

Can Transfer Learning help identify Tropical State-Dependent Bias relevant to Midlatitude Subseasonal Predictability?

Kirsten J. Mayer¹, Katherine Dagon¹, and Maria J. Molina^{1,2}

¹NSF National Center for Atmospheric Research

²University of Maryland



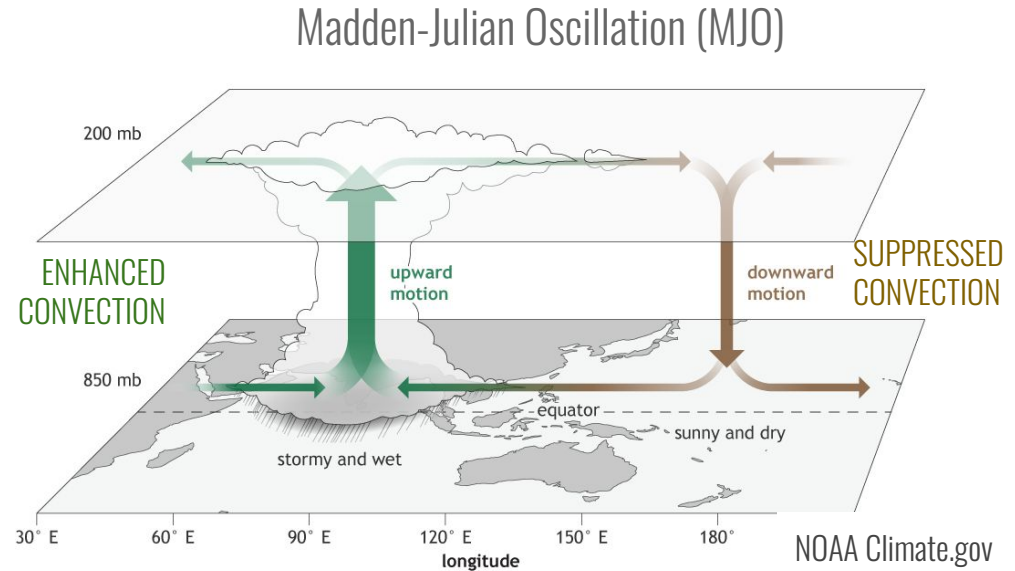
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State-Dependent Predictability

Forecasts of Opportunity - certain conditions can lead to more predictable behavior than others

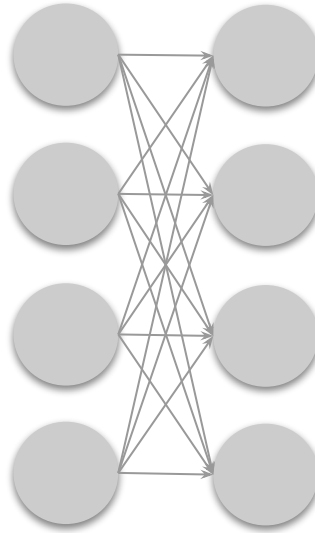
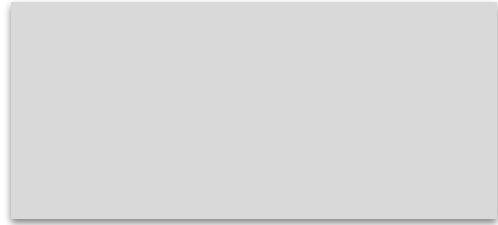
When the MJO is active, we use information about the state of the MJO today to predict what will happen to weather in the coming weeks



MJO is a convective dipole that propagates from the Indian Ocean into the central Pacific Ocean over about 20-90 days

State-Dependent Predictability

... identified by a neural network



Final layer uses the softmax function to convert the output into two values that sum to one

