

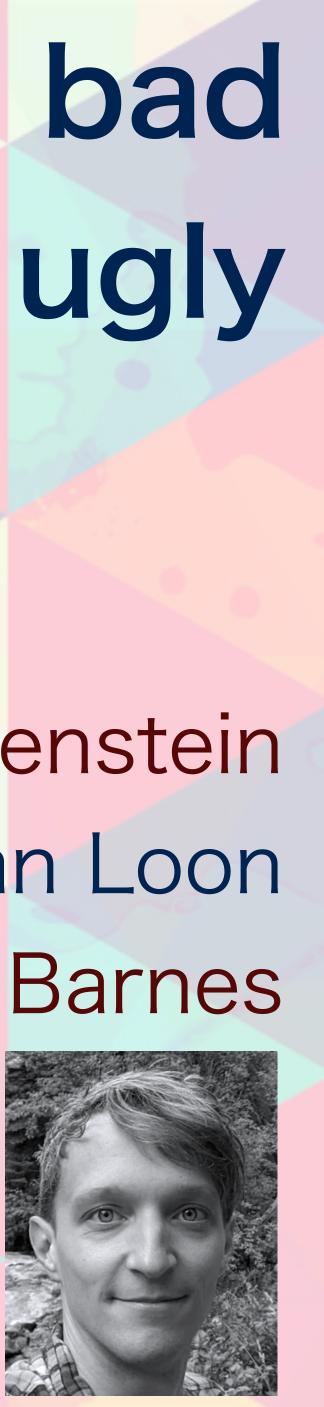
How the good and the bad conspire to the ugly

Maria Rugenstein Dirk Olonscheck, Shreya Dhame, Marc Alessi, Senne Van Loon Robb Wills, Masahiro Watanabe, Richard Seager, Libby Barnes

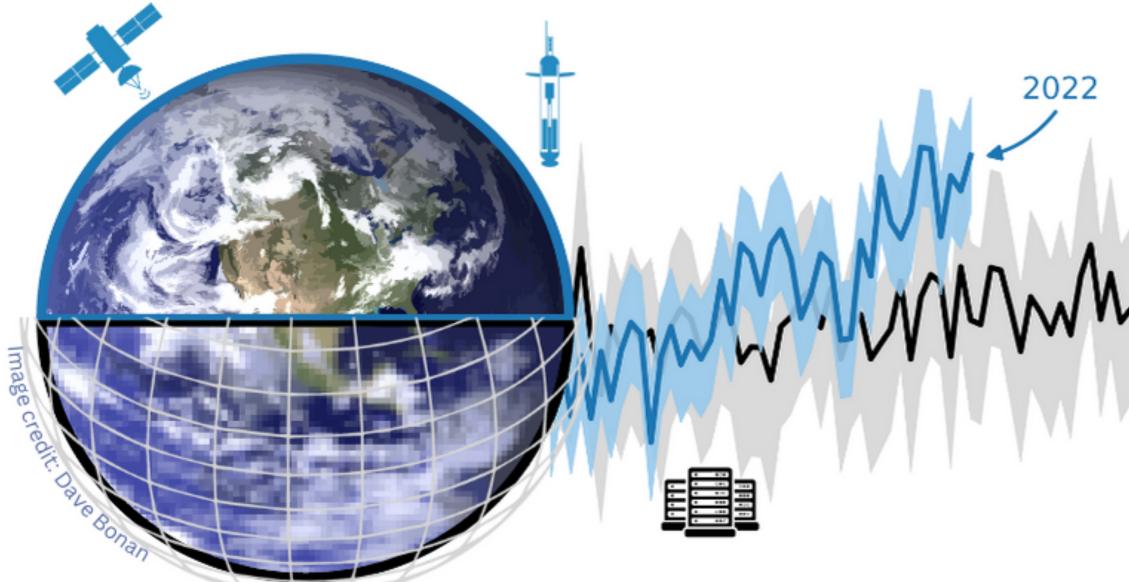








trends with observations: the good, the bad and the ugly



How the good and the bad conspire to the ugly **US Clivar workshop Confronting Earth System Model**

200 attendees from 18 countries, 120 in-person, 62% early-career

Observations www.humphand Models 2100

Confronting Earth system model trends with observations: A new era in simulating and predicting climate



How the good and the bad conspire to the ugly

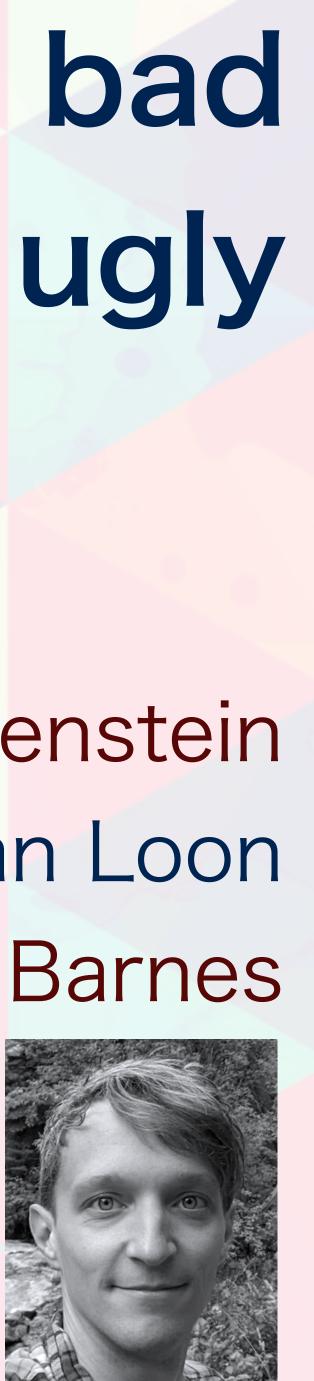
1. TOA radiation trends last 20yrs 2. Surface temperature trends last 70yrs 3. Implications for projections next 40yrs

Maria Rugenstein Dirk Olonscheck, Shreya Dhame, Marc Alessi, Senne Van Loon Robb Wills, Masahiro Watanabe, Richard Seager, Libby Barnes

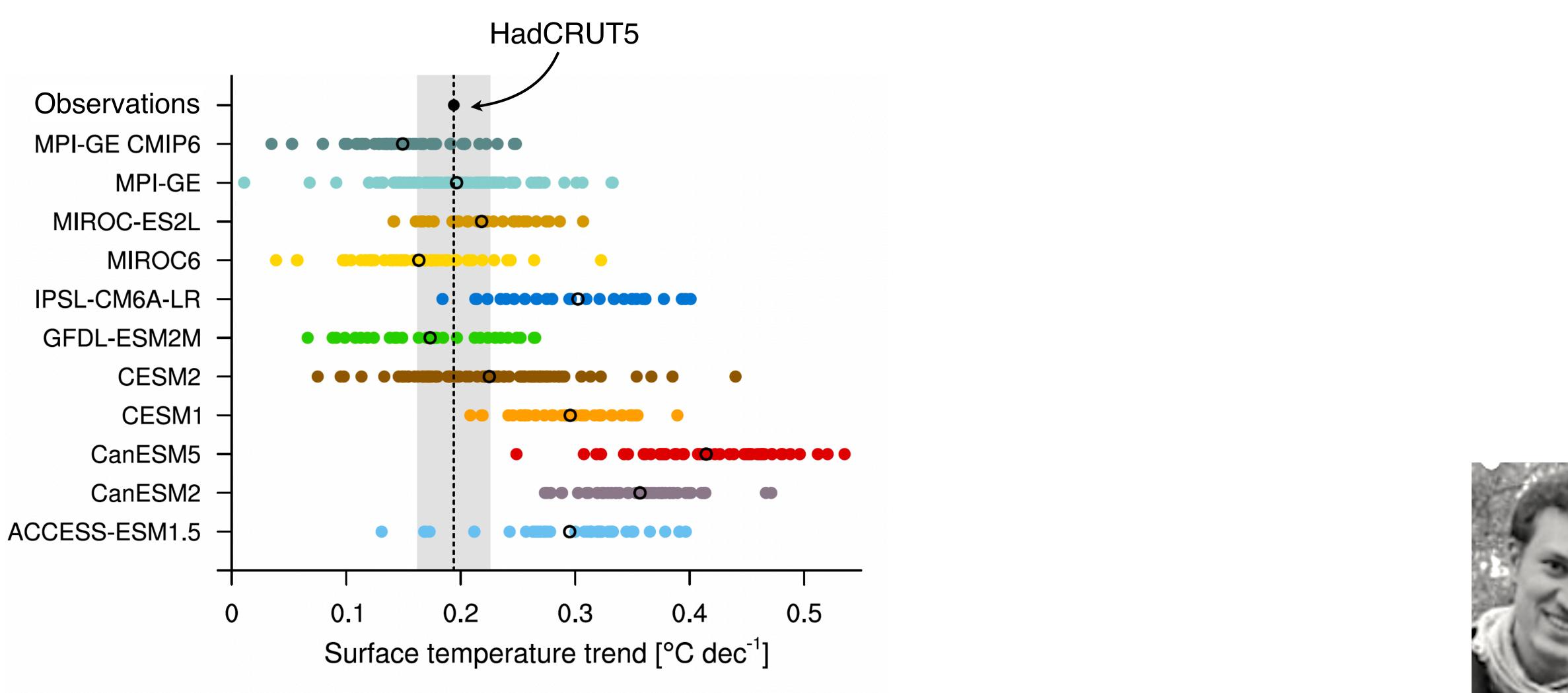








2000 - 2022 global-mean surface temperature is good



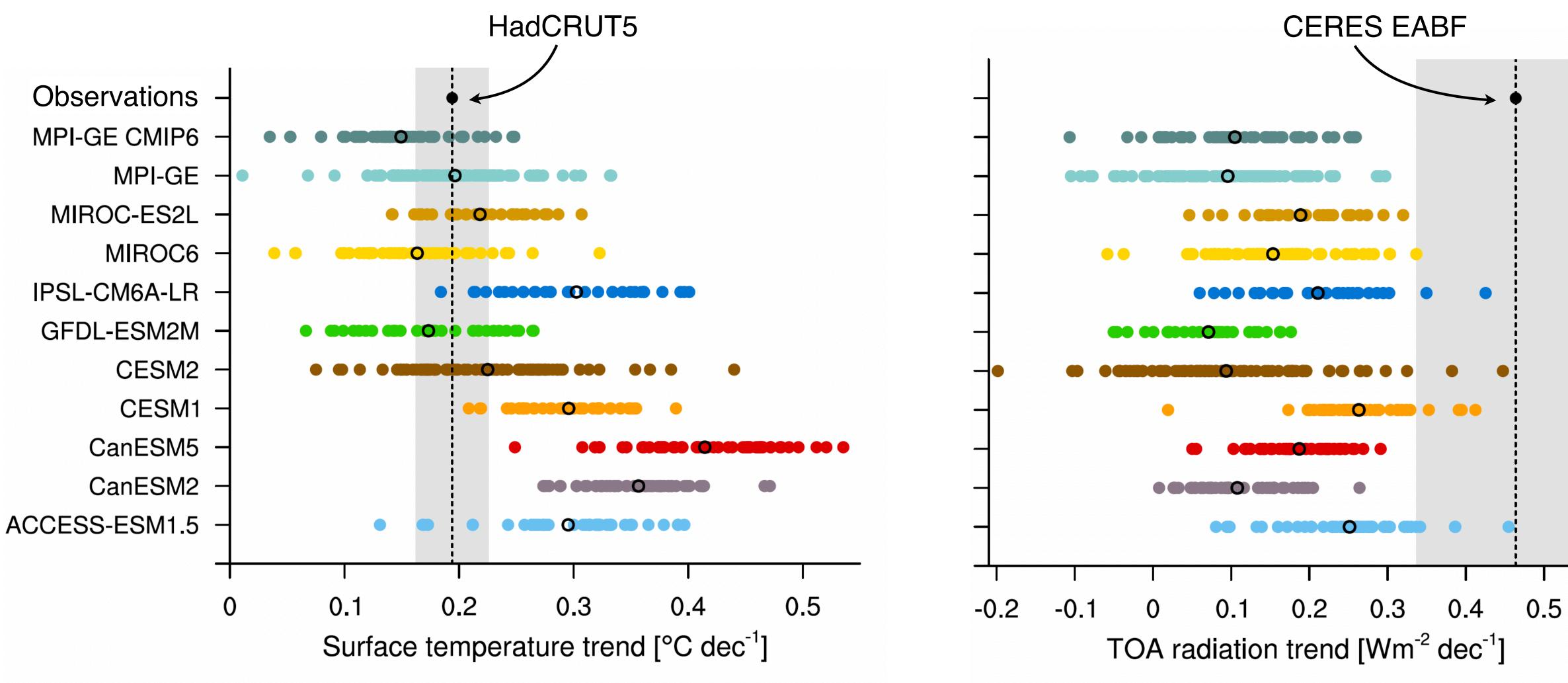
Olonscheck & Rugenstein, 2024: Coupled climate models systematically underestimate radiation response to surface warming







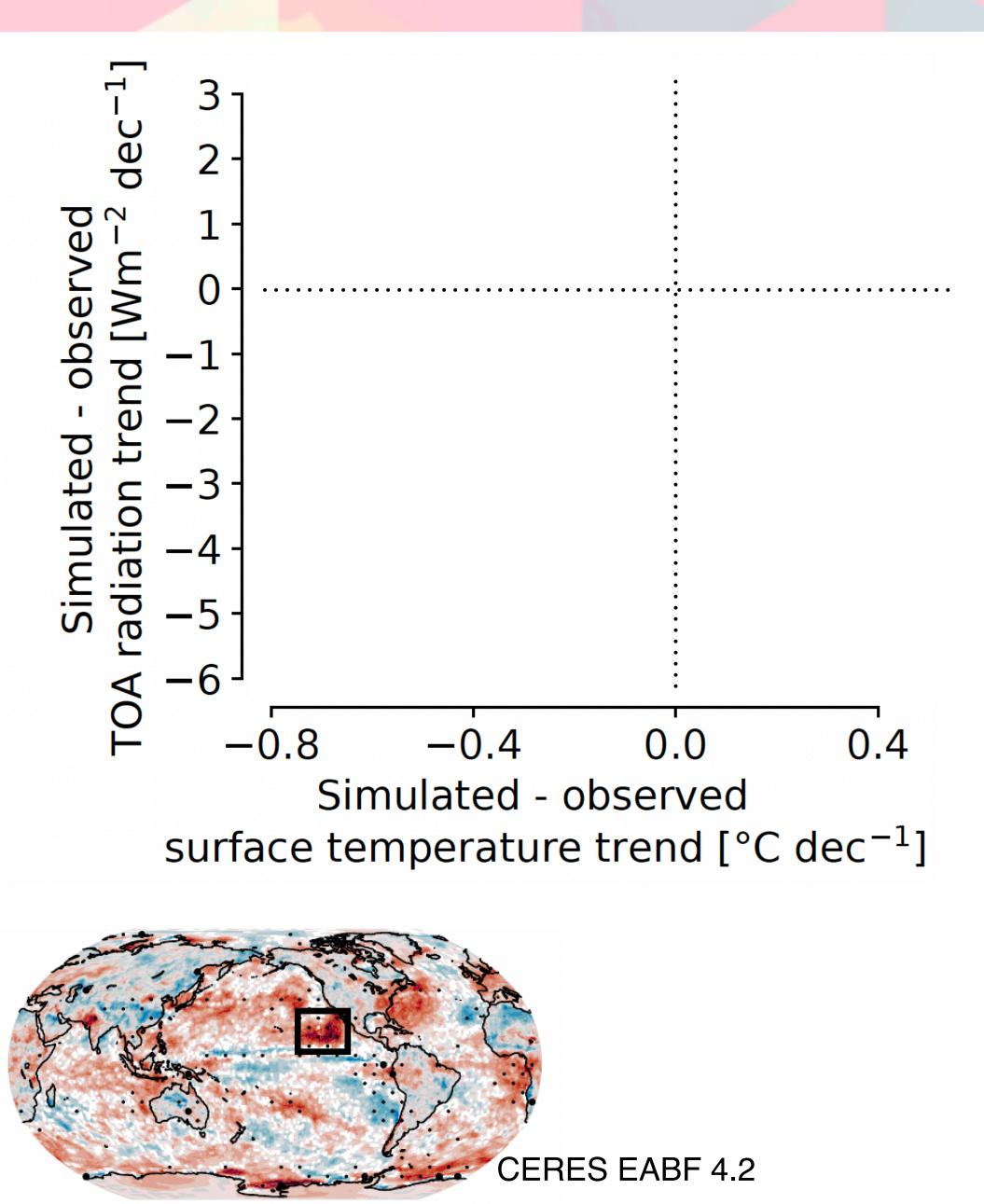
Global-mean top of the atmosphere radiation is bad





