



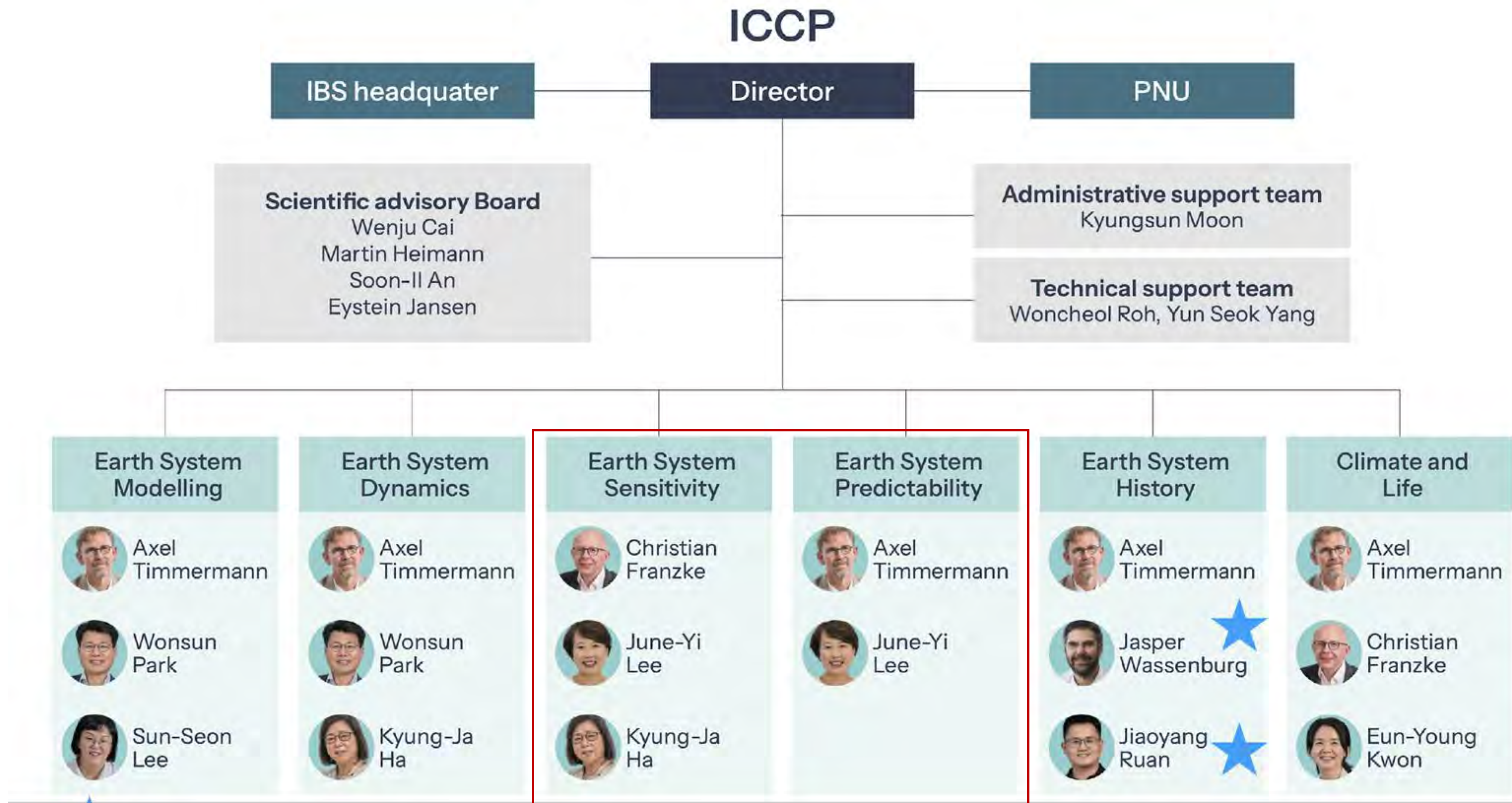
Estimating Predictability of Modes of Variability using the CESM2-based Multi-year Prediction System

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Introduction to IBS Center for Climate Physics (ICCP)



★ Young Scientist Fellow

Courtesy by Axel Timmermann



Earth System Predictability Theme in ICCP

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Session D-2



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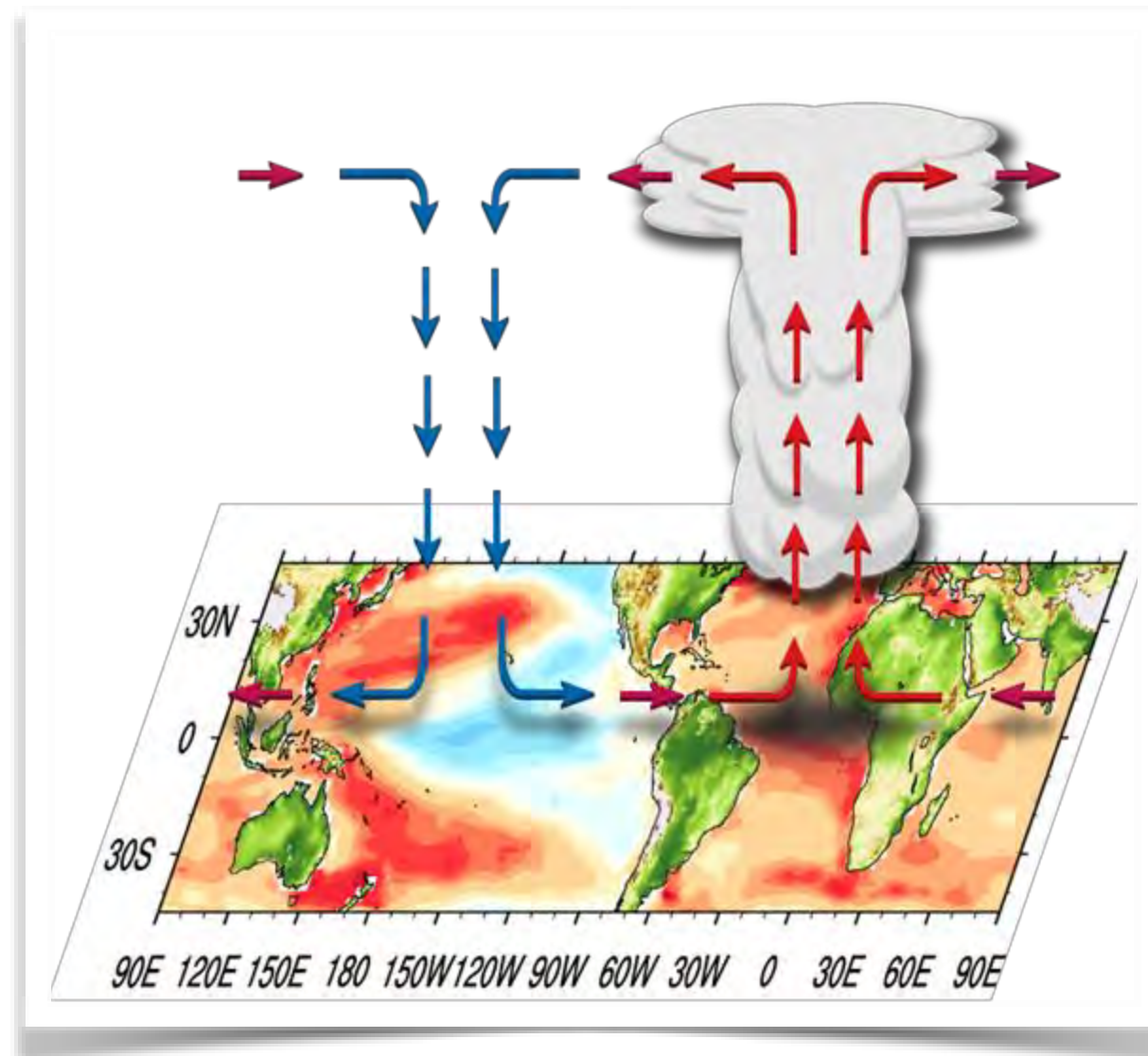
Who Kim



Noel
Keenlyside



Contents



Tropical Trans-basin Variability

Image credit: Yoshi Chikamoto

01

Introduction

02

Description of the ICCP Multi-year Prediction System

03

Multi-year Prediction of Modes of Variability

04

Modes of Variability as Sources of Multi-year Earth System Predictability

05

Summary



Objectives of This Study

- **Enhancing our predictive capability of Earth system components**, not only **physical** but also **ecological** variables, on timescales of **weeks to decades**, using improved comprehensive Earth System models and innovative technologies
- **Improving estimates of predictability** of Earth system components on seasonal to multi-year timescales
- **Better identifying the internal and external components of Modes of Variability**, including ENSO, Trans-Basin Variability, PDO, and AMV
- **Estimating and attributing predictability of Modes of Variability** based on the CESM2-based multi-year prediction system in ICCP





S2S Prediction Models

****AI Global Climate Prediction Model**



Multi-year Prediction System based on CESM2

Predictability from Forcing, Ocean Memory, & Internal Variability

Predictability of Sea Level and Ice Sheet

***Climate-Ice-Sheet Coupled Prediction System (CESM1.2.2+PSUISM)**



CESM2 Large Ensemble (LENS2)



Predictability from Forcing



Predictability of Extreme Events



* Development of a climate-ice-sheet coupled prediction system based on CESM1.2.2 and PSUISM is on progress for sea-level and ice-sheet predictability study

** Development of the AI Global Climate Prediction Model is in progress, collaborating with KIST, POSTECH, and Chonnam Univ.

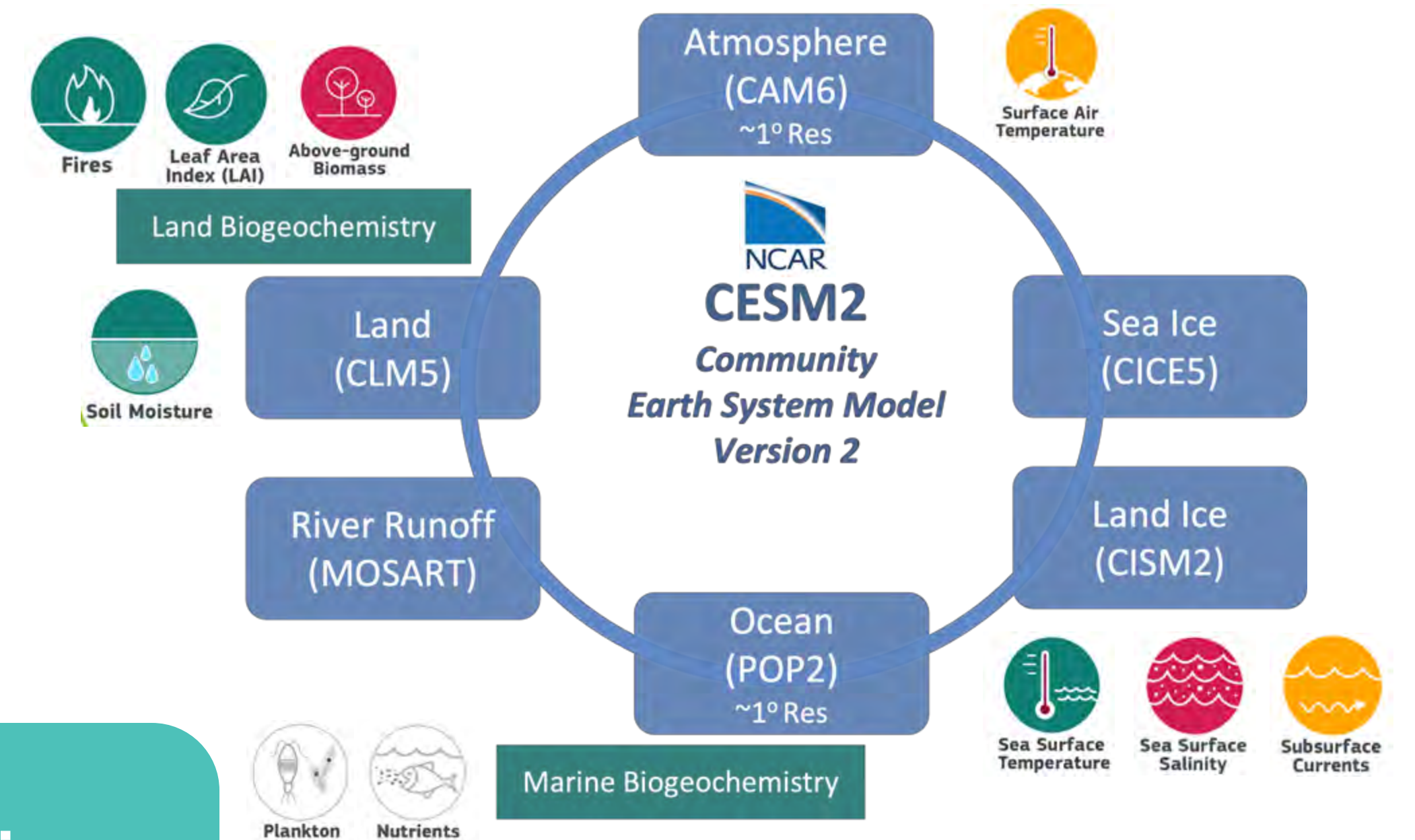


Models and Tools for the Earth System Predictability Theme



Multi-year Prediction System based on CESM2

Predictability from Forcing, Ocean Memory, & Internal Variability



Days

Weeks

Months

Seasons

Years

Decades

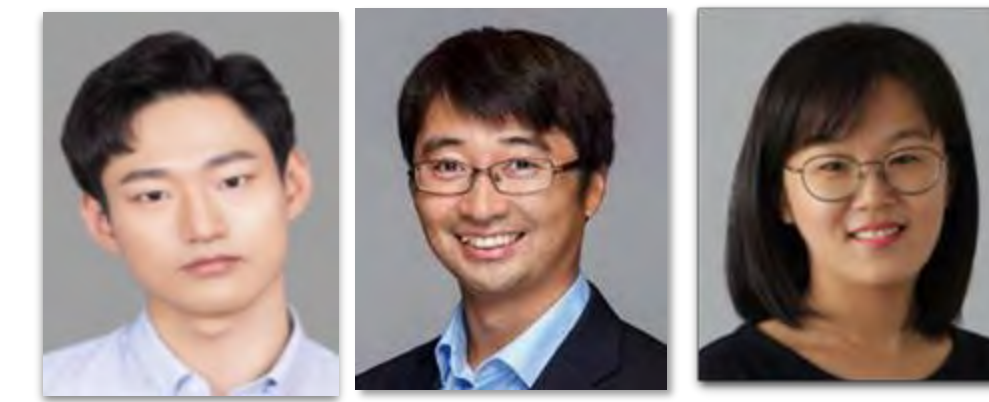
Centuries

* Development of a climate-ice-sheet coupled prediction system based on CESM1.2.2 and PSUISM is on progress for sea-level and ice-sheet predictability study

** Development of the AI Global Climate Prediction Model is in progress, collaborating with KIST, POSTECH, and Chonnam Univ.