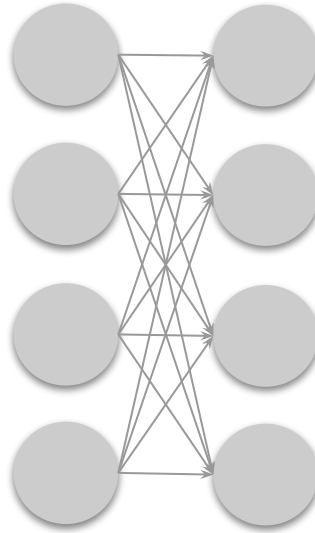
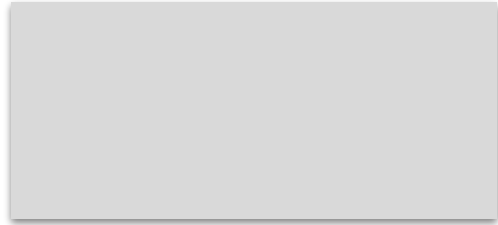


HIGH confidence



State-Dependent Predictability

... identified by a neural network



When the value is closer to one, is the network more often correct?

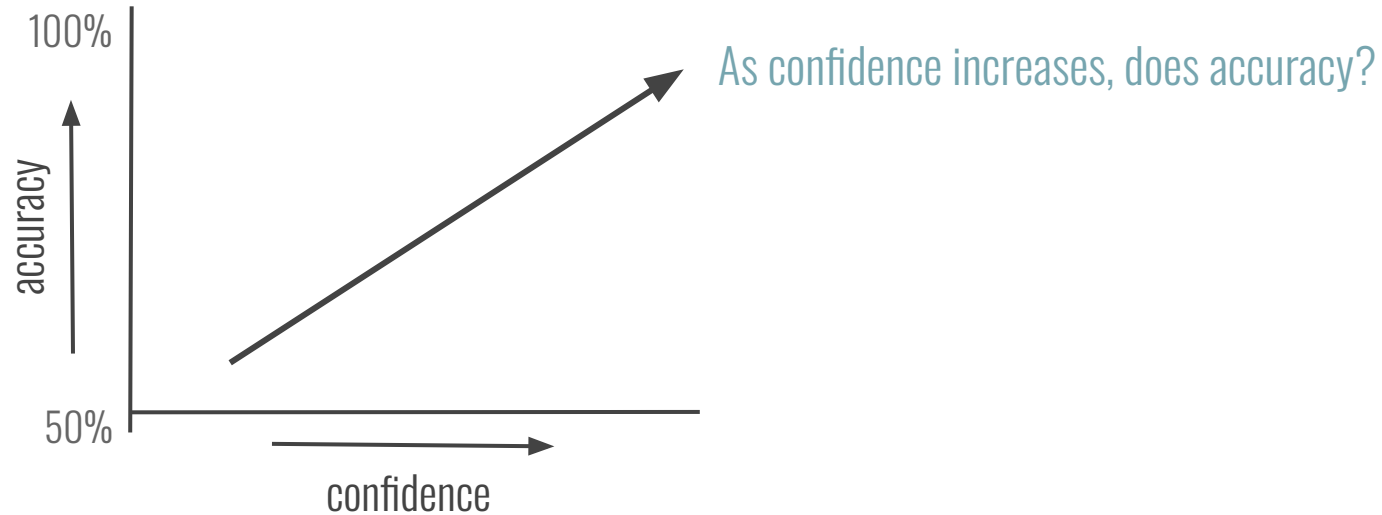


LOW confidence



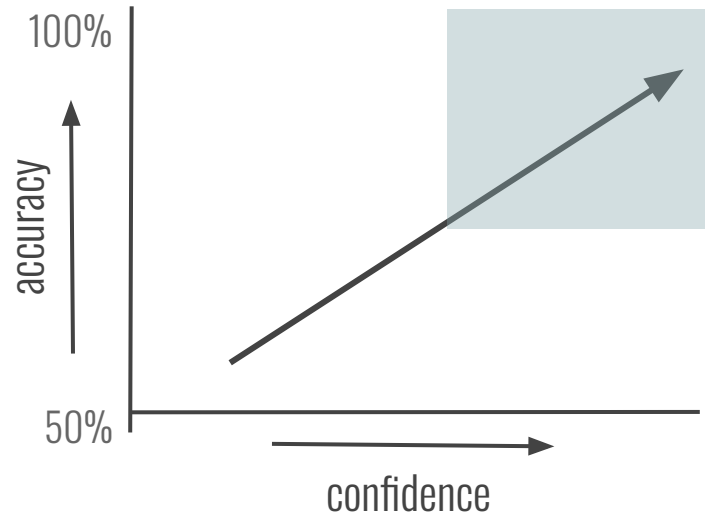
State-Dependent Predictability

... identified by a neural network



State-Dependent Predictability

... identified by a neural network

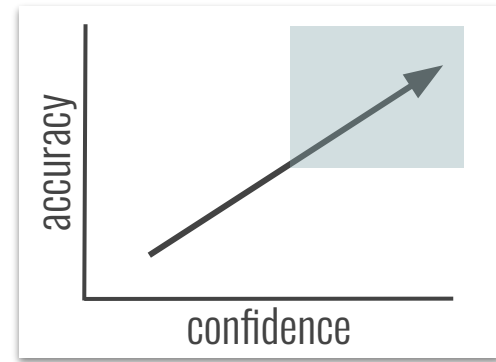


Forecasts of Opportunity

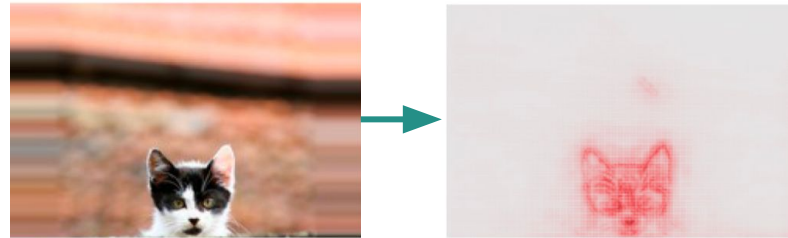
periods of enhanced predictability identified using network confidence (Mayer & Barnes 2021)

State-Dependent Predictability

... identified by an explainable neural network



with eXplainable Artificial Intelligence (XAI), we can identify sources of enhanced prediction skill