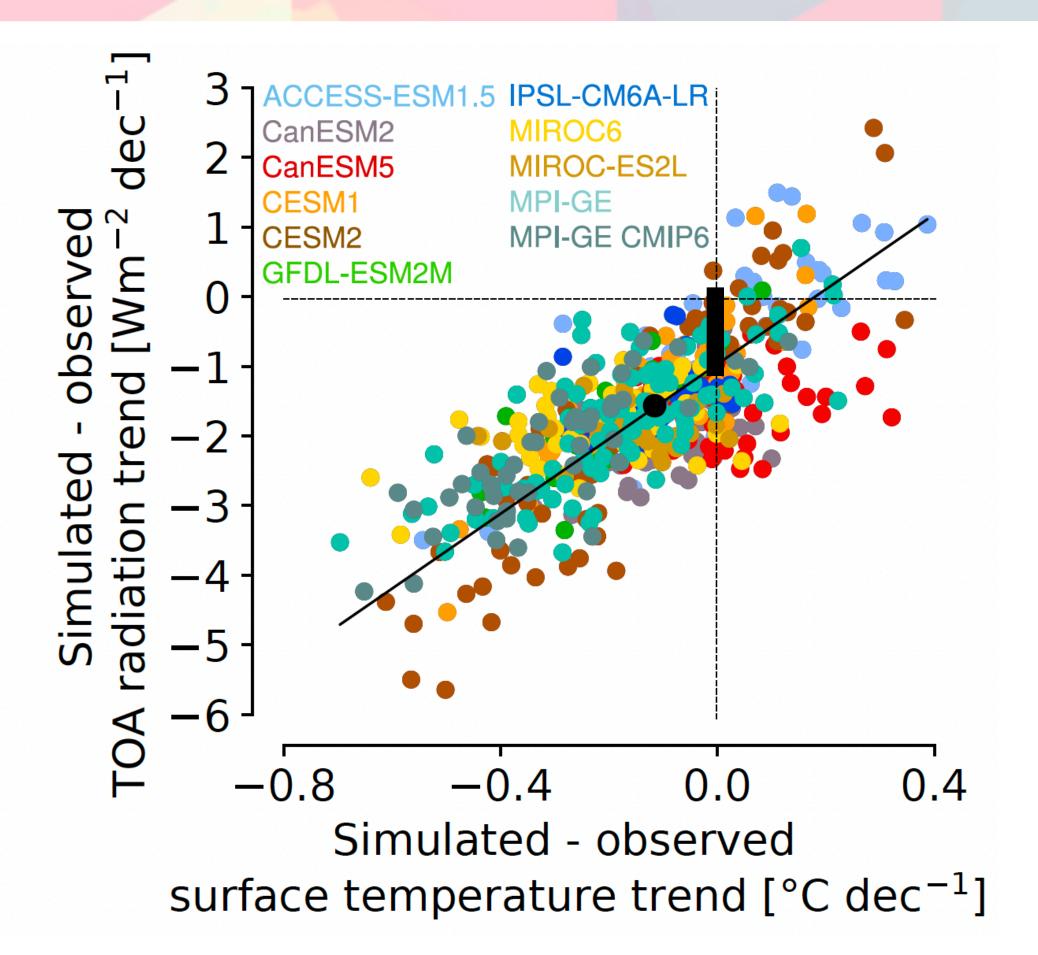


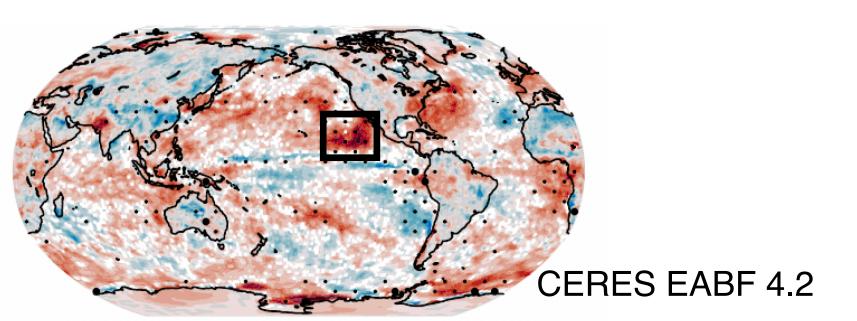
Bias measure: if surface trend is correct, how off is radiation trend



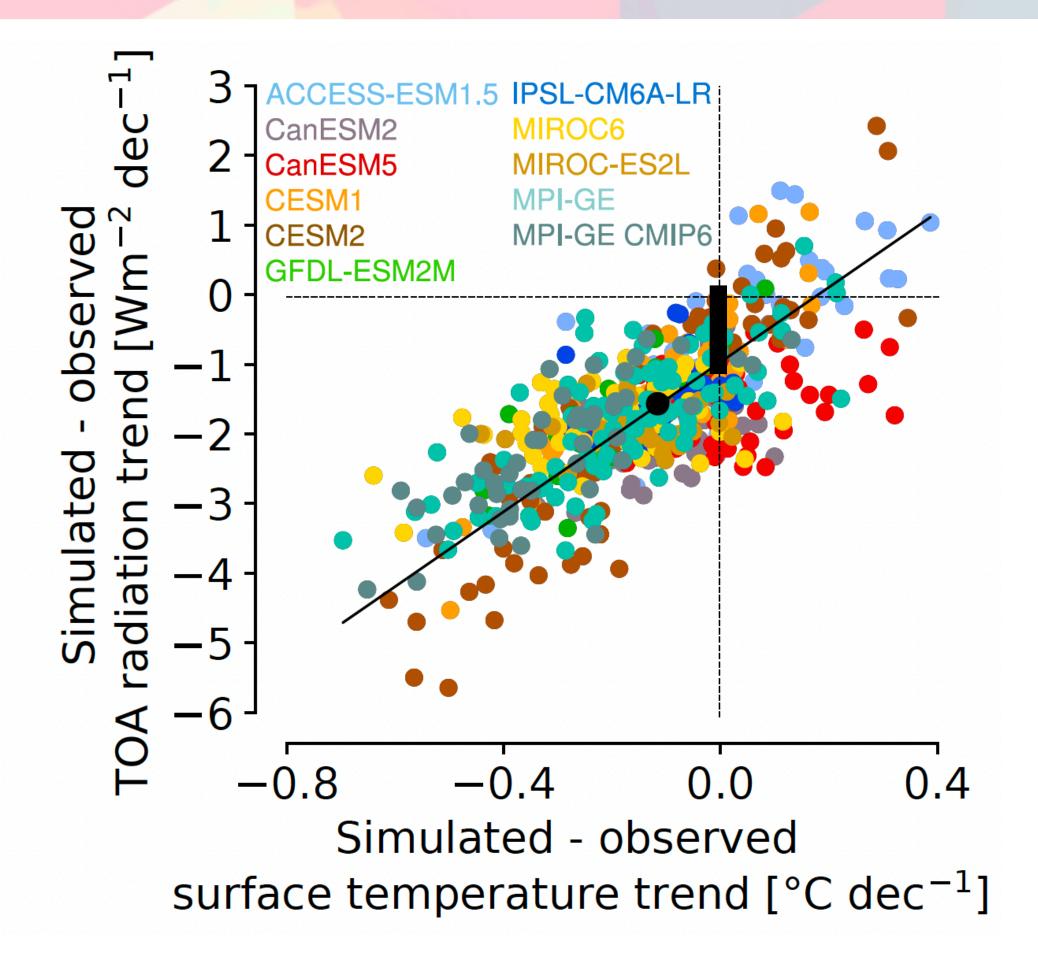


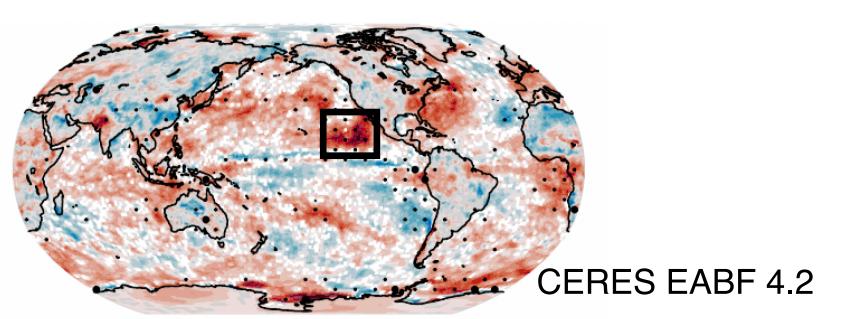
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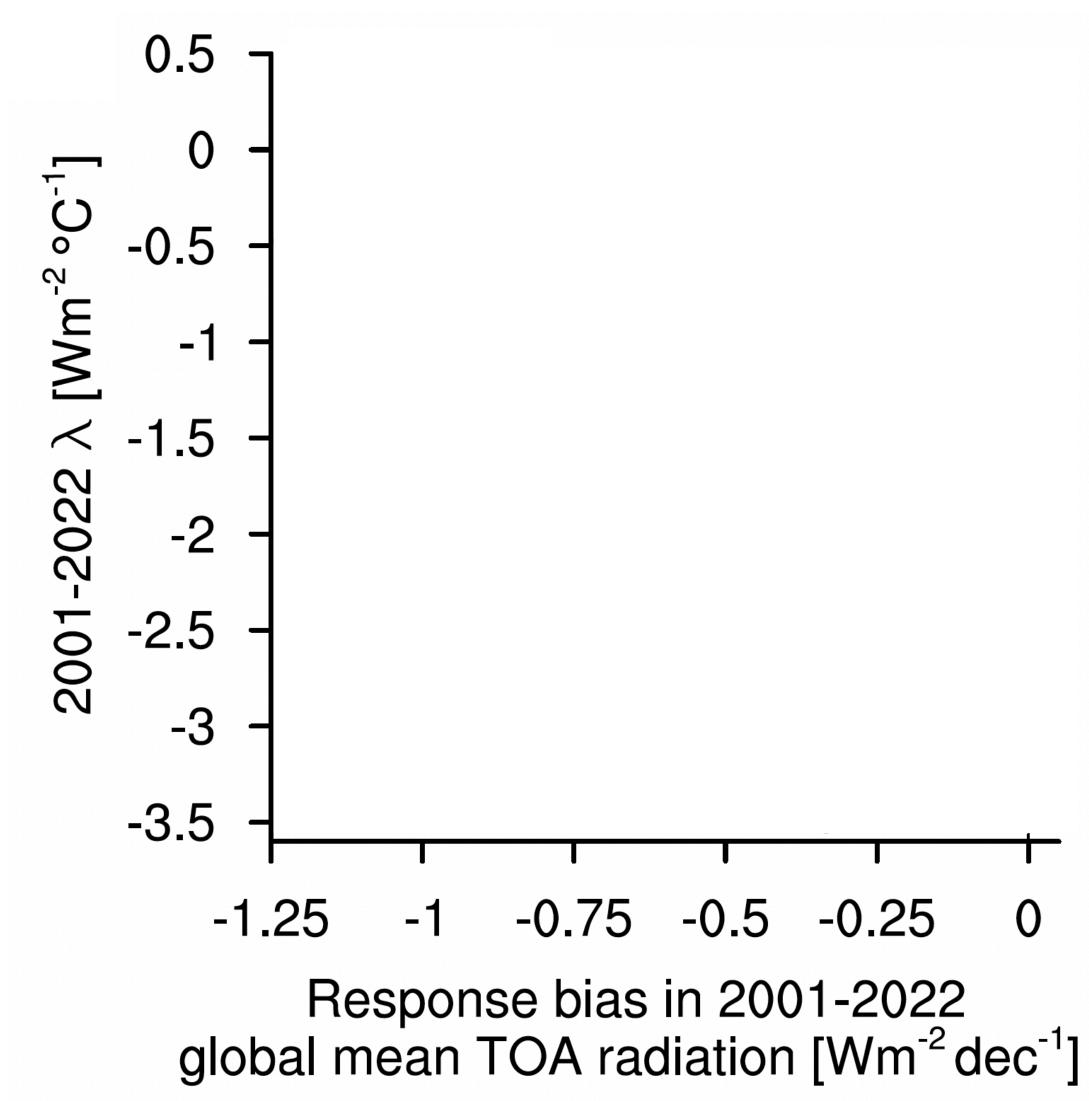