The CESM2-based Multi-year Prediction System



Ocean Reanalyses

- UKMO EN4.2.2 (Good et al., 2013)
- JMA ProjD7.3 (Ishii et al., 2017)
- ECMWF ORAS4 (Balmaseda et al., 2013)

3-Dimensional Ocean Temperature & Salinity

CESM2 Large Ensemble
(Uninitialized runs)

Full Forcing Included

Ocean Anomaly
Data Assimilation
(Temperature/Salinity)

Initialized from Jan 1

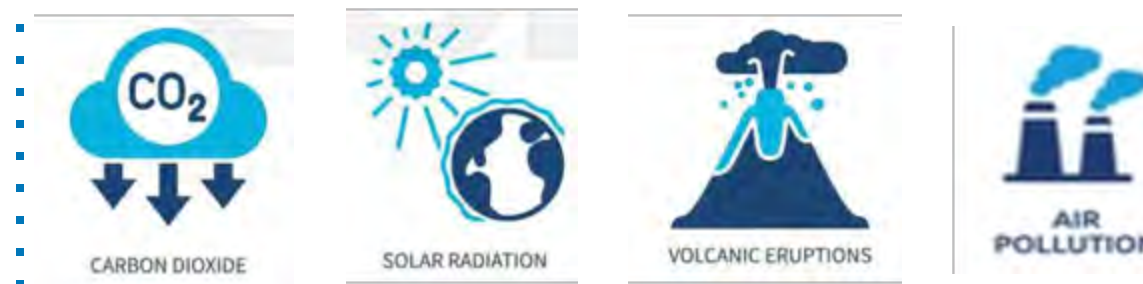
5-year Hindcast
(Initialized runs)

- Total 100 members (Rodger et al., 2021)
- 10 members for assimilation

- 10 members from EN4.2.2 (1950-2021)
- 10 members from ProjD7.3 (1955-2021)
- 10 members from ORAS4 (1958-2016)

- 20 members (EN4.2.2, ProjD7.3)
- 62 years from 1960-2021
- 6200 simulation years (>2 PB)

Predictability of the 2nd Kind
arising from **external forcings**

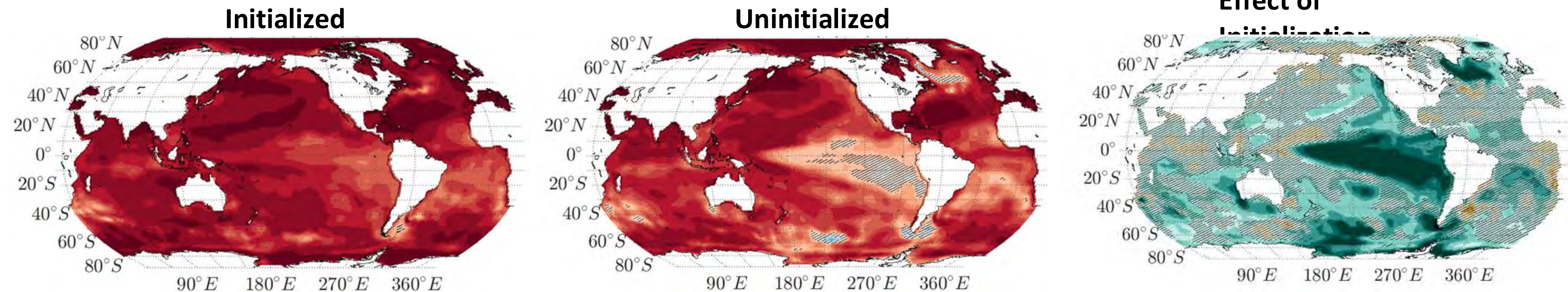


+ Predictability of the 1st Kind
arising from **ocean memory** and **modes of variability**

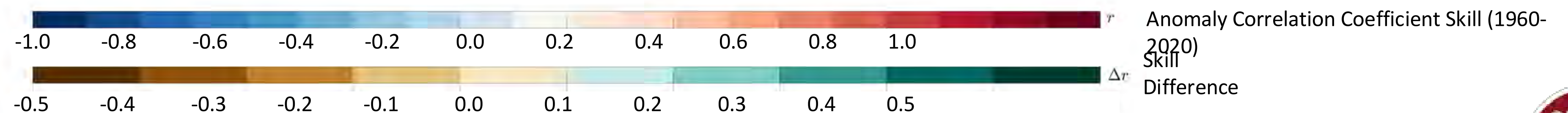


New Estimation of Predictability (ASSM vs HIND): SST

Conventional Method (Ensemble-mean Based, Lead Year 1)/ Assimilation as a reference

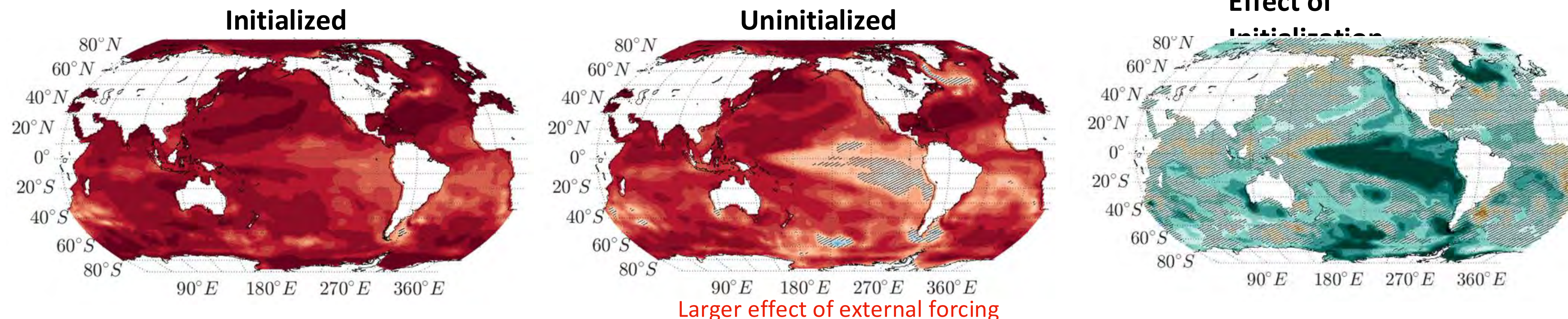


- Does ensemble mean represent the most probable predictable state for nonlinear system with non-Gaussian properties?
- A higher anomaly correlation coefficient skill in the ensemble mean than individual-member-based metrics by **removing stochastic noise** and **filtering out the less predictable internal variability component**

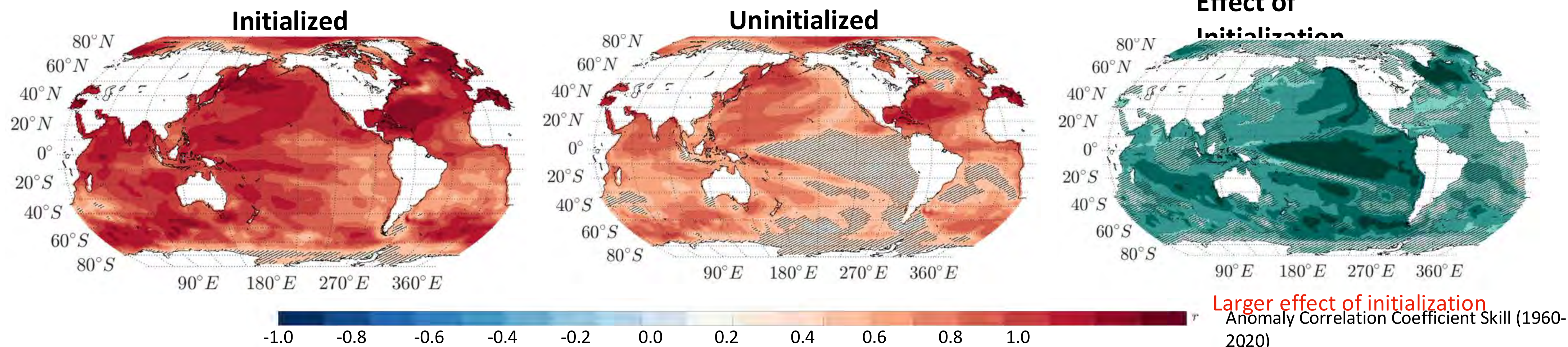


New Estimation of Predictability (ASSM vs HIND): SST

Conventional Method (Ensemble-mean Based, Lead Year 1)/ Assimilation as a reference



New Method (Individual-member Based, Lead Year1)/ Assimilation as a reference



- Initialized physical properties and the associated constraints in ocean circulation provide an important source for marine ecosystem predictability. => See Dr. Yong-Yub Kim's Poster for details