Montavon et al. 2017

State-Dependent Predictability Bias

... identified by an explainable neural network

bias: systematic differences in sources of enhanced predictability between model & obs

State-Dependent Predictability Bias

Transfer Learning

LOTS of data - biased

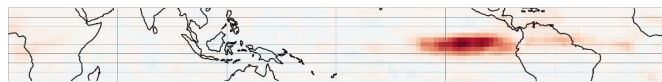
Original Neural Network
(trained with climate model data)

limited data - unbiased

Final Neural Network
(updated with reanalysis data)

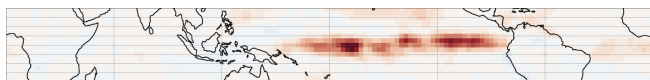
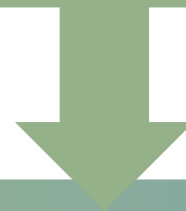
State-Dependent Predictability Bias

Transfer Learning and XAI



XAI ←

Original Neural Network
(trained with climate model data)



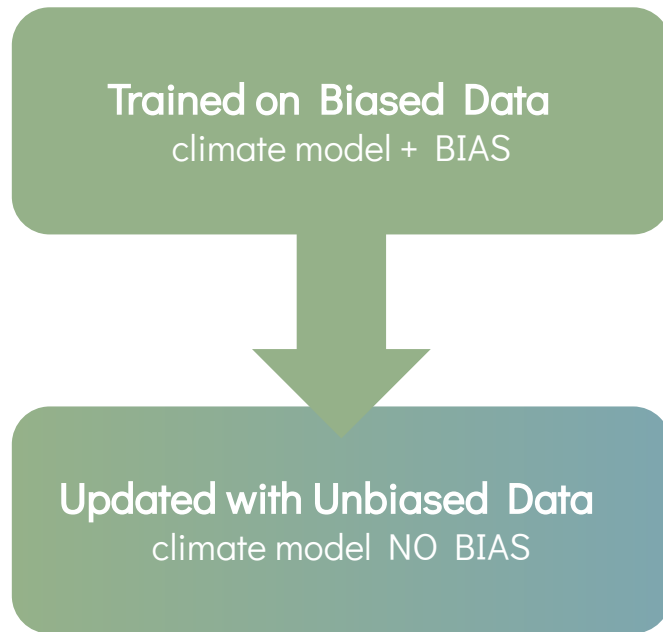
XAI ←

Final Neural Network
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State-Dependent Predictability Bias

