

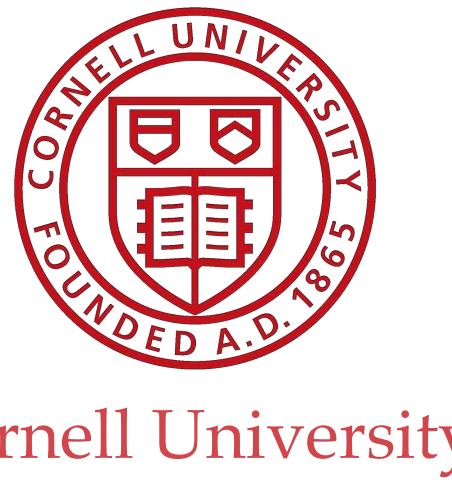
Is the forced response of precipitation timescale-dependent?

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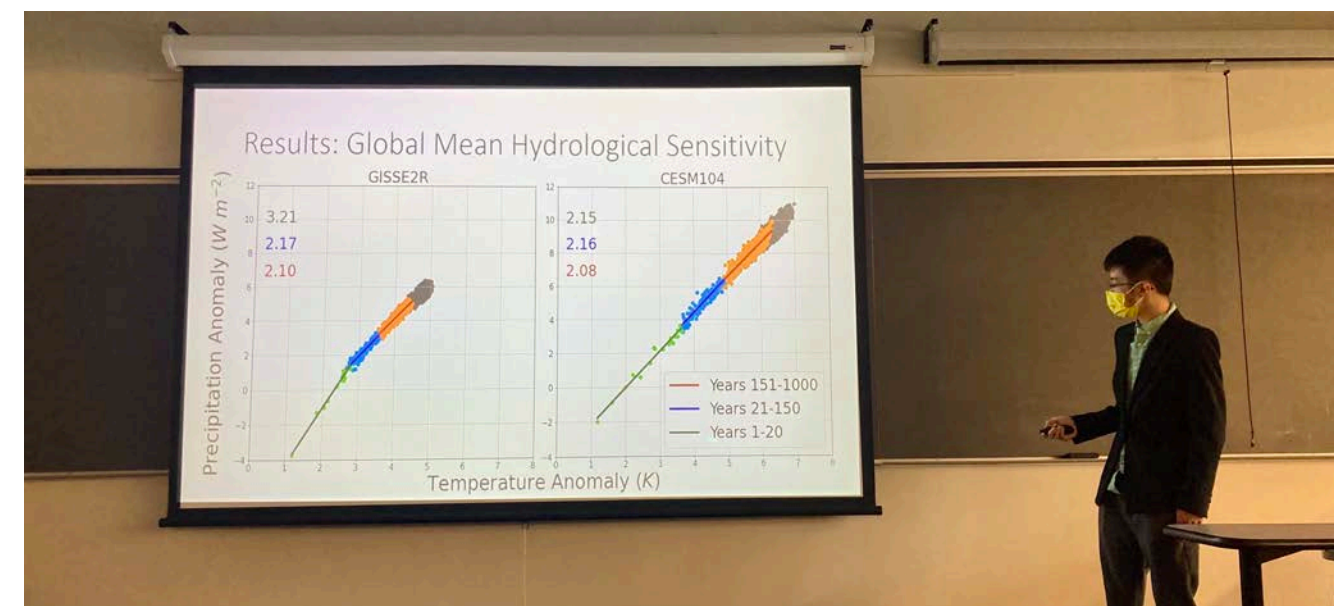
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Angie



Kinen giving his senior thesis defense on this work, 6 May 2022

The Pattern Effect: Coupling of SST Patterns, Radiative Feedbacks, and Climate Sensitivity Workshop, May 10 - May 13, 2022

Synopsis

Issue:

- The pattern effect leads to time-dependence of climate sensitivity via the multiple response timescales of TOA net radiative fluxes. Since the global-mean precipitation is in balance with the atmospheric radiative cooling (=TOA + surface fluxes) plus sensible heat flux, is its response to forcing timescale-dependent too?
- How does precipitation change in response to abrupt 4xCO2 forcing across different timescales: years 1-20, 21-150, and 151-1000?

