

Challenges in Defining the Internal Component of Atlantic Multidecadal Variability in a Changing Climate

Clara Deser

Adam S. Phillips

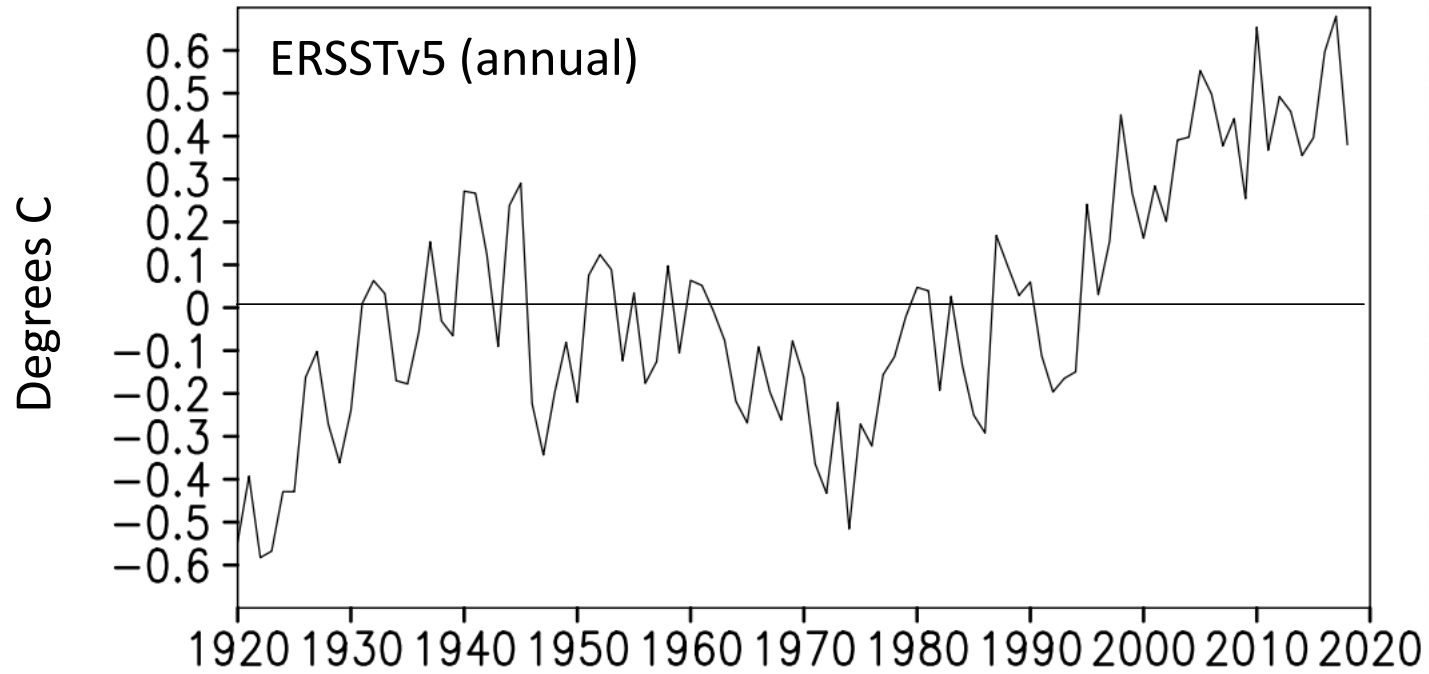
Climate Analysis Section, CGD, NCAR

CESM Climate Variability and Change Working Group Meeting
Feb. 17, 2022 (virtual)

Canonical Index of AMV

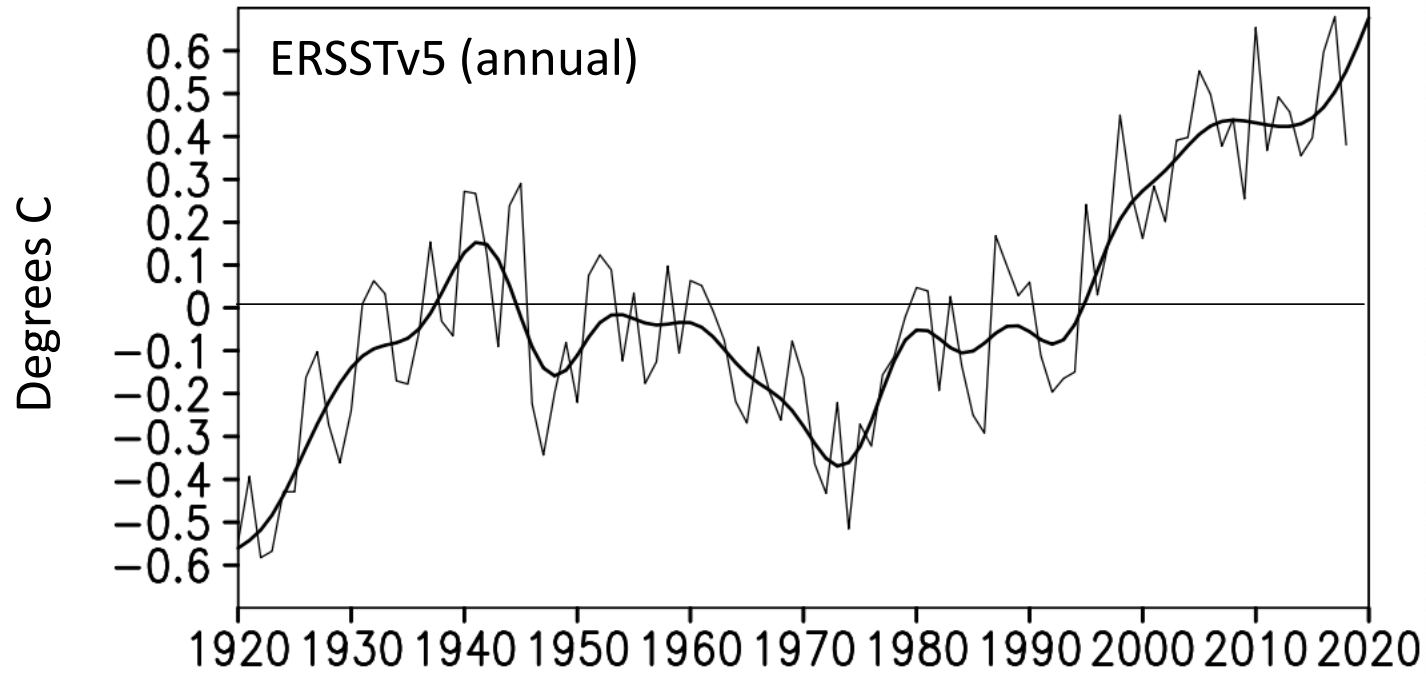
North Atlantic (0-60N) SSTA(t)

(Enfield et al. 2001)



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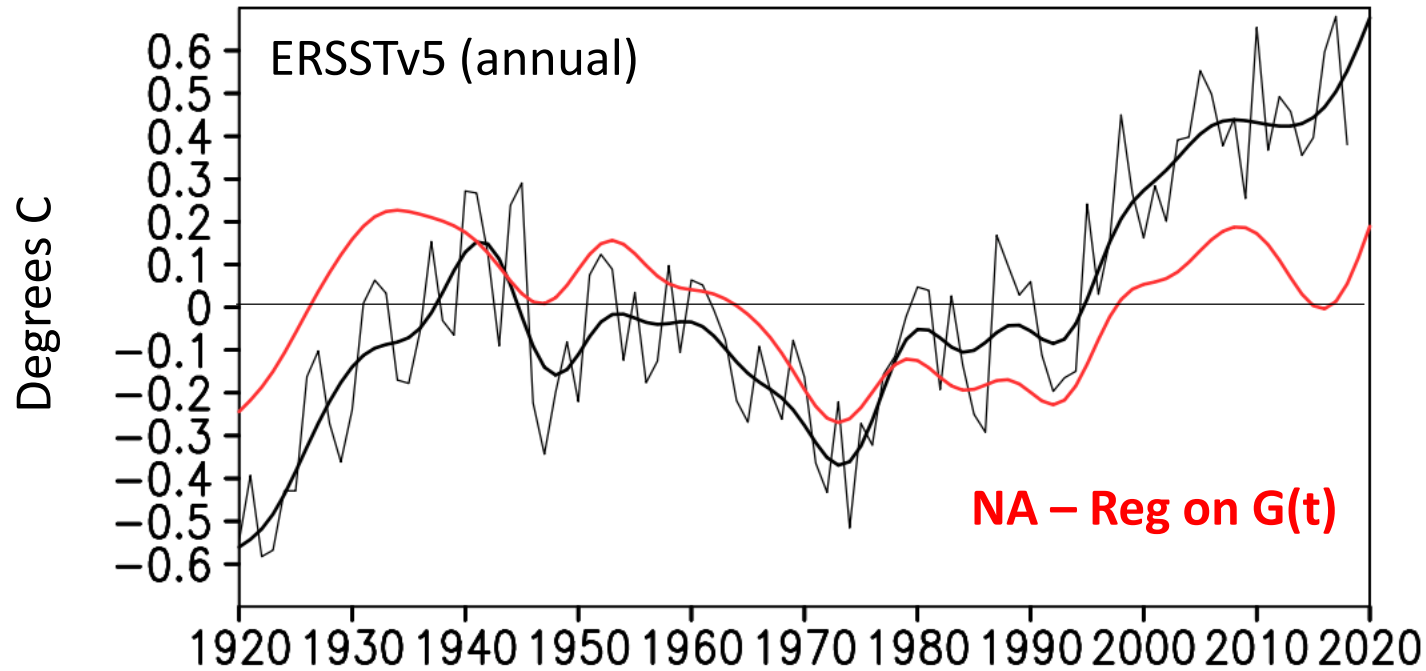
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10yr Butterworth Filter