Transfer Learning and XAI

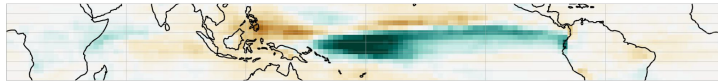
To test the feasibility of this approach...
perfect model framework



State-Dependent Predictability Bias

The prediction problem

E3SMv2



TROPICAL PRECIPITATION



+

-

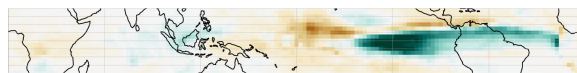


500 hPa GEOPOTENTIAL HEIGHT SIGN
days 14-21 (week 3)

State-Dependent Predictability Bias

Transfer Learning

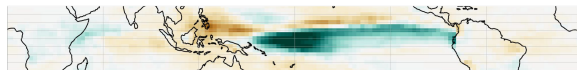
Artificially Biased E3SMv2



TROPICAL PRECIPITATION

BIAS = tropical precipitation + 60E *during ENSO*

E3SMv2



TROPICAL PRECIPITATION

Biased E3SMv2 trained weights



+

-



500 hPa GEOPOTENTIAL HEIGHT SIGN
days 14-21 (week 3)



+

-



500 hPa GEOPOTENTIAL HEIGHT SIGN
days 14-21 (week 3)

State-Dependent Predictability Bias

Transfer Learning

If transfer learning works... (once the network corrects the shift)
we expect the performance