New Estimation of Predictability (ASSM vs HIND): Terrestrial Variables

New Method (Individual-member Based, Lead Year1)/ Assimilation as a reference

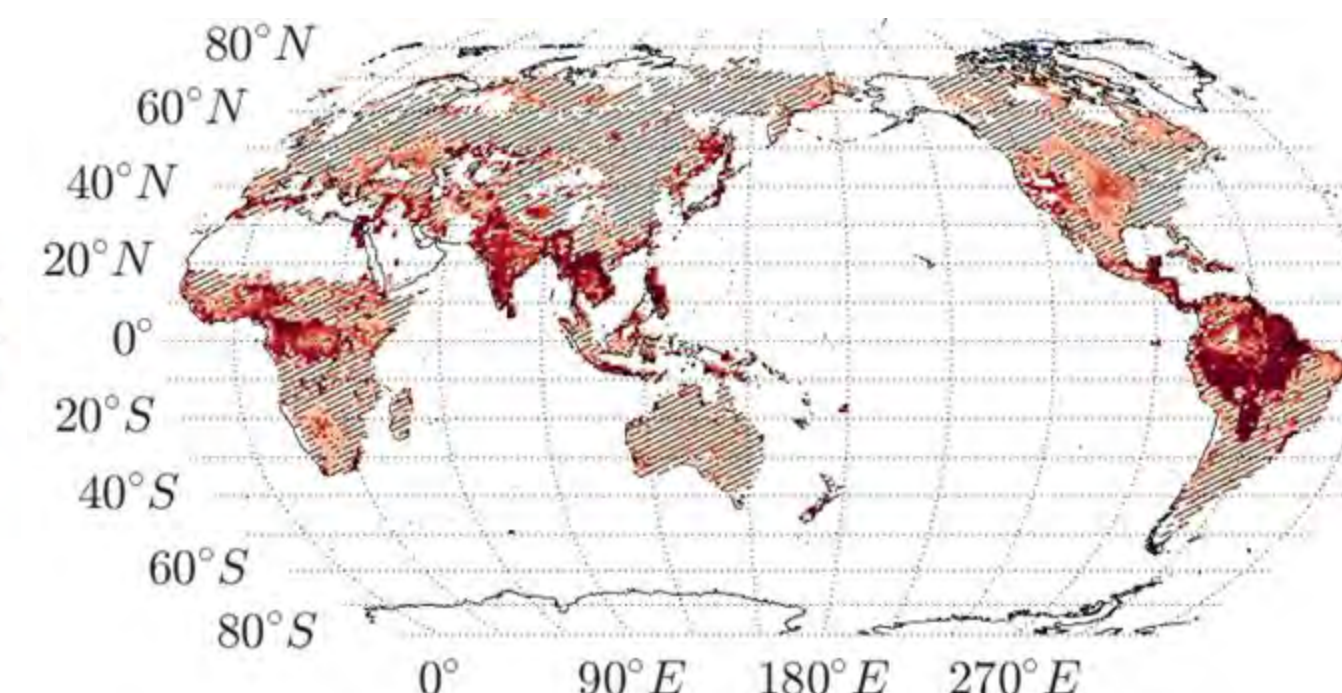
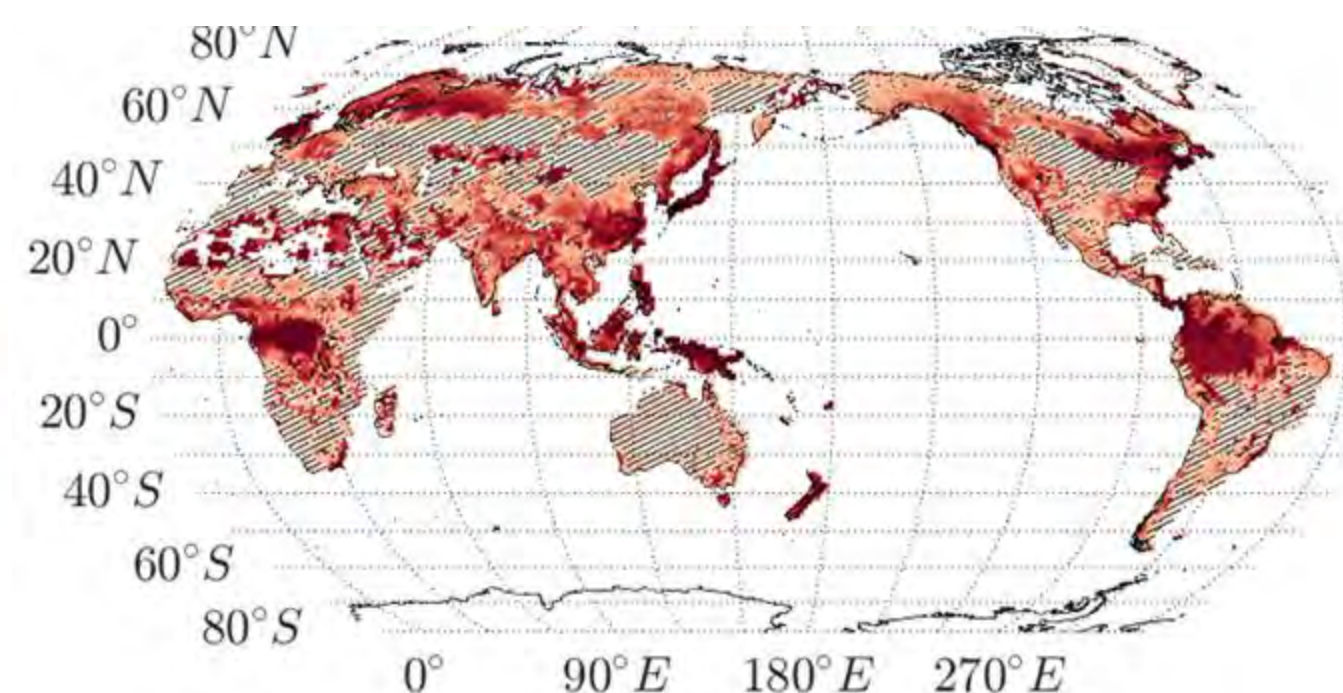
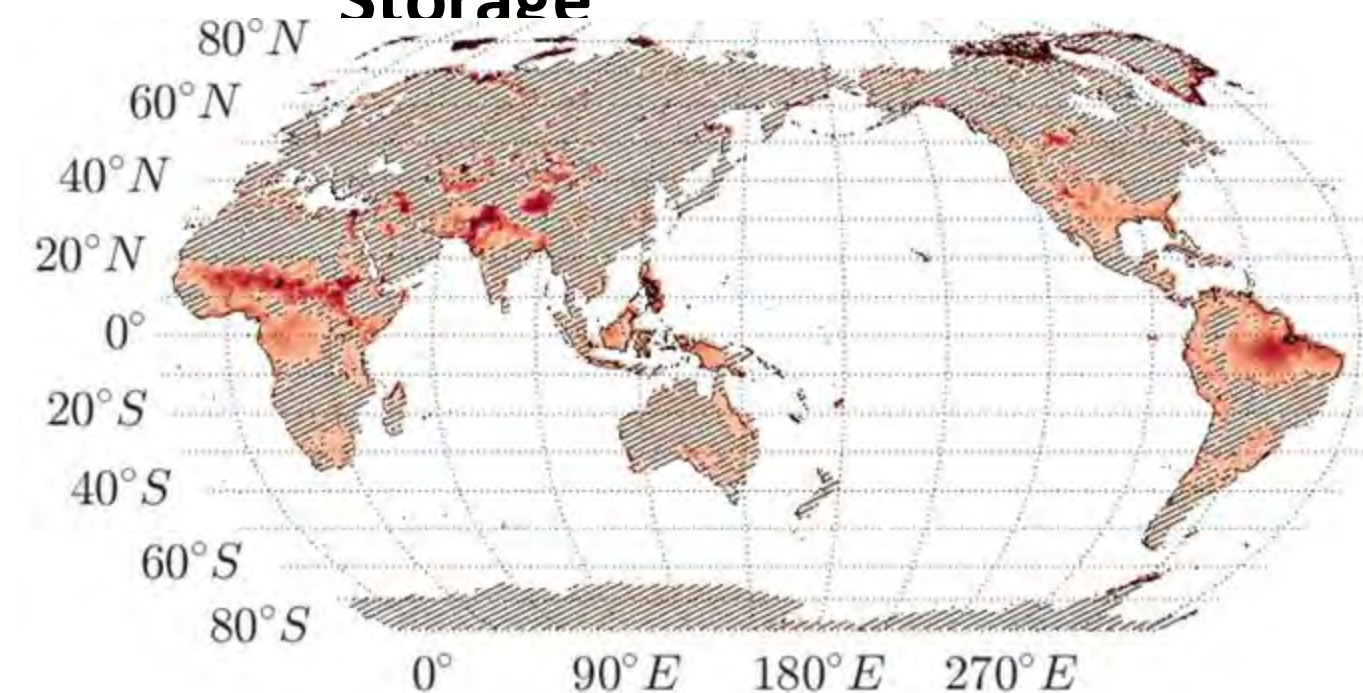
Total Water

Storage

Gross Primary Productivity

Burned Area

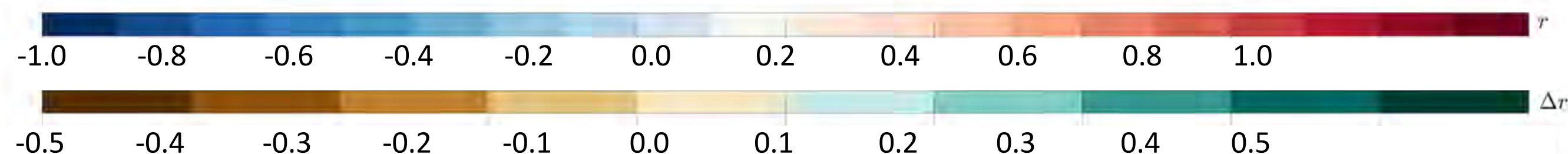
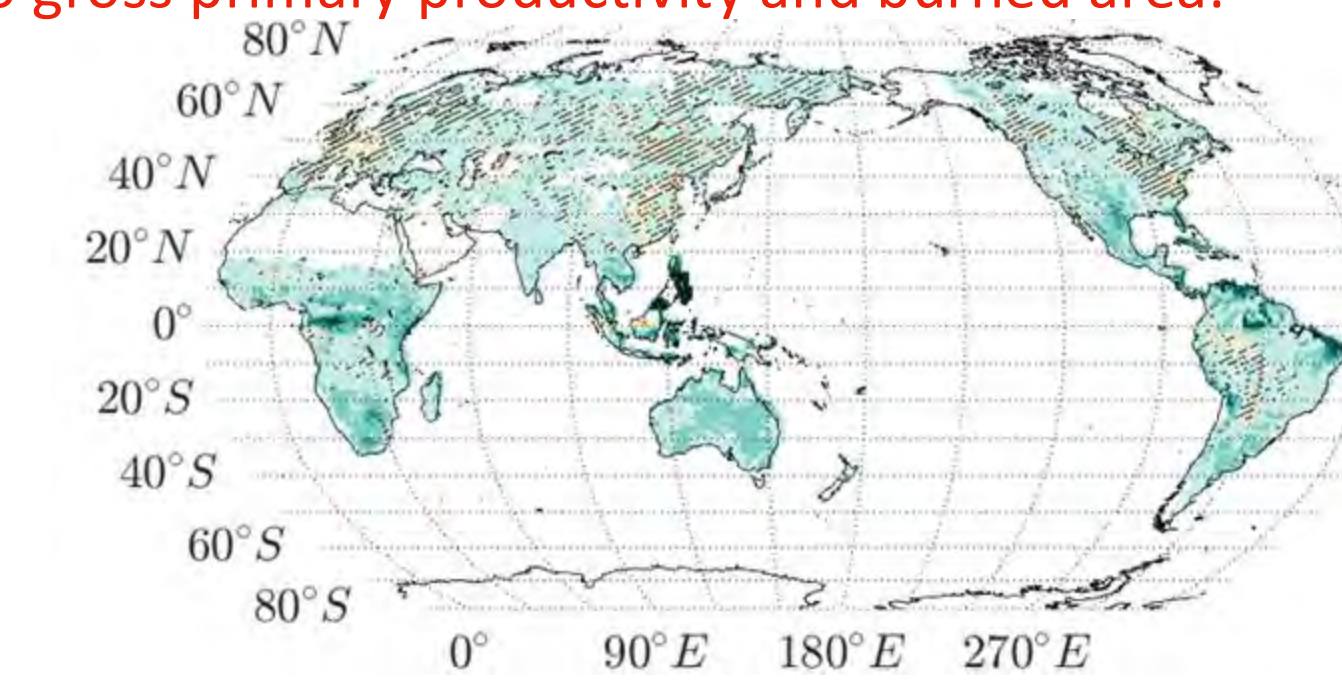
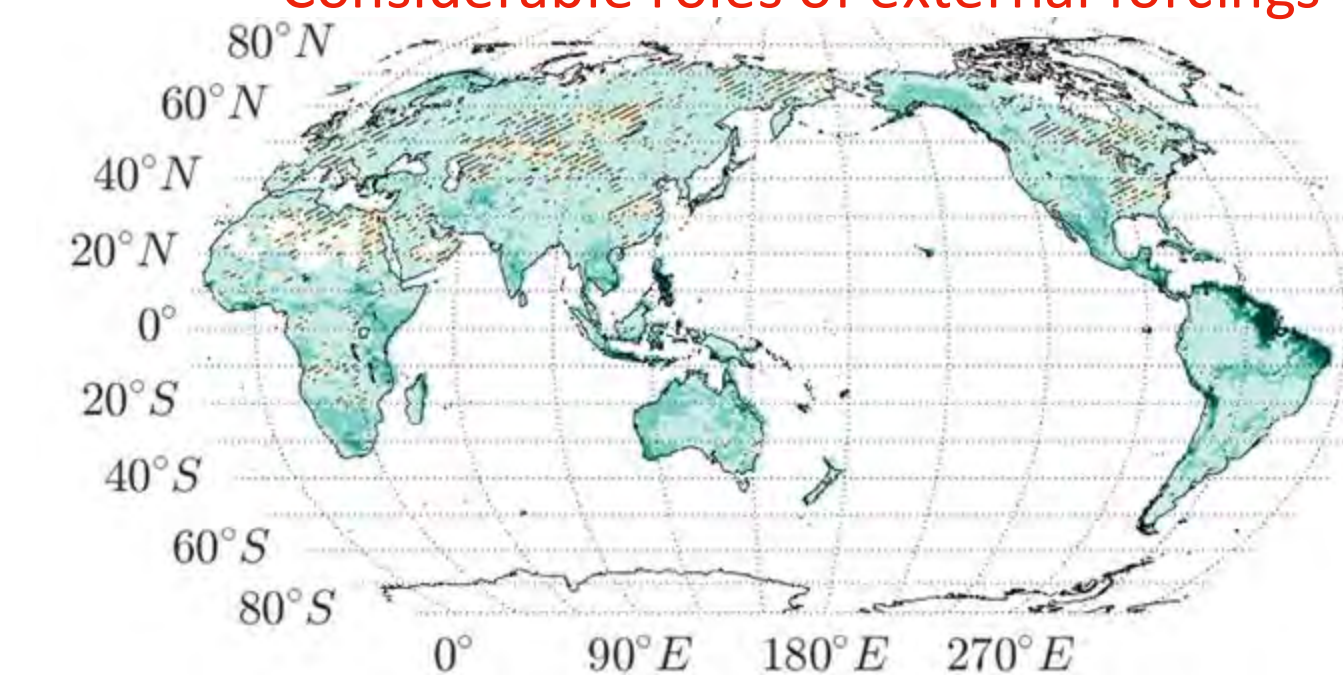
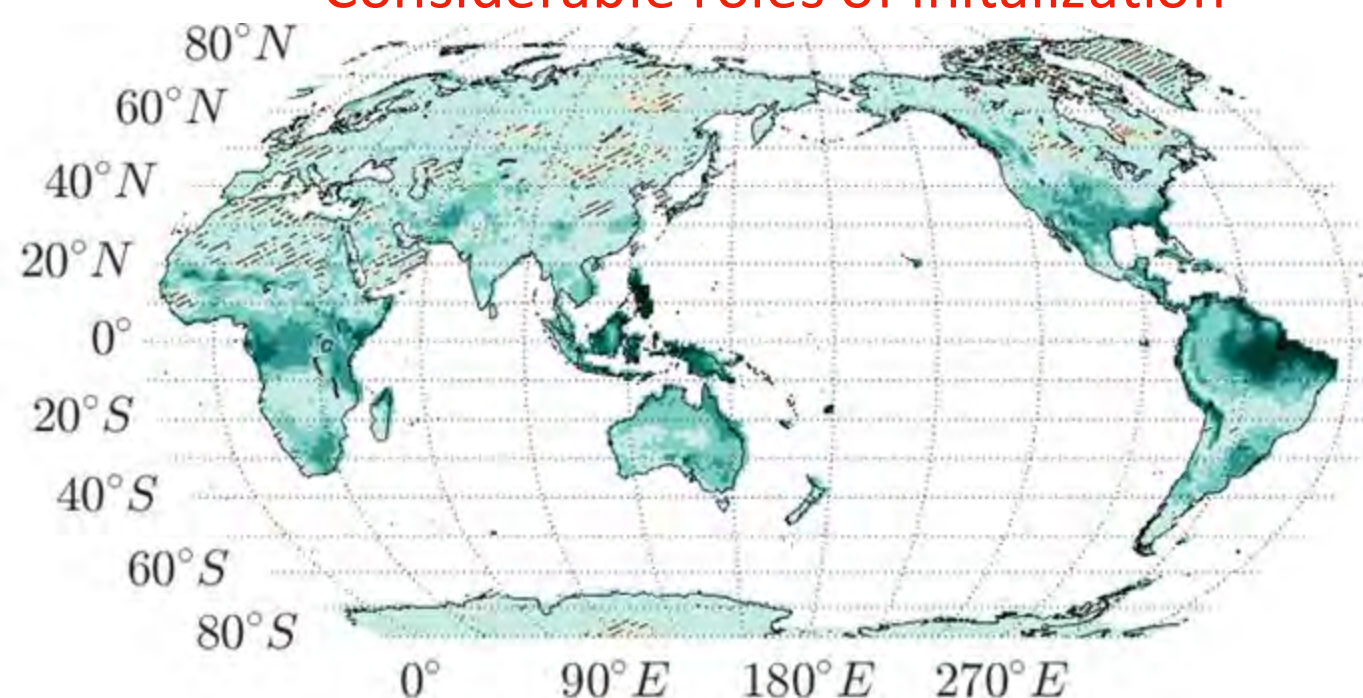
Initialized



Considerable roles of initialization

Considerable roles of external forcings to gross primary productivity and burned area.

