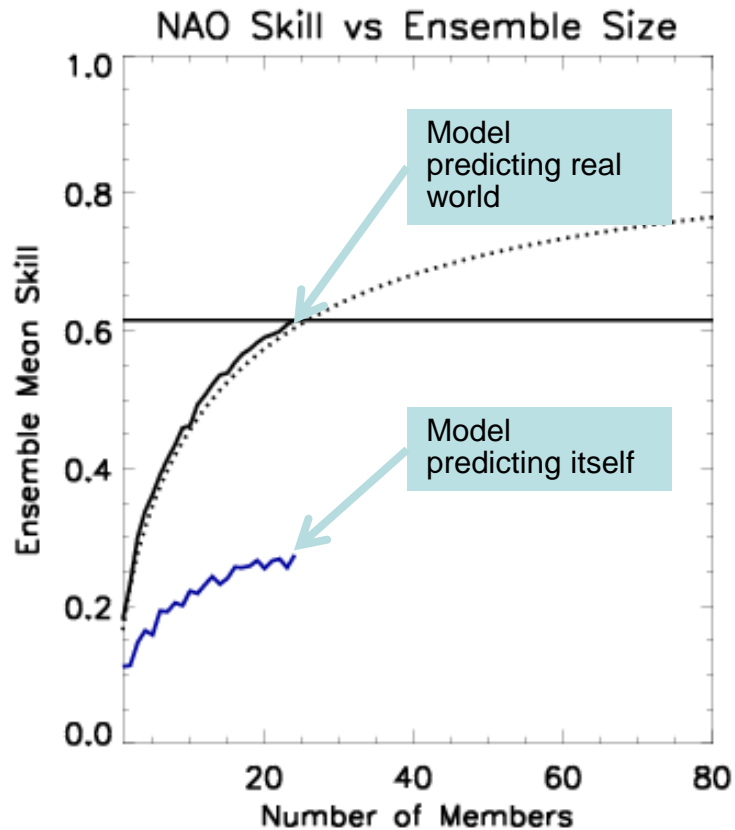


Development of the signal-to-noise paradox in
subseasonal forecasting models: After how long?
Where? Why?

Chaim I. Garfinkel, Chen Schwartz, Jeff Knight, Masakazu Taguchi,
Amy Butler, Daniela Domeisen, Wen Chen

EGU 2024

Signal to Noise paradox



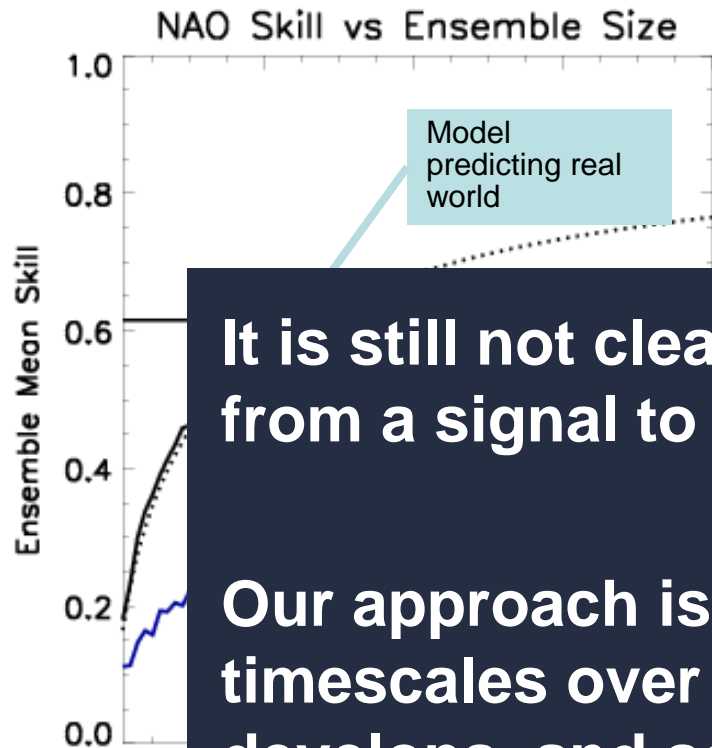
Paradox: models predict the real world better than themselves despite perfectly representing themselves

Members NOT alternate realisations of obs

Need a very large ensemble to extract the predictable signal

Models should not be taken at face value

Signal to Noise paradox



Paradox: models predict the real world better than themselves despite perfectly representing themselves

It is still not clear why models suffer from a signal to noise paradox.

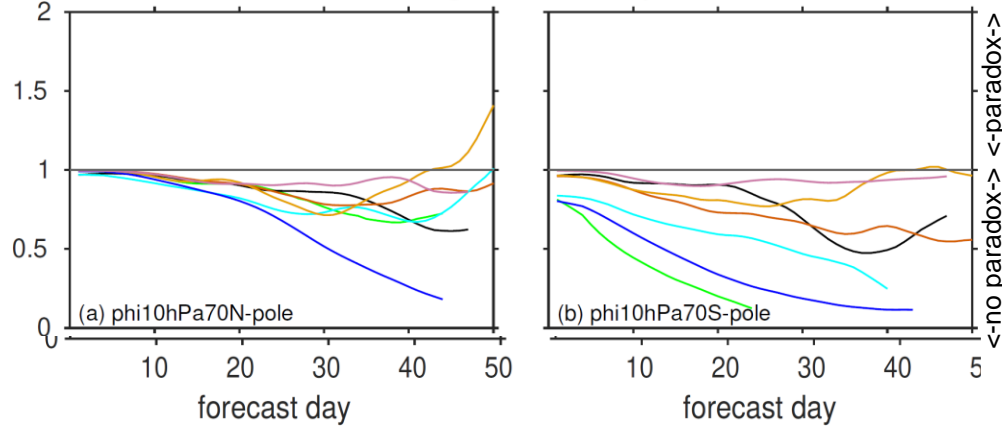
Our approach is to analyze the timescales over which the paradox develops, and so hope to provide some insight into its possible causes.

sations of obs
to extract the
al
at face value

Eade et al 2014, Scaife et al 2014, Dunstone et al 2016, 2018, Siegert et al 2016, Baker et al 2018, Scaife and Smith 2018

Where is the signal-to-noise paradox present on subseasonal timescales?

Where and when does the signal to noise paradox manifest?