Approach:

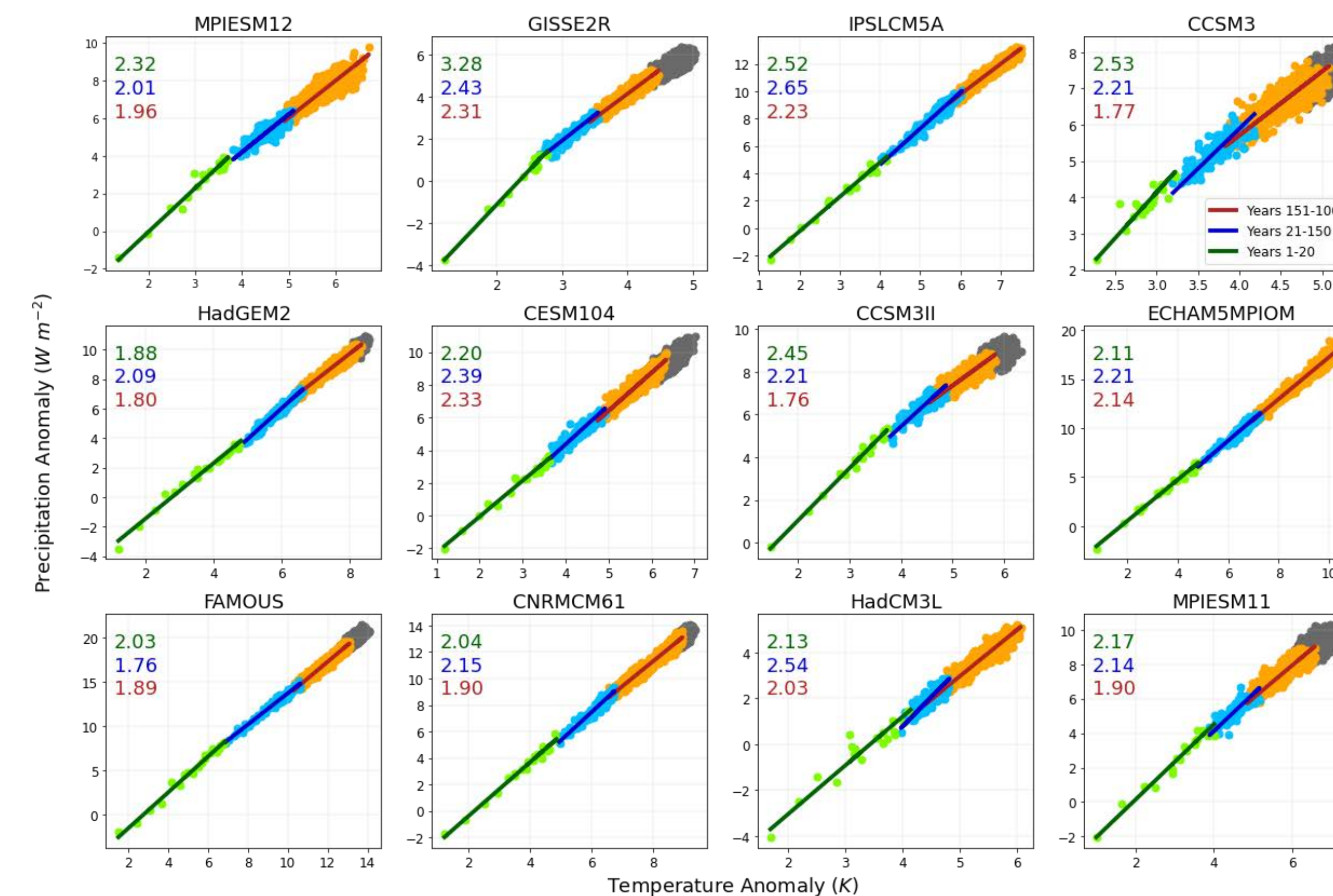
- Investigate with LongRunMIP simulations (Rugenstein et al., 2019)
- Abrupt 4xCO2 forcing, 1000+ year simulations, and pre-industrial control