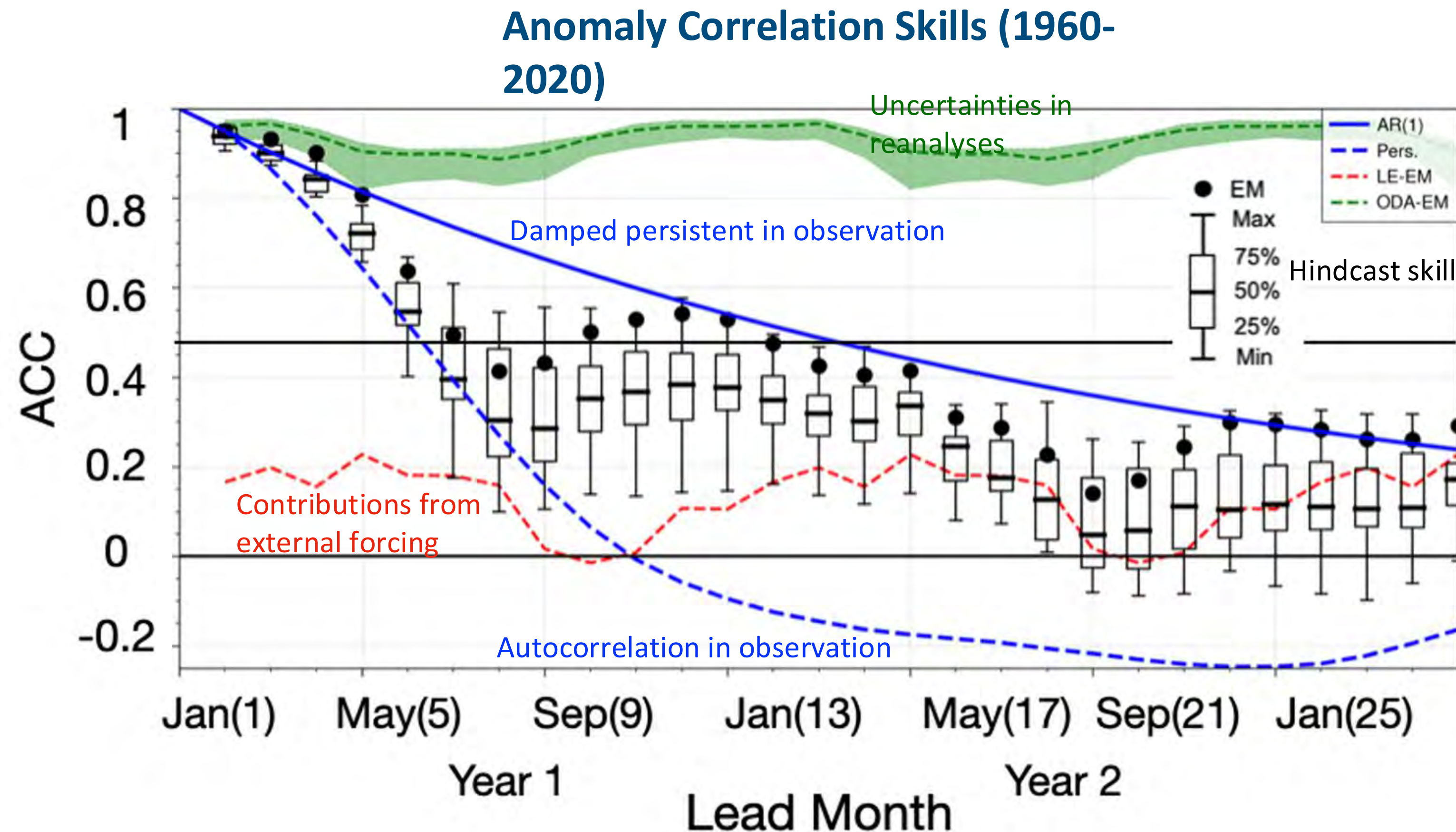
Effect of
Initialization



Anomaly Correlation Coefficient Skill (1960-2020)
Skill
Difference

- Predictability for **gross primary productivity** arising from variations in surface temperature, solar radiation, CO₂ levels, and land water storage.
- Predictability for **wildfire** arising from large-scale climate patterns and external forcings.

Prediction of Modes of Variability: El Nino-Southern Oscillation



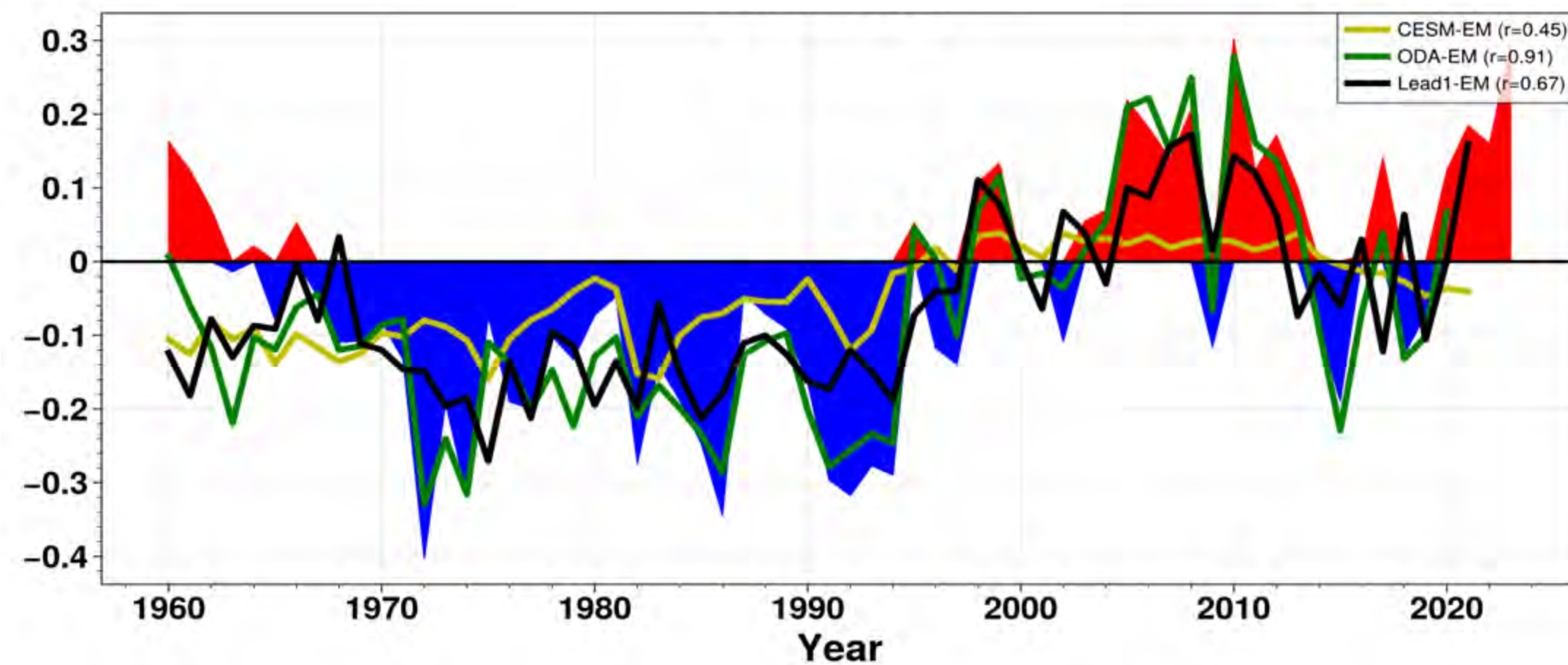
- Major modes of climate variability such as **El Niño-Southern Oscillation (ENSO)** are important sources of earth system predictability.
- Their corresponding predictive skill due to initialization may be translated into other **societally relevant physical and biogeochemical variables**.
- For ENSO, the individual-member based skill is a more robust metrics for predictability of the 1st kind since the probability distribution of ENSO is highly skewed with considerable nonlinearities.
- The long-lead predictability of the ENSO is also contributed by its damped persistent characteristics.

* The ENSO index based on Niño 3.4 SST index

* The prediction skill is against observation



The AMV index

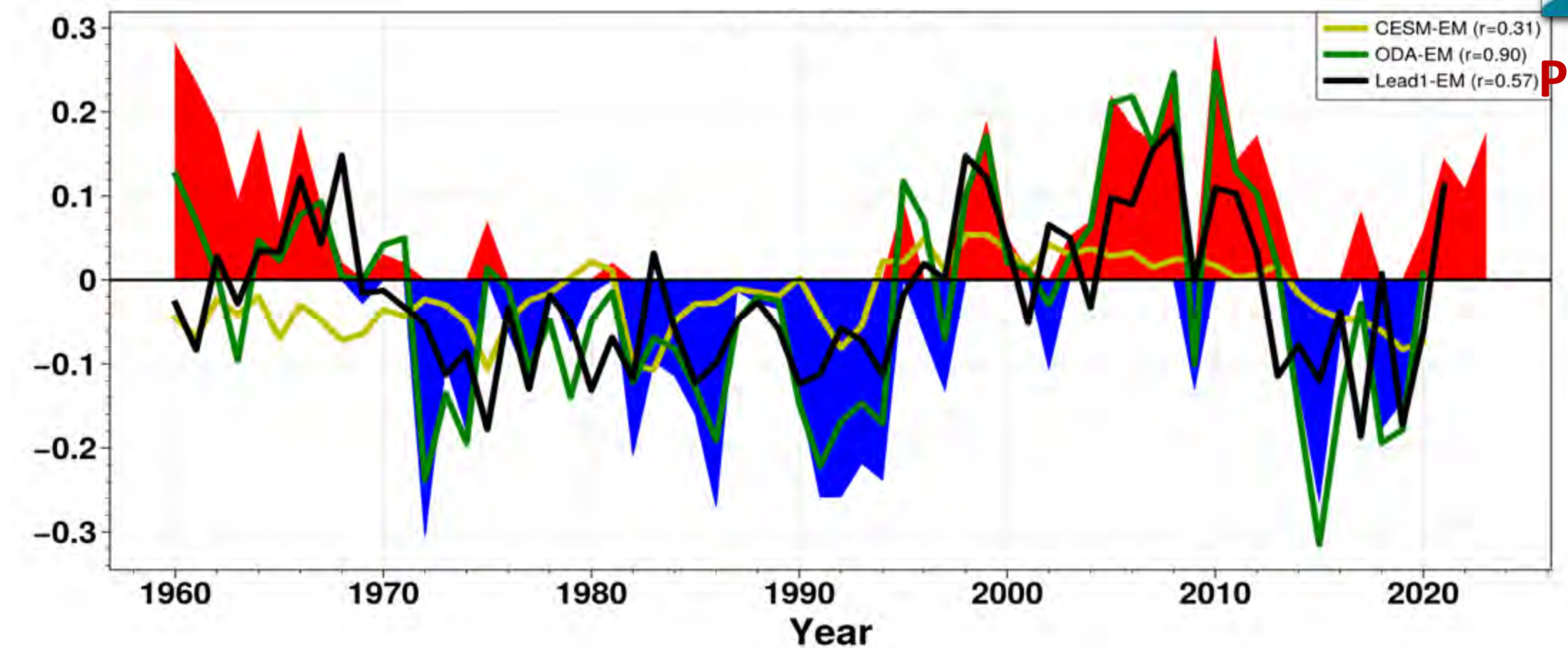


Based on Trenberth and Shea (2006)