post-TL network \cong pre-TL network

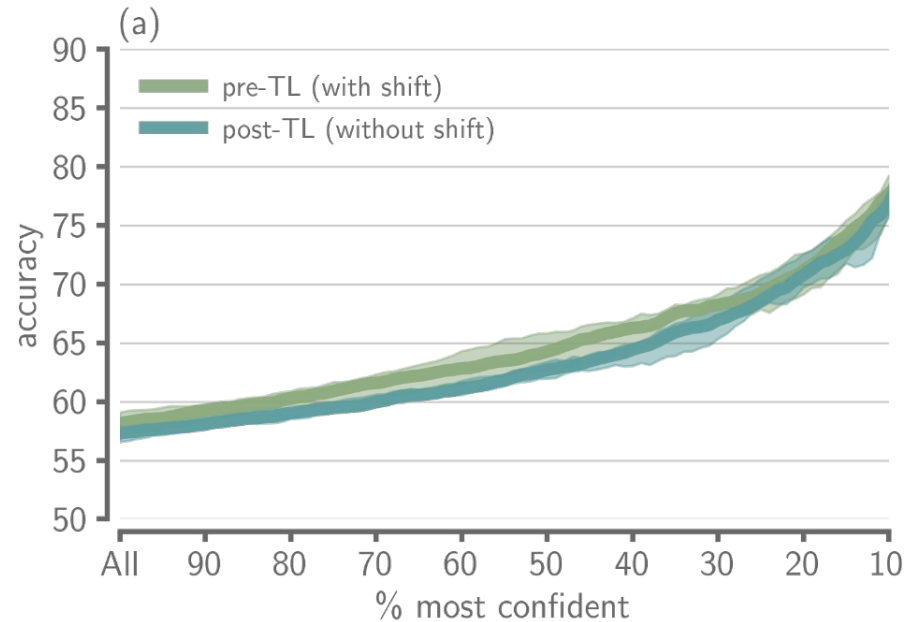


State-Dependent Predictability Bias

Transfer Learning

Both networks are skillful
and able to identify
forecasts of opportunity

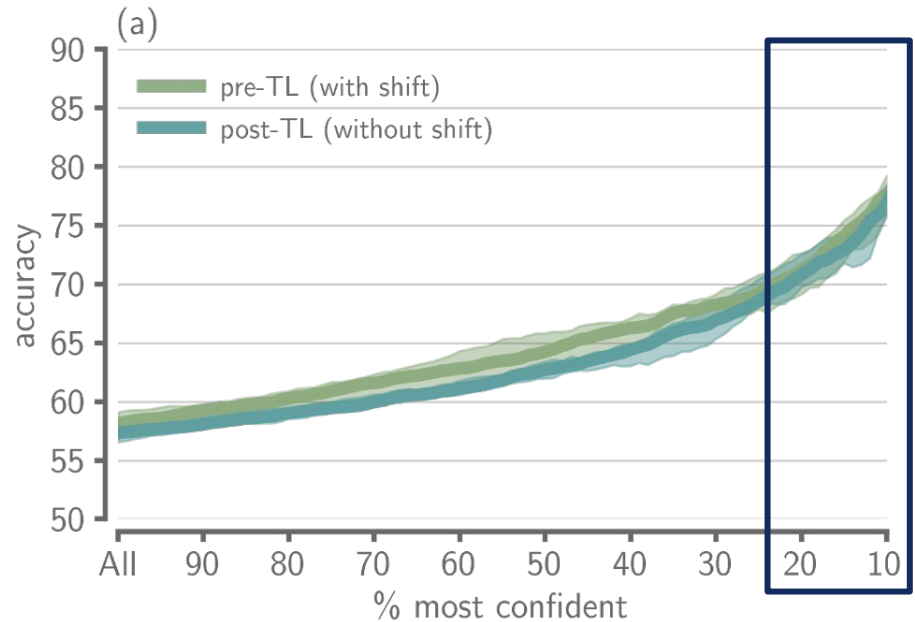
Transfer learning worked!



State-Dependent Predictability Bias

Transfer Learning

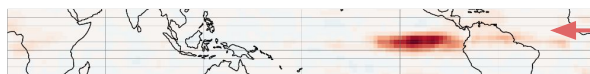
Let's pretend we don't know the true bias & look at network-identified forecasts of opportunity



State-Dependent Predictability Bias

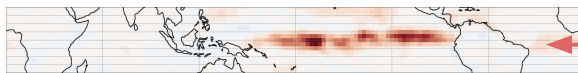
Transfer Learning with XAI

Biased E3SMv2 Trained Network
(before transfer learning)



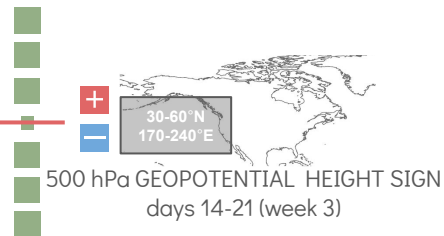
XAI HEATMAP

Compare XAI Heatmap
before & after transfer learning



XAI HEATMAP

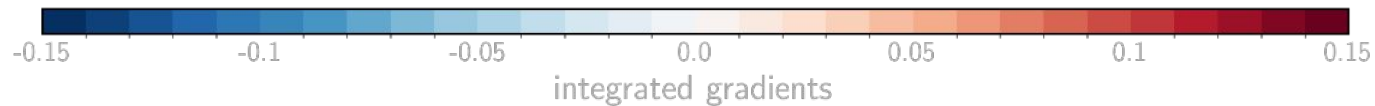
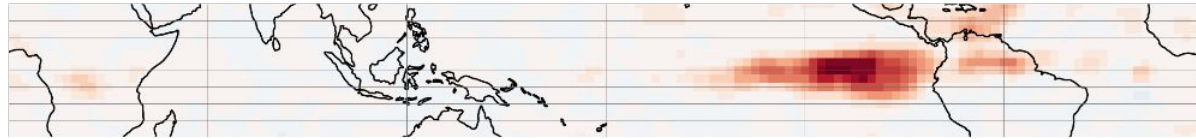
E3SMv2 Updated Network
(after transfer learning)



State-Dependent Predictability Bias

Transfer Learning with XAI

(a) Pre-TL: Biased Test Member (N=7768)