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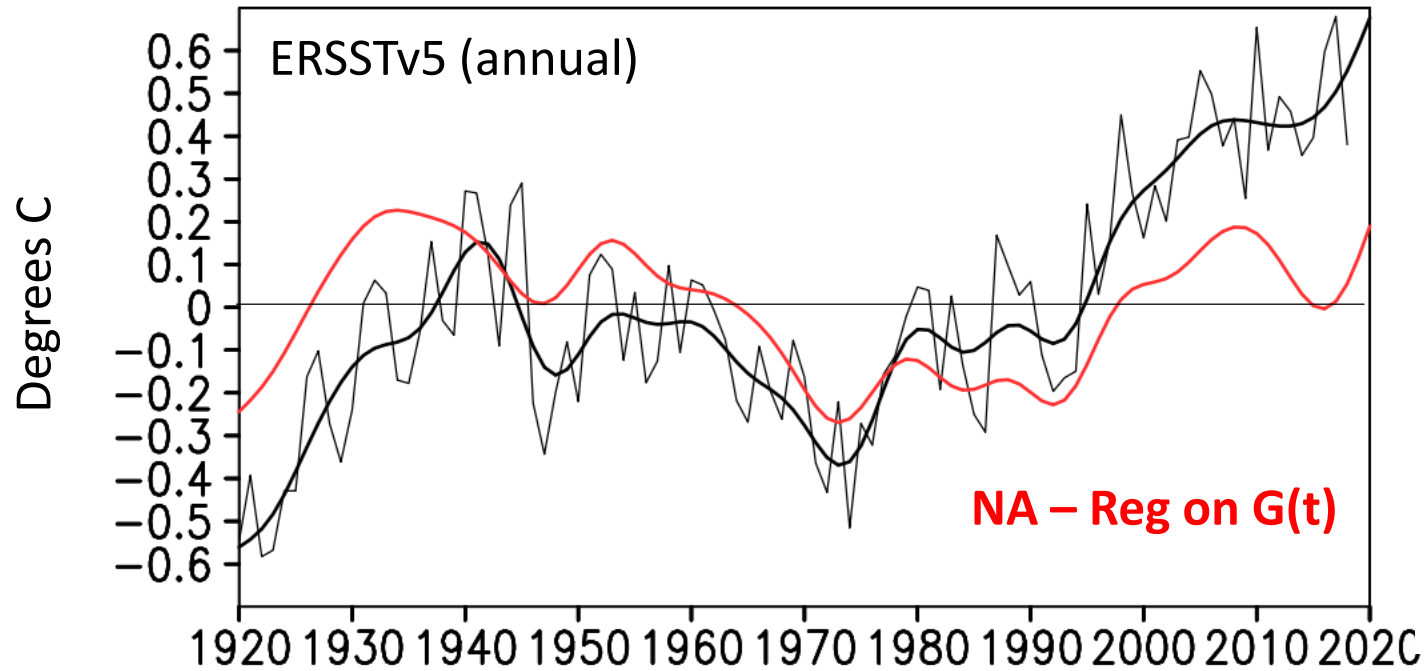
**After removing regression
on Global Mean SST
(Zhang et al. 2019)**



Internal Component

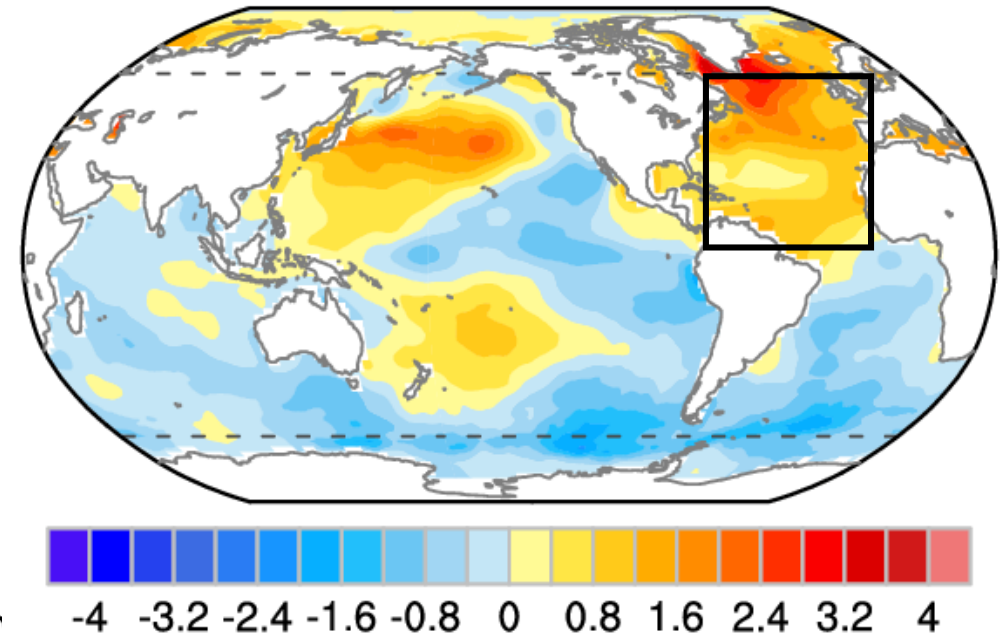
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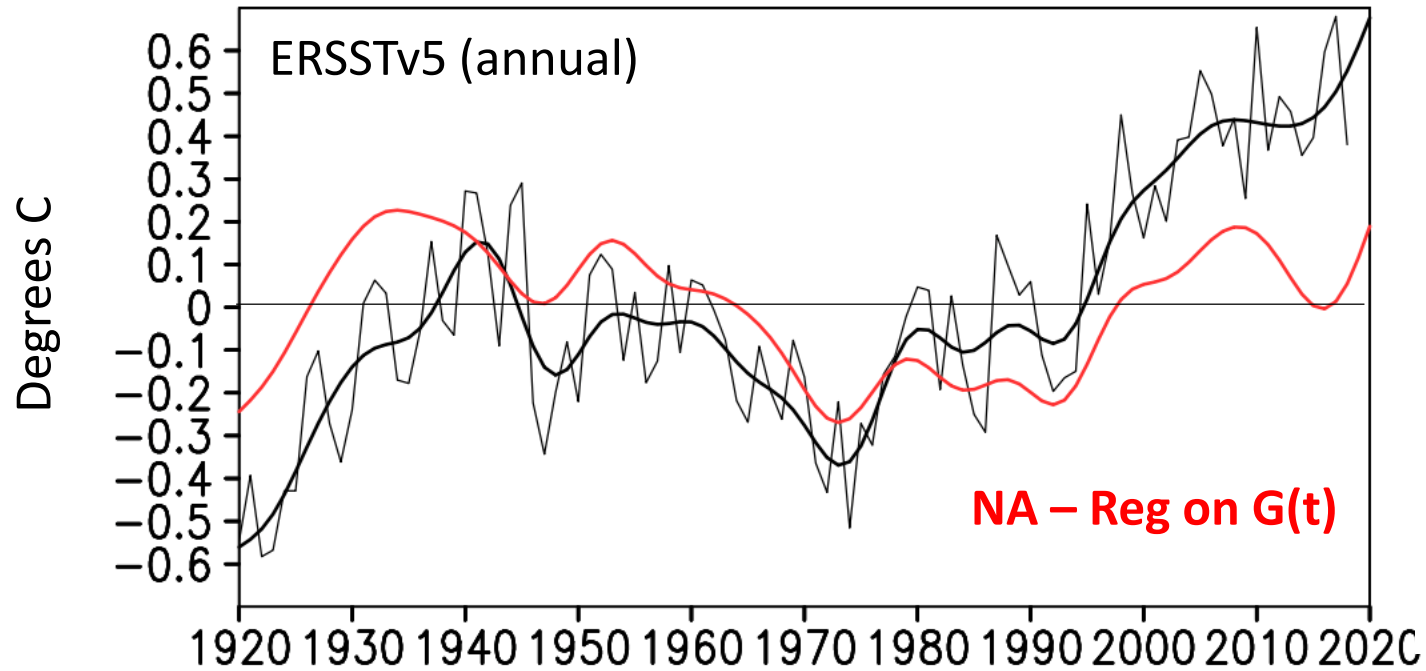
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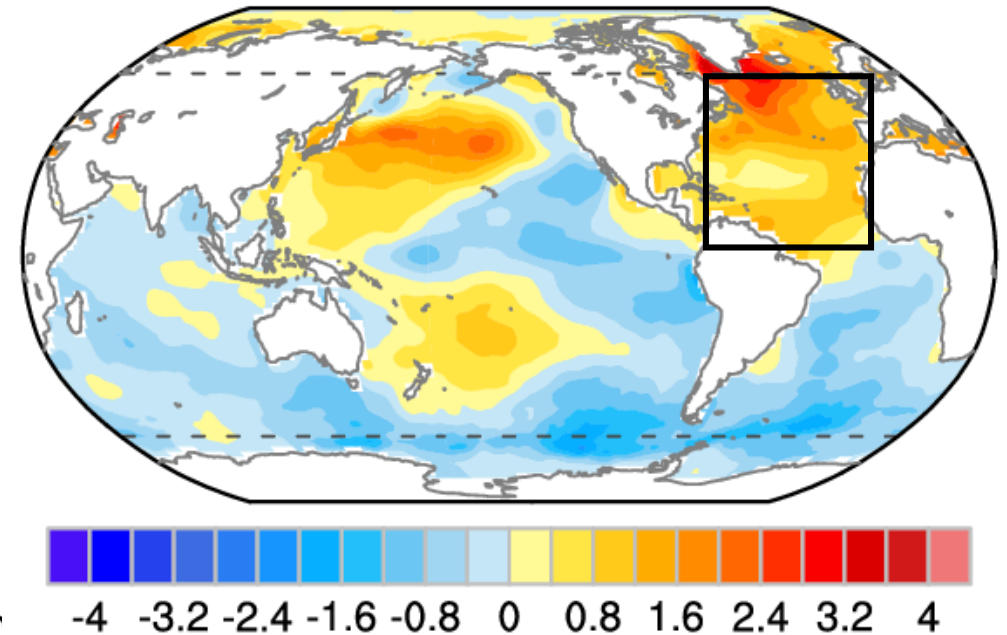
Deser and Phillips (GRL, 2021)