$$\text{AMV Index} = \text{AVG}(\text{NASST} - \text{Gt})$$

* Forced Response for $\text{AVG}(\text{NASST}) \sim 45\%$

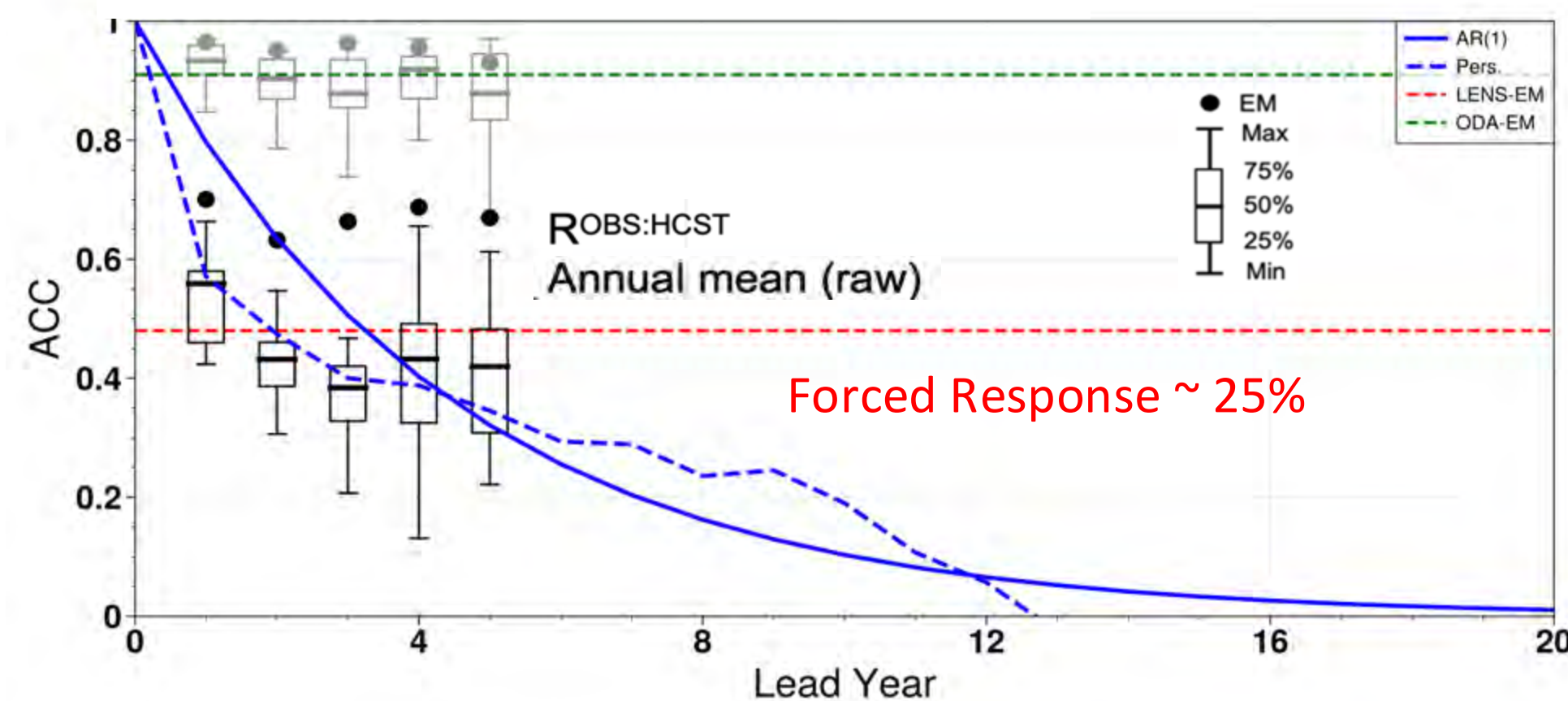
The internal AMV (iAMV) index



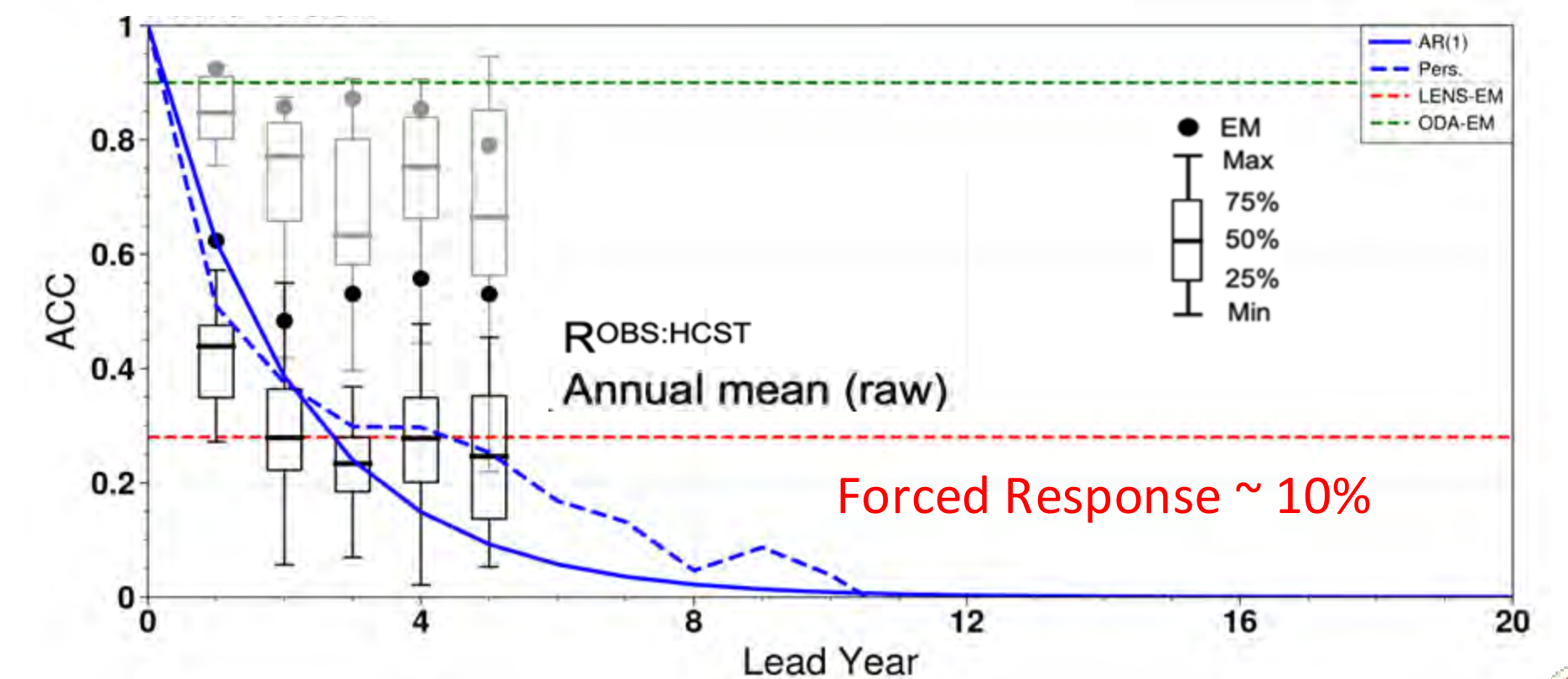
Based on Global Residual Method (Deser and Phillip, 2021)

$$\text{iAMV Index} = \text{AVG}(\text{NASST} - \text{GT} * \text{GSSTreg})$$

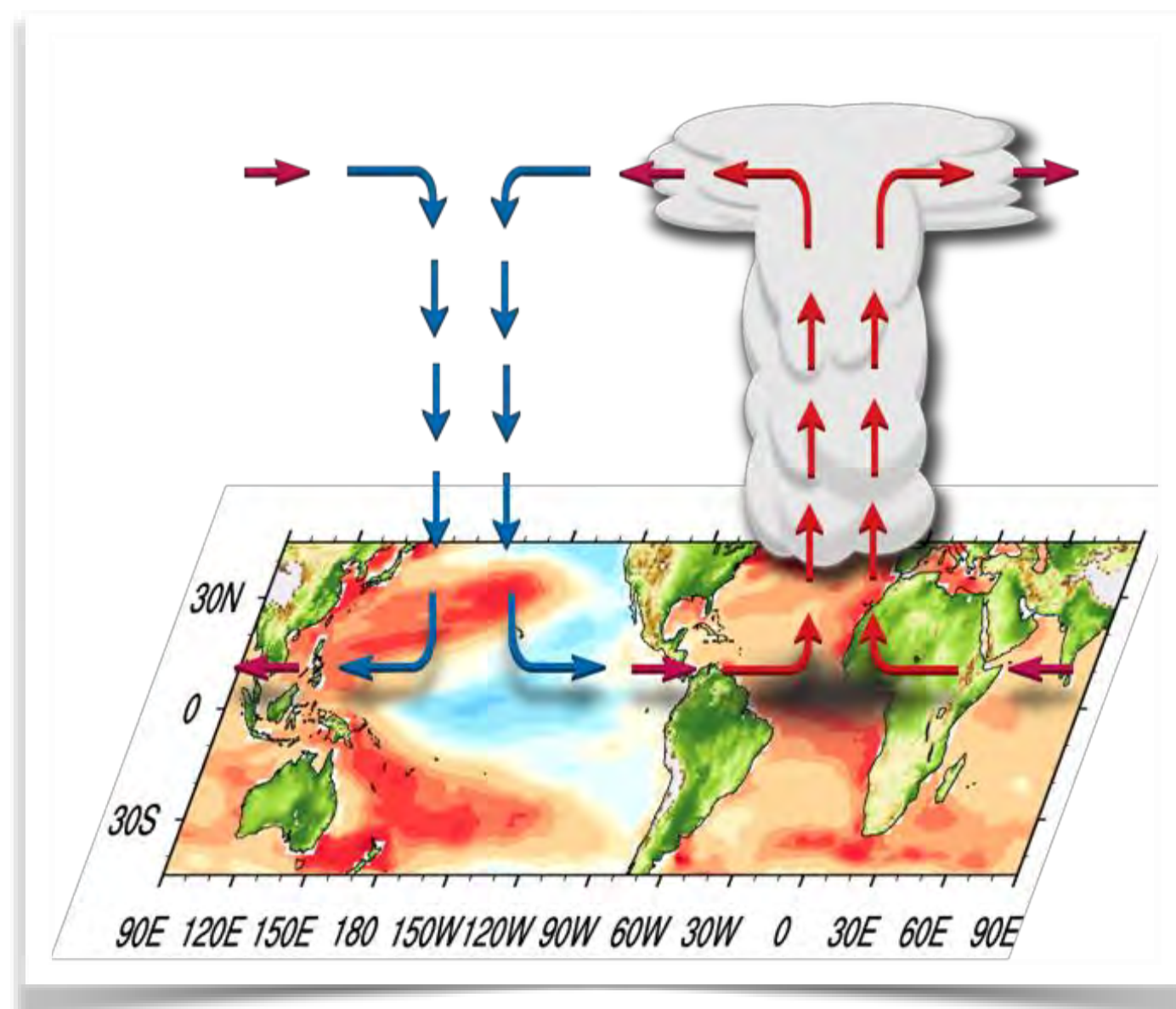
The ACC Skill for the AMV index



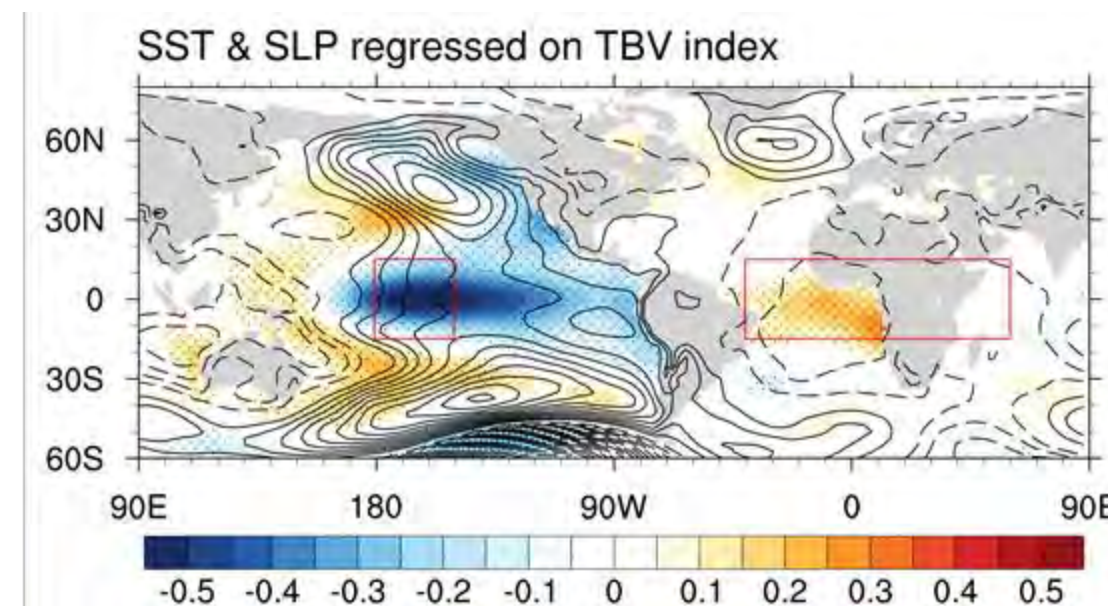
The ACC Skill for the iAMV index



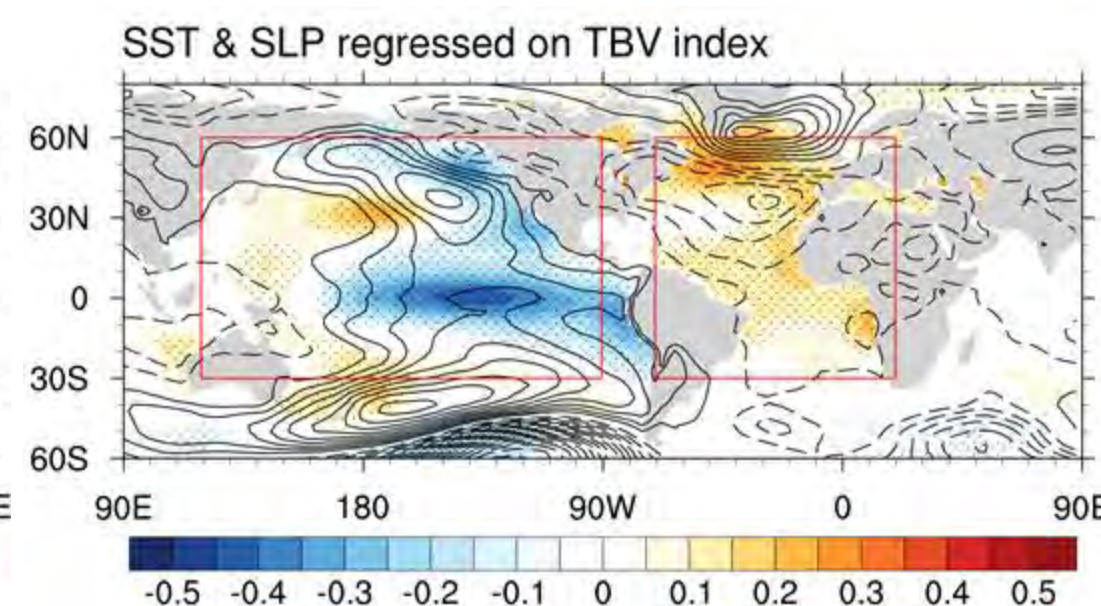
Prediction of Modes of Variability: Trans-basin Variability