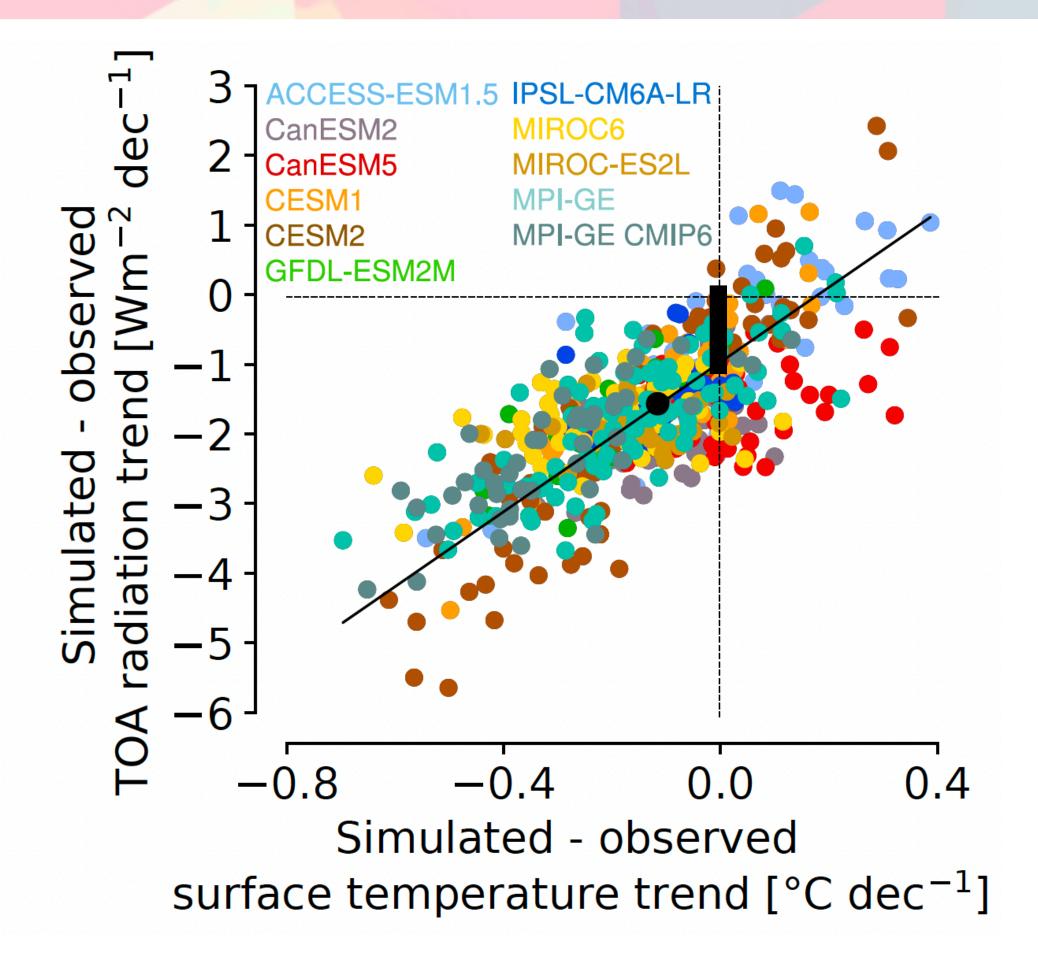


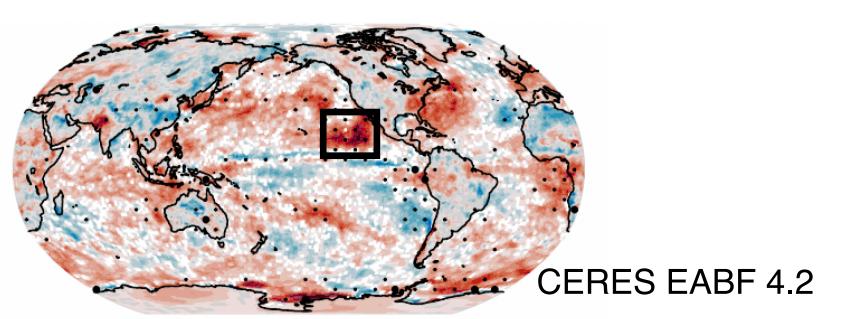


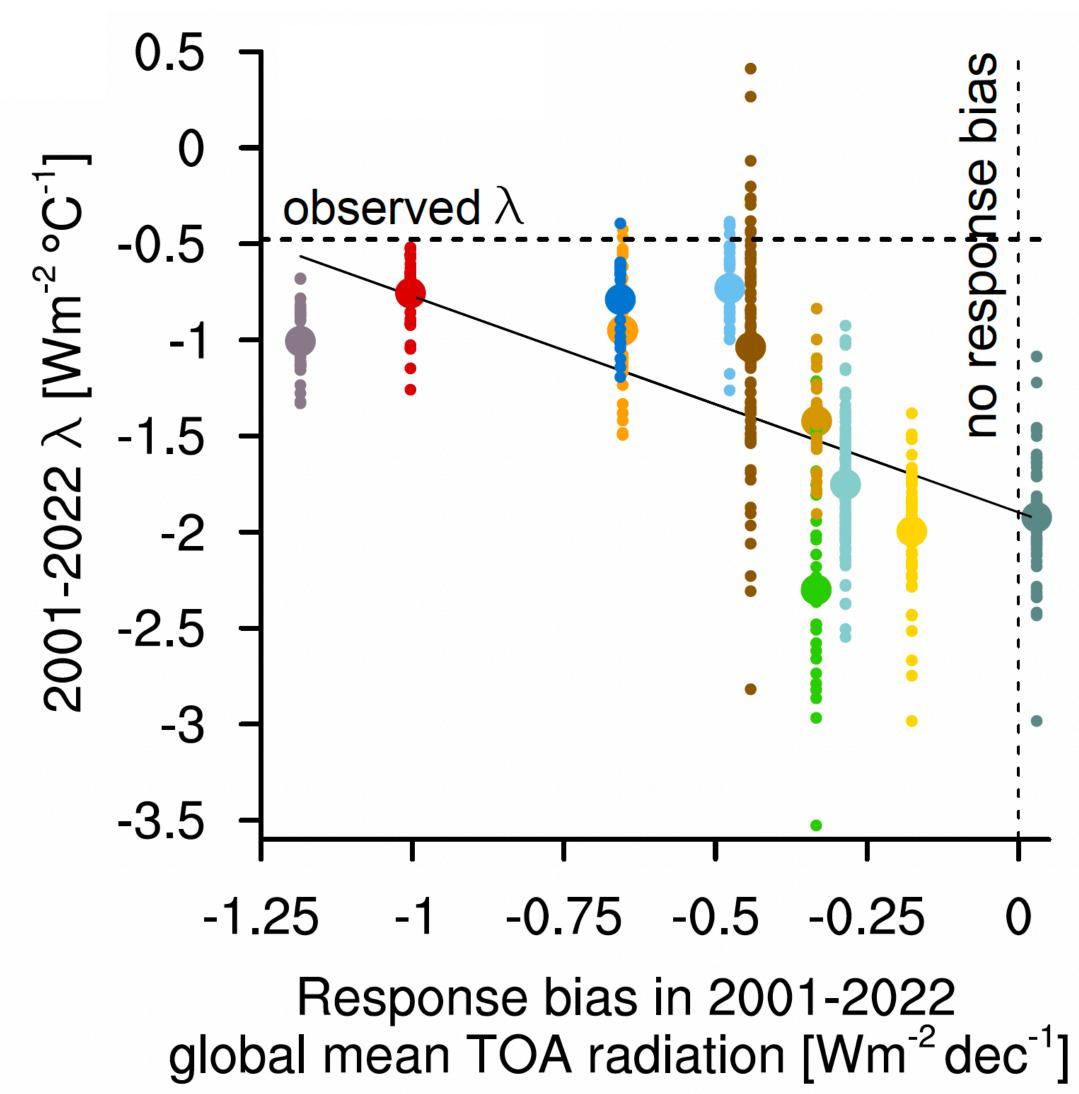






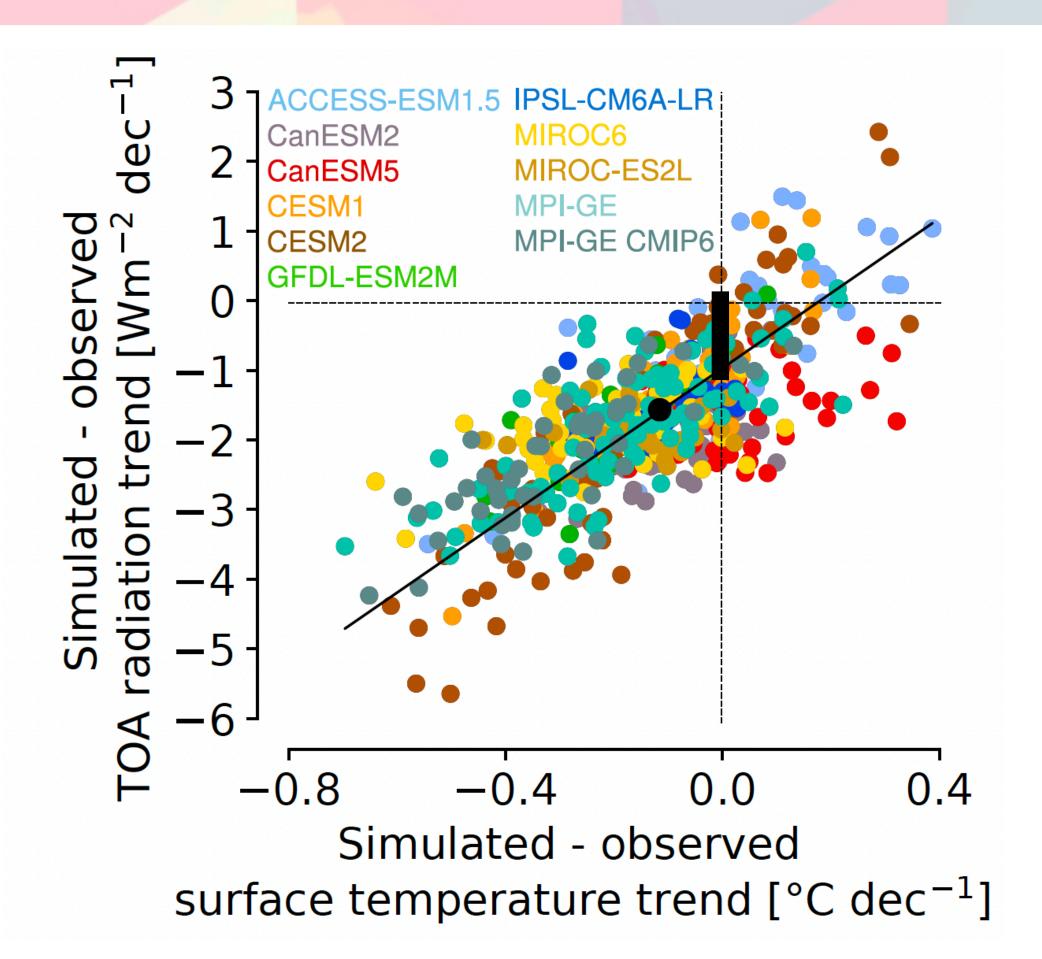


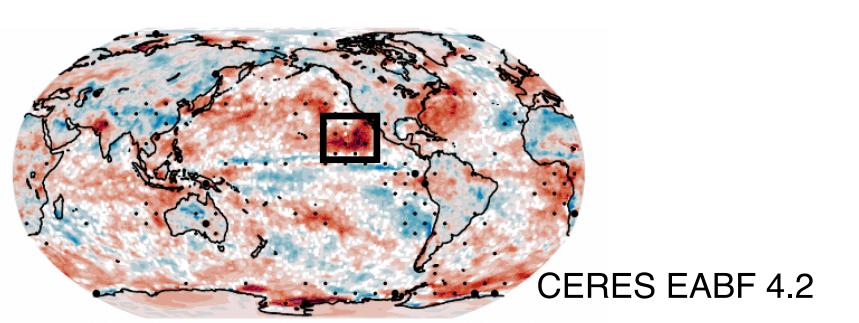


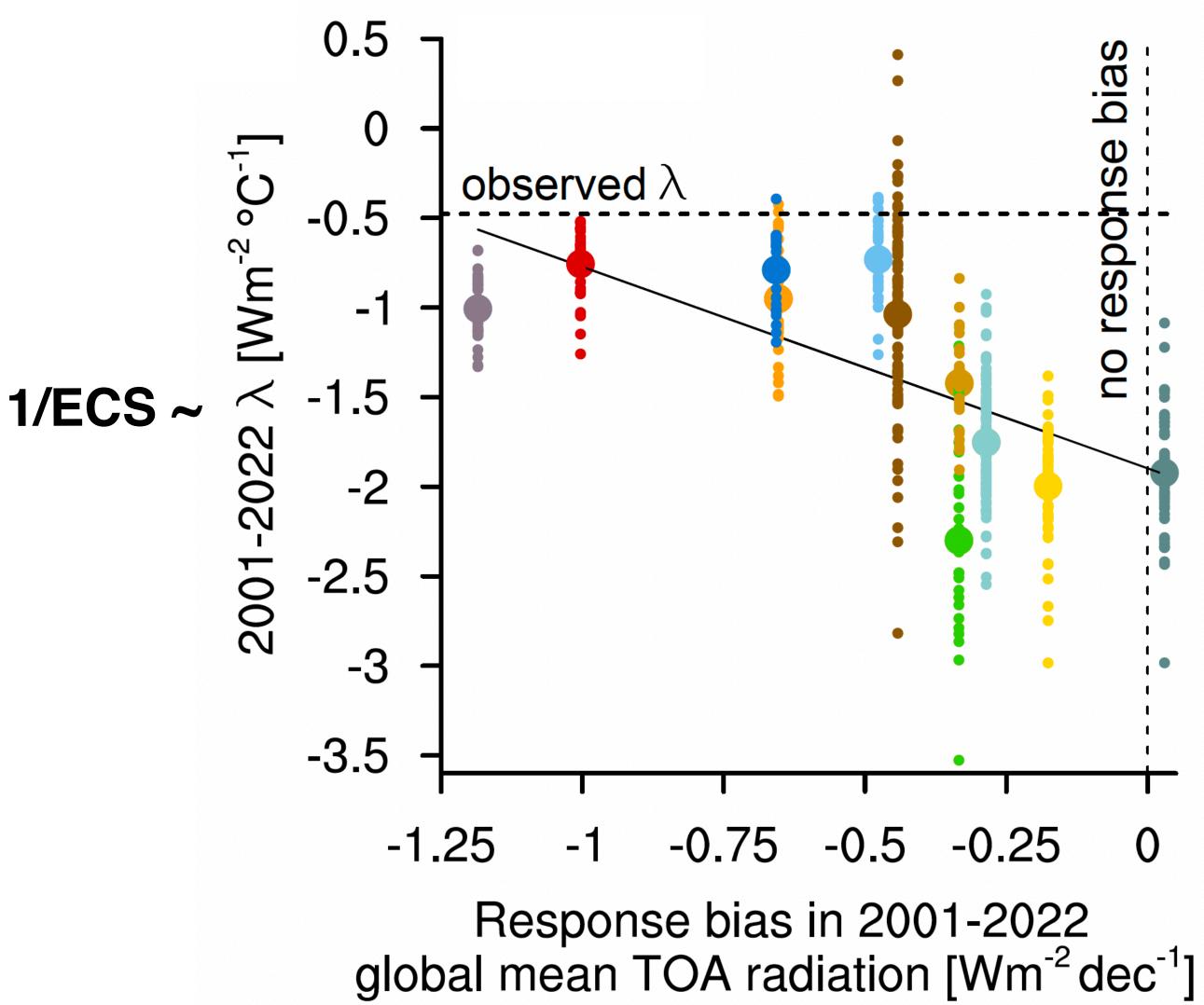










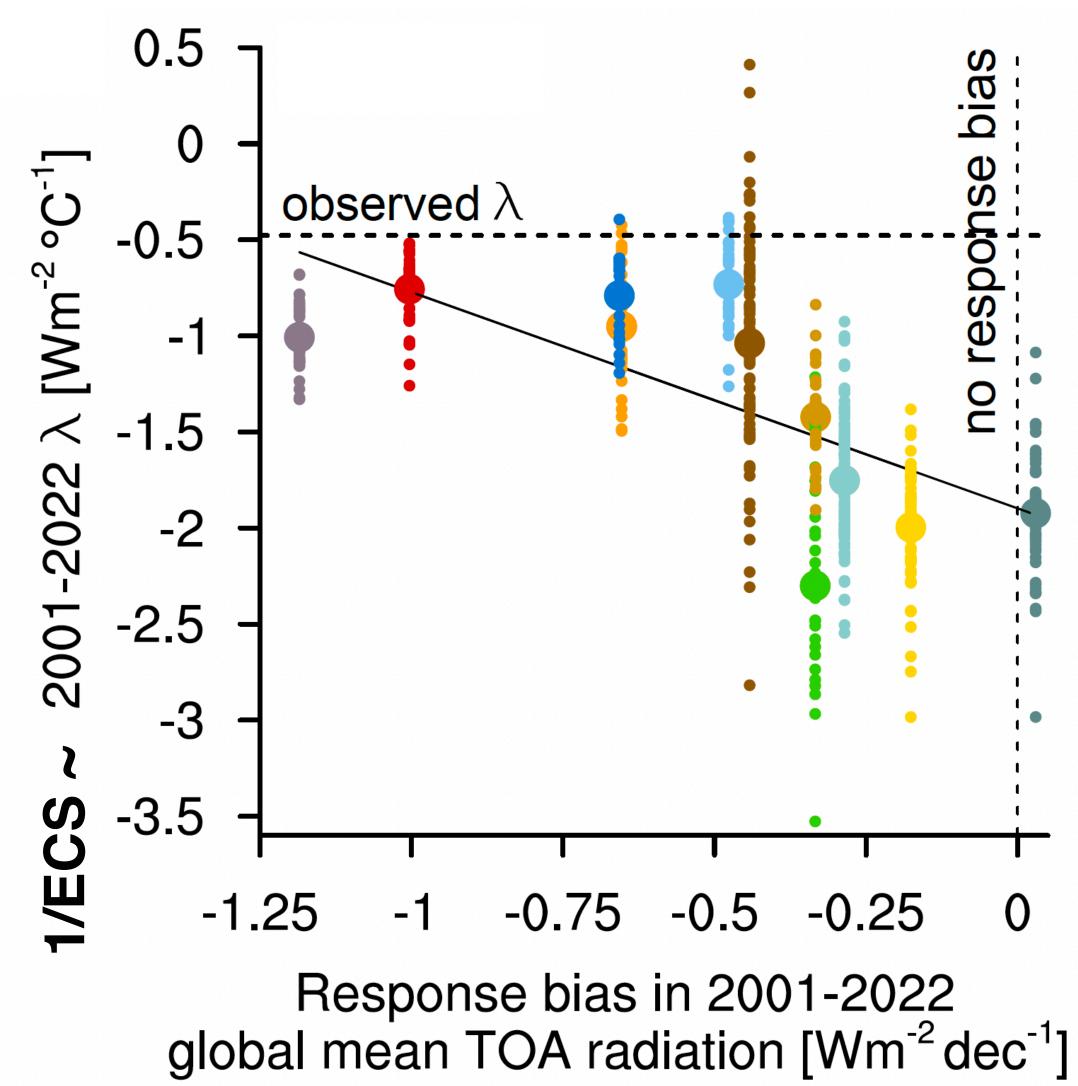




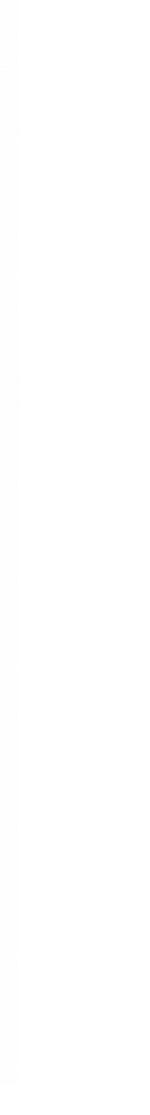


- Surface temperature trends = good
- OA radiation trends = bad
- Observed still too noisy to constrain models feedbacks = ugly
- Radiation response bias indicates models with a more negative, stabilizing, radiative feedback more readily reproduce observations
- Better models may have better SST patterns, better cloud physics, better forcing, or better compensating biases
- Implications for ocean heat uptake to follow