Findings:

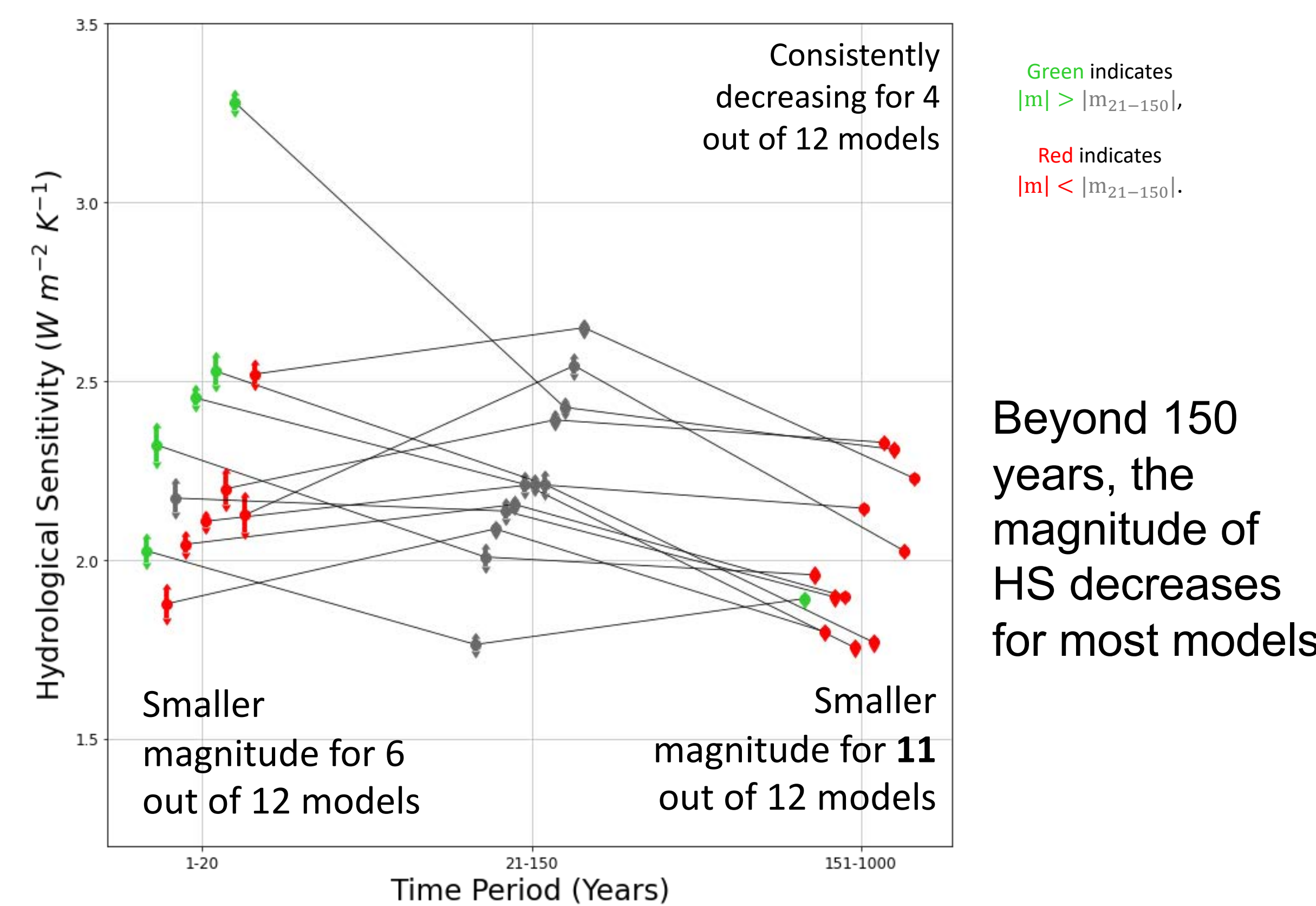
- The hydrological sensitivity has a **less consistent trend** across time scales compared to climate sensitivity.
- Timescale dependence of climate sensitivity can be overestimated when calculated with OLS regression for 1 ensemble member because of differing bias across timescales.