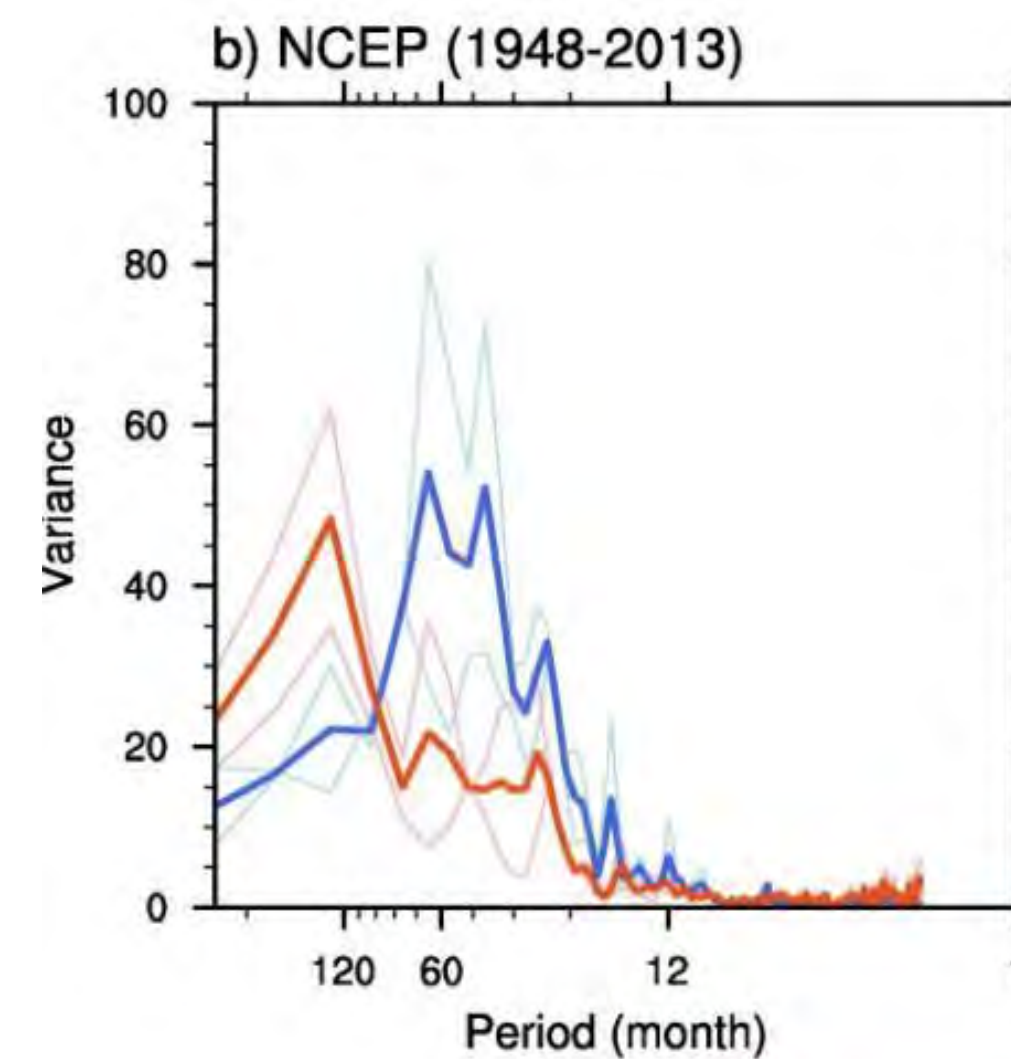
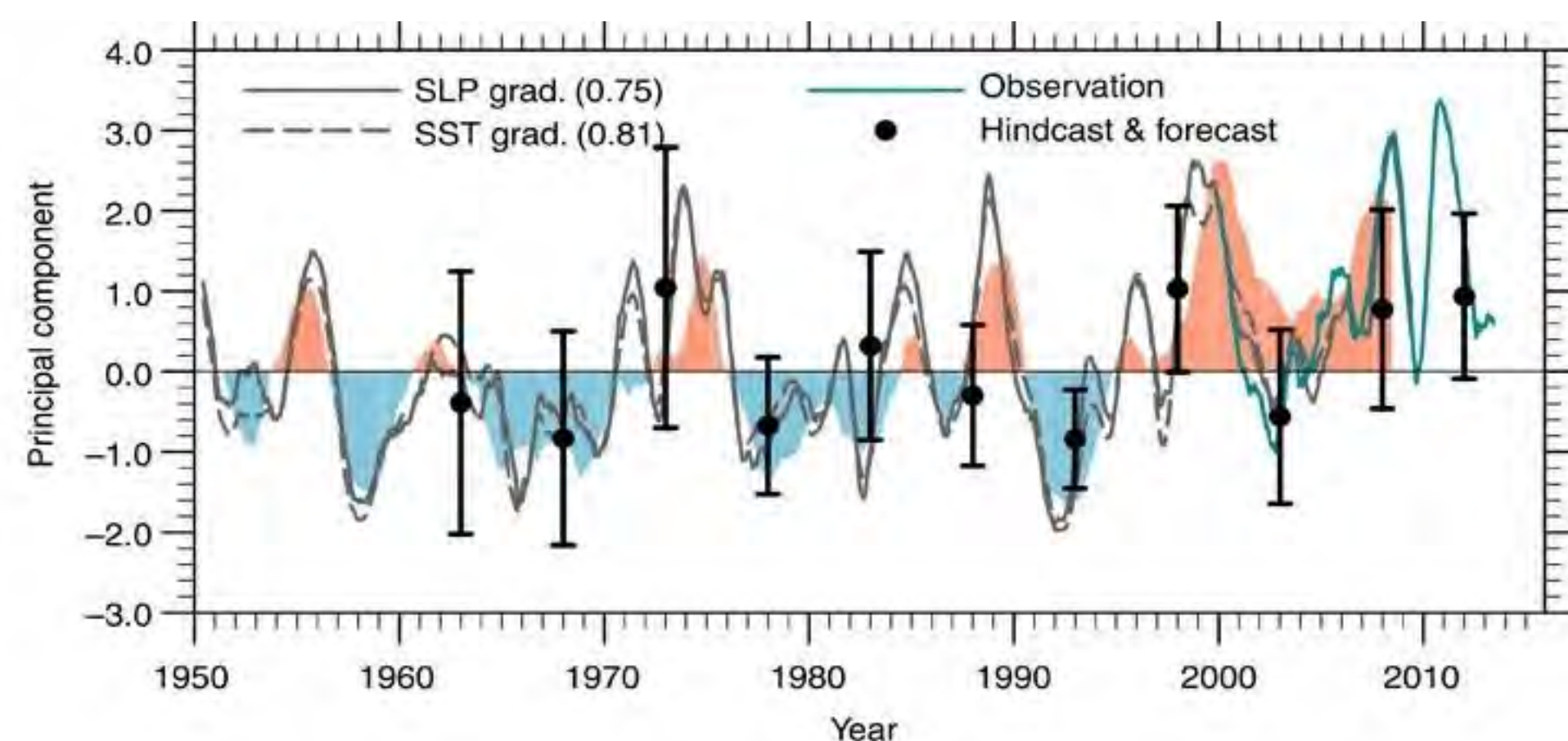
SST-based index
(Chikamoto et al. 2015)



SST-based index
(McGregor et al. 2014)



Tropical Trans-Basin Variability Indices



Dominant Periods

Blue line for ENSO:
2~7 years

Red line for TBV:
10~20 years

Prediction of Modes of Variability: Trans-basin Variability

Definition

Tropical Central Pacific

Tropical Atlantic-Indian Ocean

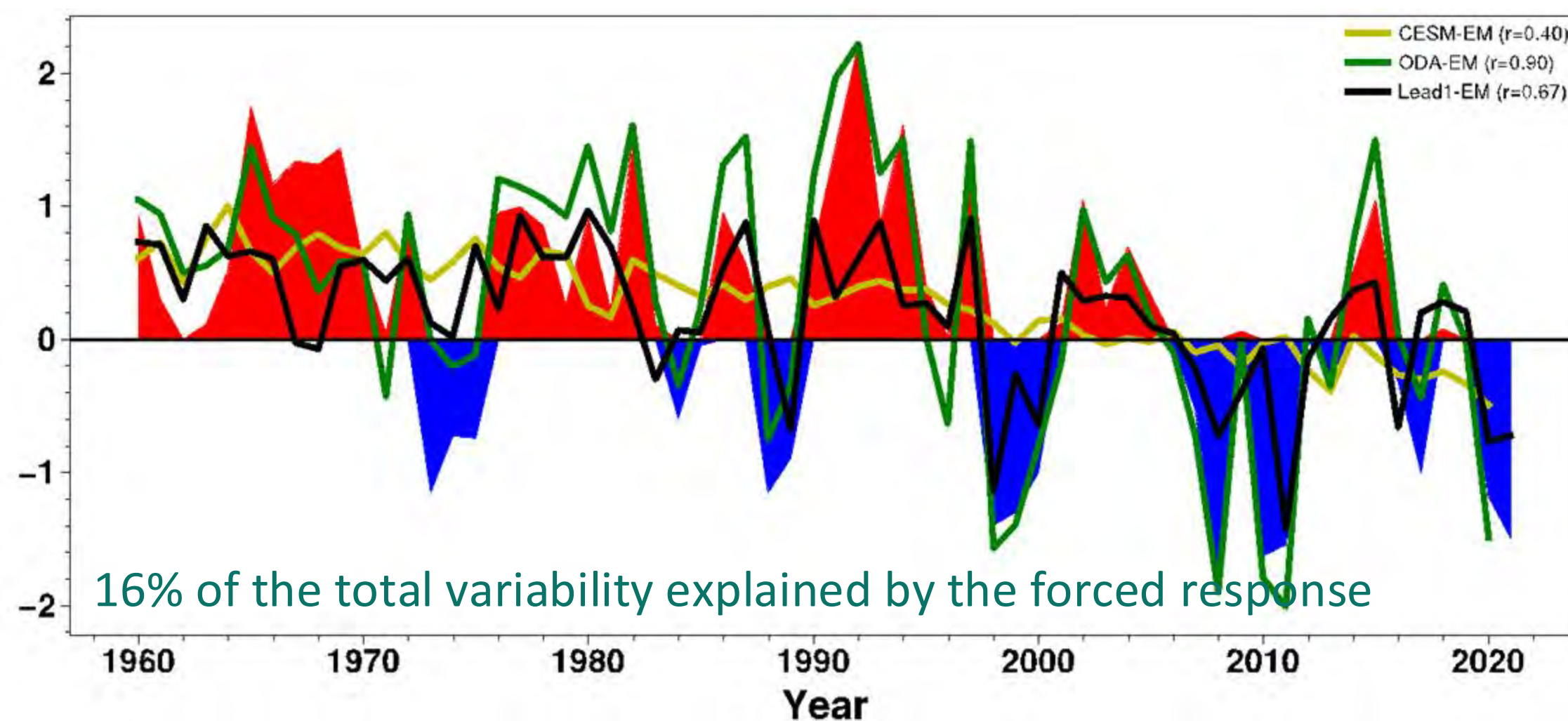
The TBV index: $SST_{\text{norm}} (15^{\circ}\text{S}-15^{\circ}\text{N}, 180^{\circ}\text{W}-150^{\circ}\text{W}) - SST_{\text{norm}} (15^{\circ}\text{S}-15^{\circ}\text{N}, 40^{\circ}\text{W}-60^{\circ}\text{W})$

$SST_{\text{norm}} = \text{SST anomaly normalized by standard deviation at each grid}$
(based on Chikamoto et al. (2015))