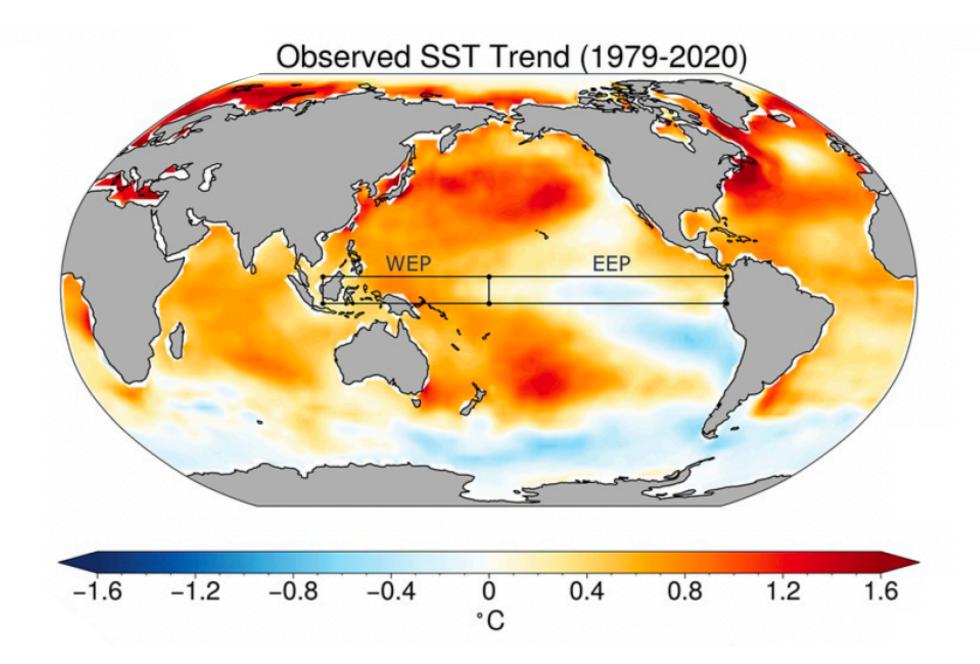
Uribe et al. 2022, Raghuraman et al. 2021, 2023, Loeb et al. 2021, Schmidt et al. 2024, CERES MIP, Hodnebrog et al. 2024











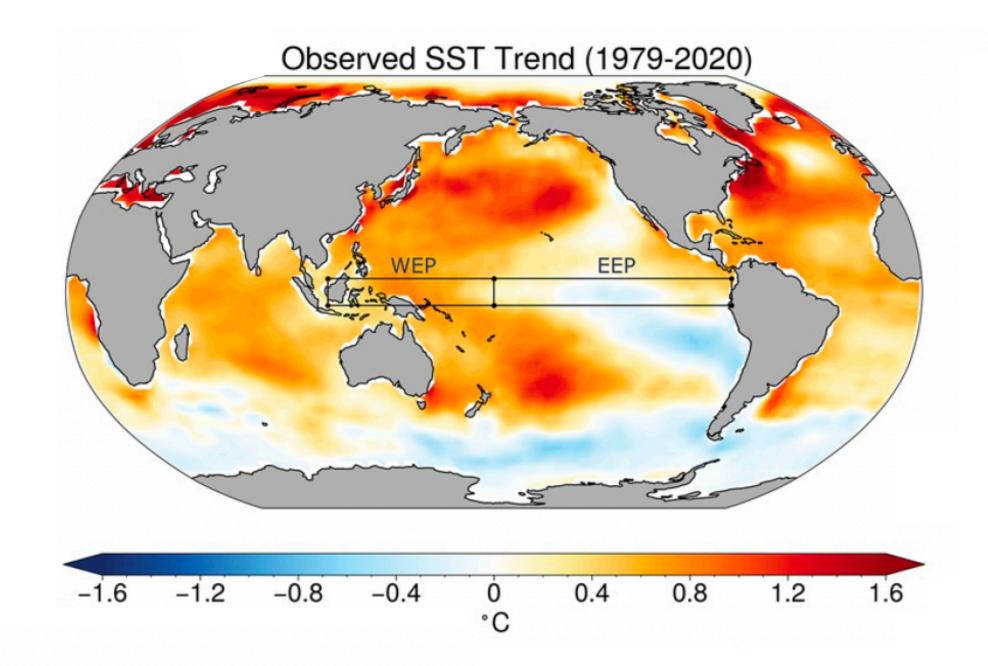
Rugenstein, Dhame, Olonscheck, Wills, Watanabe, Seager, 2023 Connecting the SST pattern problem and the hot model problem