NCEP
 ECMWF2020
 UKMO2020
 CMA
 BOM
 KMA
 CNRM2019

$$RPC^2 = r_{mo}^2 / r_{mm}^2$$

r_{mo} : correlation of ensemble mean with obs

r_{mm} : correlation of ensemble mean with randomly chosen member left out



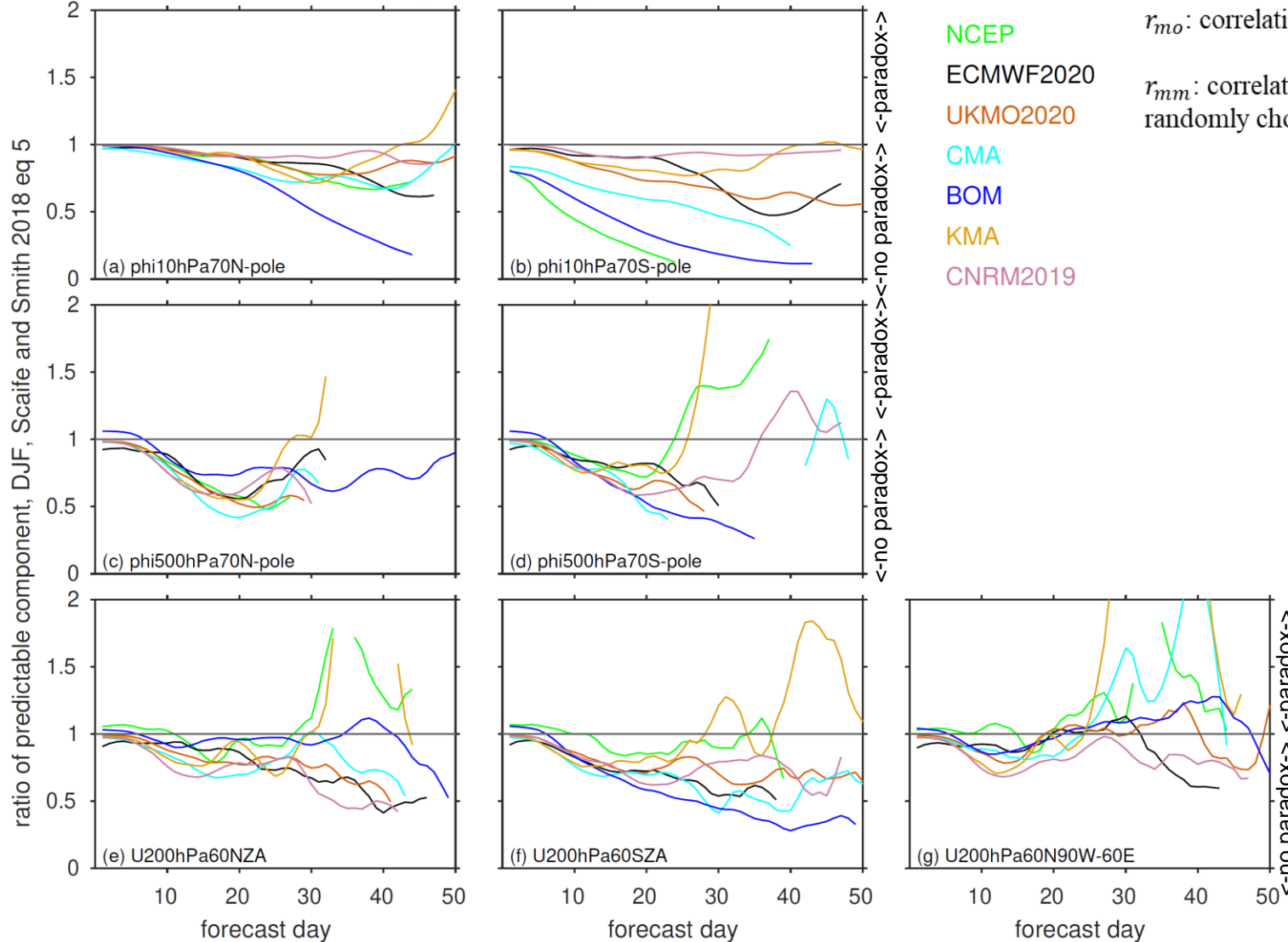
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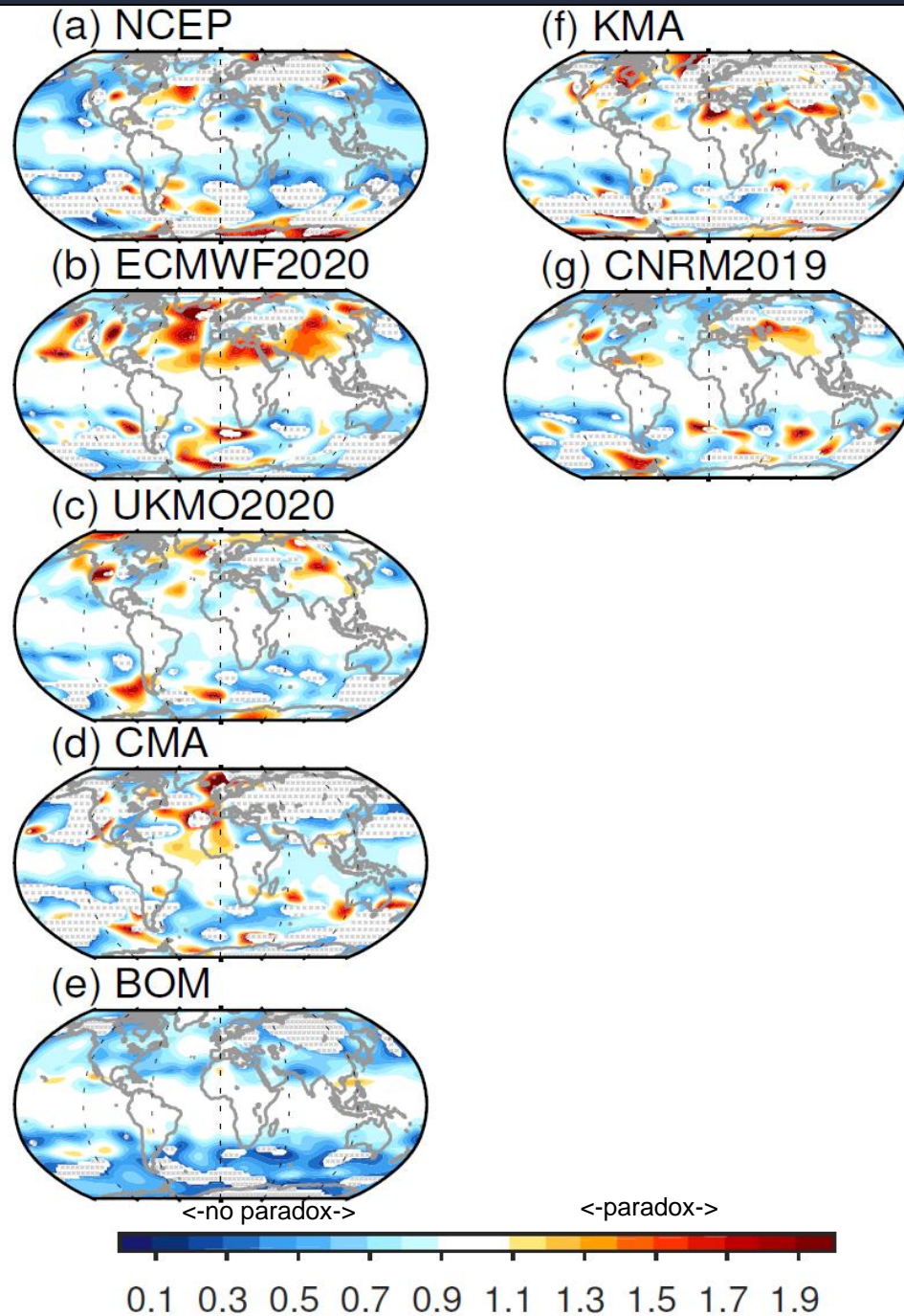
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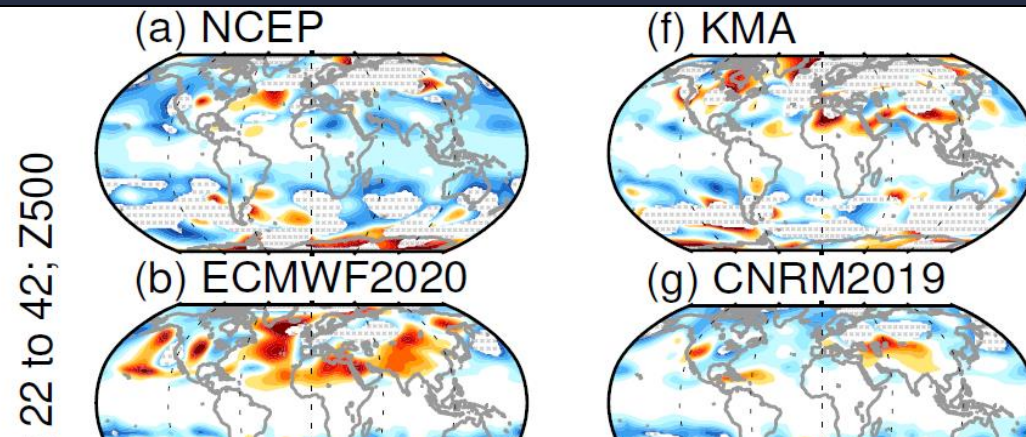


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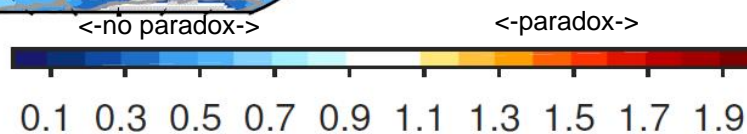
Where is there a signal to noise paradox? day: 22 to 42; Z500