Hydrologic Sensitivity

$$HS = \frac{L\Delta P}{\Delta T}$$

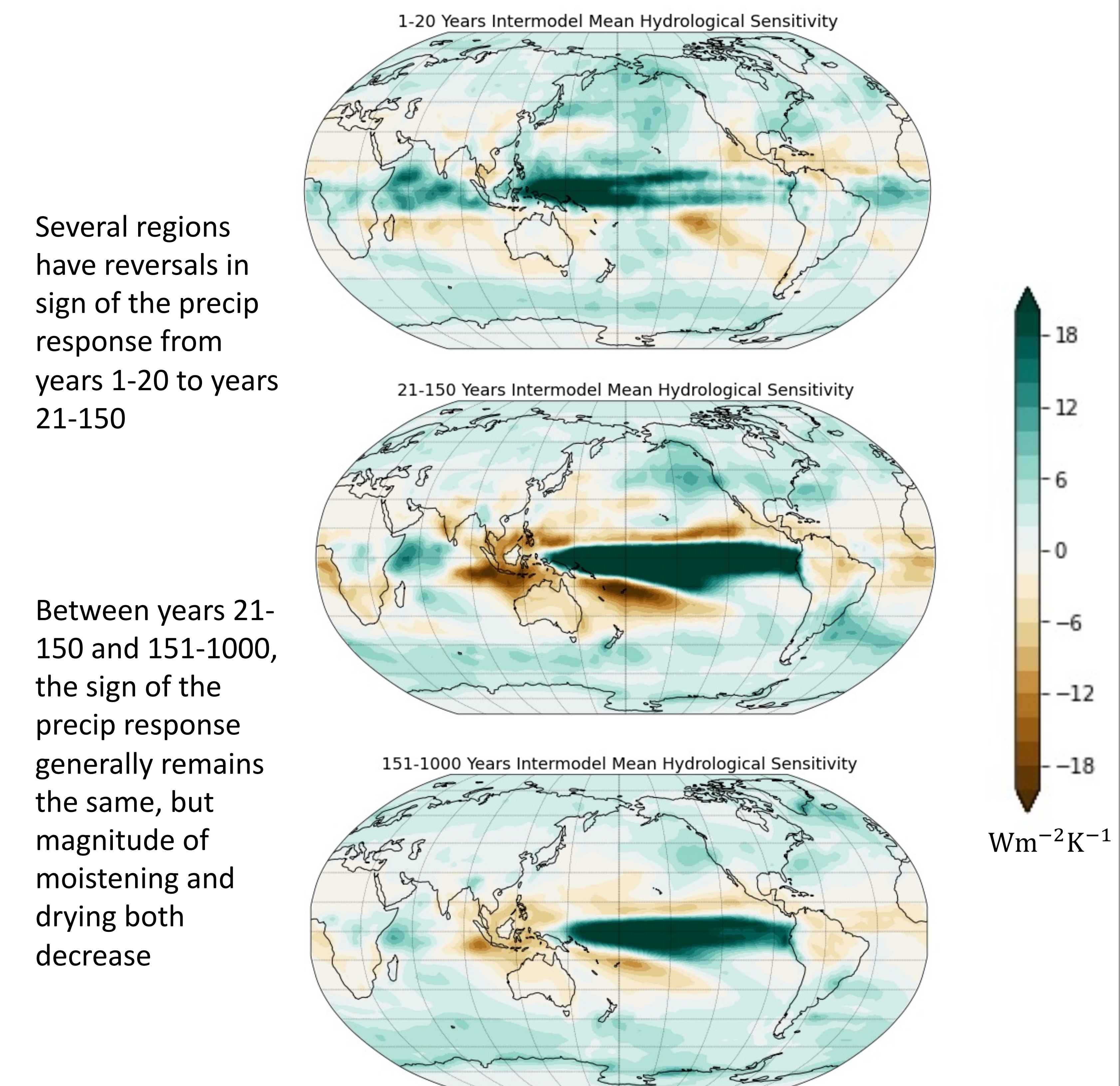


In contrast to climate sensitivity, HS for the first 20 years lack a consistent response among models compared to the next 130 years