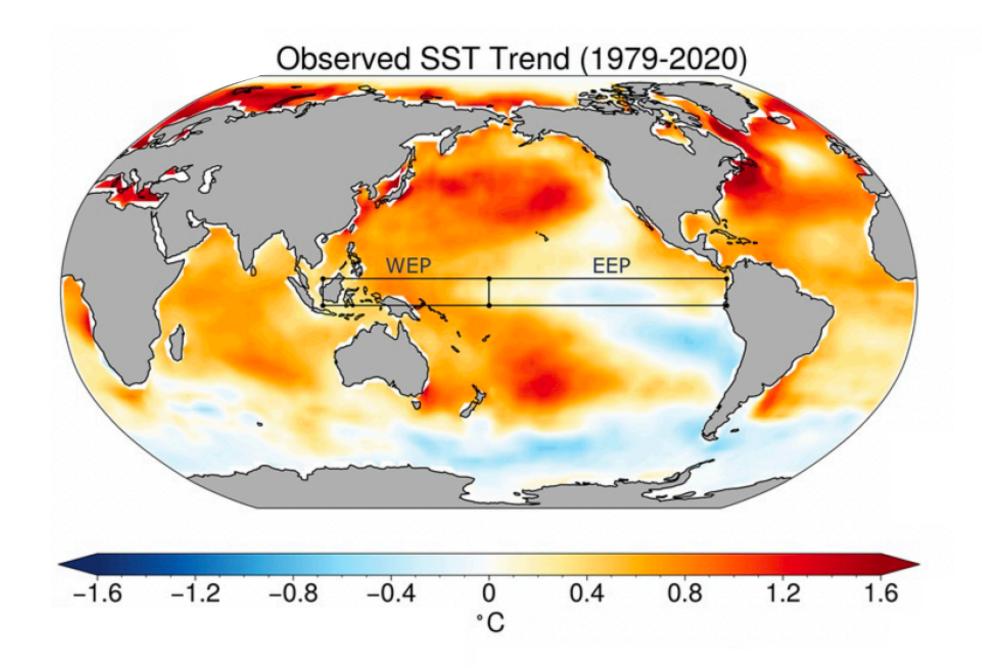








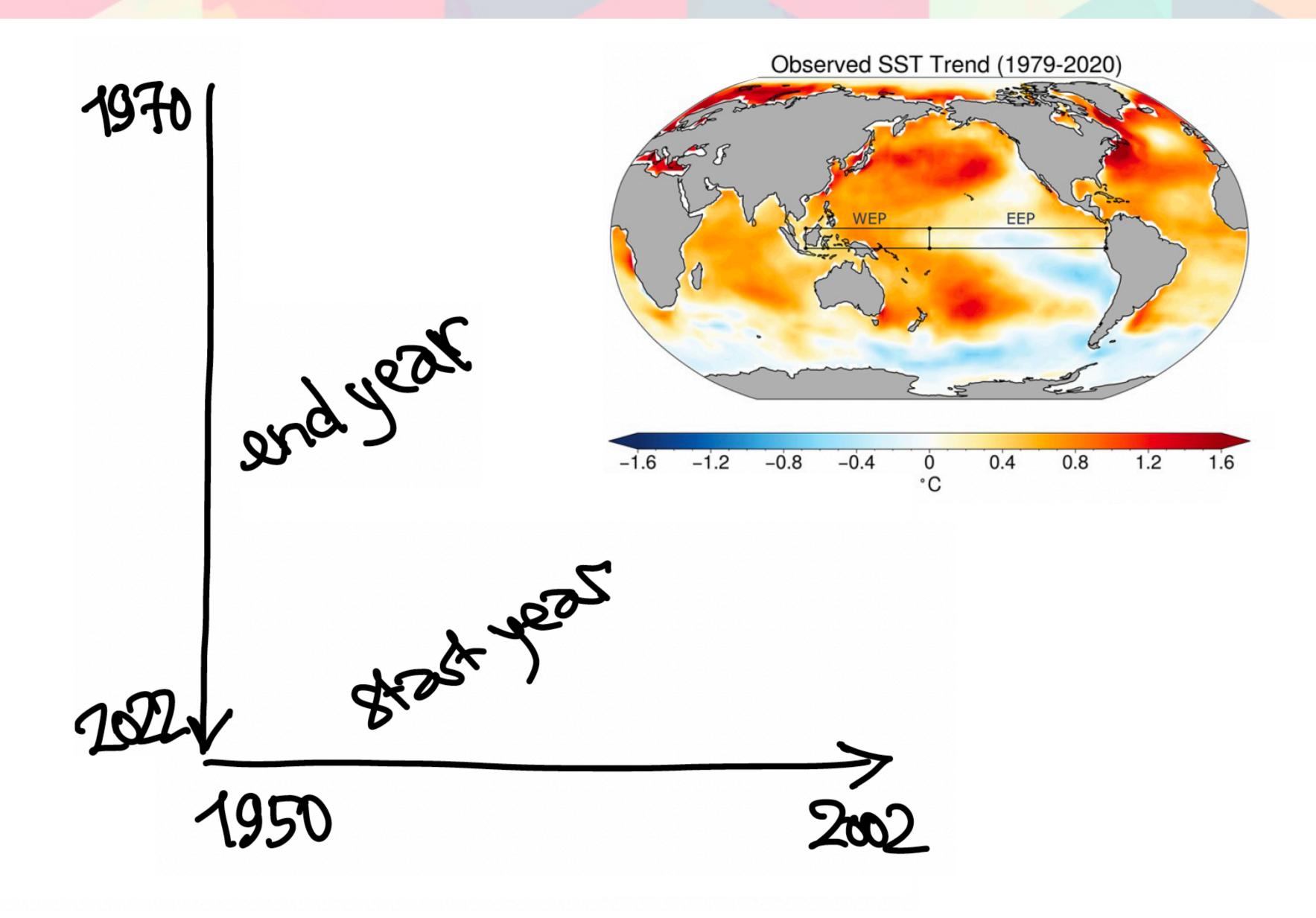






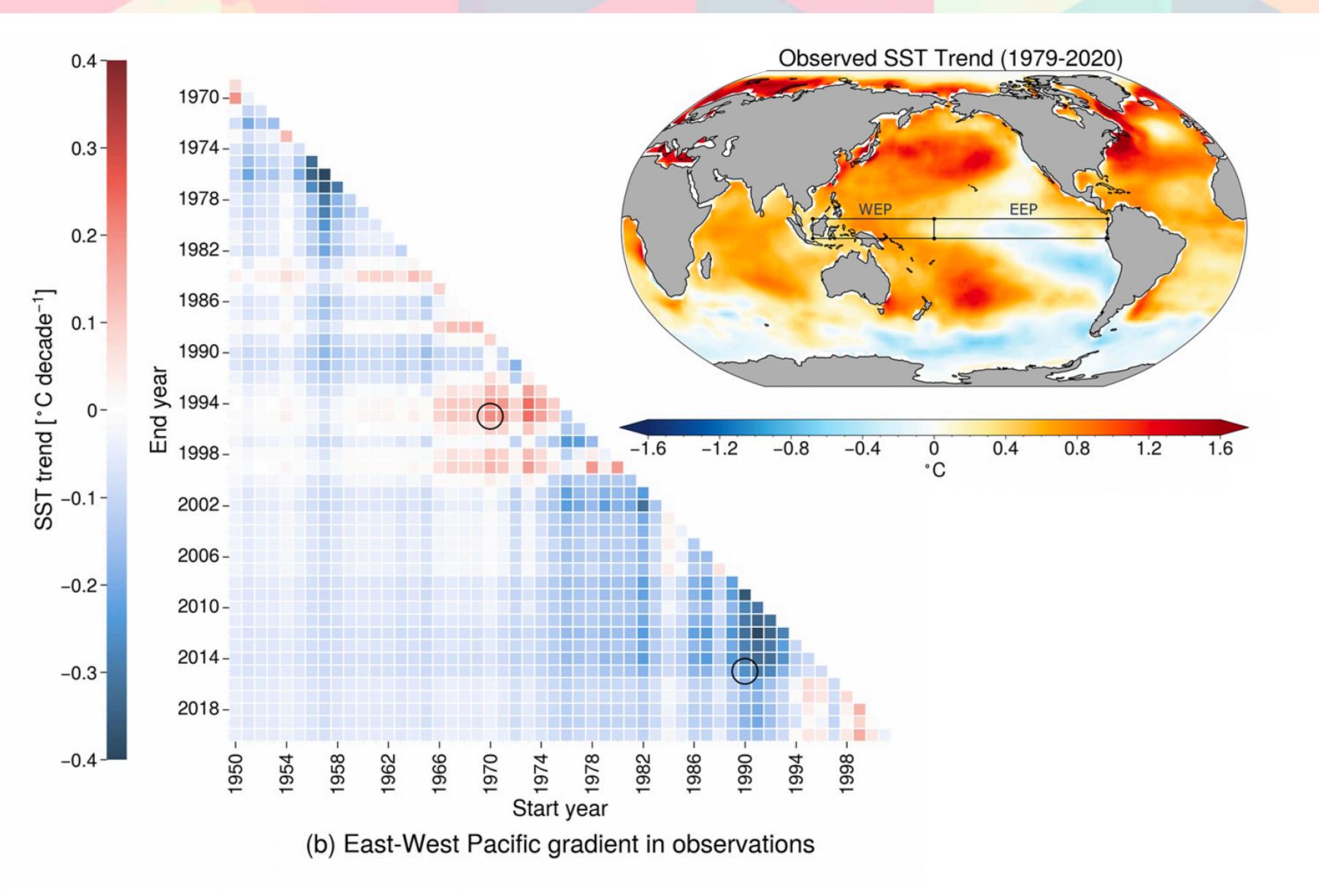






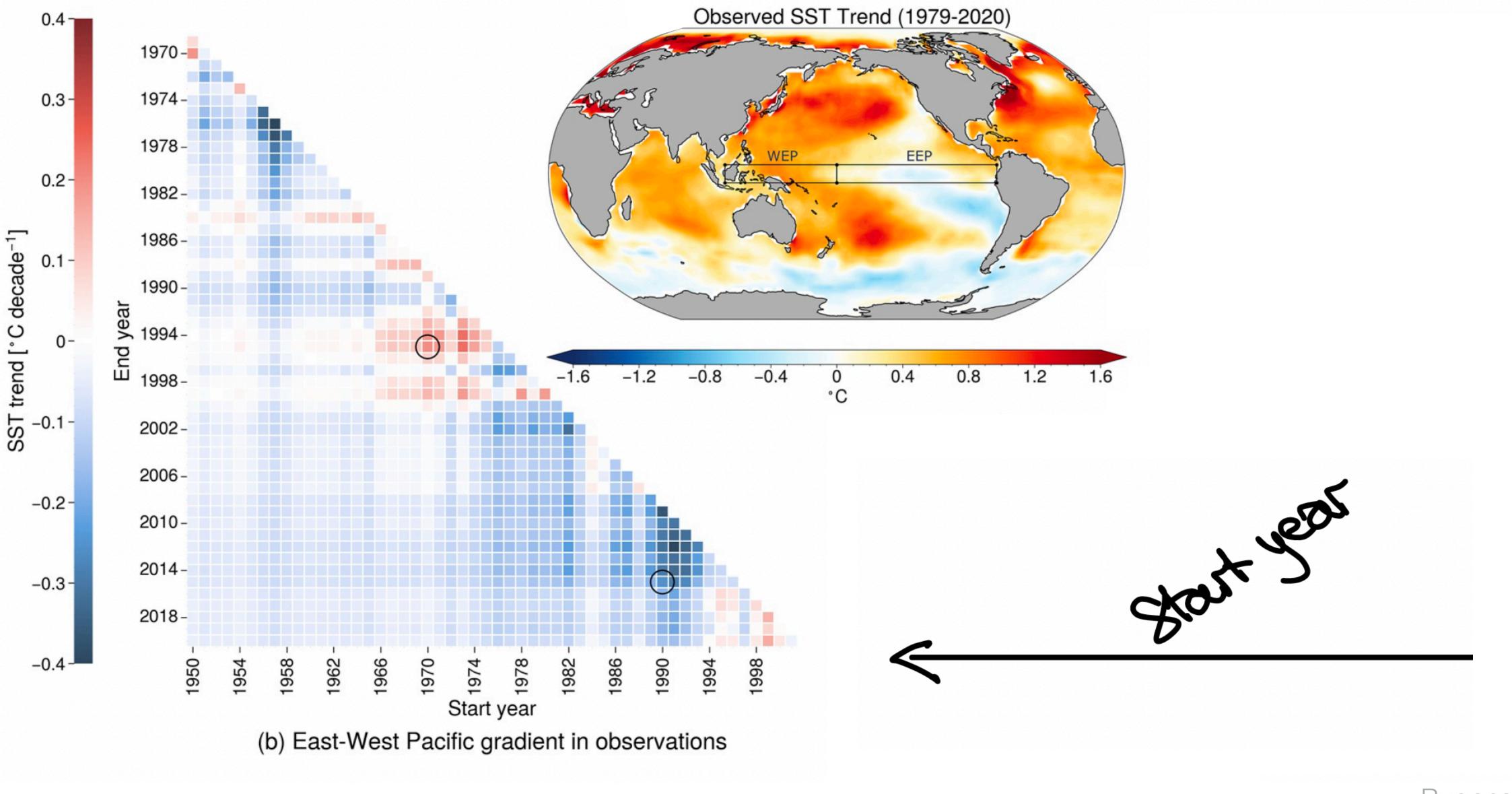






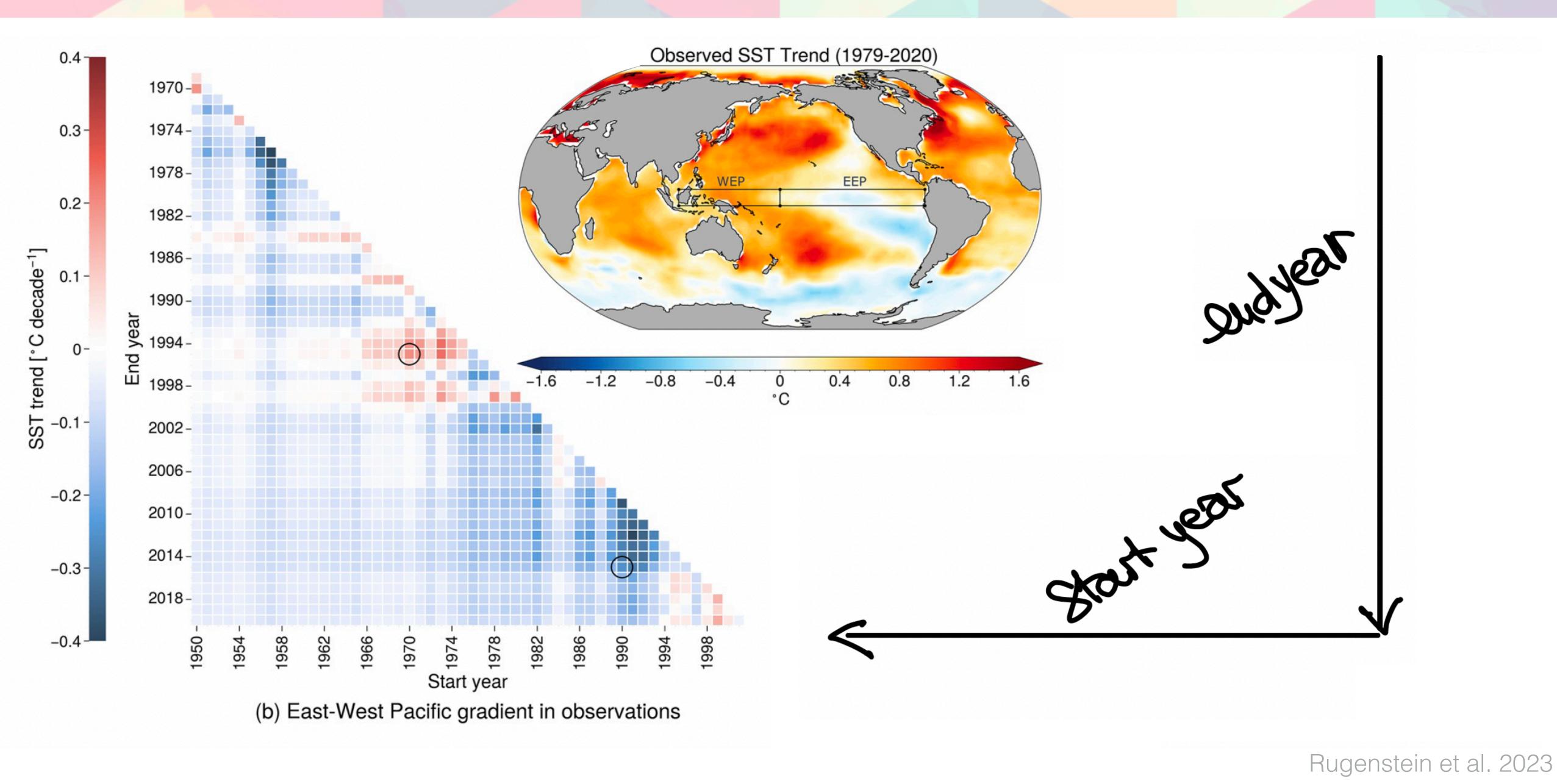






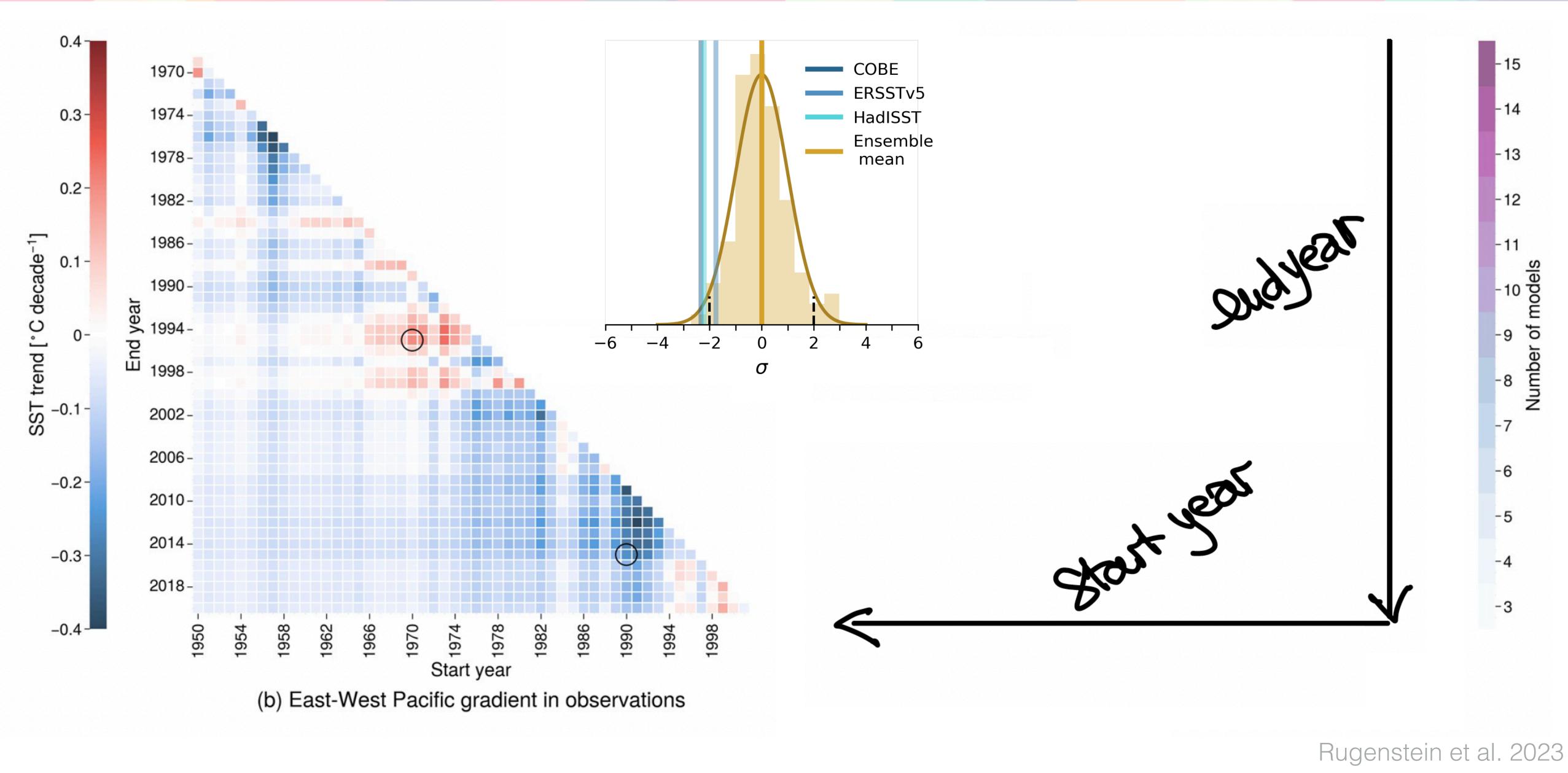




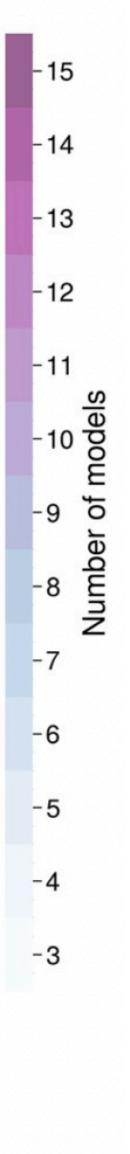


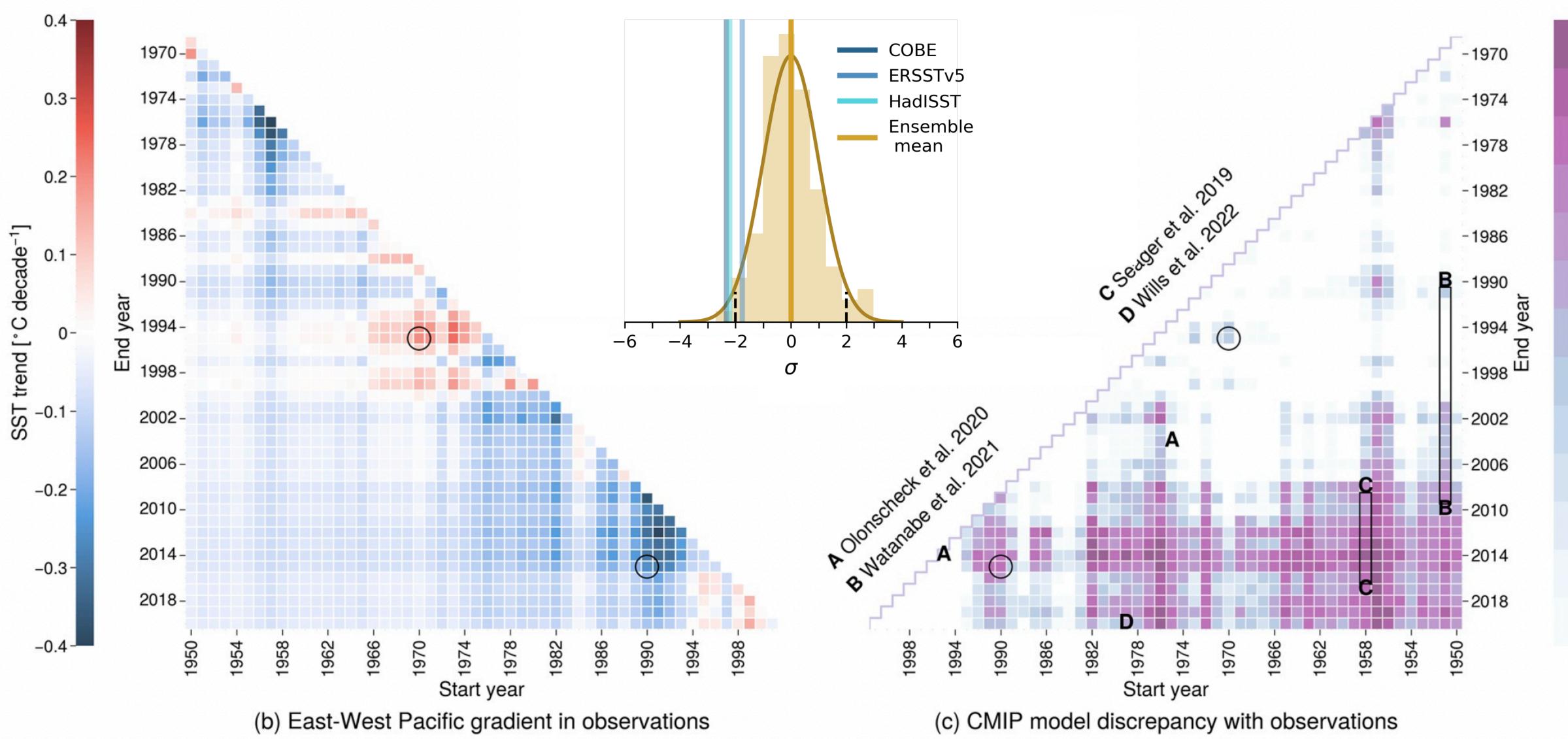




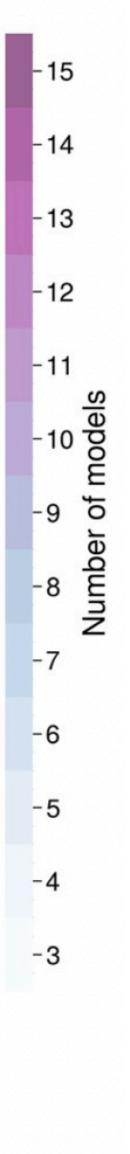




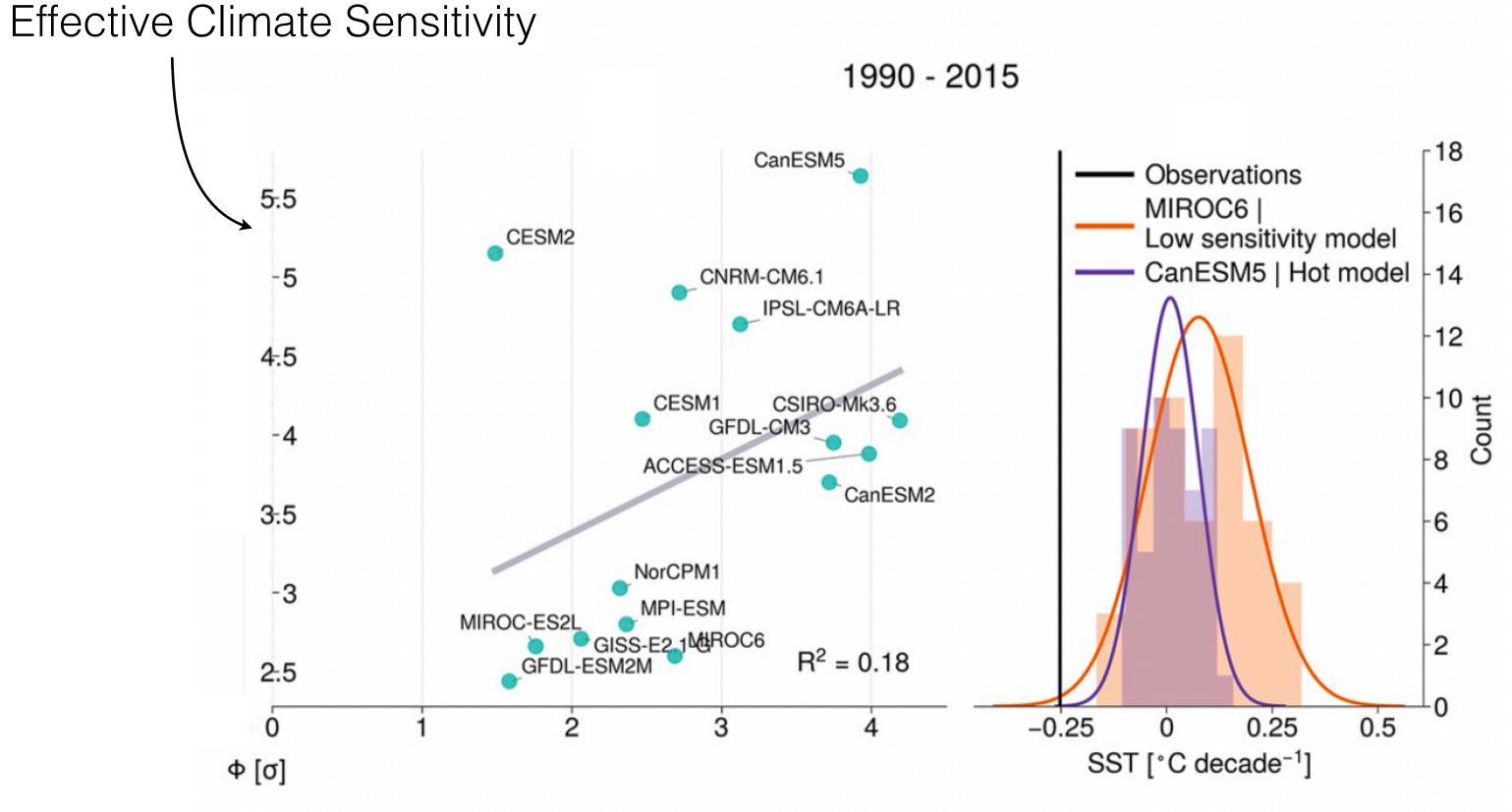






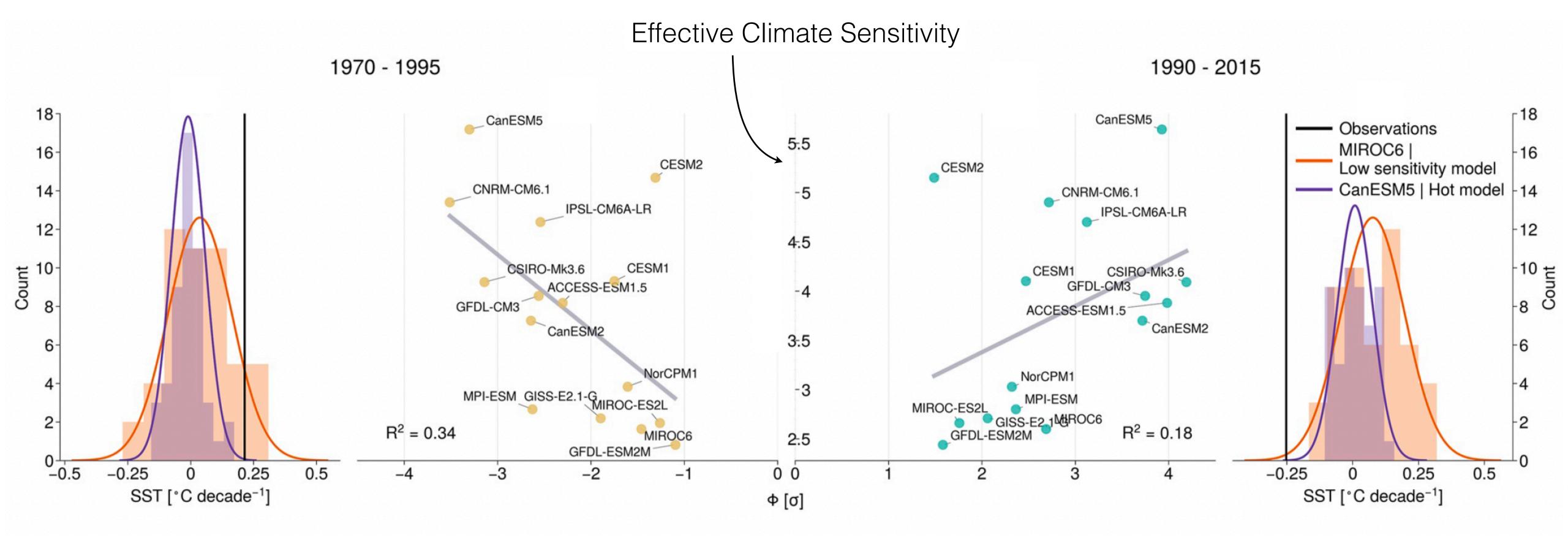


Models which cannot reproduce swings tend to have very high ECS





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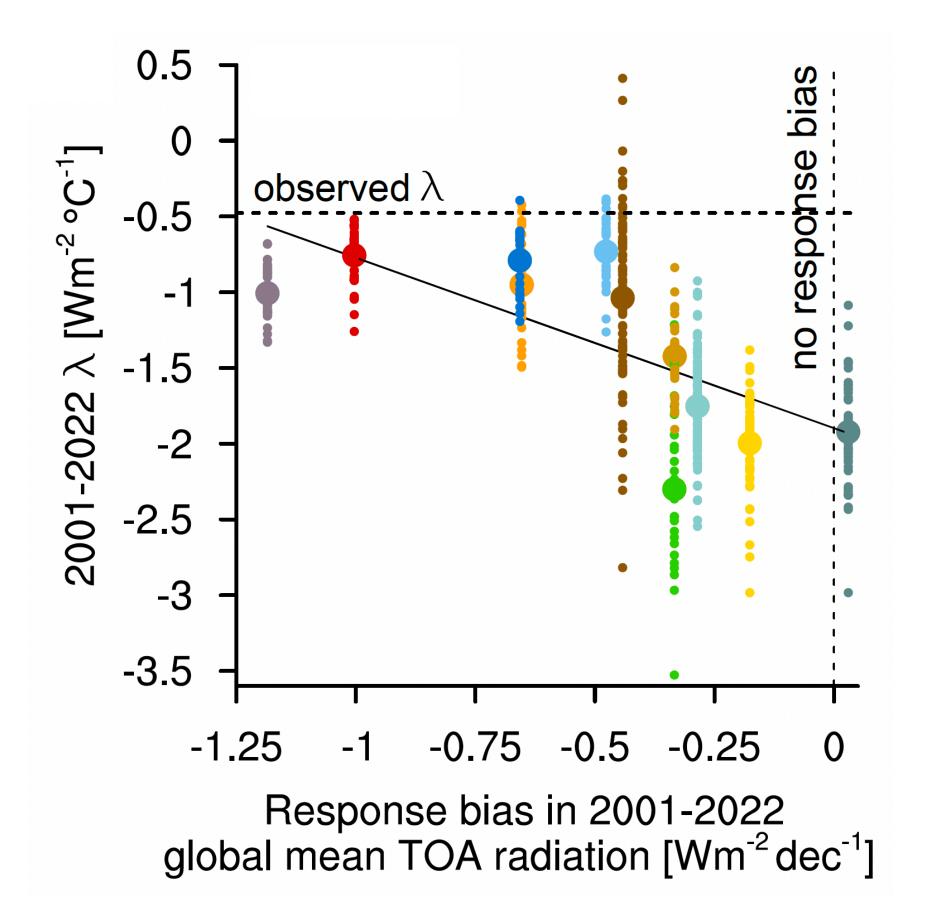


- Models all underestimate the slightly negative equatorial SST gradient on long timescales
- Half of the models underestimate the decadal-scale positive and negative swings