**Does this procedure really isolate
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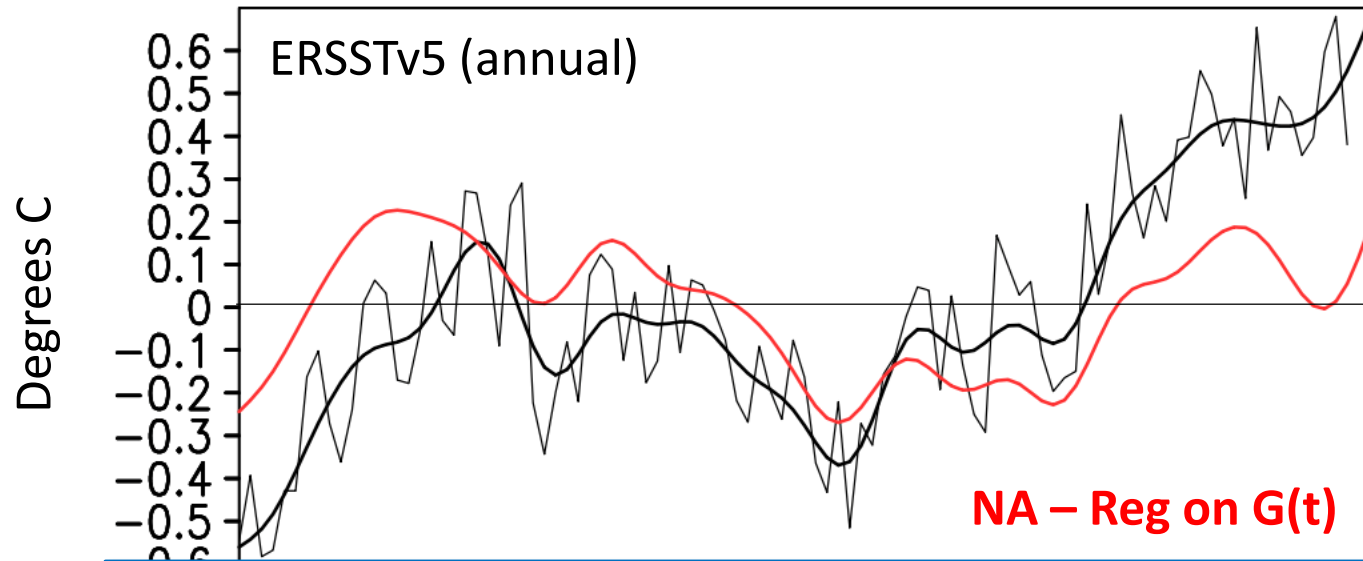
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Internal Component



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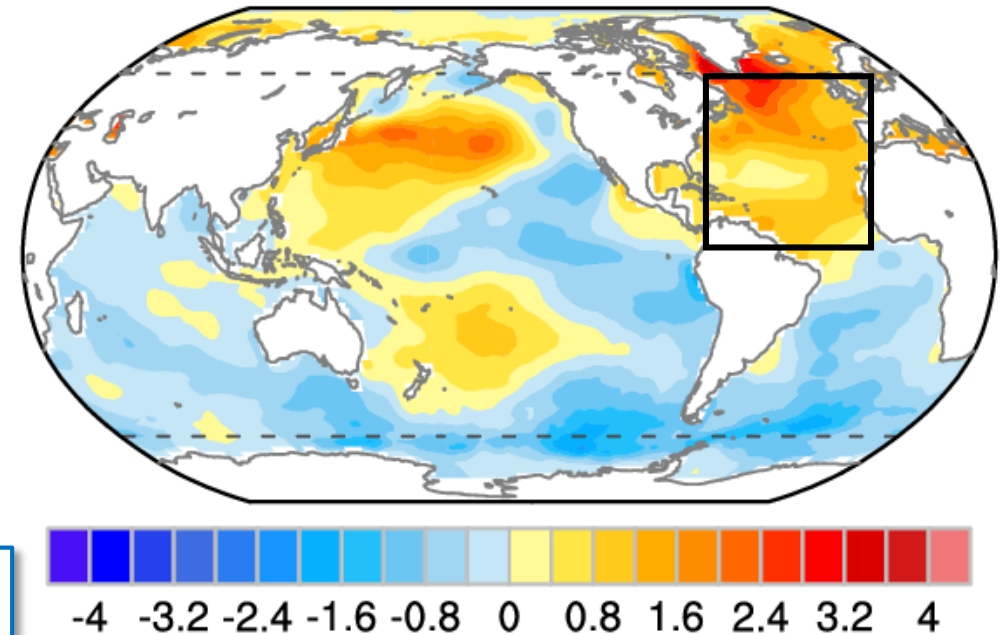
**Does this procedure really isolate
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**Test with model Large Ensembles,
where the true internal variability
can be well estimated.**

**After removing regression
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Internal Component



Deser and Phillips (GRL, 2021)

Roadmap

Aim is to isolate the internal component of AMV.

- Step 1) Use the Residual Method to isolate internal variability (iSSTA).
- Step 2) Compute AMV index and associated spatial pattern from iSSTA.
- Step 3) Test method accuracy using model Large Ensembles where the true internal variability is known *a priori*.

Roadmap

Aim is to isolate the internal component of AMV.

Step 1) Use the Residual Method to isolate internal variability (iSSTA).

Residual Method

$$\text{iSSTA}(x,y,t) = \text{SSTA}(x,y,t) - [\text{Reg of SST}(x,y,t) \text{ on } G(t)]$$

Global mean SSTA

Estimated forced component

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**Note that $G(t)$ contains both
forced and internal components**
 $G(t) = fG(t) + iG(t)$

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7 Model Large Ensembles (430 Members Total)

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Procedure for each model

1a) **True** Internal Variability

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e = ensemble member

EM = ensemble mean

1b) **Estimated** Internal Variability

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2) Regress iSSTA(x,y,t,e) onto iAMV_index(t,e).

3) Average the regression maps across all members.

4) Average the 7 models' average regression maps.

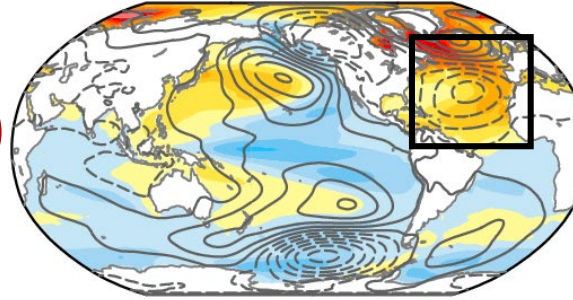
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1950-2020

Estimated
SST & SLP

iNA (-iG)



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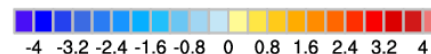
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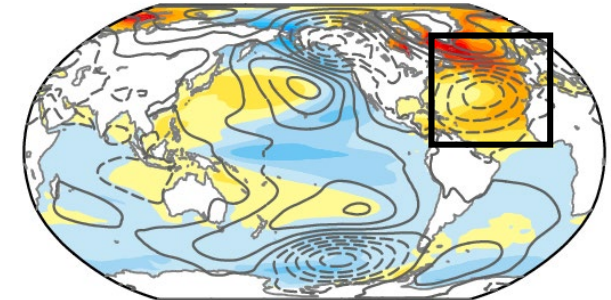
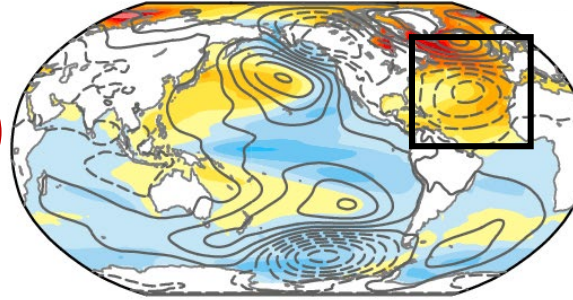
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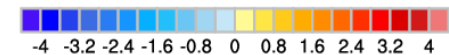
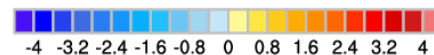
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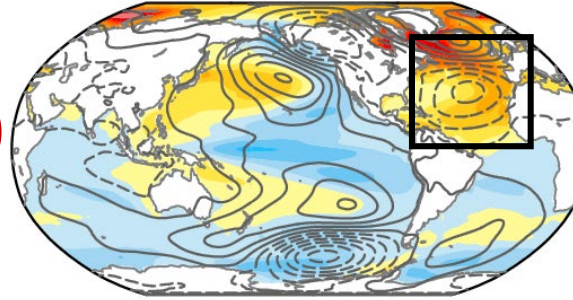
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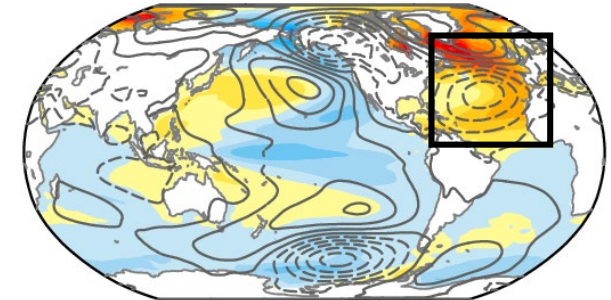
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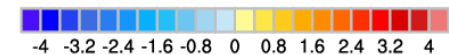
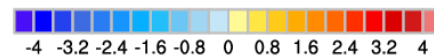
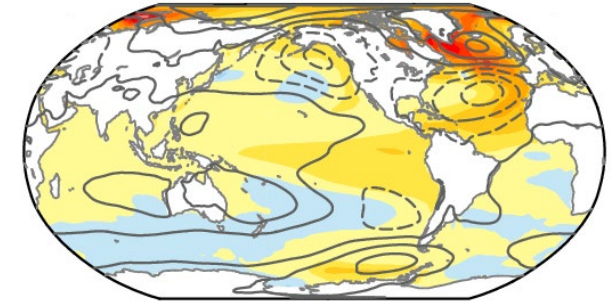


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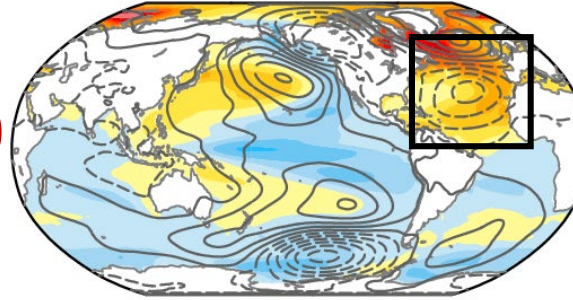
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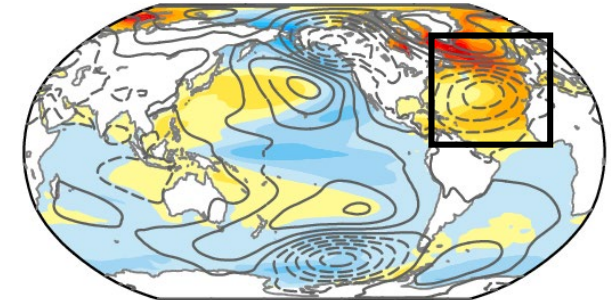
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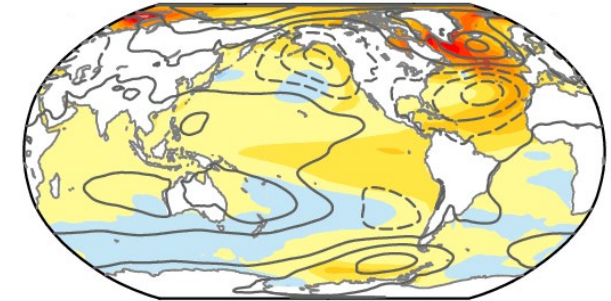