Beyond 150 years, the magnitude of HS decreases for most models

Spatial Patterns in Hydrological Sensitivity

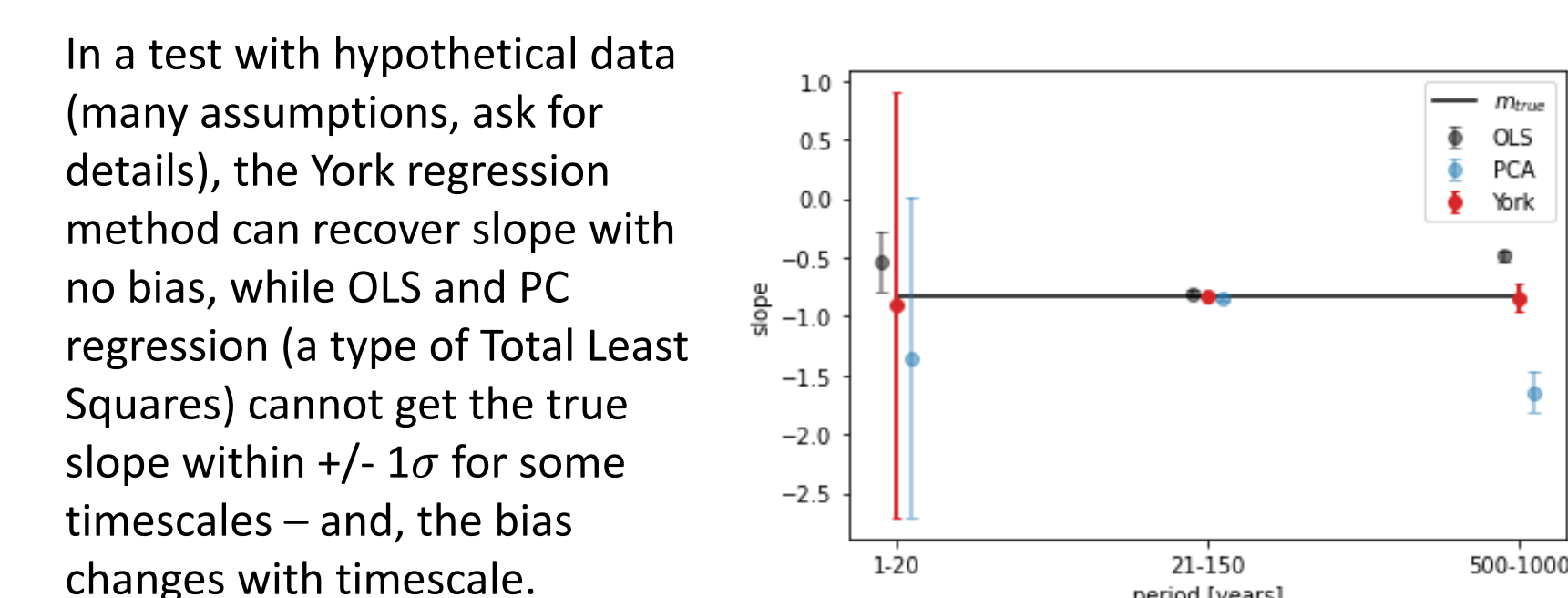


Several regions have reversals in sign of the precip response from years 1-20 to years 21-150

Between years 21-150 and 151-1000, the sign of the precip response generally remains the same, but magnitude of moistening and drying both decrease