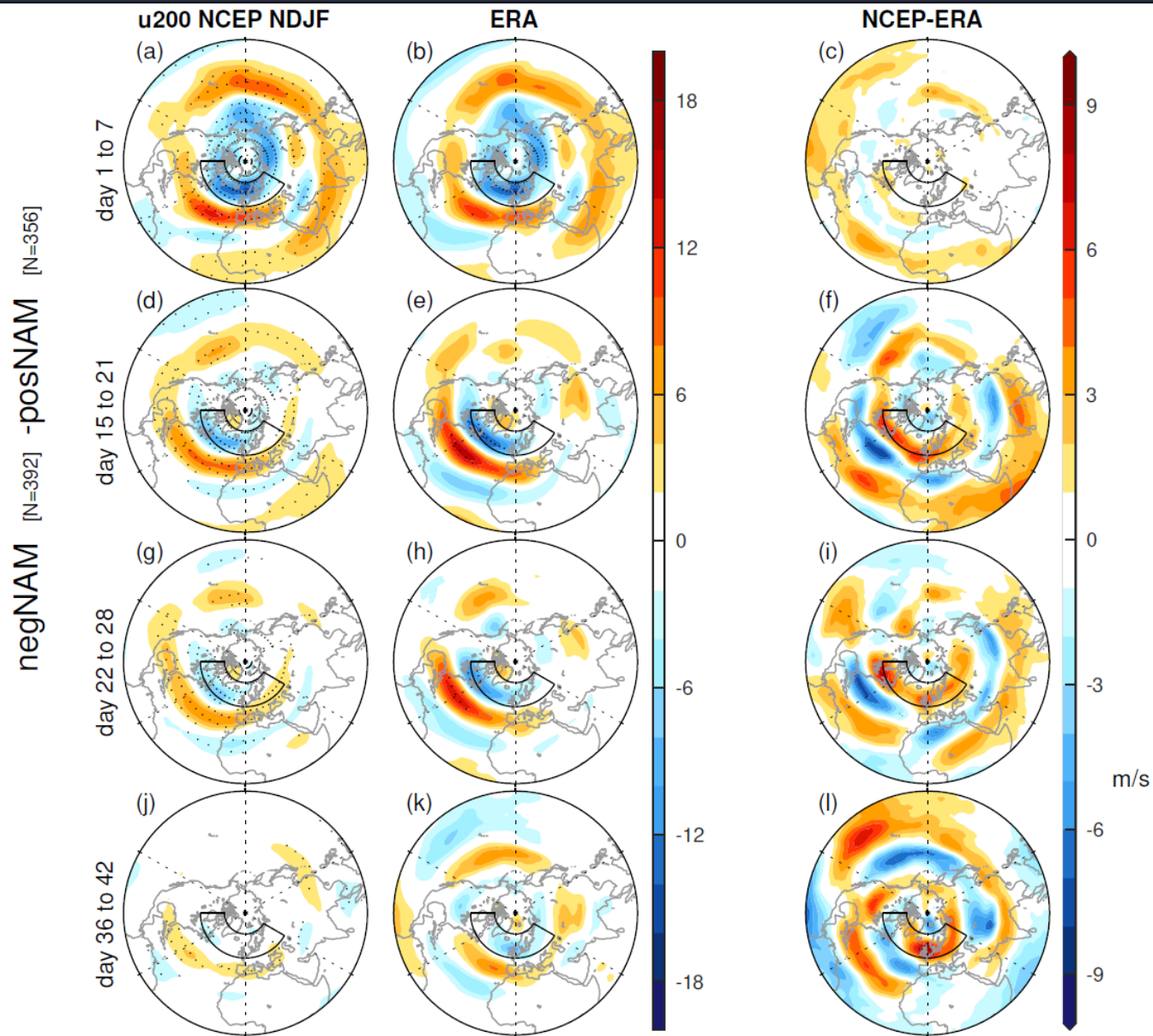
Where is the signal-to-noise present on subseasonal timescales?



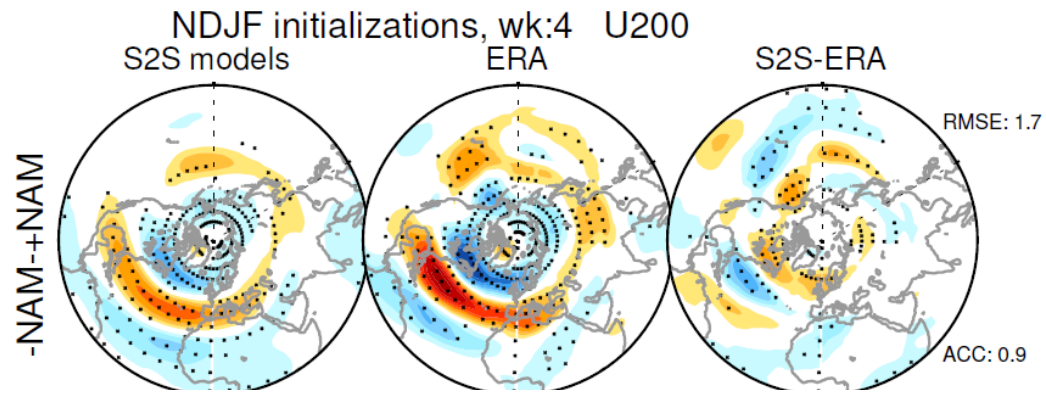
Given the concentration of $RPC > 1$ in the Atlantic sector, the next step is create composites of initializations in which polar cap Z at 500hPa is anomalously low (+NAM) and anomalously high (-NAM), and then composite U at 200hPa