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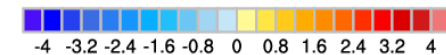
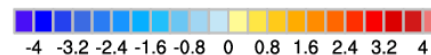
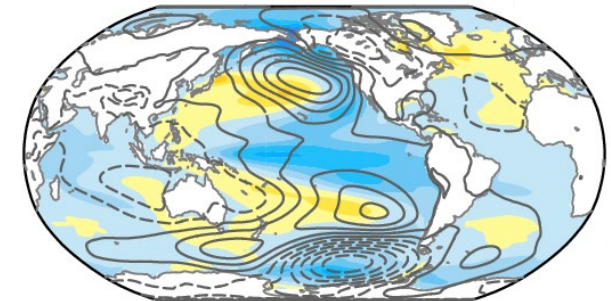
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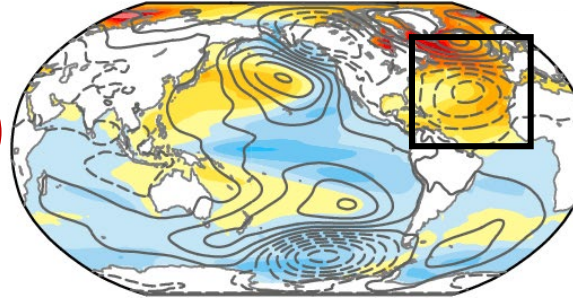
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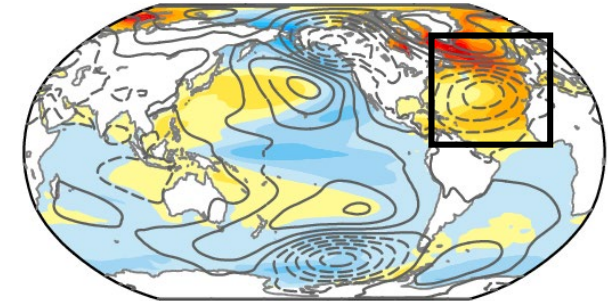
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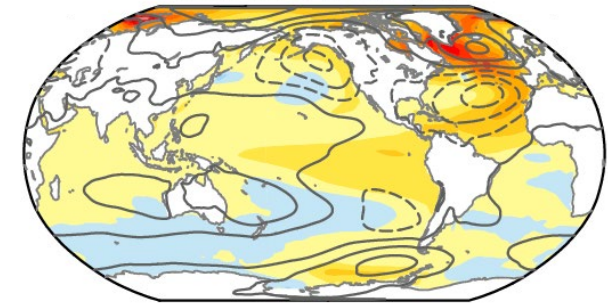


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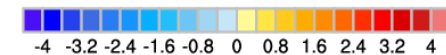
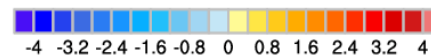
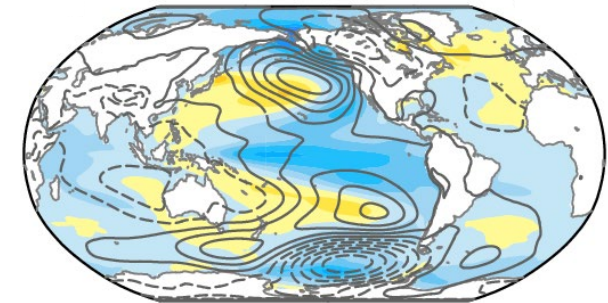
iNA



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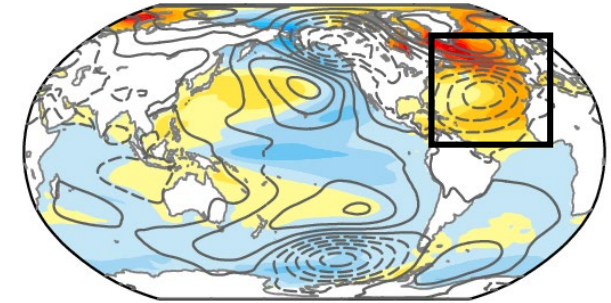
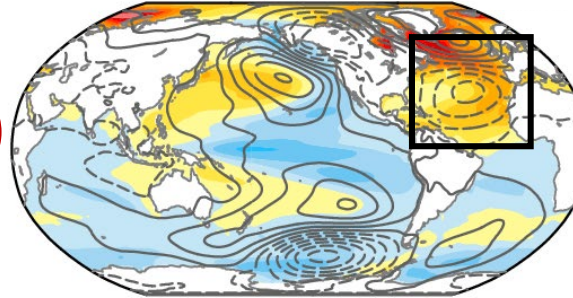
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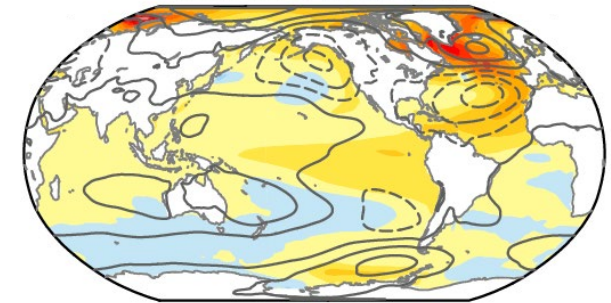
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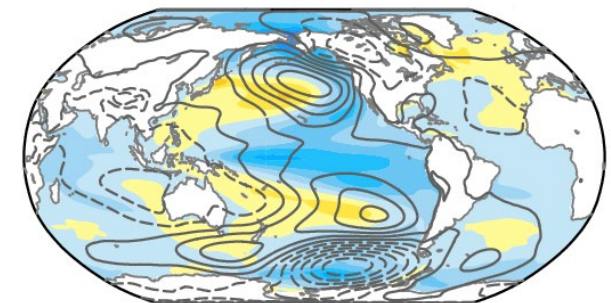
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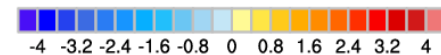
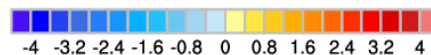
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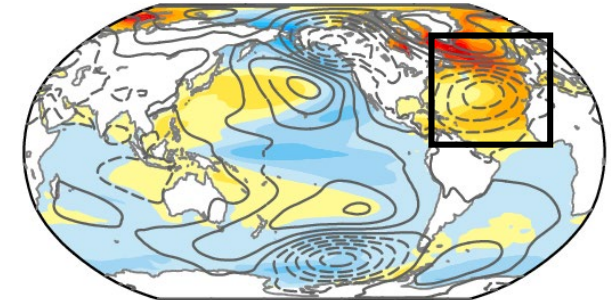
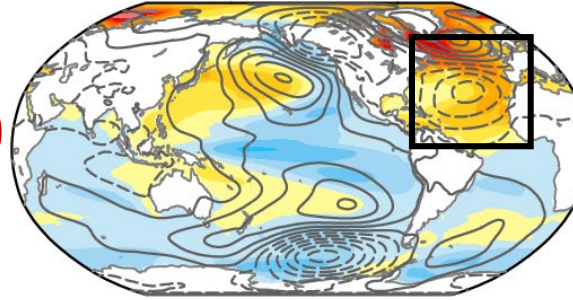
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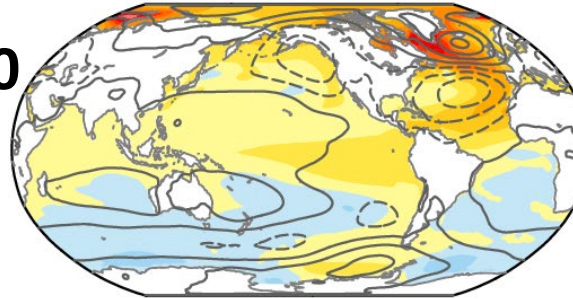
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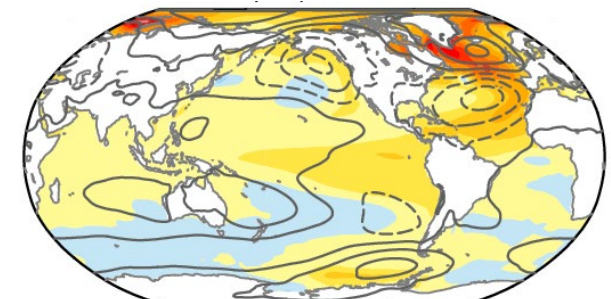
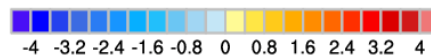
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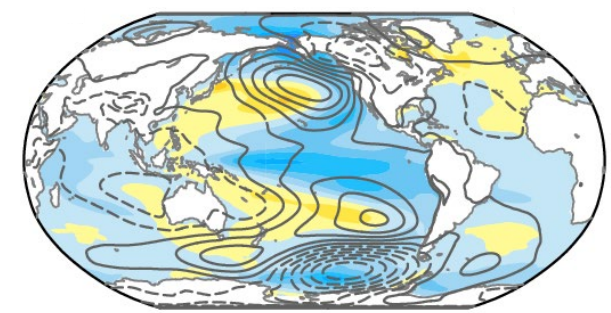
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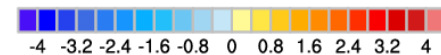
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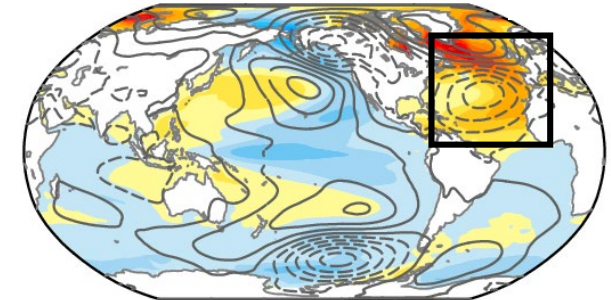
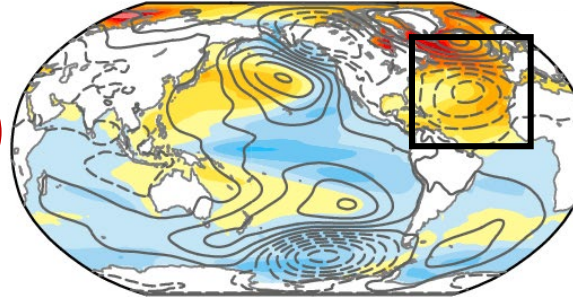
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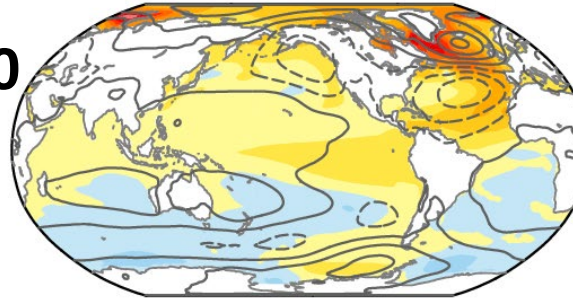
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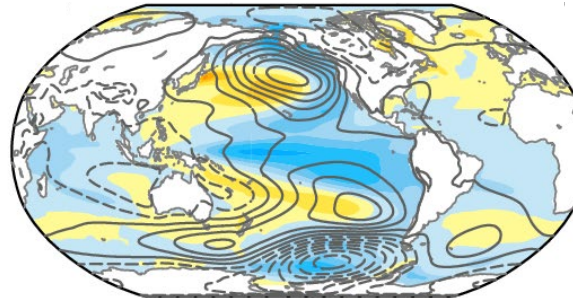
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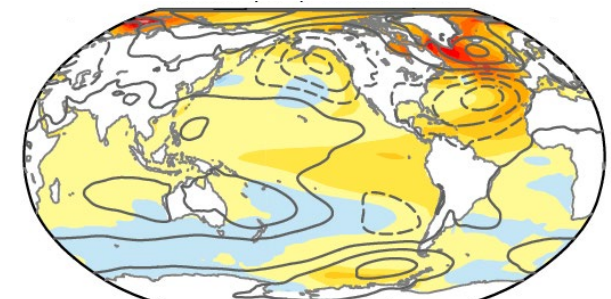
iNA-iG



