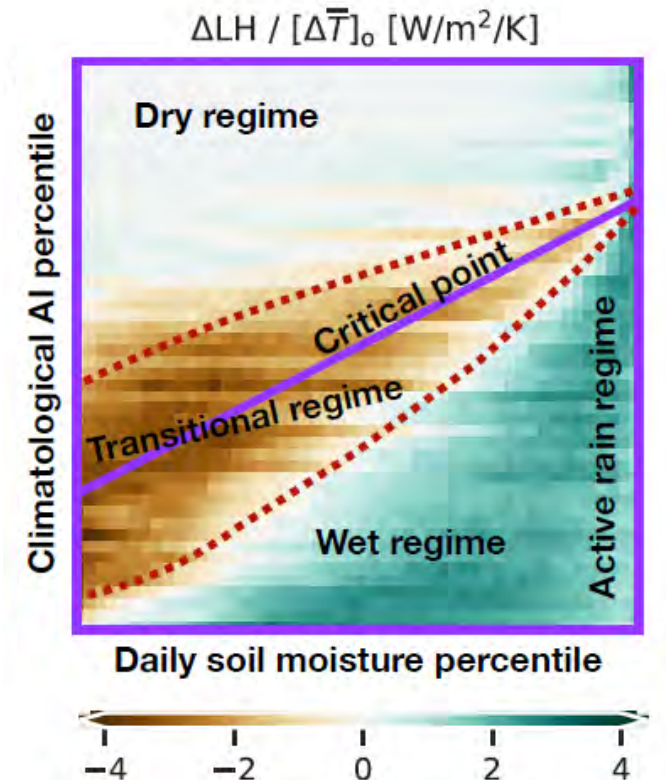
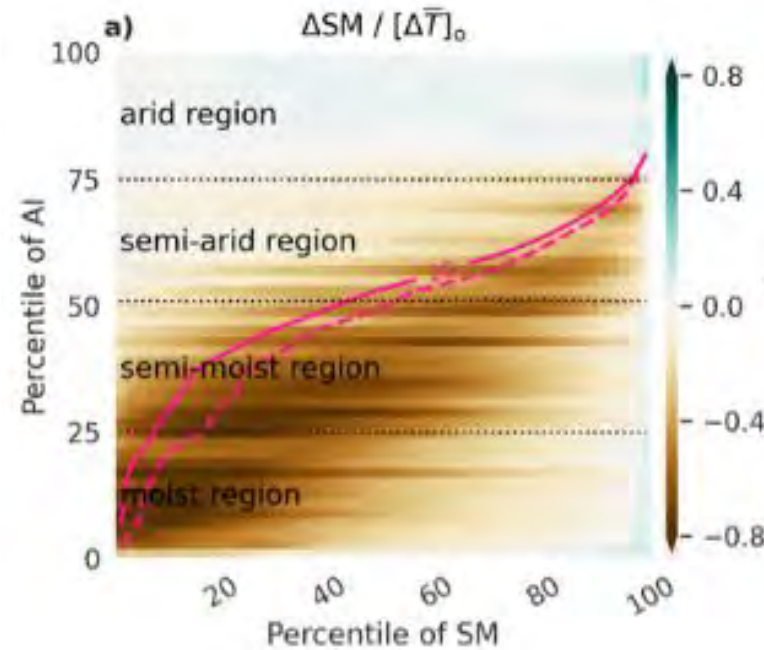
Following Milly and Dunne (2016, 2017):

$$\text{Aridity Index} = 0.8 \frac{\text{Radiation}_{\text{net}}^{\text{sfc}}}{L_v \cdot \text{Precipitation}}$$

The SM-AI Percentile phase space



Duan, Findell, and
Fueglistaler, 2023, *GRL*



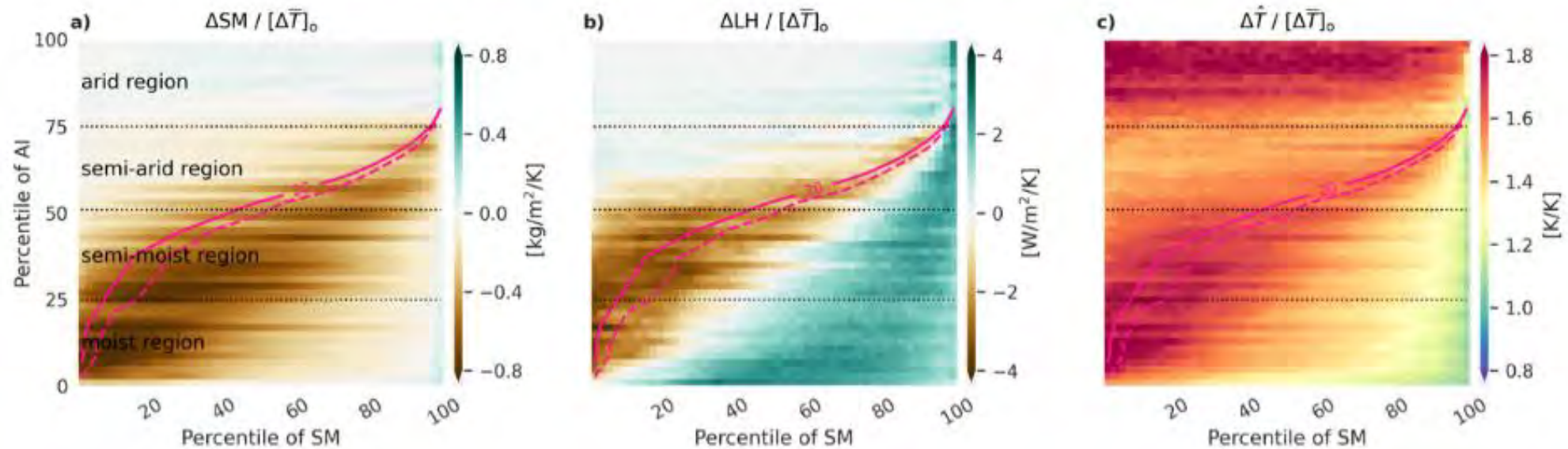
Using 8 CMIP5 models, $4\times\text{CO}_2$ – preindustrial control experiments

20 kg/m^2 loosely aligns with the critical point

Temperature response

Duan, Findell, and Fueglistaler, 2023, *GRL*

Multi-Model Mean, 4xCO₂ – piCtrl, Tropics 30S–30N, Warm Season



Daily maximum temperature increases are most pronounced:

- In arid regions: uniform increase on all days
 - PDF change: uniform shift to warmer temperatures
- In moist and semi-moist regions: on days with moisture limitation
 - PDF change: warm shift + lengthening of the warmest tail