Overly weak NAM persistence

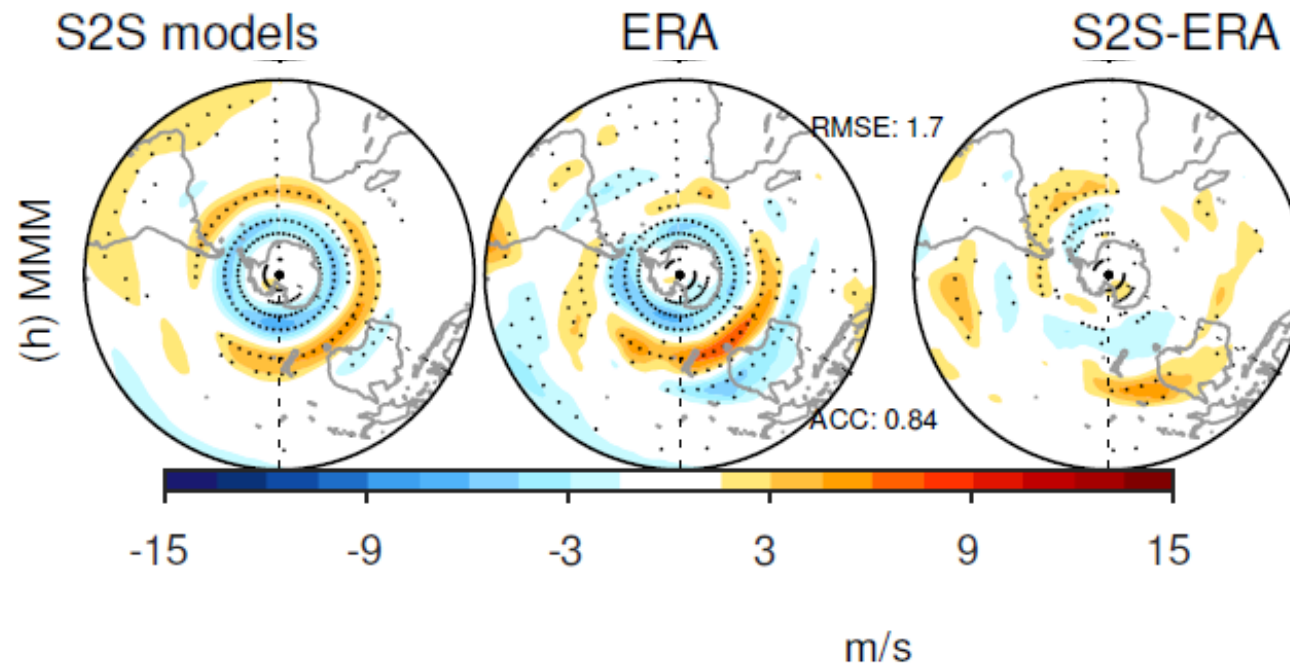


Overly weak NAM persistence



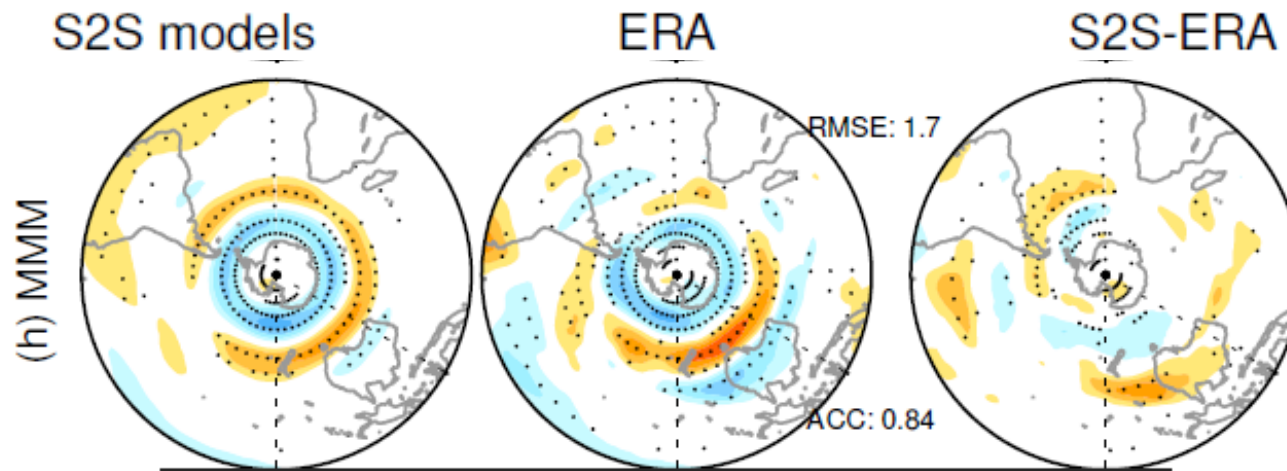
The SAM is too persistent, if anything

-SAM minus +SAM ND initializations; U200 in week 4



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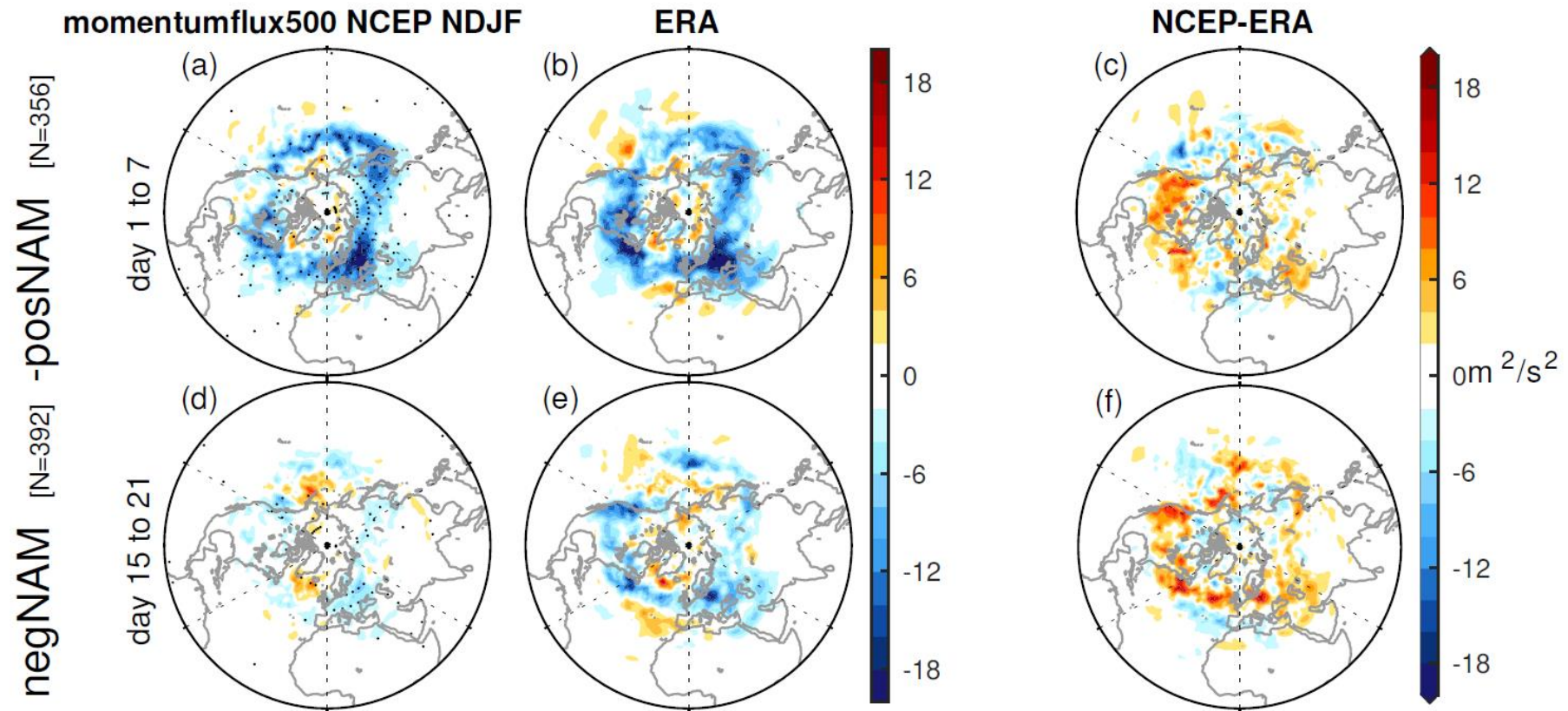


Is the problem the stratosphere?

Is the problem synoptic eddy feedback?



Eddy feedback too weak (transient eddy $u'v'$)



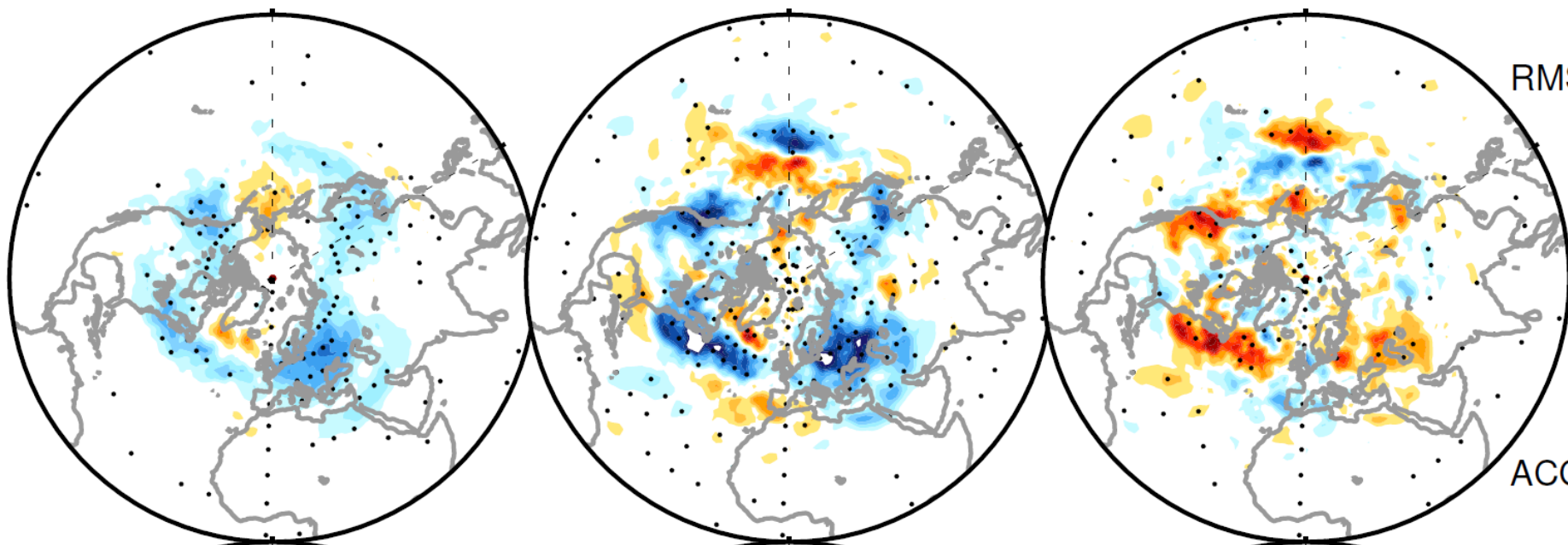
Eddy feedback too weak (transient eddy $u'v'$)