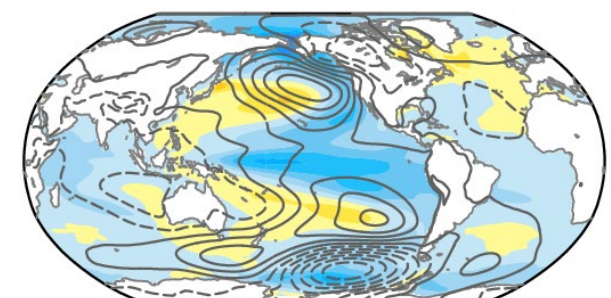
Diff



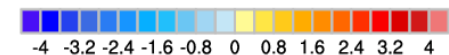
$$[iNA(t) - iG_{NA}(t)] - iG^*(t)$$



iNA

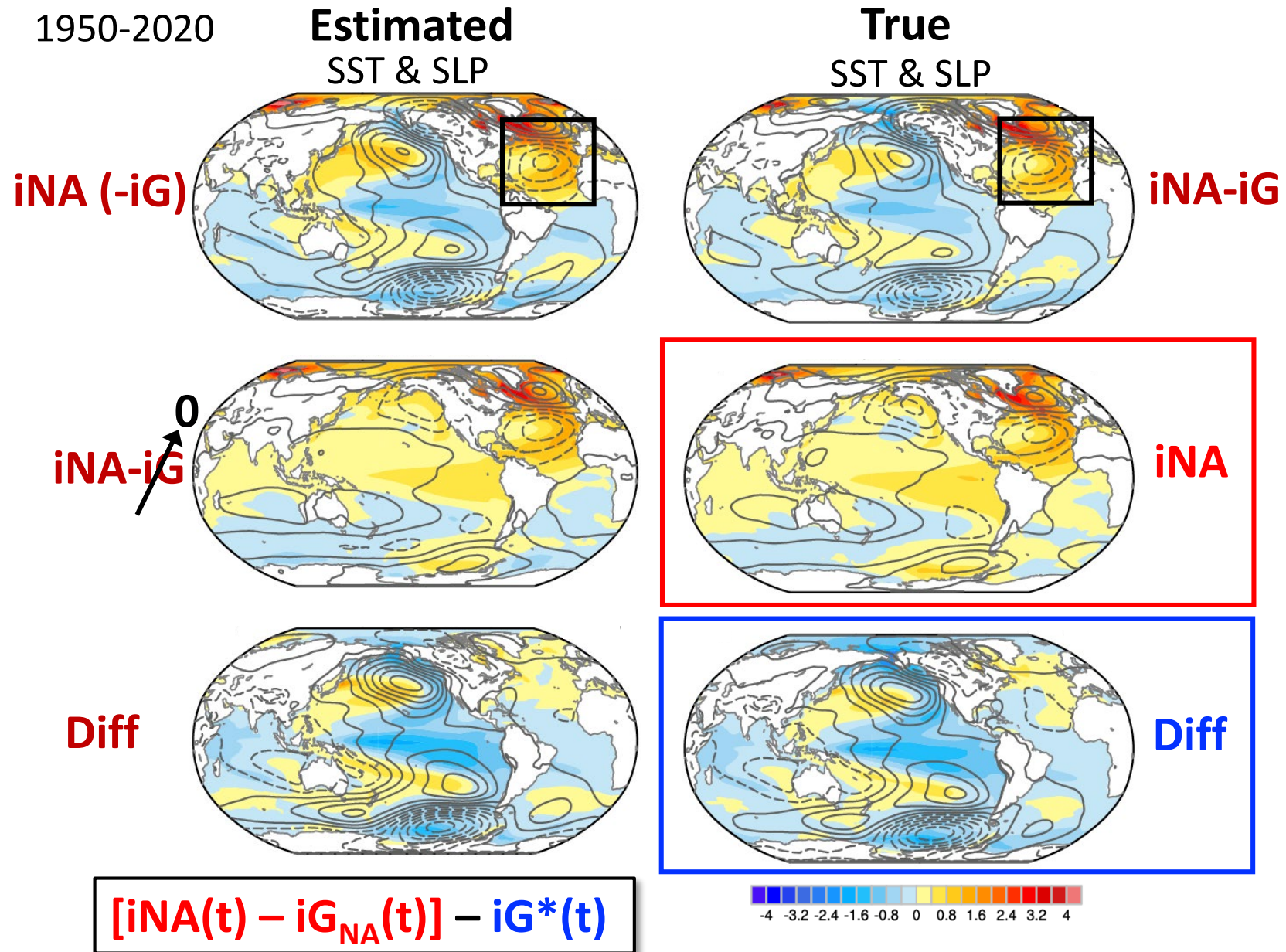


Diff



CONCLUSIONS

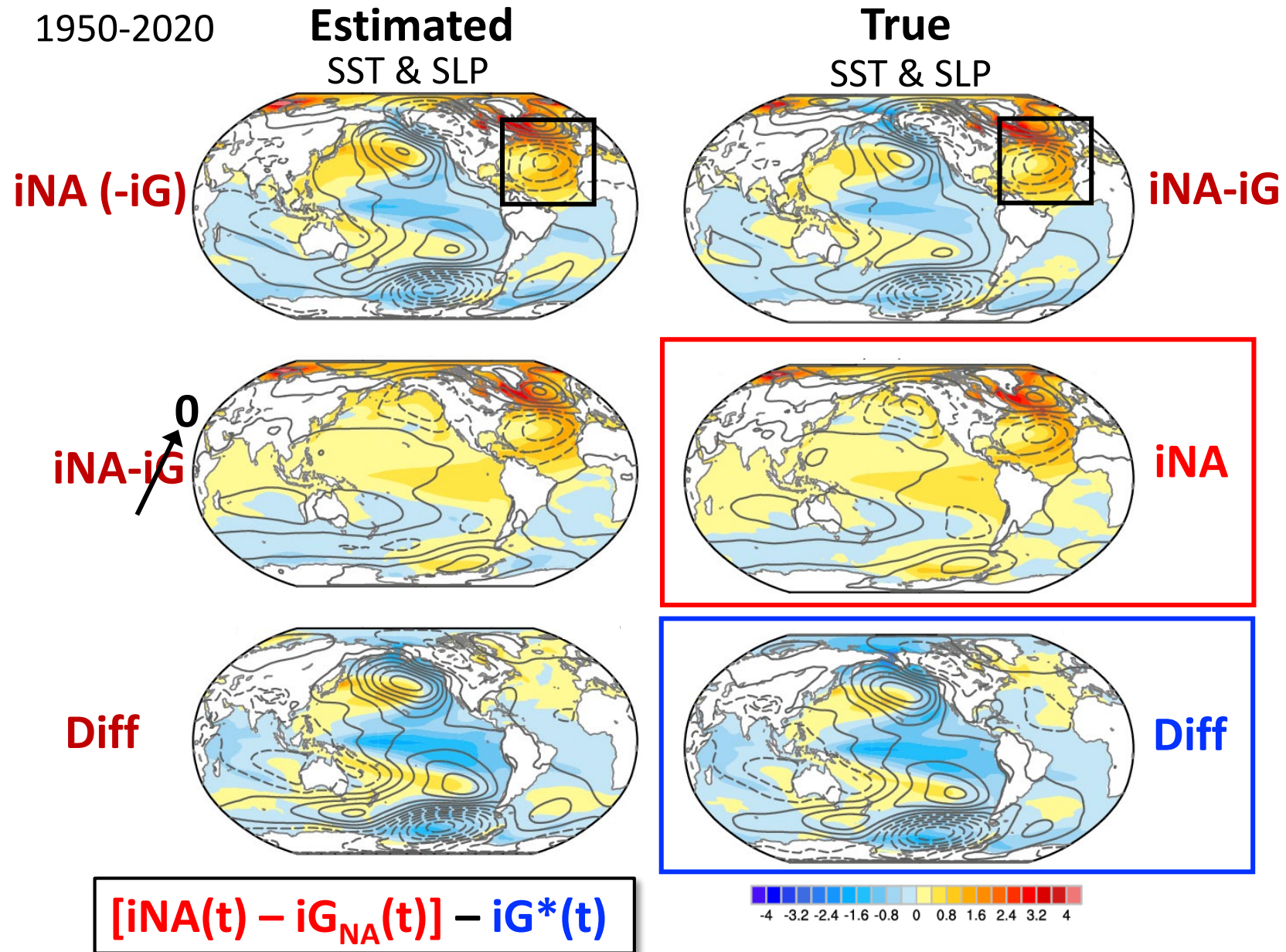
7 Model Large Ensembles (430 Members Total)



CONCLUSIONS

1) Estimating the forced component with the global-mean T regression method introduces a spurious connection between the Atlantic and Pacific in the definition of internal AMV in models.