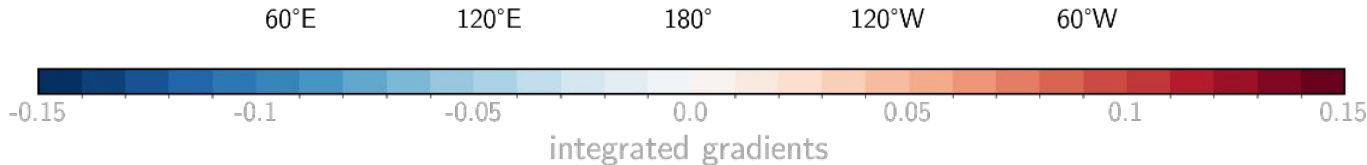
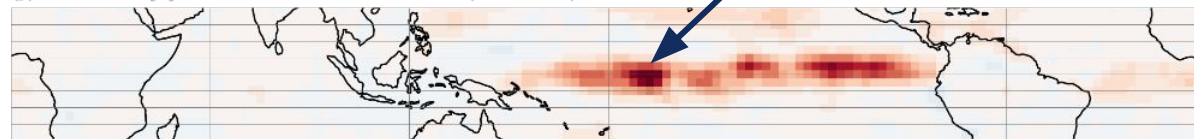
State-Dependent Predictability Bias

Transfer Learning with XAI

(a) Pre-TL: Biased Test Member (N=7768)



(j) Post-TL [9]: Unbiased Test Member (N=7100)



State-Dependent Predictability Bias

Transfer Learning with XAI

These results use **NINE** ensemble members to re-tune...
what if we only had **ONE** ensemble member