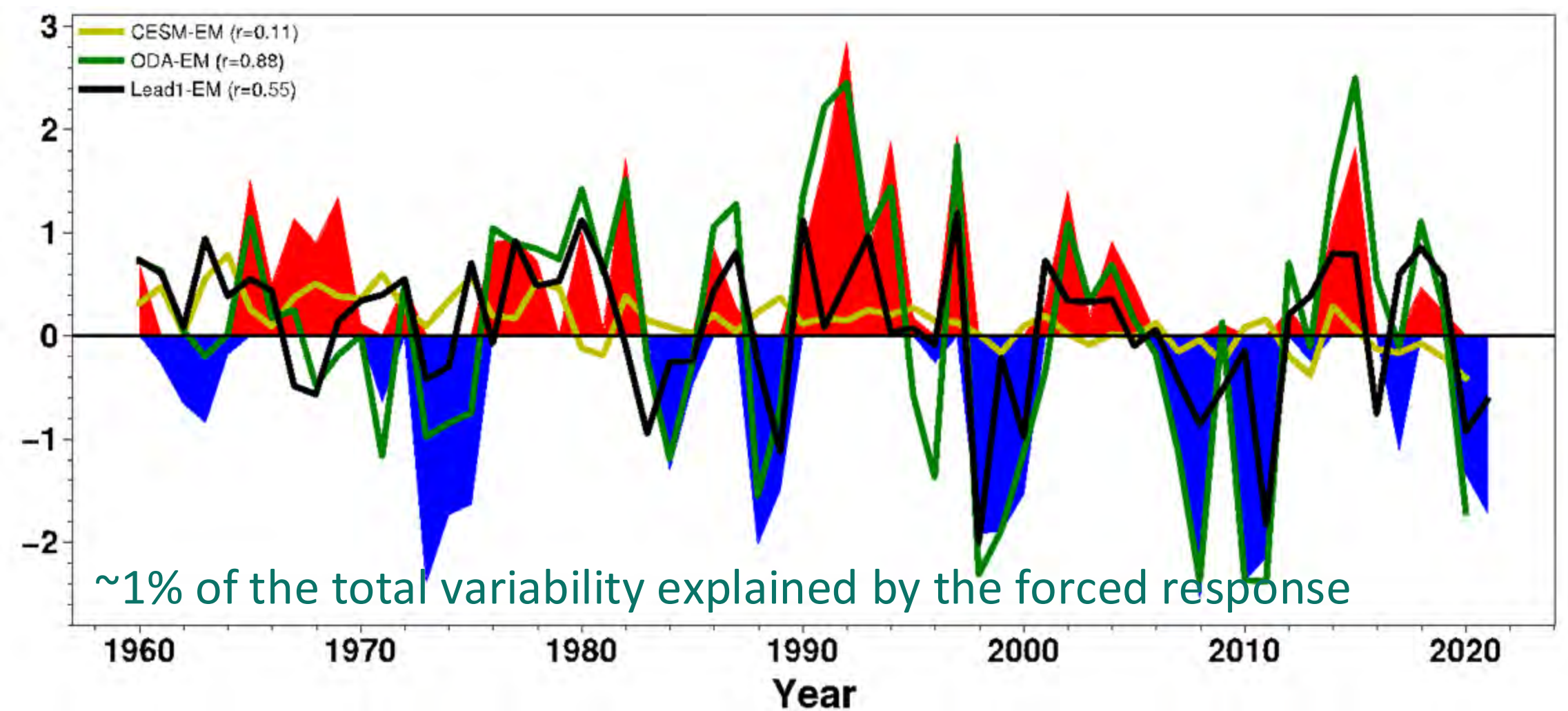
The iTBV index: $iSST_{\text{norm}} (15^{\circ}\text{S}-15^{\circ}\text{N}, 180^{\circ}\text{W}-150^{\circ}\text{W}) - iSST_{\text{norm}} (15^{\circ}\text{S}-15^{\circ}\text{N}, 40^{\circ}\text{W}-60^{\circ}\text{W})$

$iSST \text{ anomaly} = \text{SST anomaly } (i,j) - \text{GM Regression } (I,j)$ (based on Deser and Phillips (2021))

The TBV index

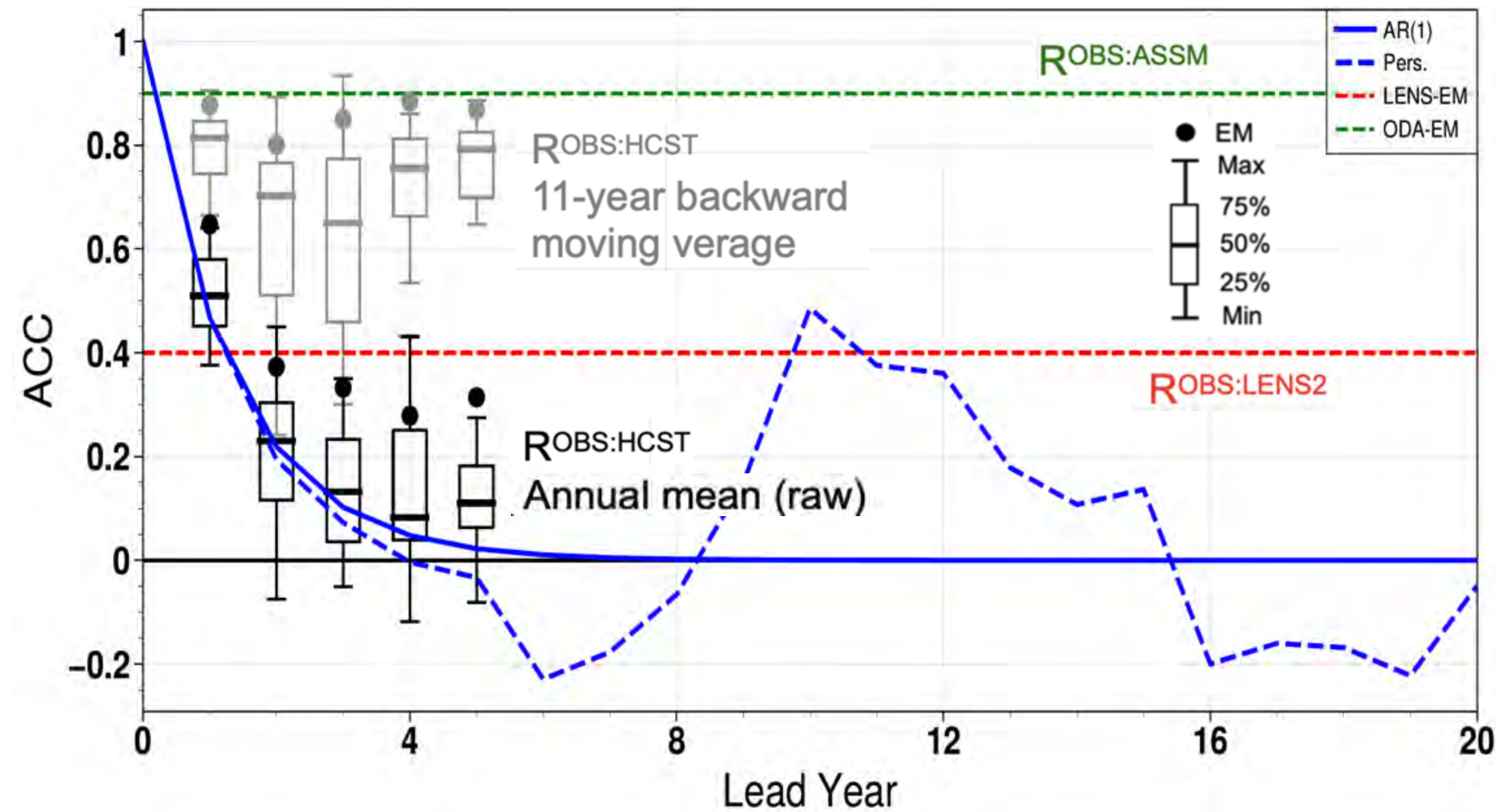


The iTBV index

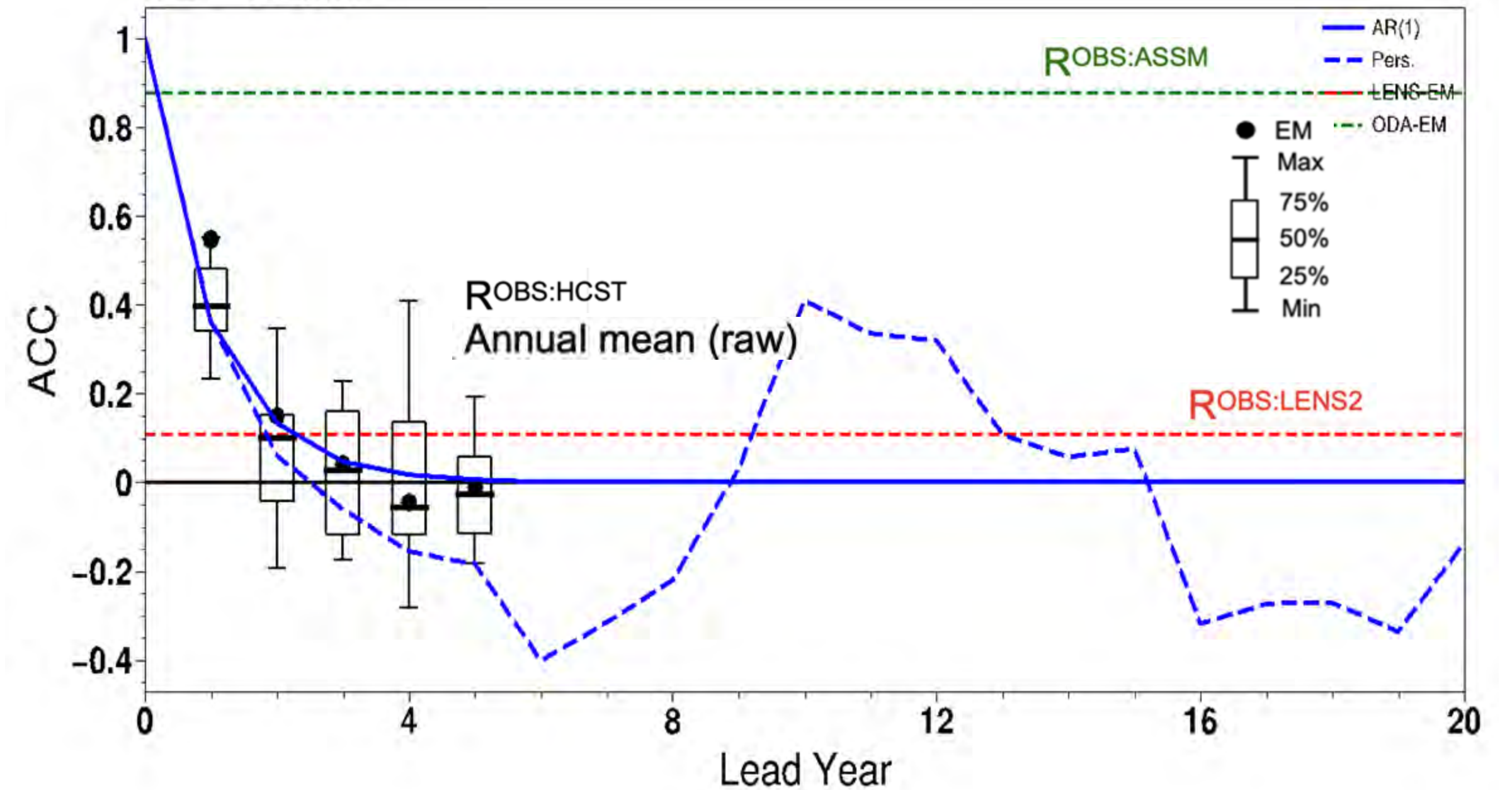


Prediction of Modes of Variability: Trans-basin Variability

The TBV index



The iTBV index



- The long-term predictability mainly comes from the externally forced response
- The CESM2 hindcast has a useful skill for the internal component of the TBV only at 1-year forecast lead

Modes of Variability as Important Sources for Climate Predictability

The AMV and iAMV Teleconnection: Sea Surface Temperature

OBS

Only 1 sample

LENS2

Average of
50 samples

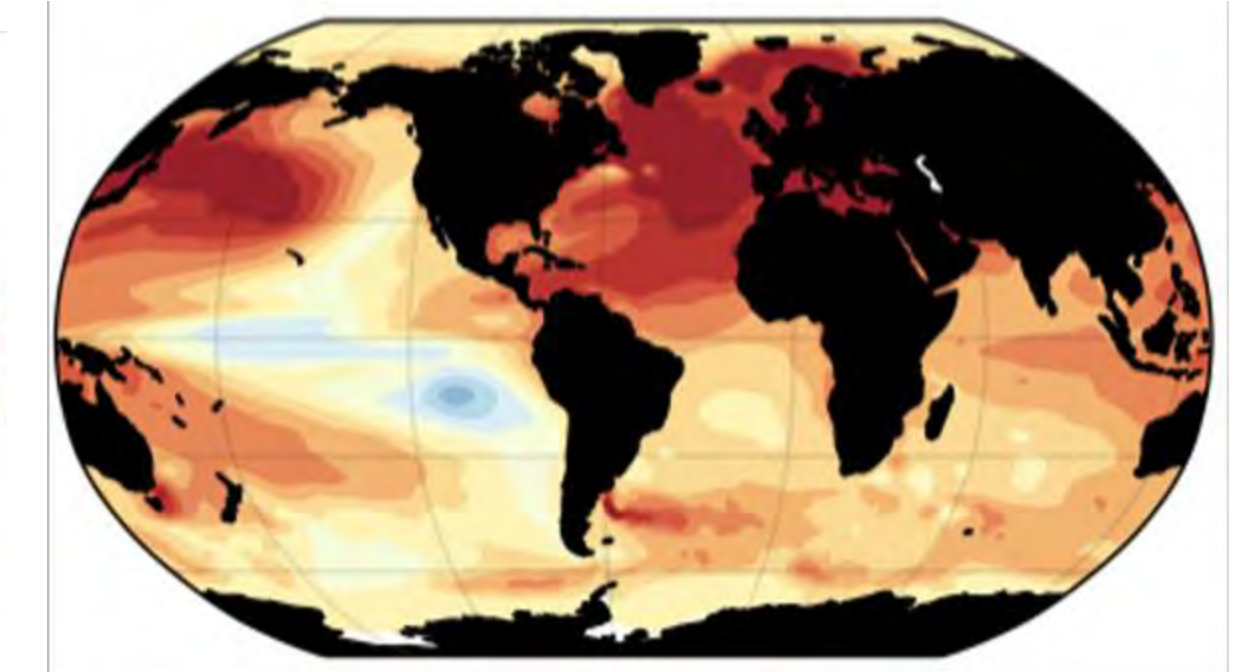
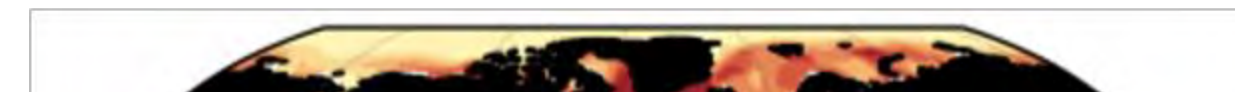
ODA