NDJF initializations, wk:3
S2S models

momentumflux500
ERA

S2S-ERA

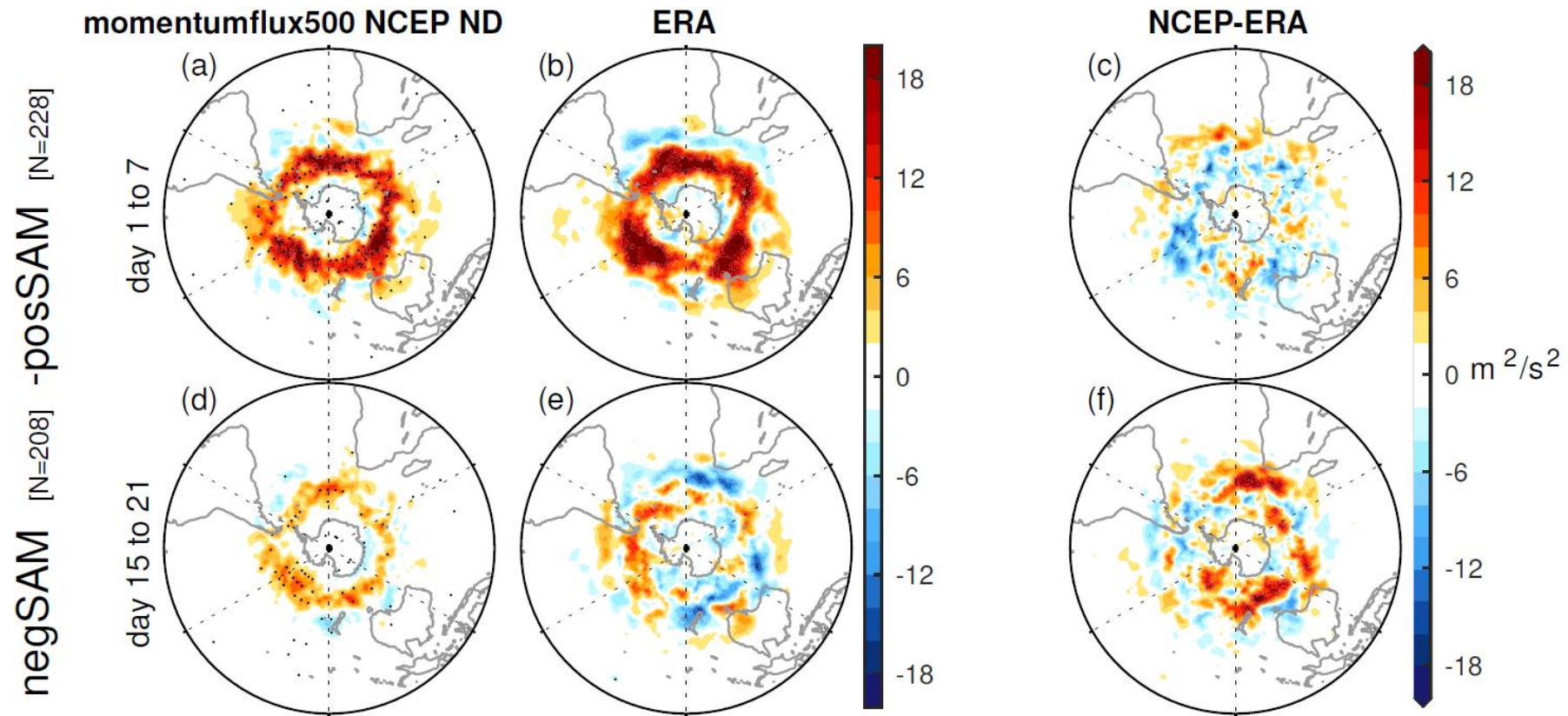
-NAM--+NAM



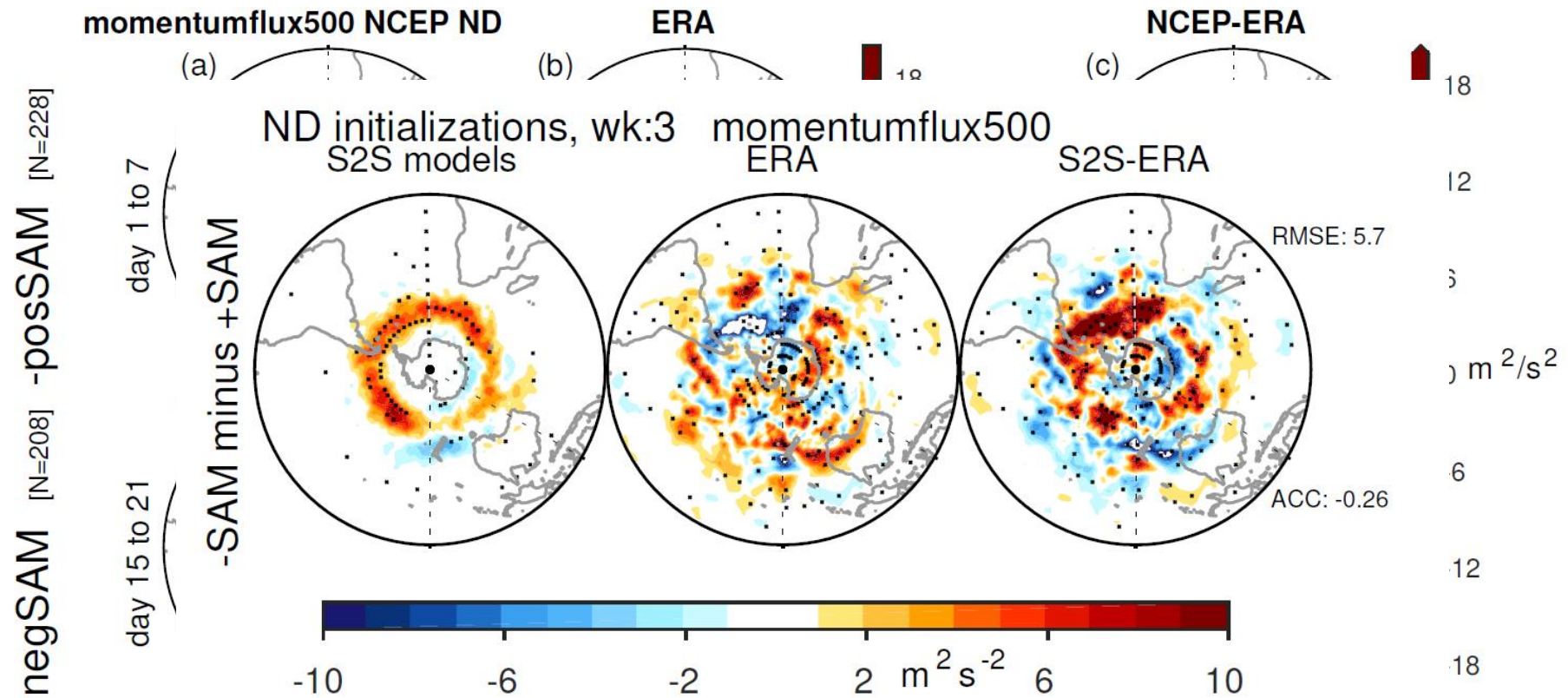
RMSE: 2.7

ACC: 0.68

In SH, eddy feedback too strong if anything



In SH, eddy feedback too strong if anything



Conclusions

S2S models develop RPC much greater than 1 by week 4

Paradox is particularly pronounced in the North Atlantic sector

The causes of this can be diagnosed by contrasting initializations during +NAM with initializations during -NAM