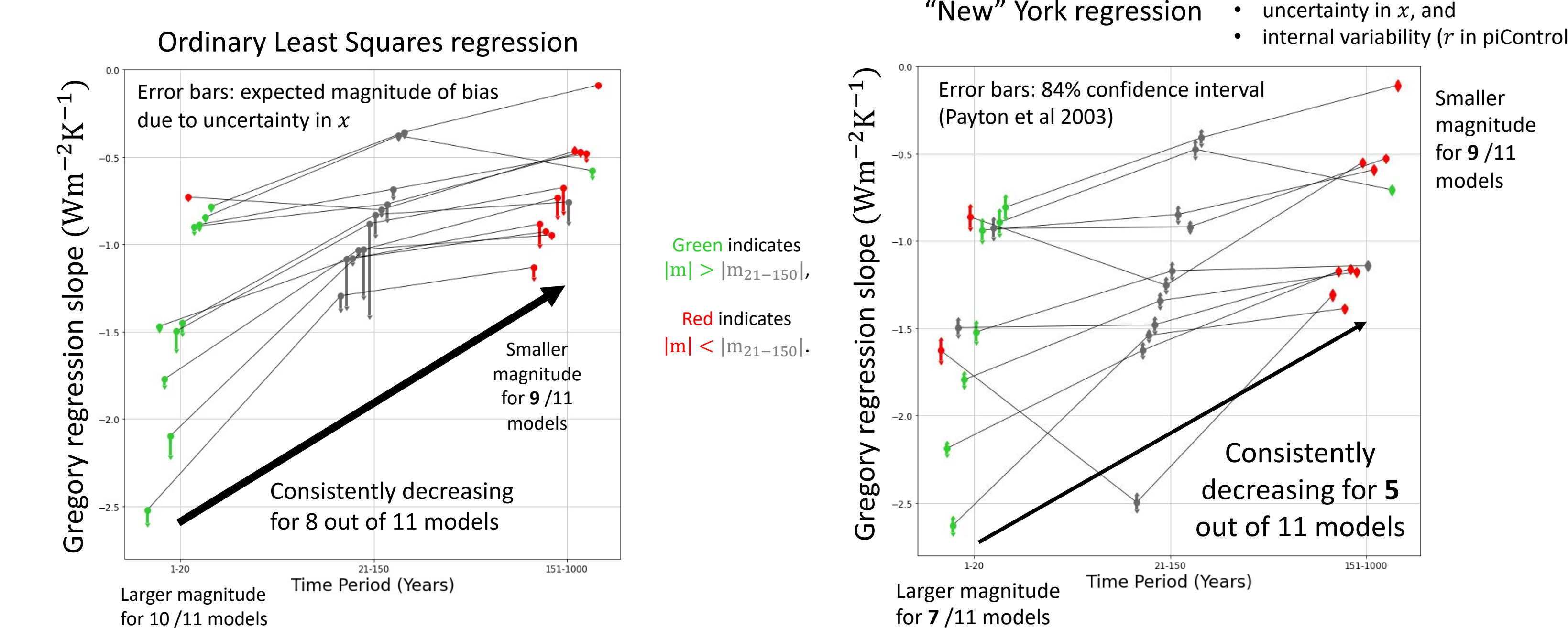
Regression methods applied to TOA radiation

Slope estimates are not always unbiased; Ordinary Least Squares (OLS) regression assumes the x -variable is known exactly (Gregory et al., 2020 discuss implications for climate sensitivity) and there is no correlation between x - and y - internal variability. Can these biases affect different timescales differently?

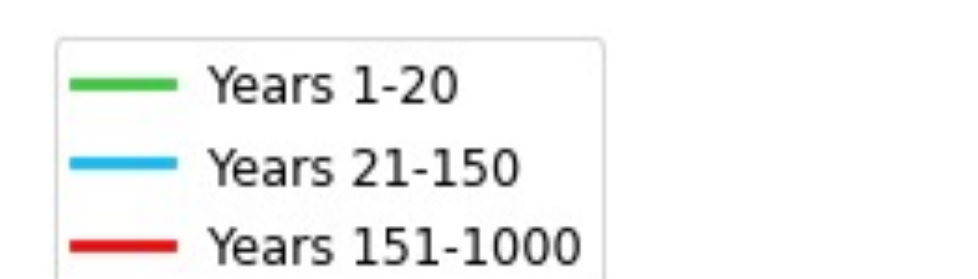


piControl, multi-model mean
 $r(T, R_{TOA}) = -0.2$
 $r(T, P) = 0.6$
 Correlation of internal variability is larger for precip than net TOA radiation