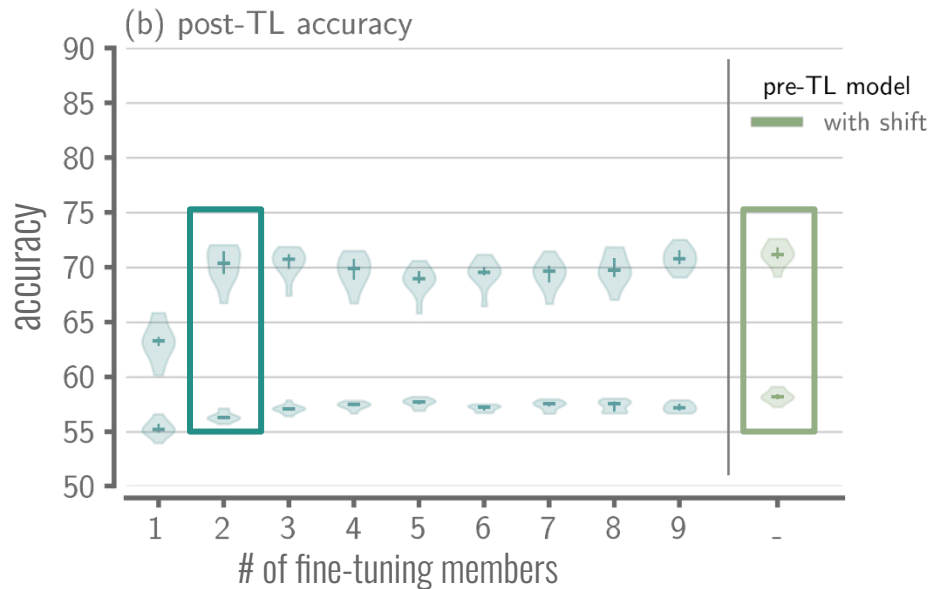


State-Dependent Predictability Bias

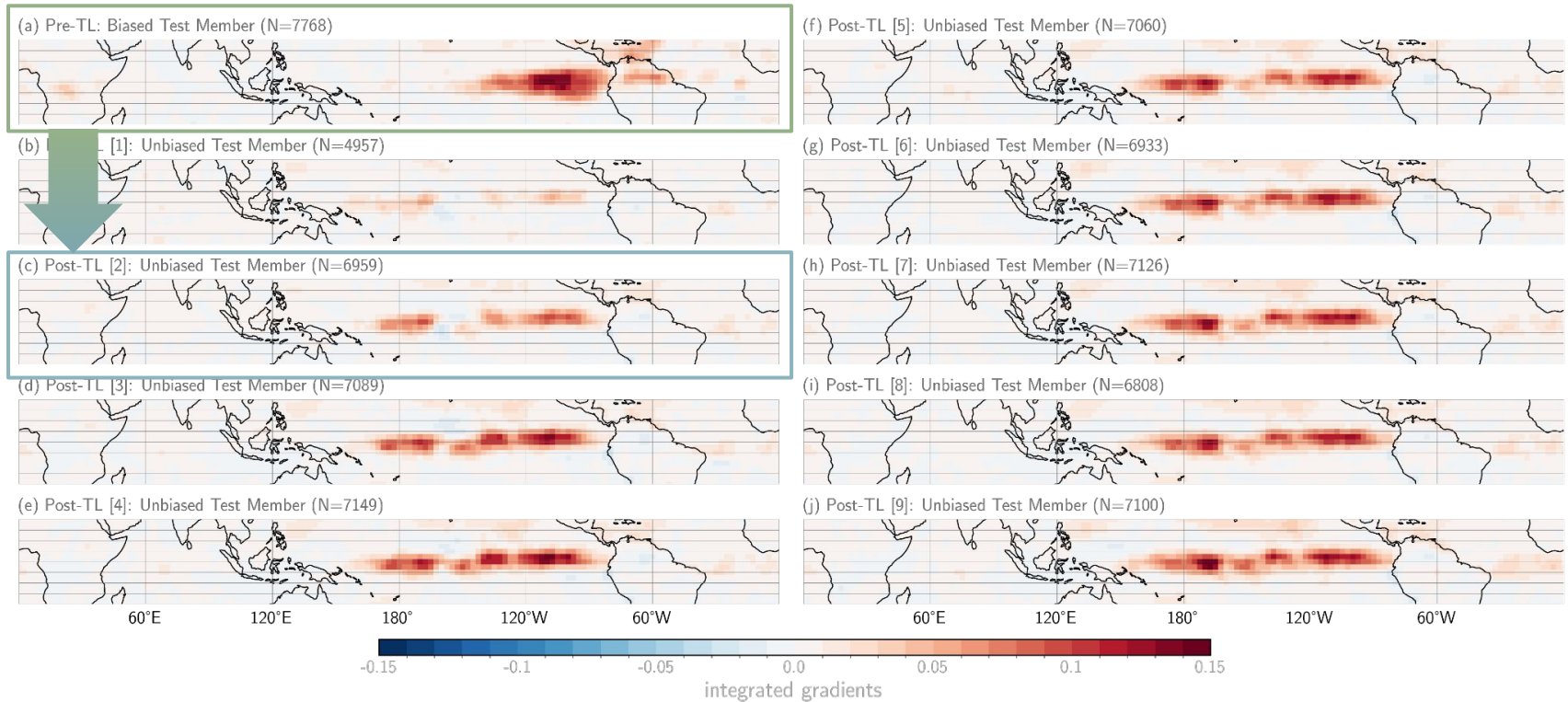
Transfer Learning

Need ~2 tuning members
to get to similar accuracy
as the original model....

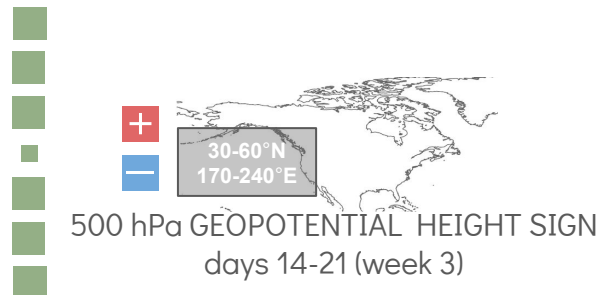
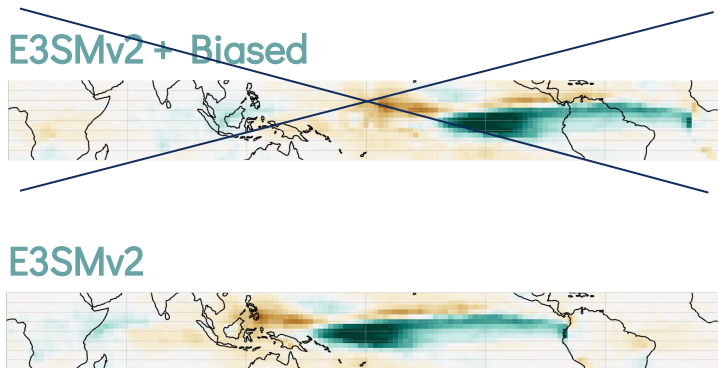
128 years of data
(<1900-Today)