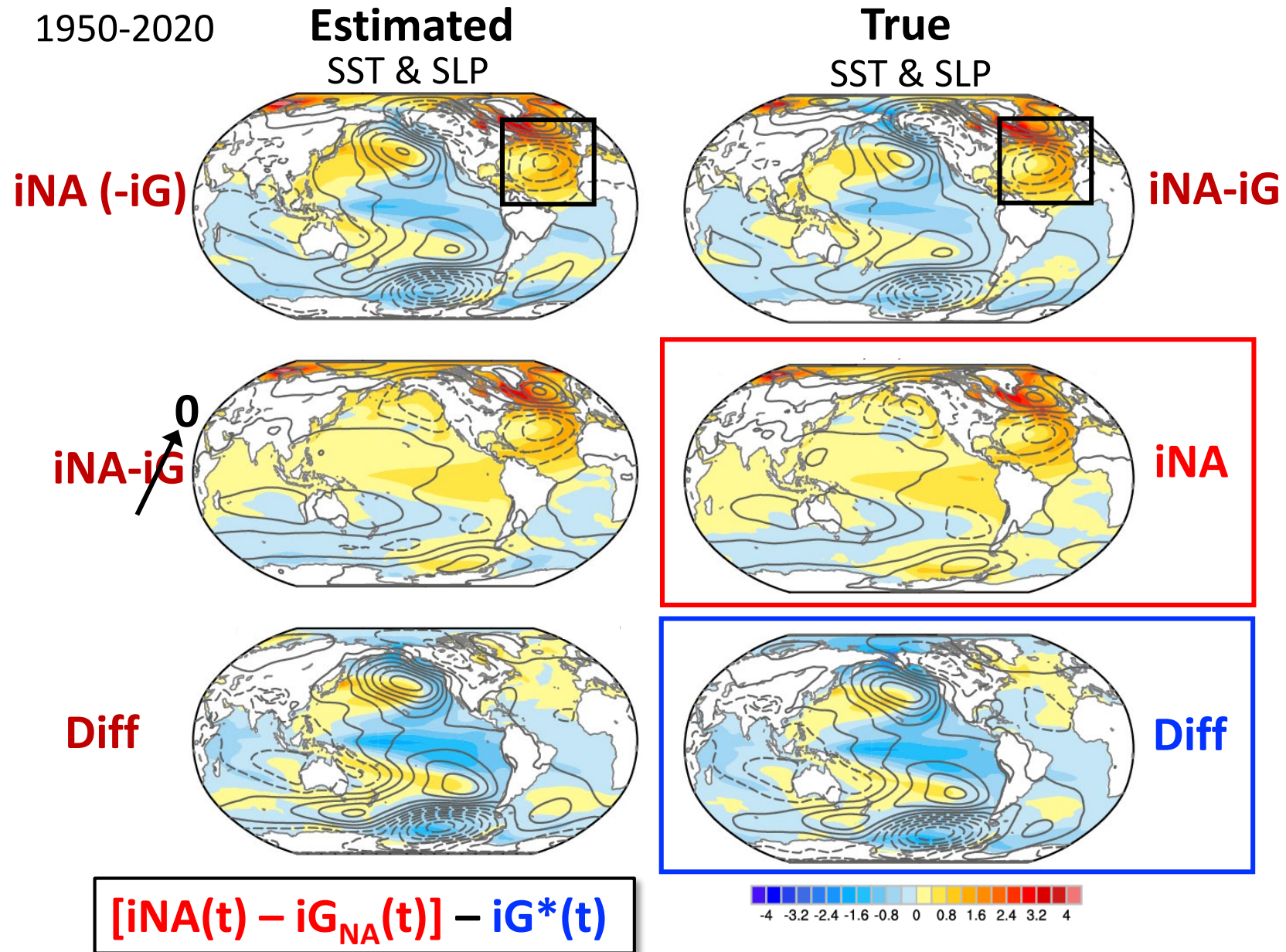
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CONCLUSIONS

- 1) Estimating the forced component with the global-mean T regression method introduces a spurious connection between the Atlantic and Pacific in the definition of internal AMV in models.
- 2) Redefining the method to use the ensemble-mean global-mean T alleviates the problem.

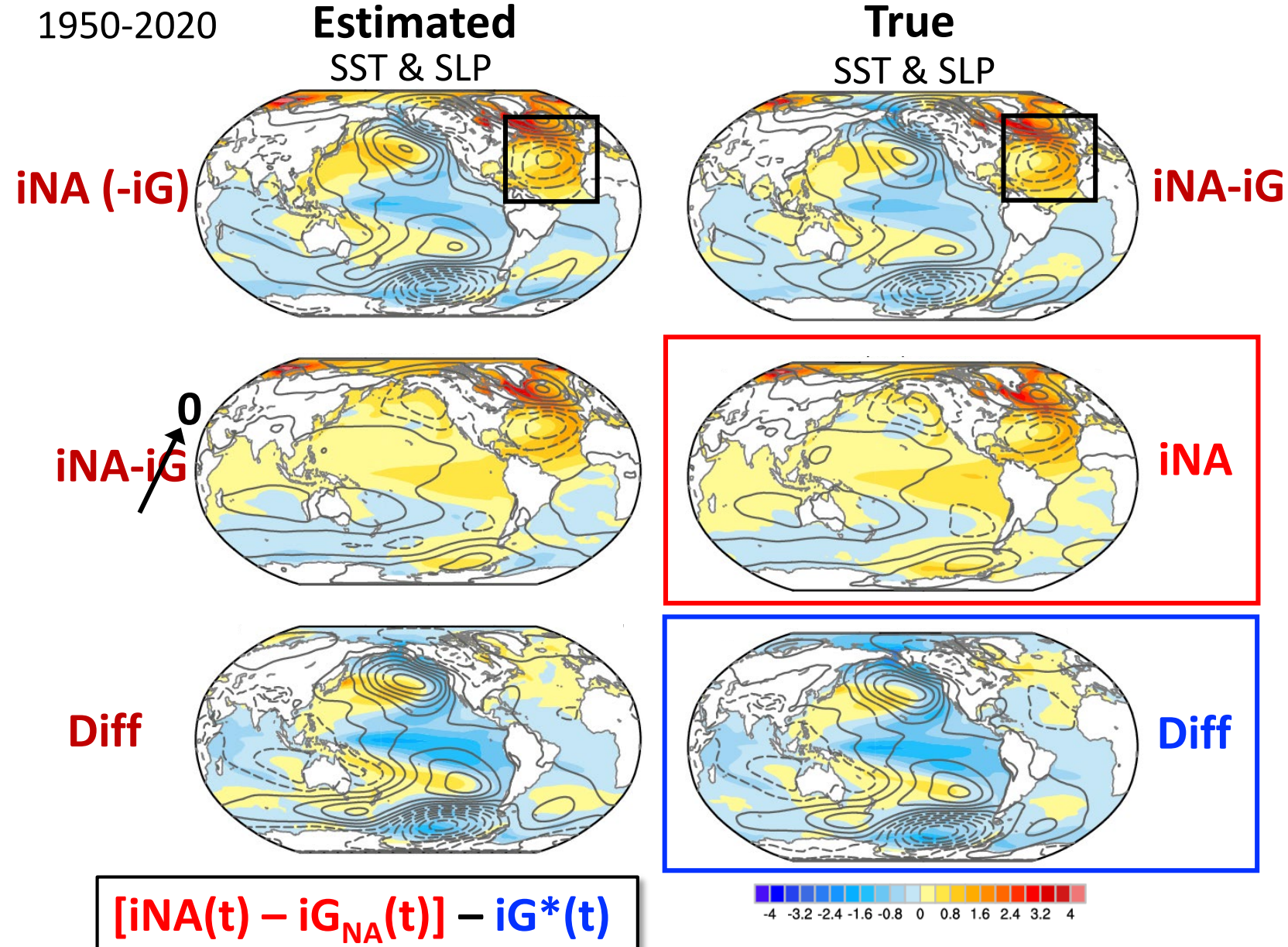
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- 3) Results are robust across models and time periods (not shown).

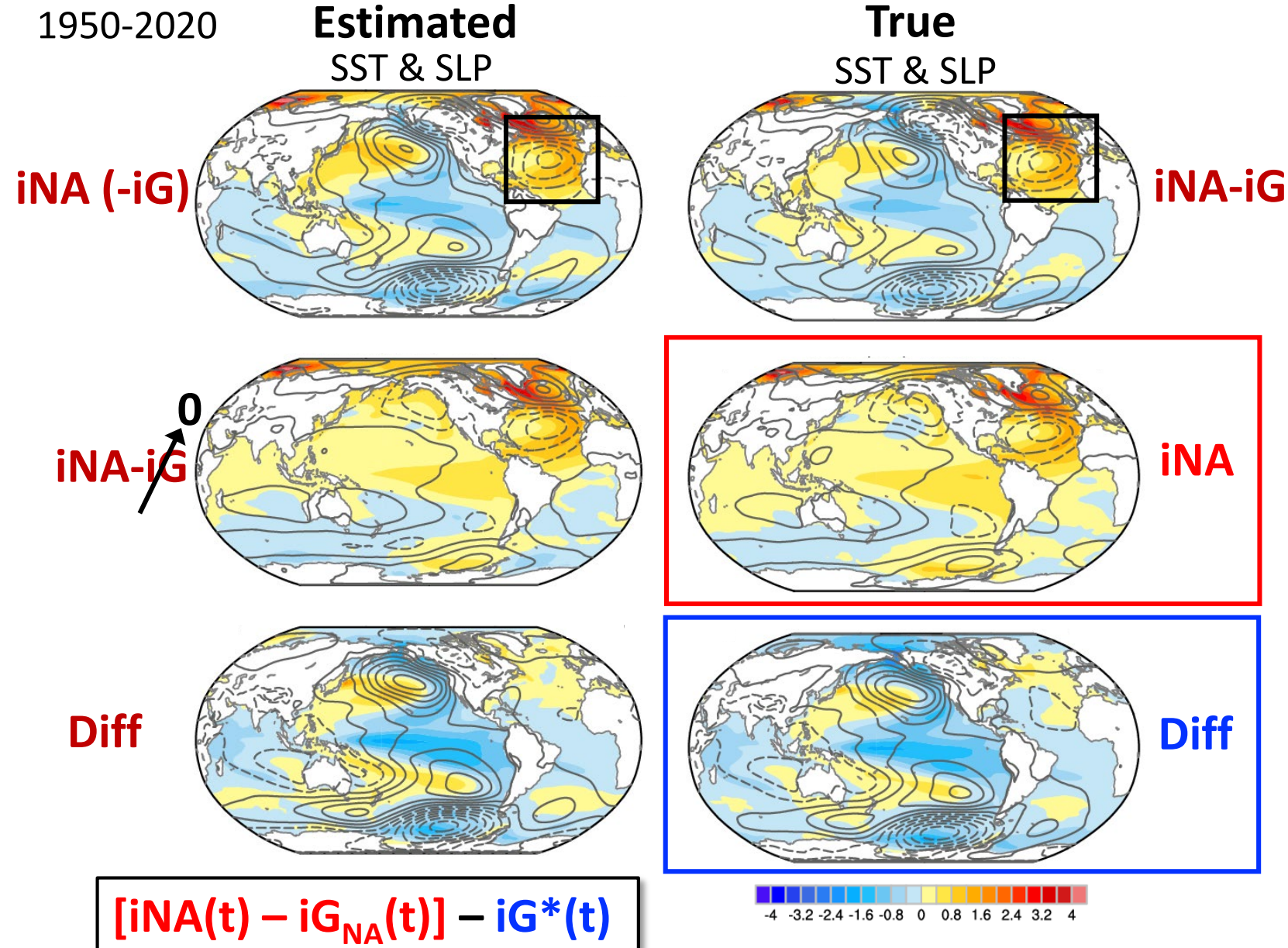
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- 4) iG^* related to IPV/PDV