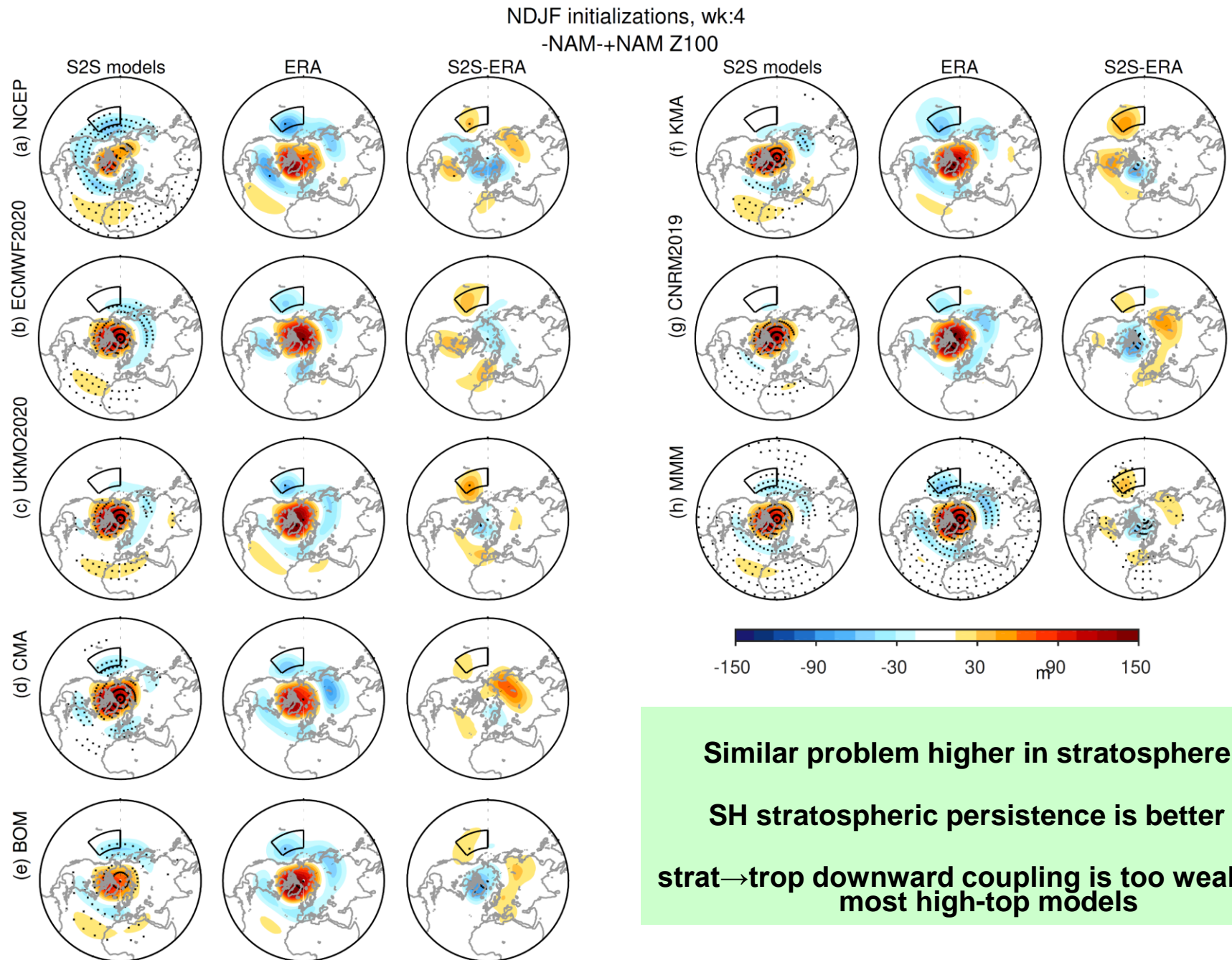
The NAM signal decays too quickly in all models, with the bias more pronounced for +NAM

Possible causes include the stratospheric signal decaying too quickly as well as overly weak NH eddy feedback (and these two causes may be linked)

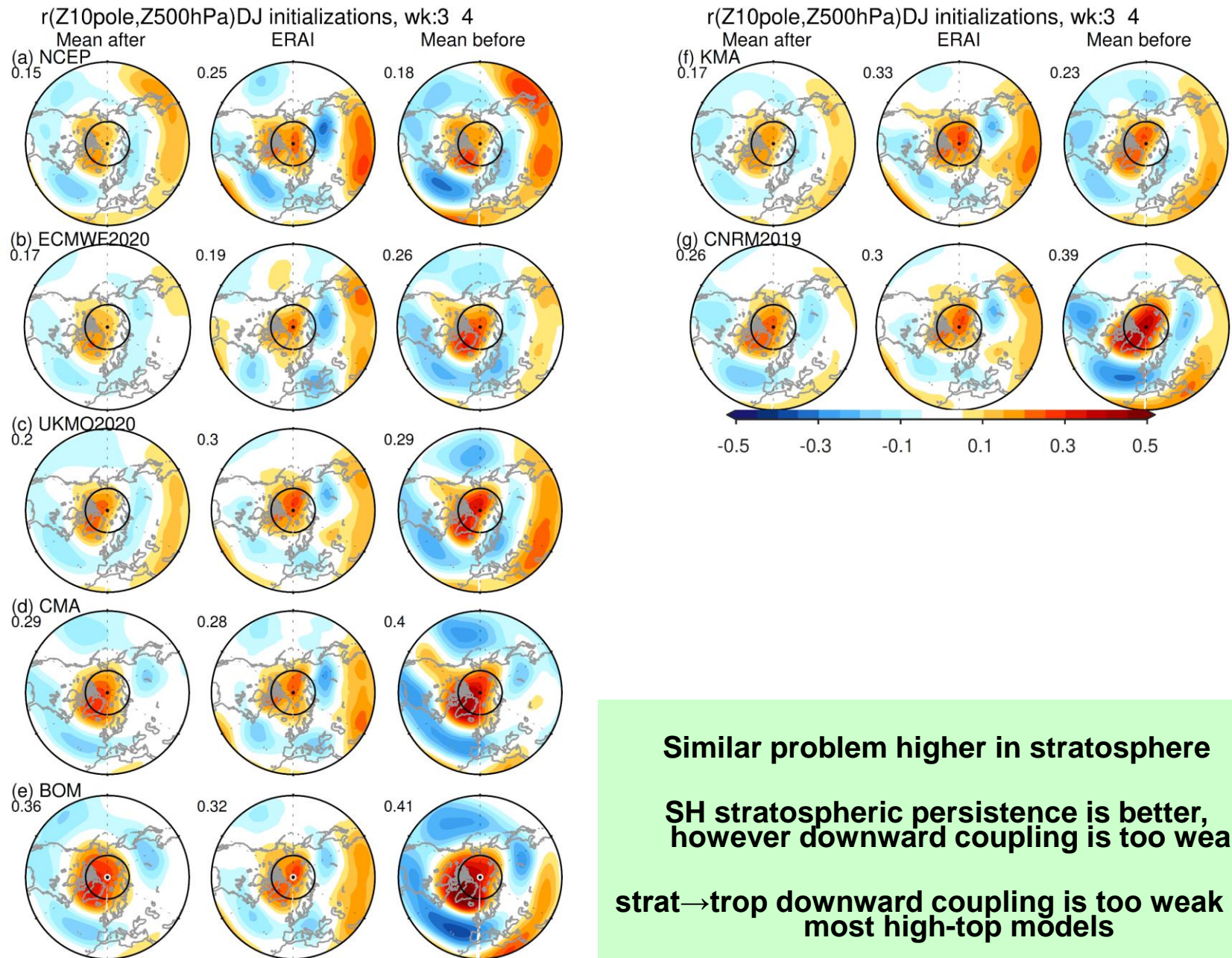
In contrast, the stratospheric signal in the SH persists realistically, and SH eddy feedback is too strong. This is consistent with a weaker signal to noise paradox in the SH.