- Models with low variability tend to have a high climate sensitivity possibly due to a mismatch between SST and EIS short-wave cloud feedback

Wills et al. 2022, Seager et al. 2019, 2022, Watanabe et al. 2021, Olonscheck et al. 2020, Lee et al. 2022, Heede et al. 2020, 2021, Hwang et al. 2024, Kim et al. 2022, ...

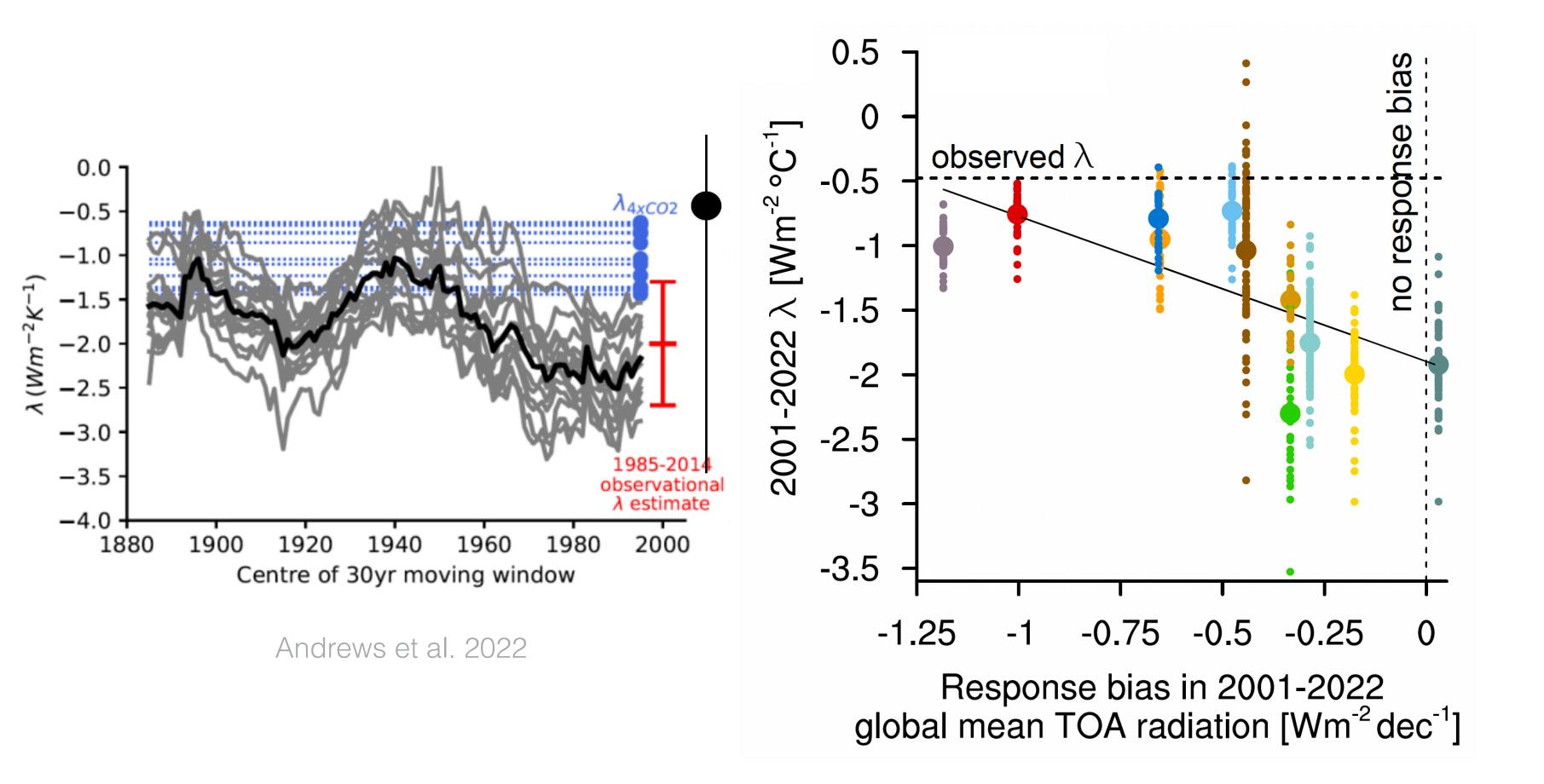


Radiative feedbacks change in time (unverifiably though)



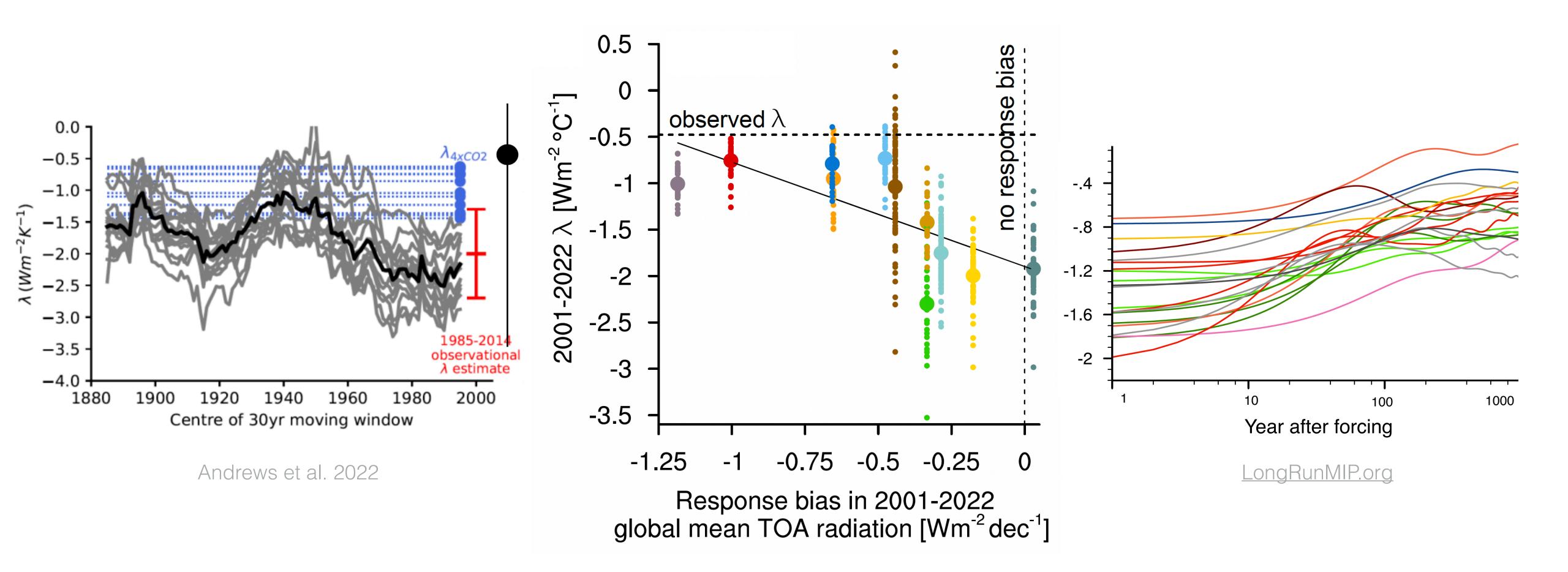


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Andrews et al. 2015, 2017, 2018, 2022; Williams et al. 2008, Winton et al. 2010, Armour et al. 2013, 2017; Zhou et al. 2016, 2017; Dong et al. 2019, 2020, Williams et al. 2023, Gregory et al. 2018, Ceppi and Gregory 2019, Rugenstein et al. 2016, 2020; Salvi et al. 2022, ...





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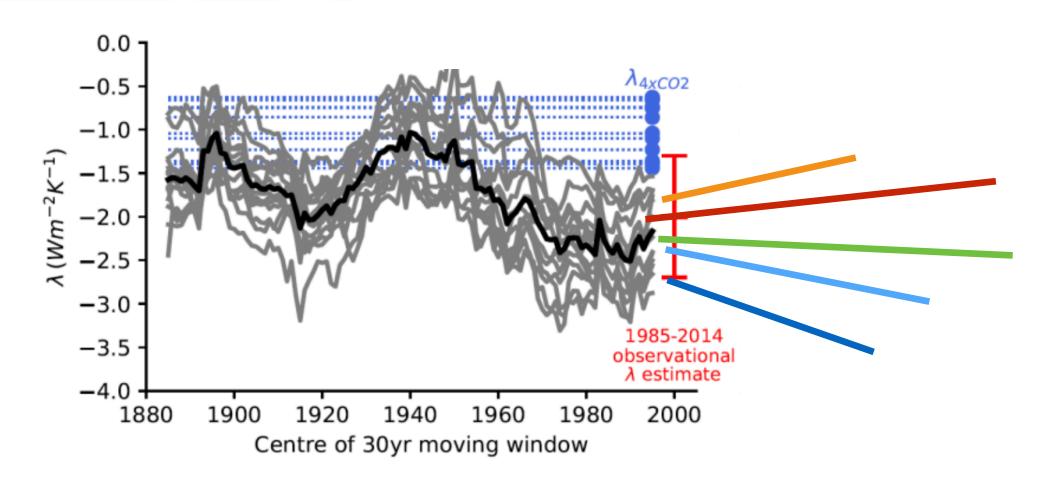