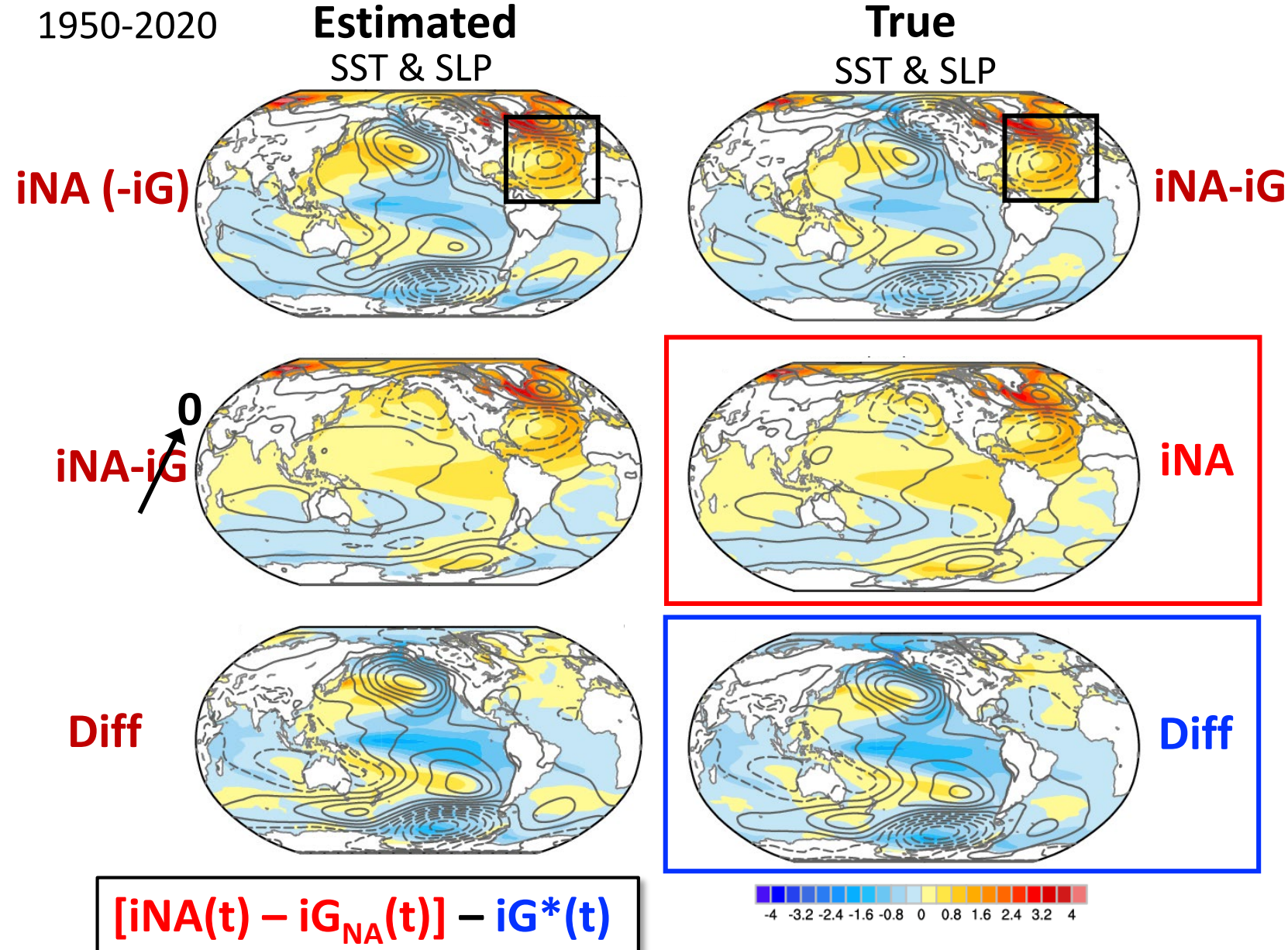
7 Model Large Ensembles (430 Members Total)



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- 5) iNA related to AMOC?

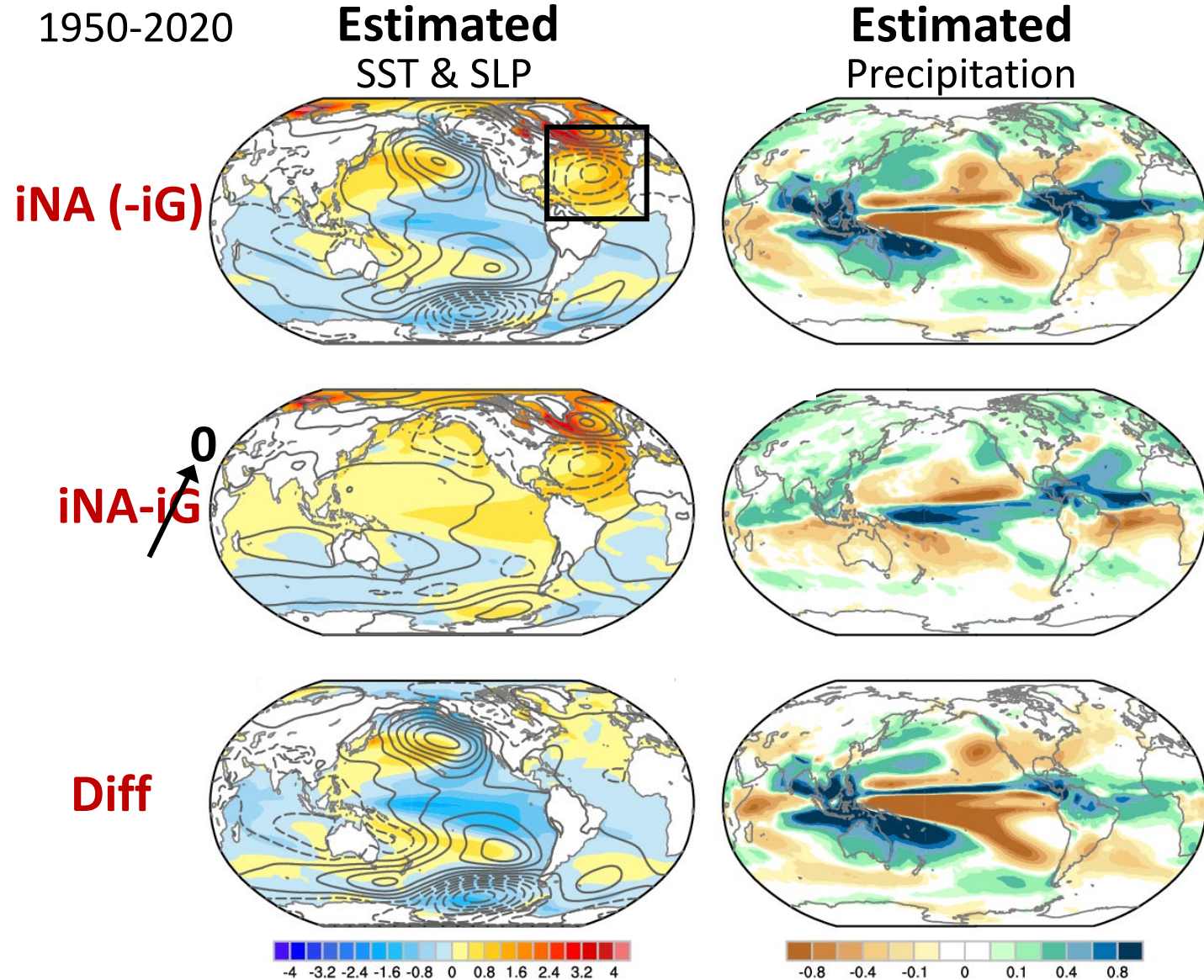
7 Model Large Ensembles (430 Members Total)



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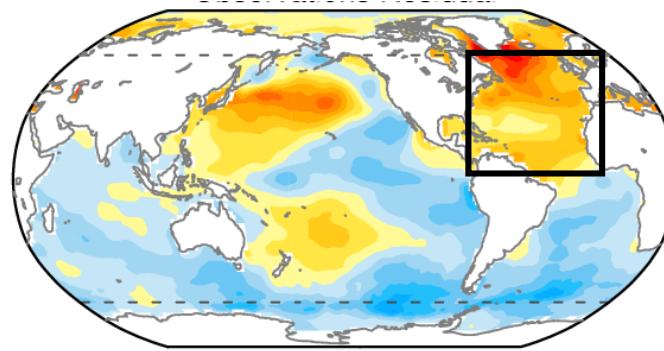
EXTRA SLIDES

Observations

(ERSSTv5, 1950-2020, 10yrLP)

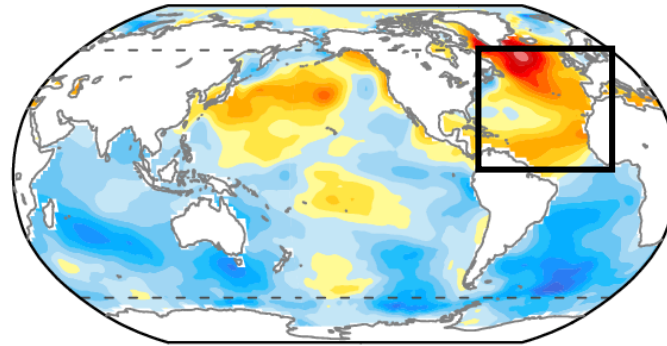
Residual Method
using Obs $G(t)$

iNA (-iG)