State-Dependent Predictability Bias

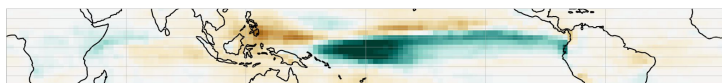


State-Dependent Predictability Bias



State-Dependent Predictability Bias

E3SMv2

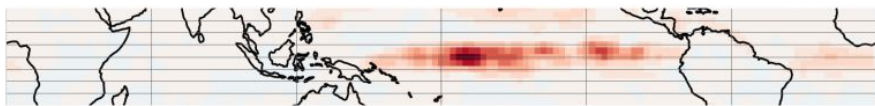


TROPICAL PRECIPITATION

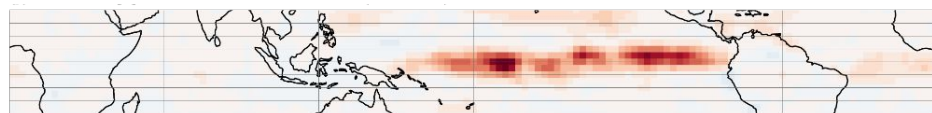


30-60°N
170-240°E

500 hPa GEOPOTENTIAL HEIGHT SIGN
days 14-21 (week 3)



No Transfer Learning



Post-Transfer Learning

Can Transfer Learning identify Tropical State-Dependent Bias Relevant to Midlatitude Subseasonal Predictability?

... kinda ...
but not with reanalysis



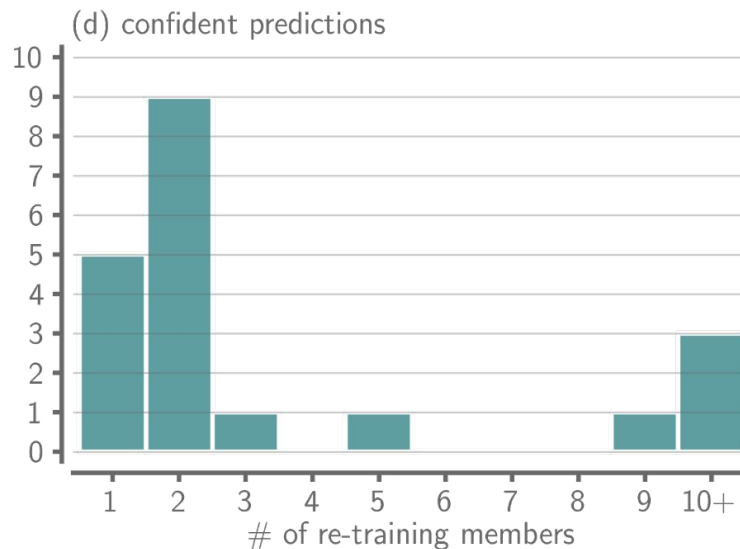
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State-Dependent Predictability Bias

Transfer Learning

This is common across different testing data.



State-Dependent Predictability Bias

Convolutional Neural Network

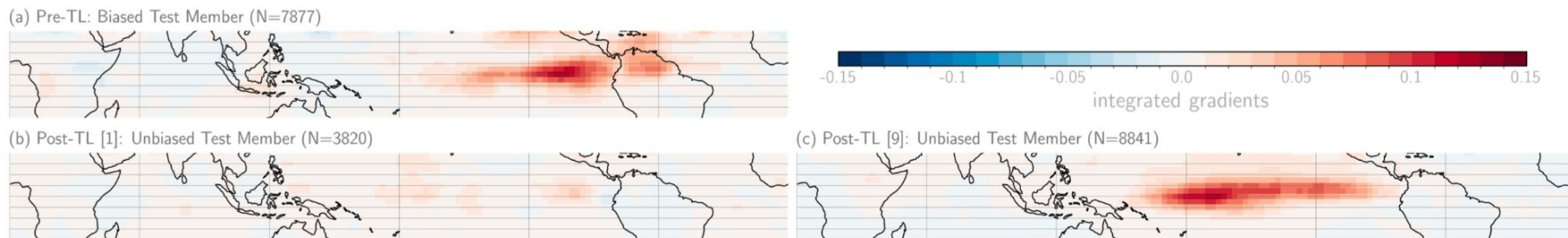


FIG. A1. As in Figure 3, but for negative predictions from a convolutional neural network for (b) one or (c) nine retraining members.