The stratospheric signal decays too quickly



Biases in downward coupling

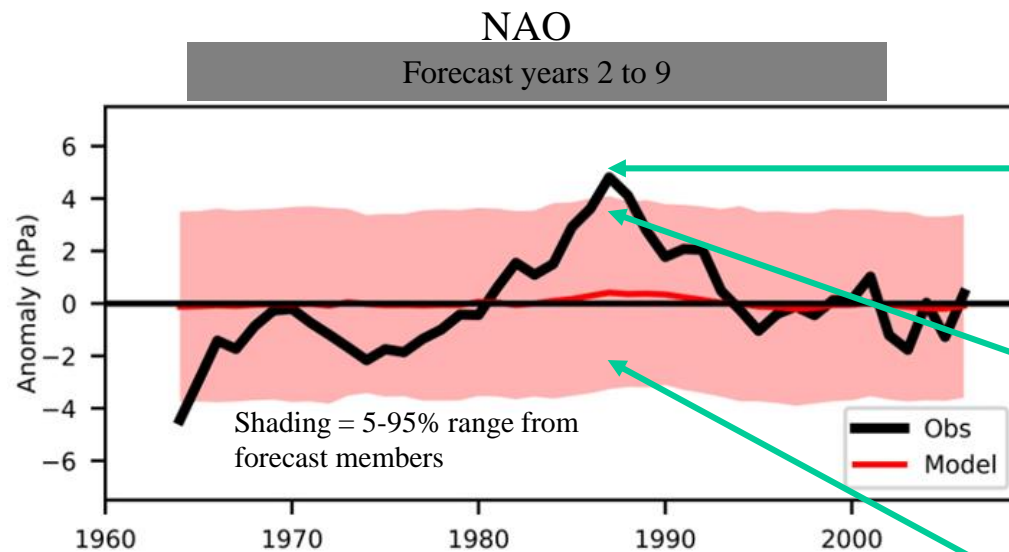


Similar problem higher in stratosphere

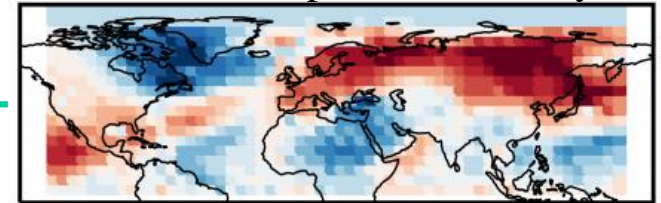
SH stratospheric persistence is better,
however downward coupling is too weak

strat→trop downward coupling is too weak in
most high-top models

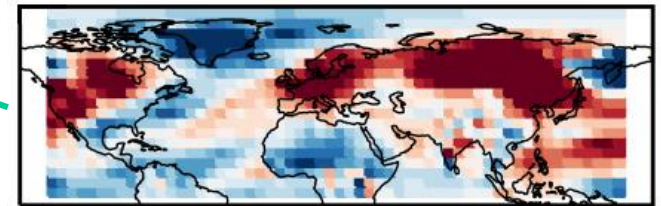
Irreducible internal variability?



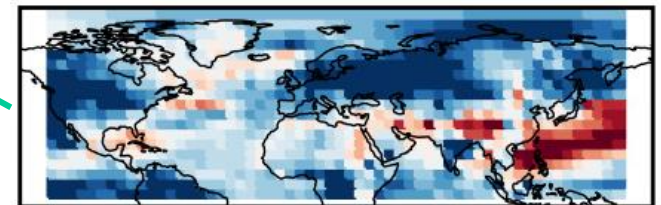
Observed temperature anomaly



Forecast member 3



Forecast member 670



CMIP5 + CMIP6 decadal predictions

Almost no signal in ensemble mean (red curve)

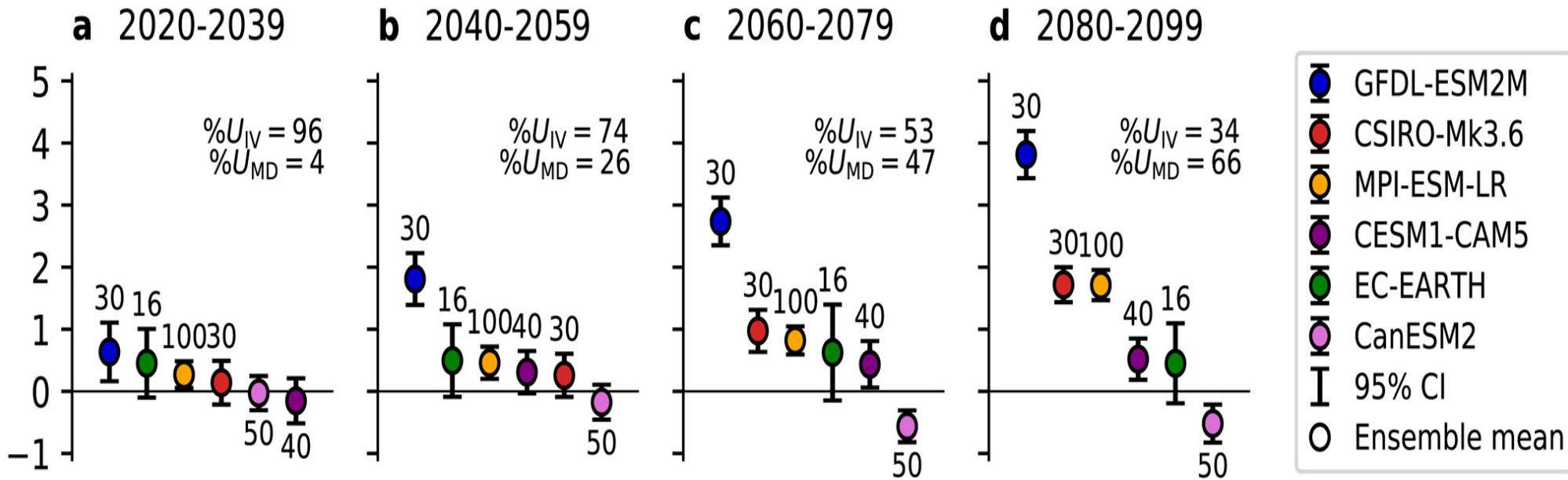
Irreducible internal variability if models taken at face value

BUT this can be tested...