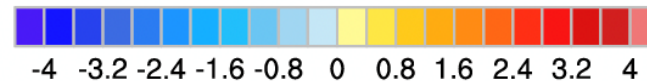
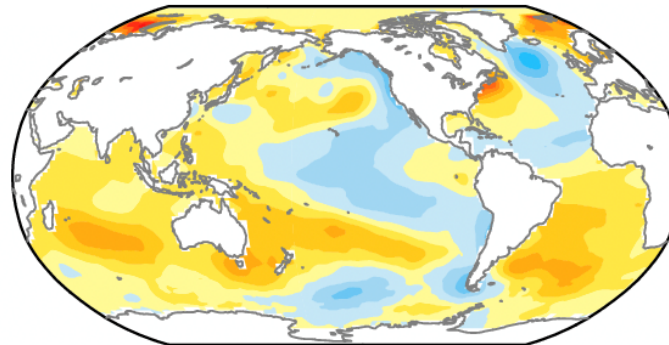


Residual Method
using ensemble-mean
 $G(t)$ averaged across
the 7 model LEs

iNA - $\overset{0}{\nearrow}$ iG



Diff

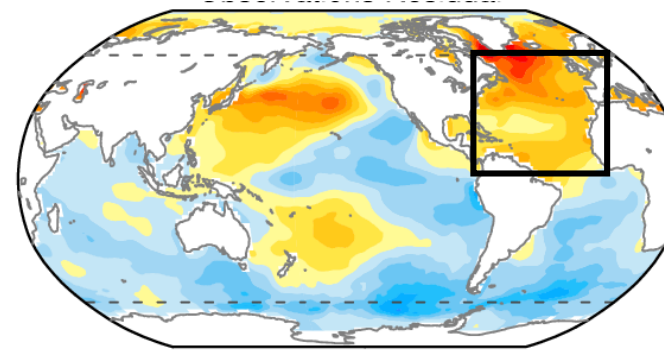


Observations

(ERSSTv5, 1950-2020, 10yrLP)

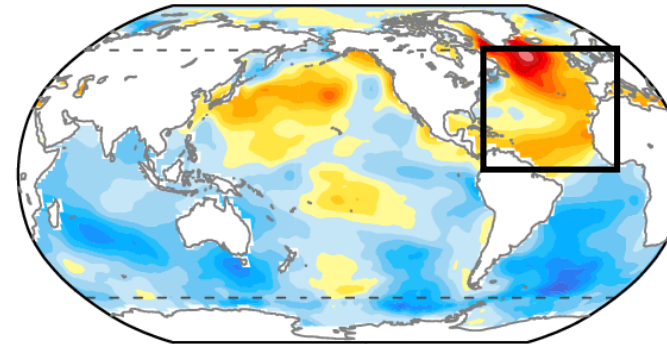
Residual Method
using Obs G(t)

iNA (-iG)

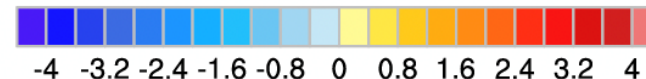
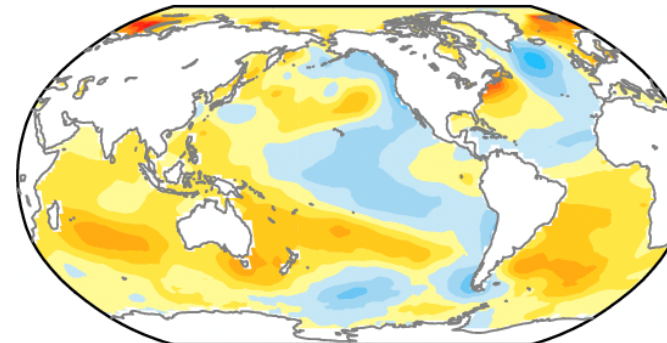


Residual Method
using ensemble-mean
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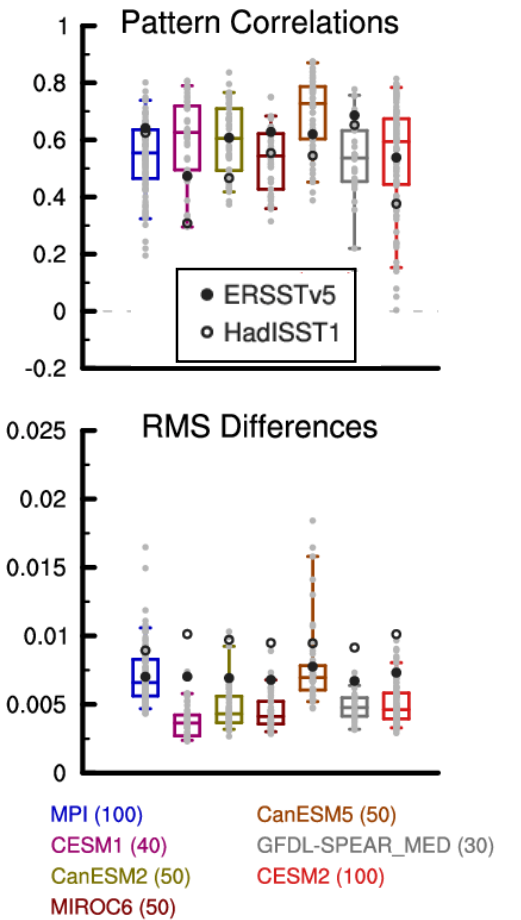
iNA - iG



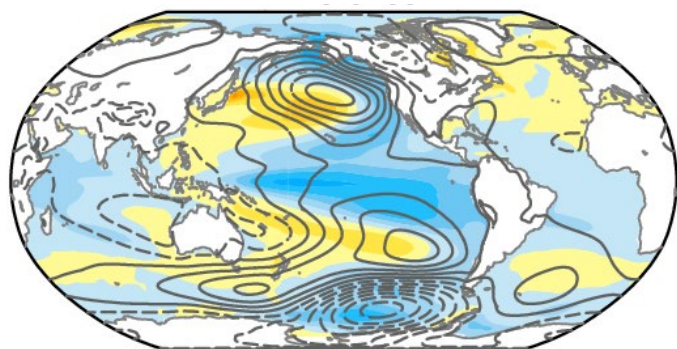
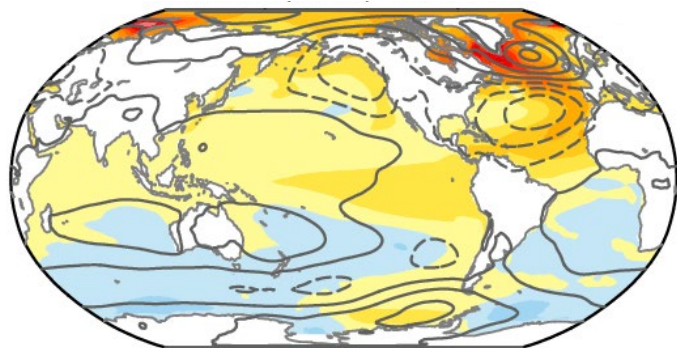
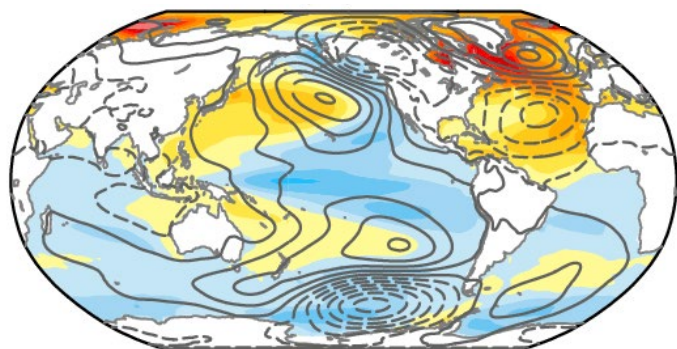
Diff



Sampling variability of AMV pattern



Model Avg



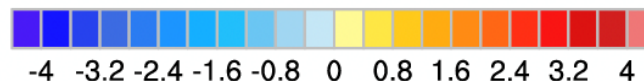
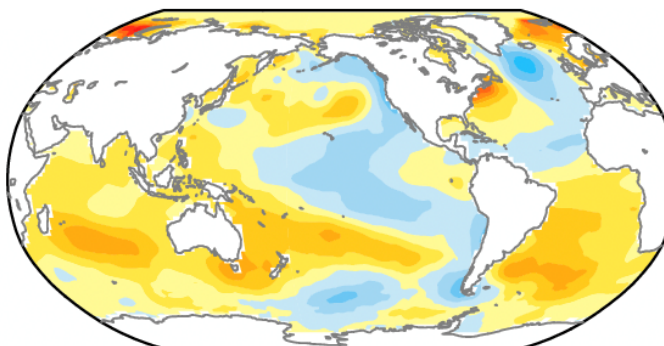
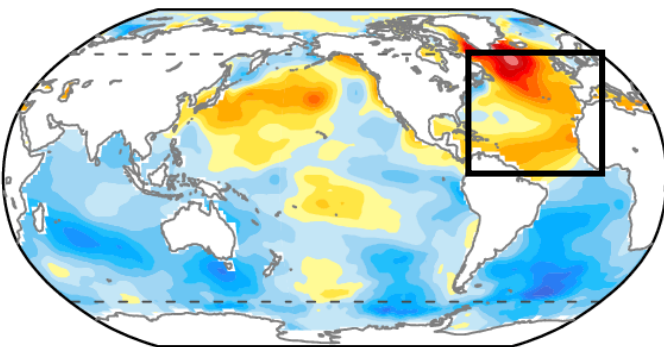
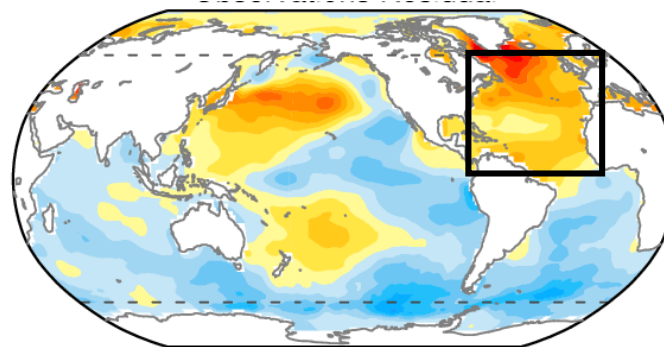
iNA (-iG)

iNA - $\overset{0}{\nearrow}$ iG

Diff

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(ERSSTv5, 1950-2020, 10yrLP)



Sampling variability of AMV pattern