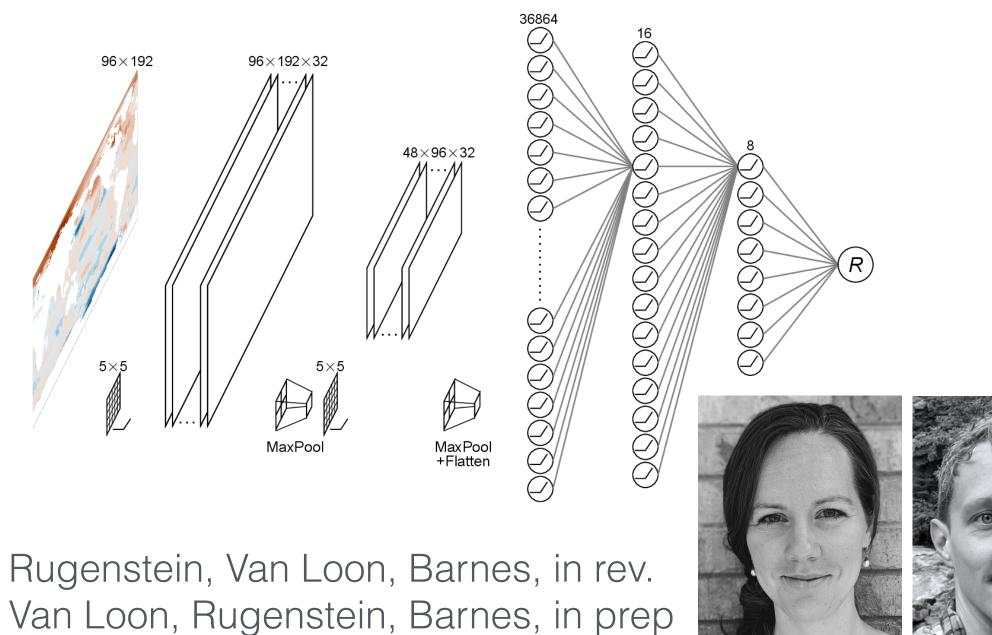




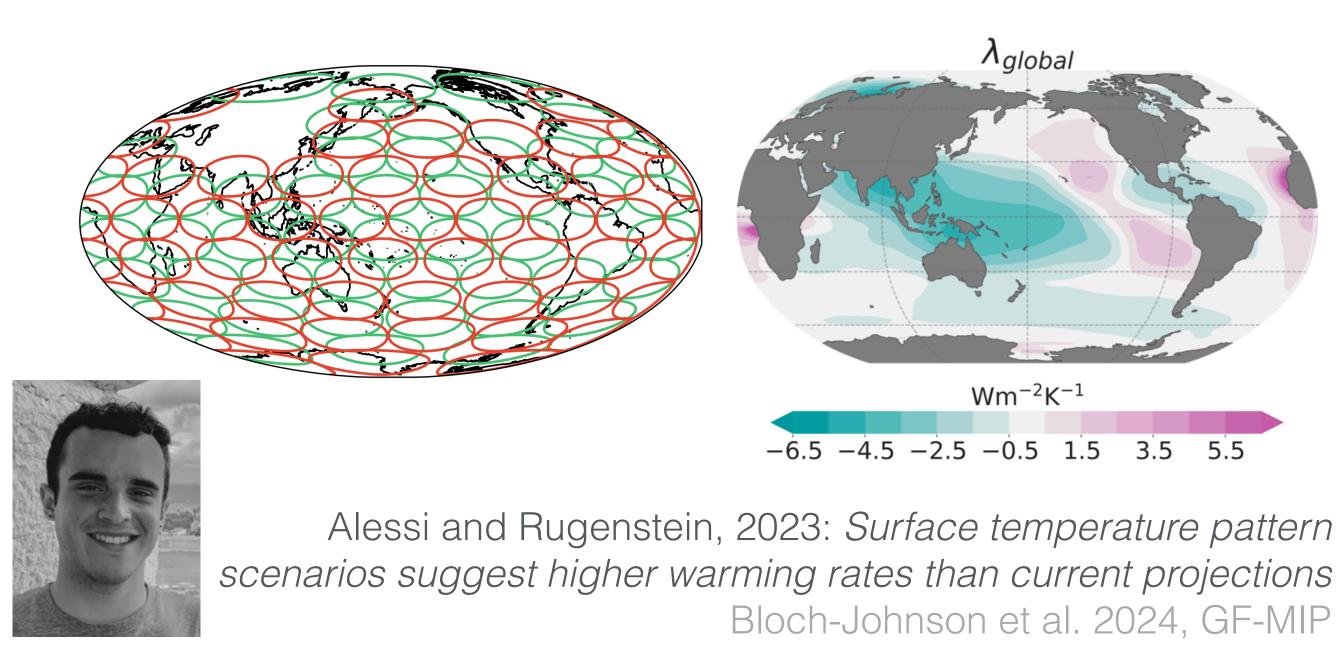




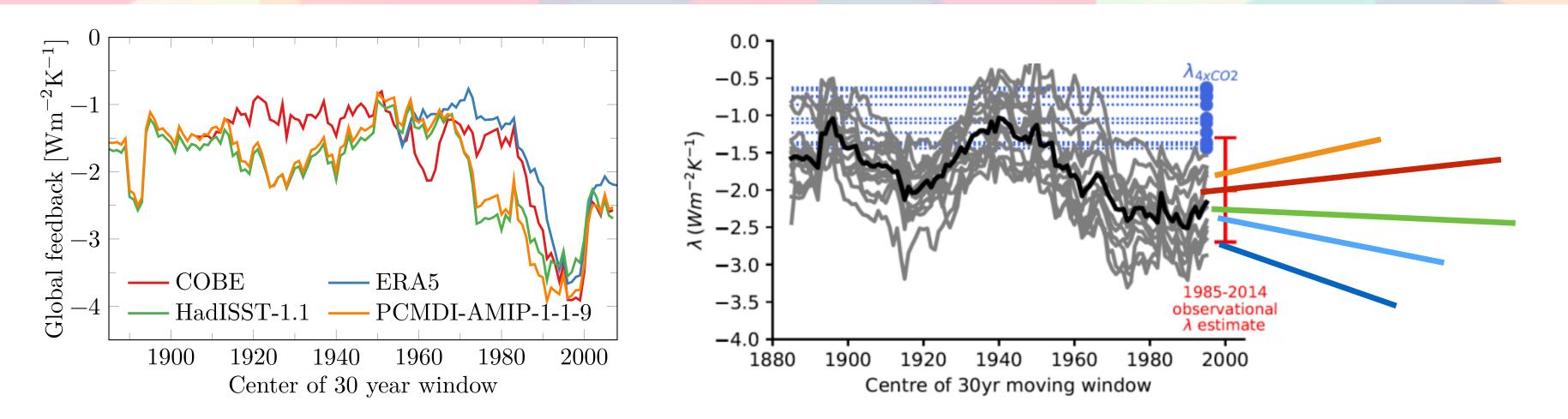
Convolutional Neural Networks nonlinear



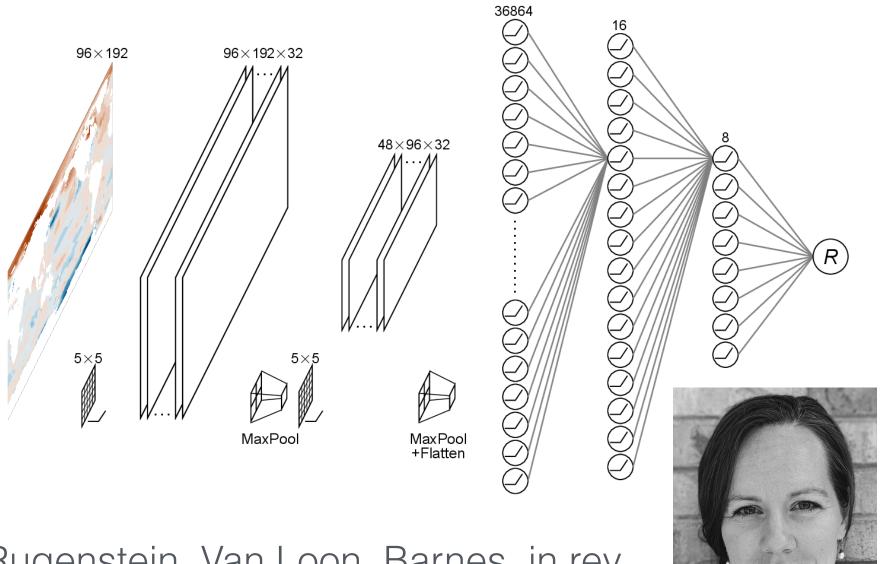
Green's function ~ SST patch simulations linear







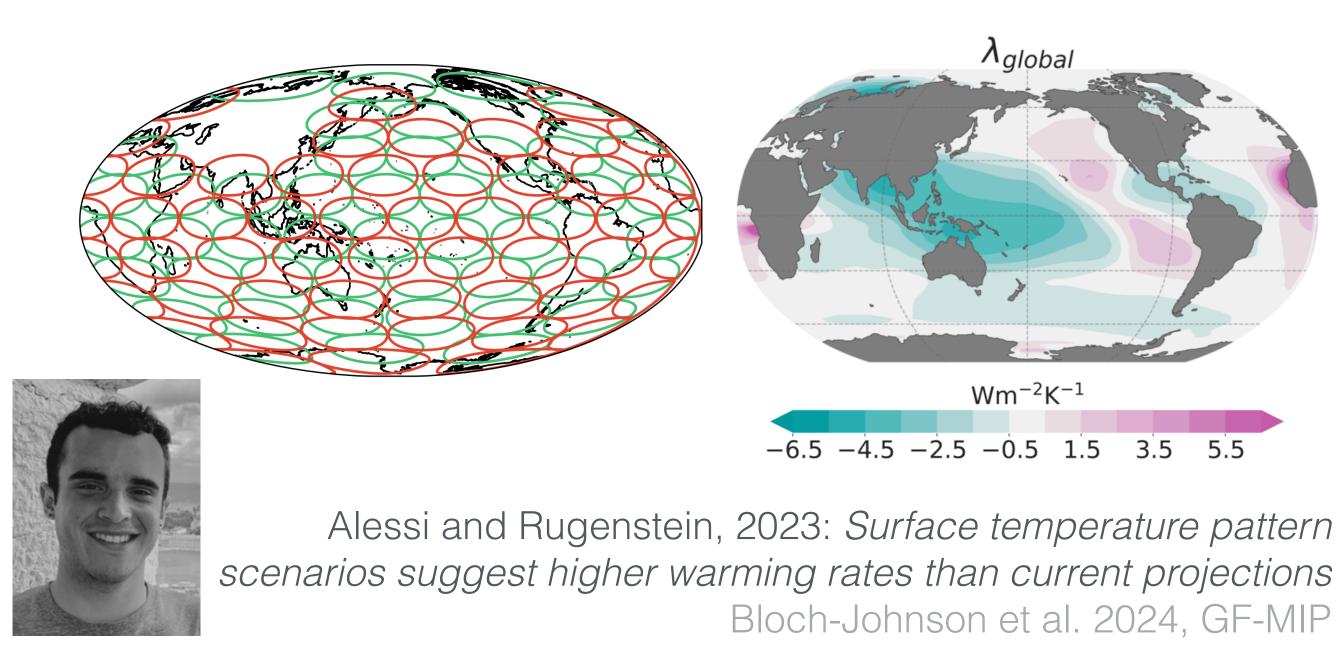
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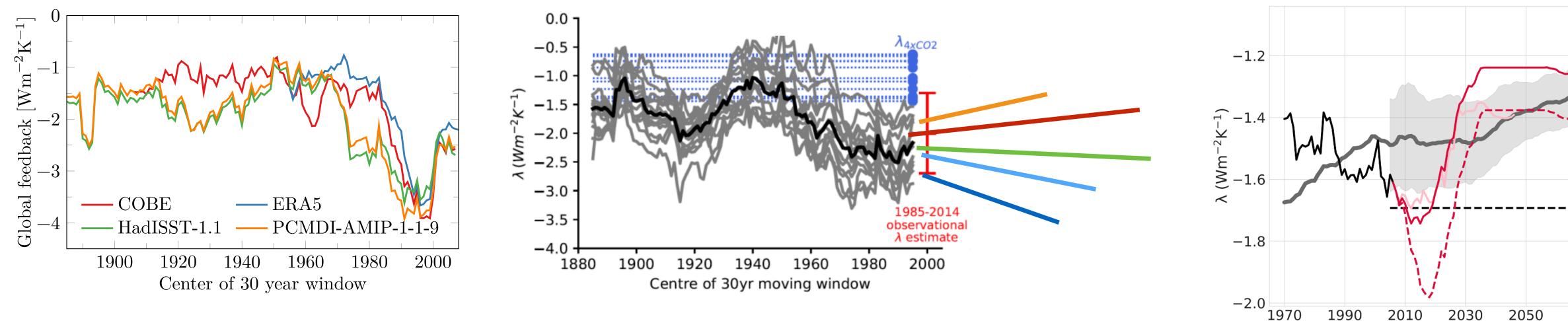
Rugenstein, Van Loon, Barnes, in rev. Van Loon, Rugenstein, Barnes, in prep



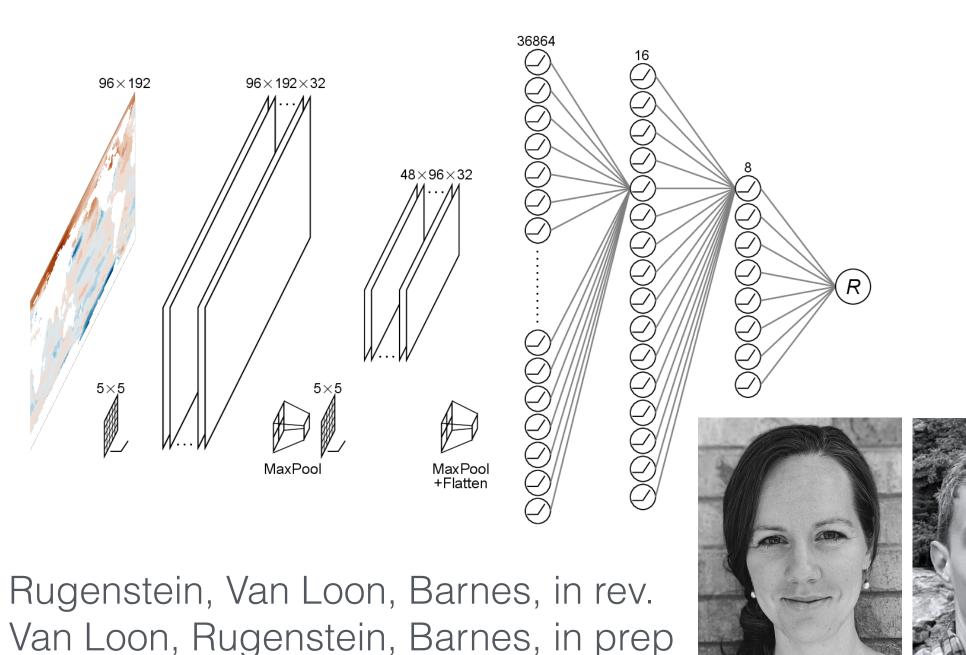
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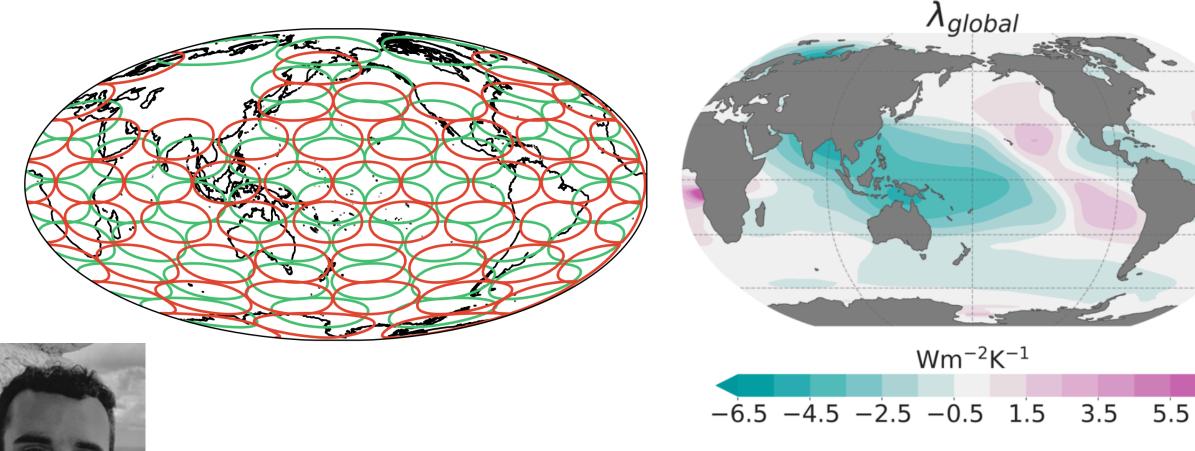




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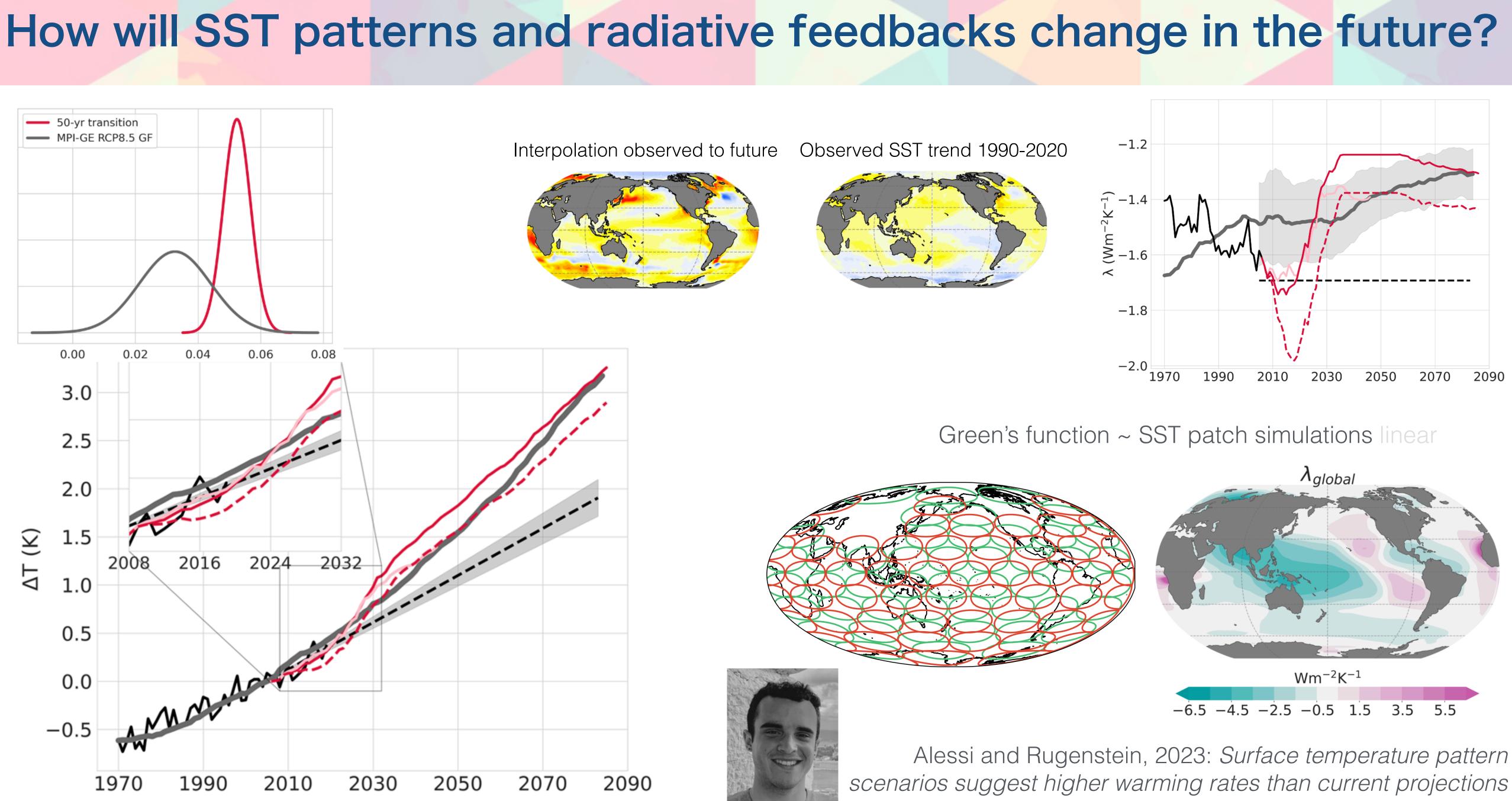