Uncertainty in dynamical response to climate change

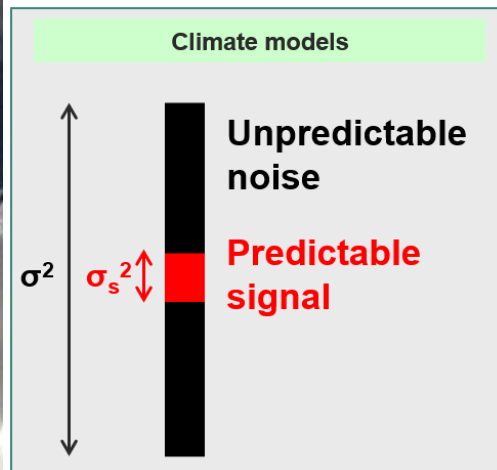
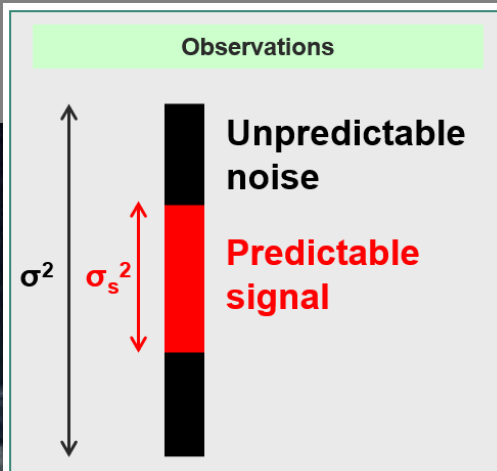
Detecting forced differences in the DJF NAO index

NAO index relative to 1995-2014 [hPa]



McKenna and Maycock 2021, GRL

Ratio of predictable components



Observations: Predictable Component $PC \geq r$ (anomaly correlation)

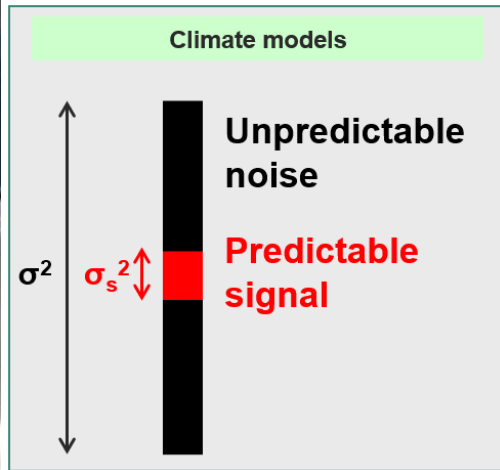
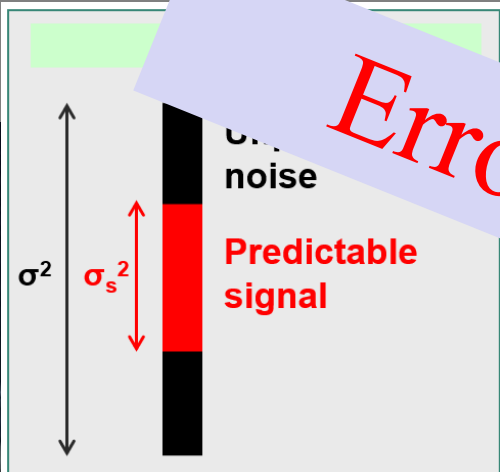
$$\text{Models: } PC = \frac{\sigma_{ensemble\ mean}}{\sigma_{ensemble\ members}}$$

$$\text{Ratio of predictable components (RPC)} \geq \frac{r}{\sigma_{ensemble\ mean} / \sigma_{ensemble\ members}}$$

RPC should be one

RPC > 1 shows the signal to noise error

Ratio of predictable components



Observations: Predictable Component $PC \geq r$ (anomaly correlation)

$$\text{Models: } PC = \frac{\sigma_{ensemble\ mean}}{\sigma_{ensemble\ members}}$$

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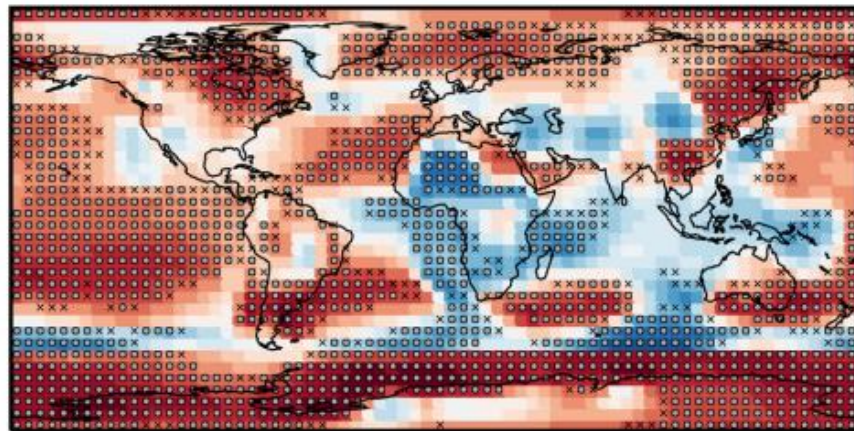
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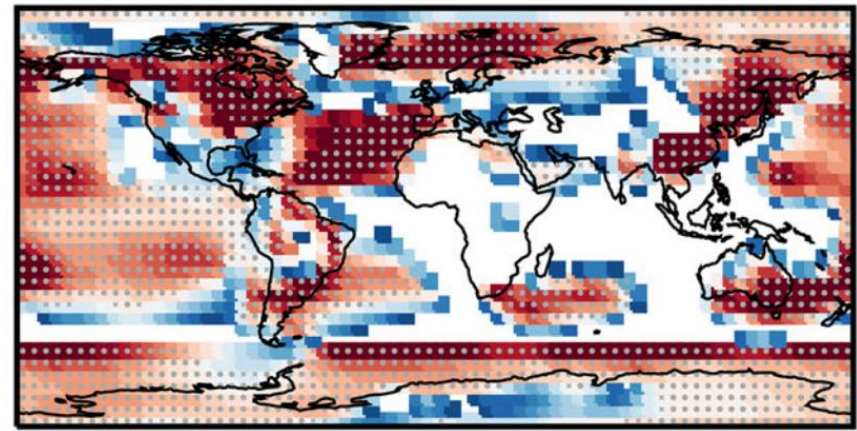
Error not seen in total variability

A key issue

MSLP skill (years 2-9)

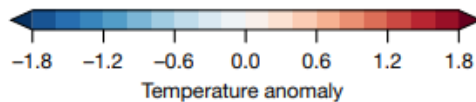
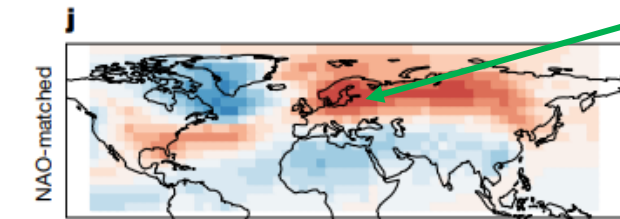
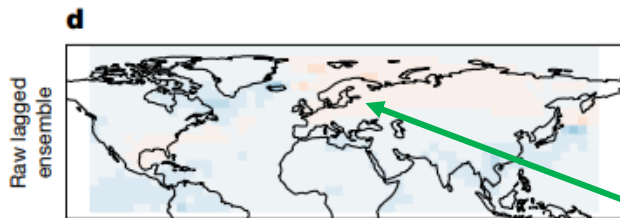
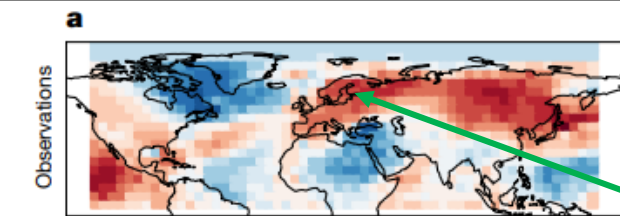
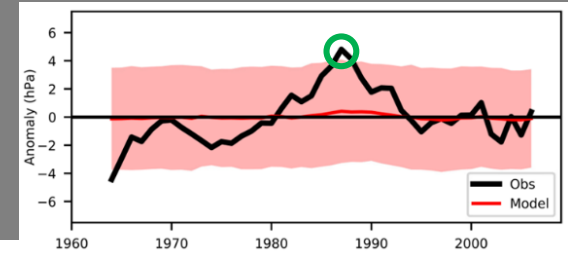


Error in magnitude of signal (RPC)



Wherever there is skill the modelled signals are too small!

NAO impacts not captured



$$T = T_{\text{DYN}} + T_{\text{THERMO}} + \epsilon$$

Real world: $T_{\text{DYN}} \gg T_{\text{THERMO}}$

Ensemble mean: $T_{\text{DYN}} \ll T_{\text{THERMO}}$ because NAO signal too small

Scaling retains the incorrect ratio $T_{\text{DYN}}/T_{\text{THERMO}}$

Can be overcome by selecting ensemble members with correct magnitude of NAO

Standard D&A approach will not work
Need to look at models in new ways