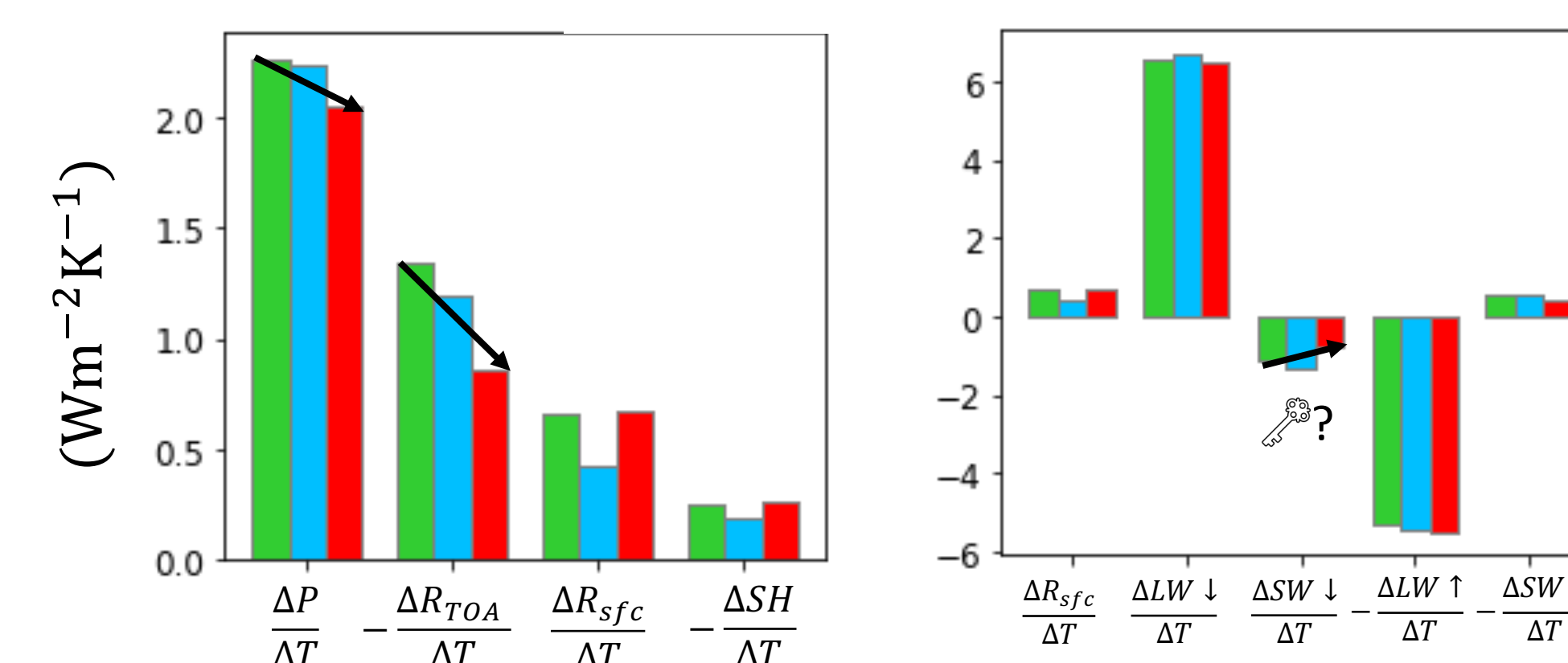
"New" York regression Accounts for:
 • uncertainty in x , and
 • internal variability (r in piControl)



Energy budget



The only energy budget component changing with the right sign to explain the difference between HS and climate sensitivity seems to be downwelling shortwave radiative flux at surface



References and Acknowledgements

Kao (senior thesis; in review) / Kao and Pendergrass (in prep.)

References

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Acknowledgements

Thanks to everyone involved in running and coordinating the LongRunMIP simulations and for generously sharing them with us, to Jonathan Gregory for insightful discussions about the regression technique, and to Mike Town for bringing the "New" York regression Python code to our attention and for simplifying the Python function for calculating it. The original release of the "New" York regression Python code is publicly available at github.com/LNL/MahonFitting/