Green's function ~ SST patch simulations linear



Alessi and Rugenstein, 2023: *Surface temperature pattern* scenarios suggest higher warming rates than current projections Bloch-Johnson et al. 2024, GF-MIP



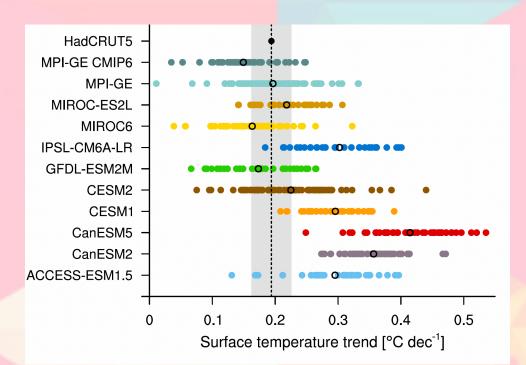


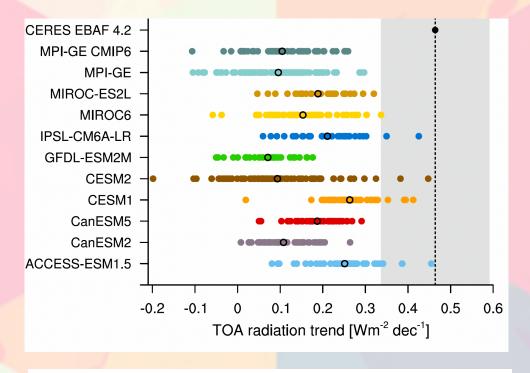
Bloch-Johnson et al. 2024, GF-MIP

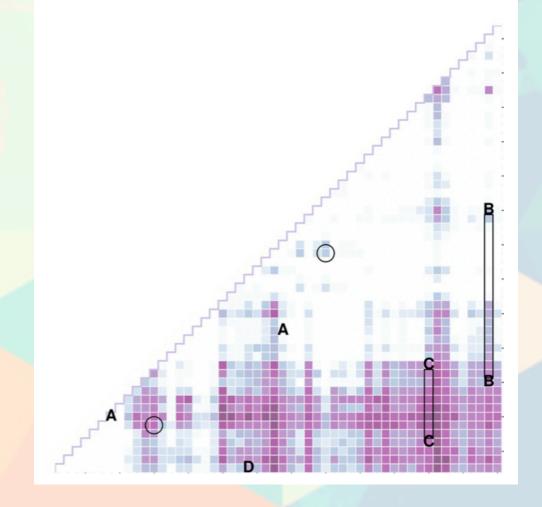


How the good and the bad conspire to the ugly **Observable TOA radiation trends** are seriously underestimated even though surface temperature is OK, feedbacks are unmeaningful Observable surface temperature trends are locally and for certain timescales systematically off, radiation is unknown and feedbacks unverifiable Implications for projections of climate change

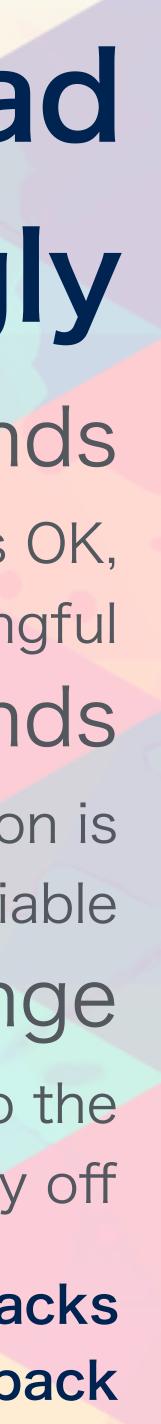
Climate models: weak forcing &v strong OHU balanced by too positive feedbacks **Observations: strong forcing** &v weak OHU balanced by strongly restorative feedback

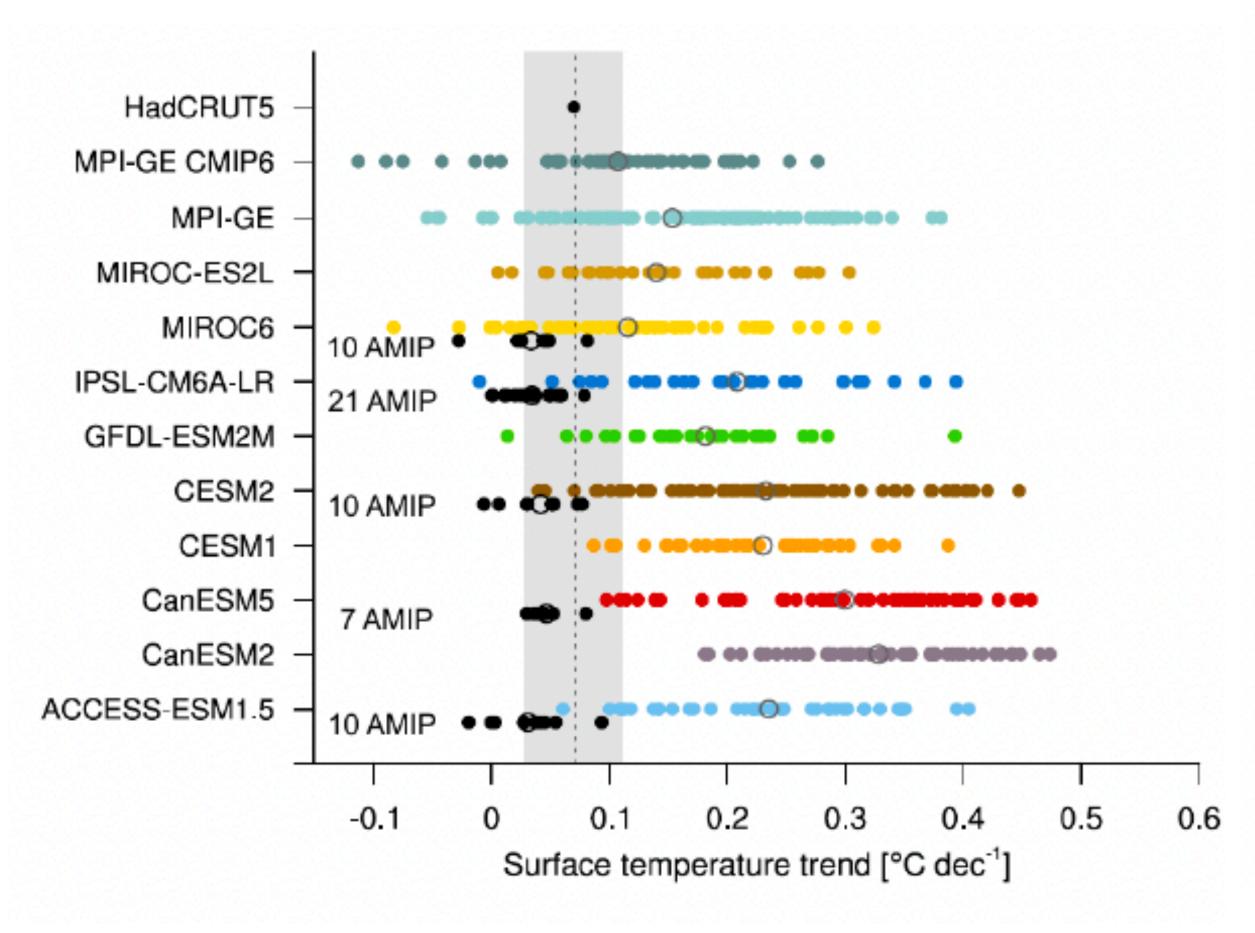


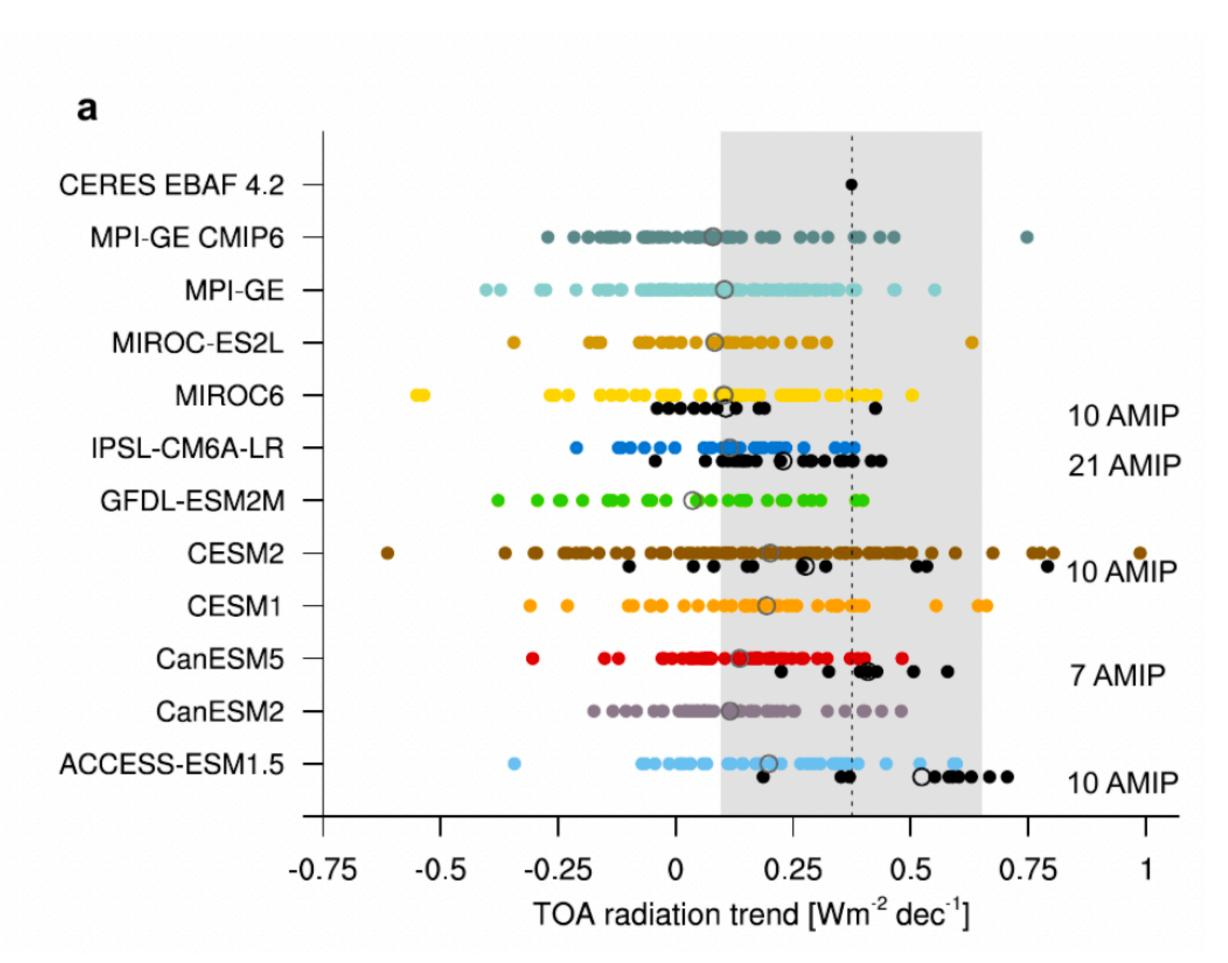


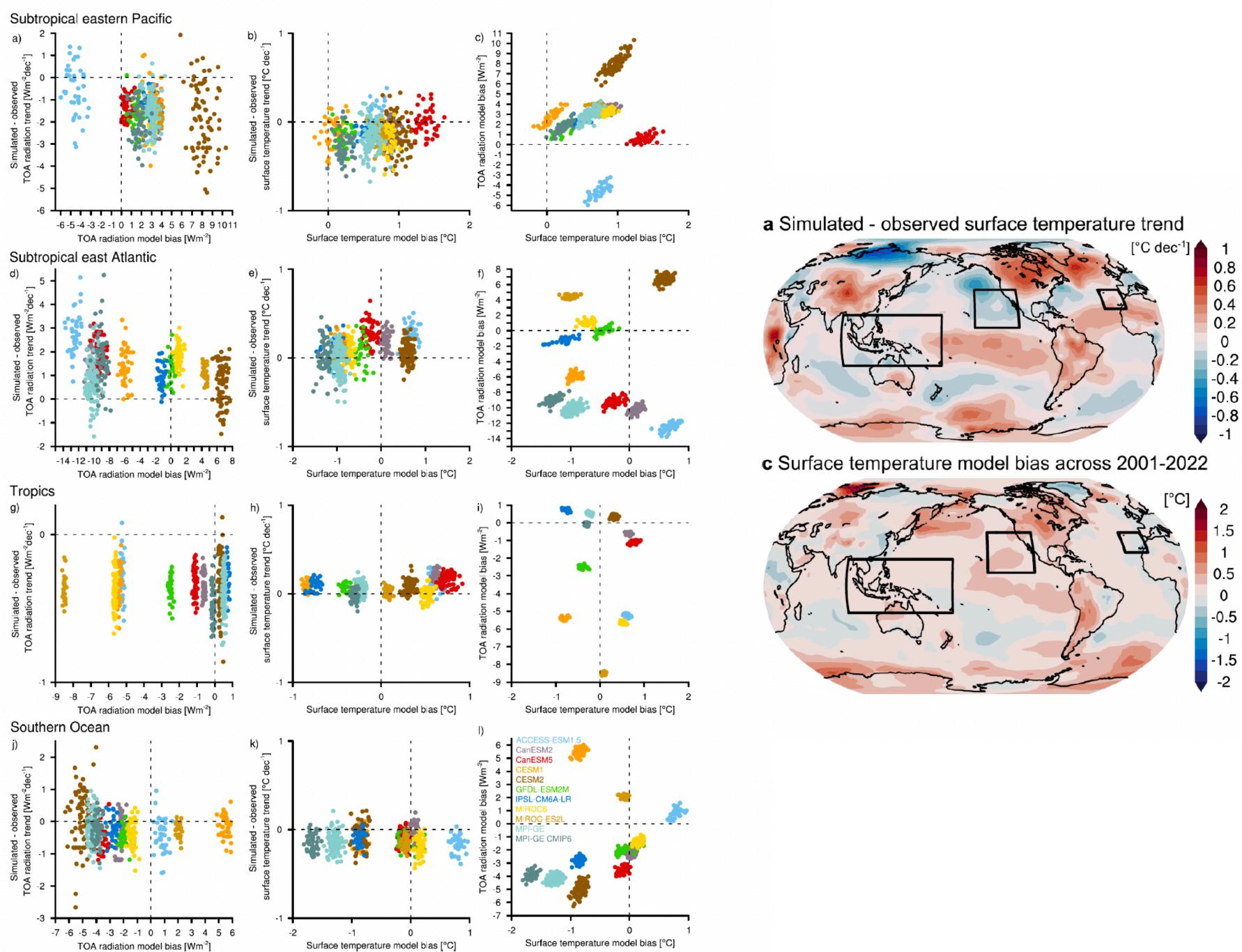


If the SST pattern problem and the heat uptake problem persist into the future projections of global-mean temperature might be seriously off

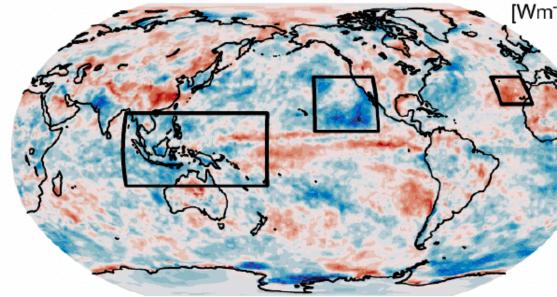








b Simulated - observed TOA radiation trend



d TOA radiation model bias across 2001-2022