Average of
20 samples

1-yr HCST

Average of
20 samples

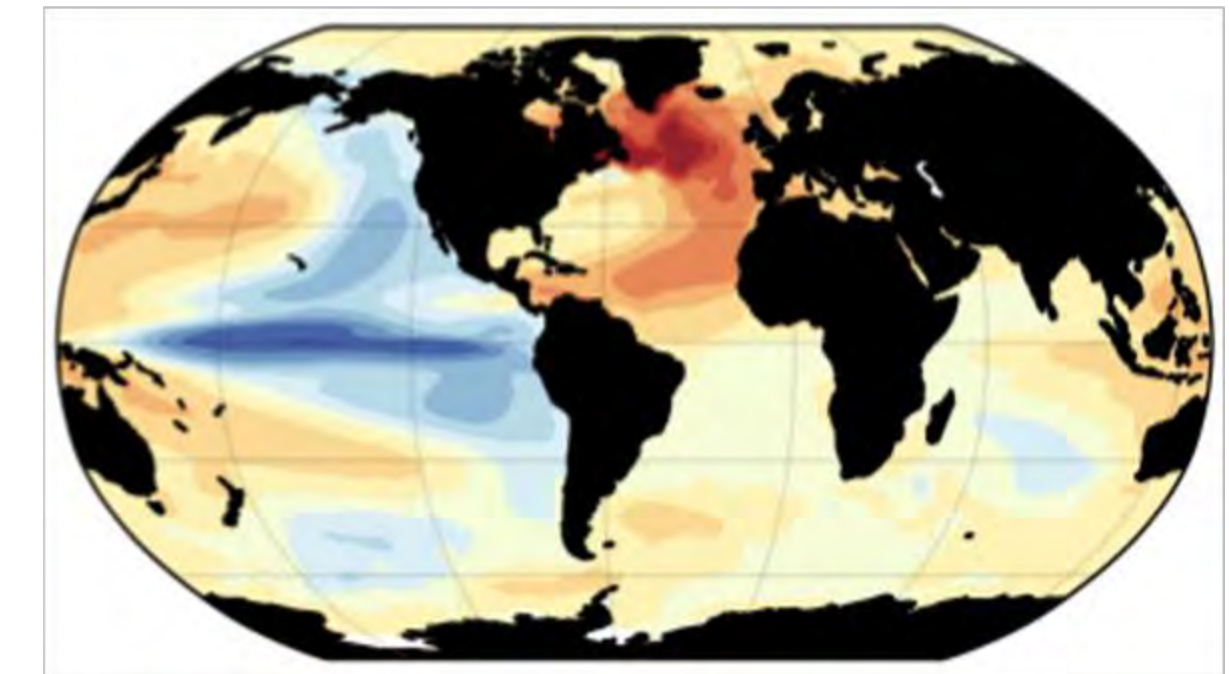
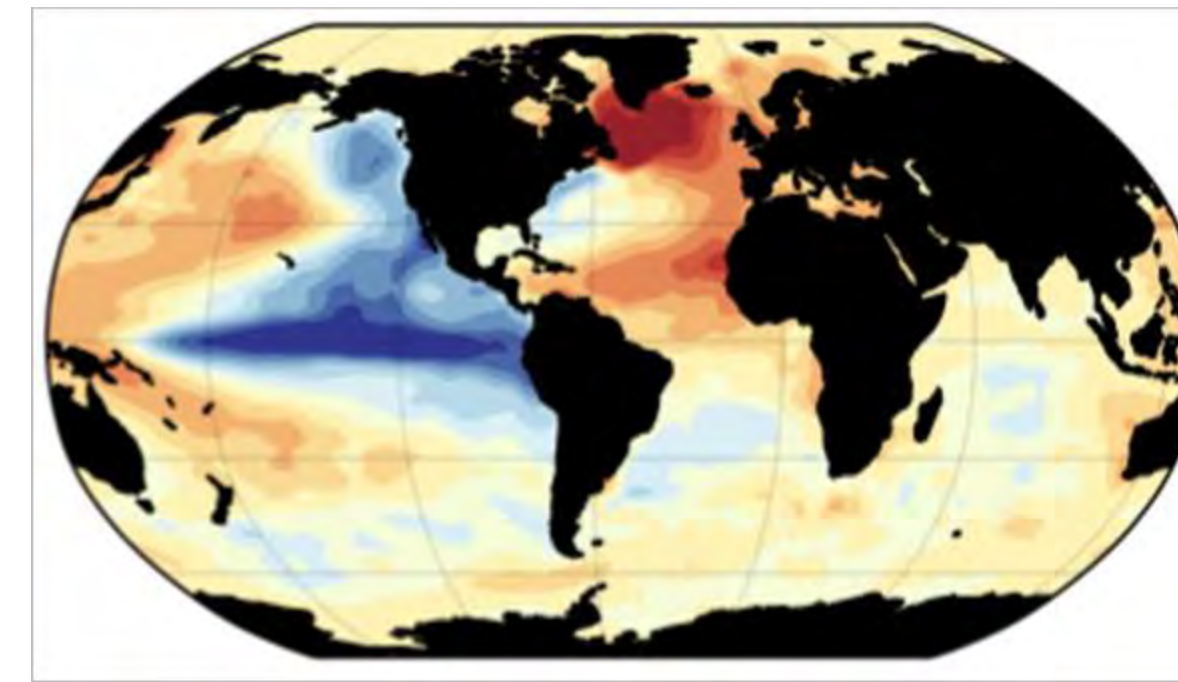
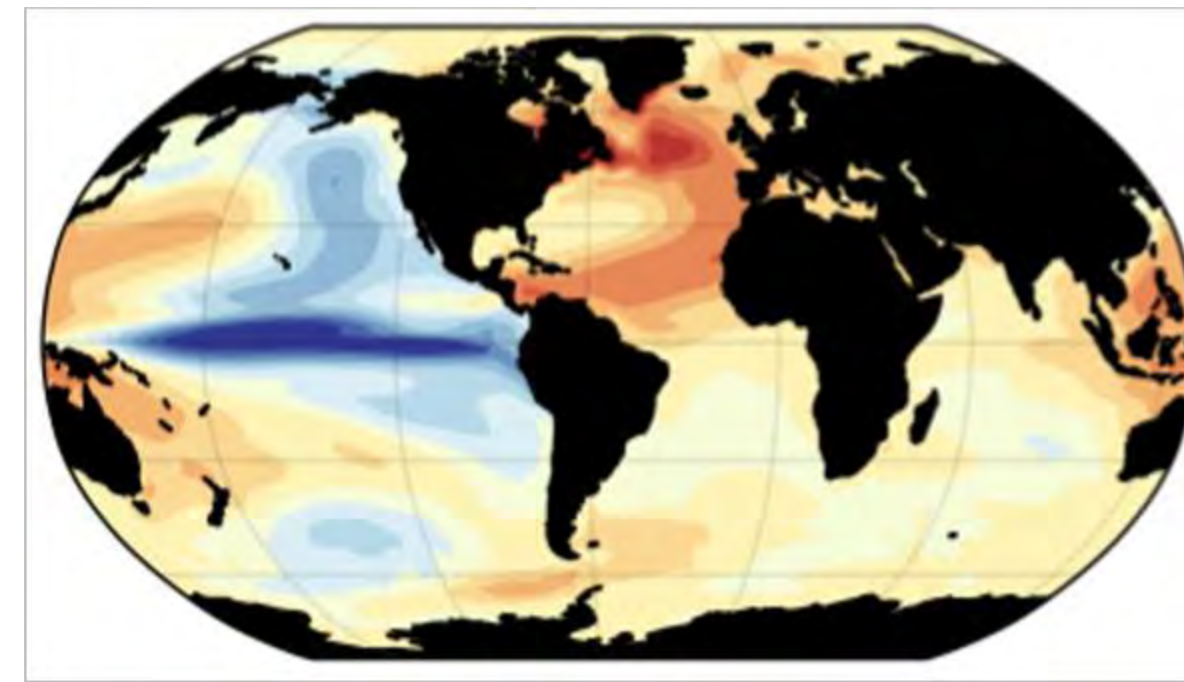
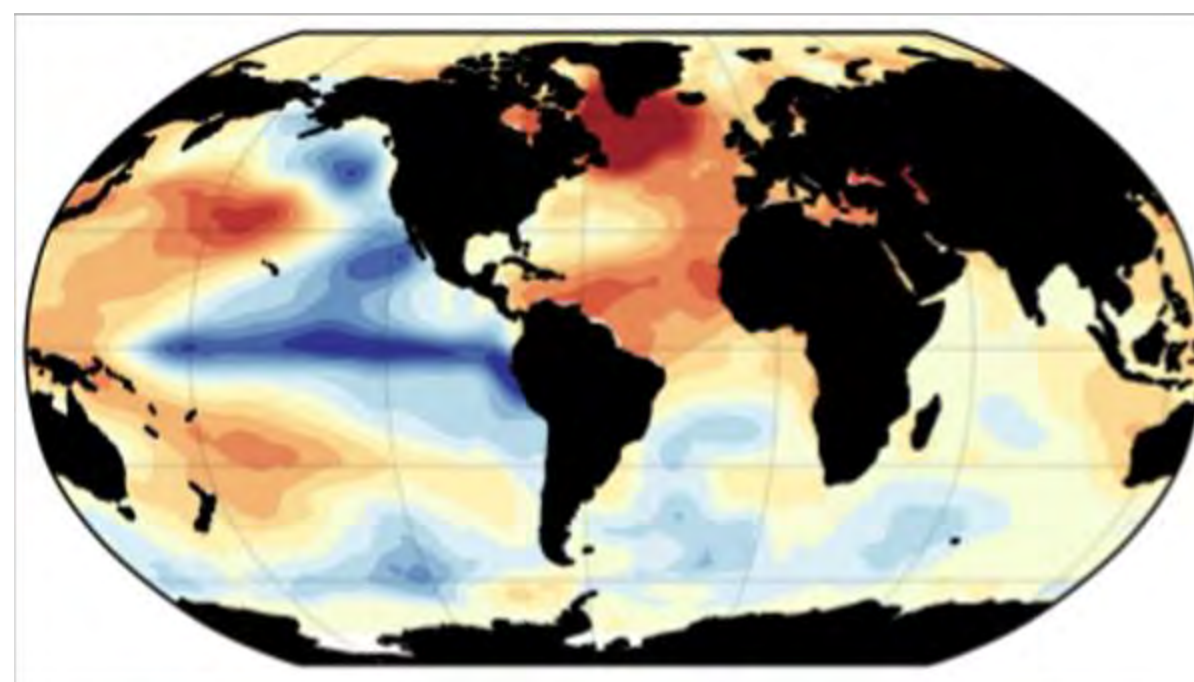
The AMV Index



In many cases, the externally forced responses tend to reinforce the teleconnection of internal components of modes of variability.

The model's systematic biases in the modes of variability and their teleconnection still hinder our prediction capability.

The iAMV Index



Analysis Period: 1961-2020



Regression (°C)

Modes of Variability as Important Sources for Climate Predictability

The AMV and iAMV Teleconnection: Sea Level Pressure

OBS

Only 1 sample

LENS2

Average of
50 samples

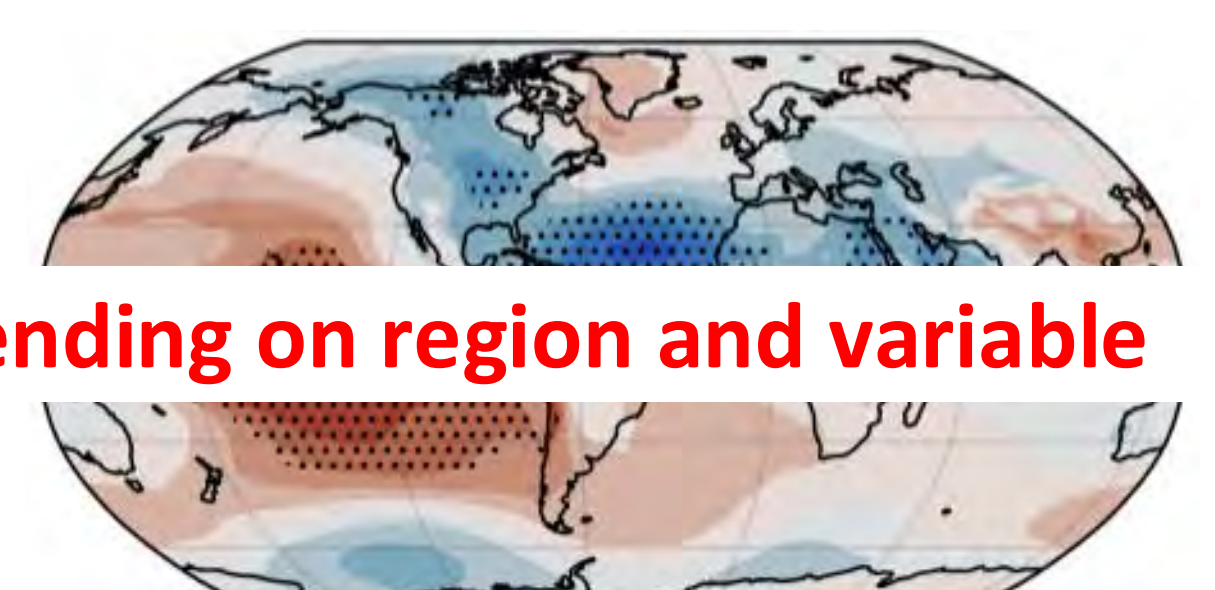
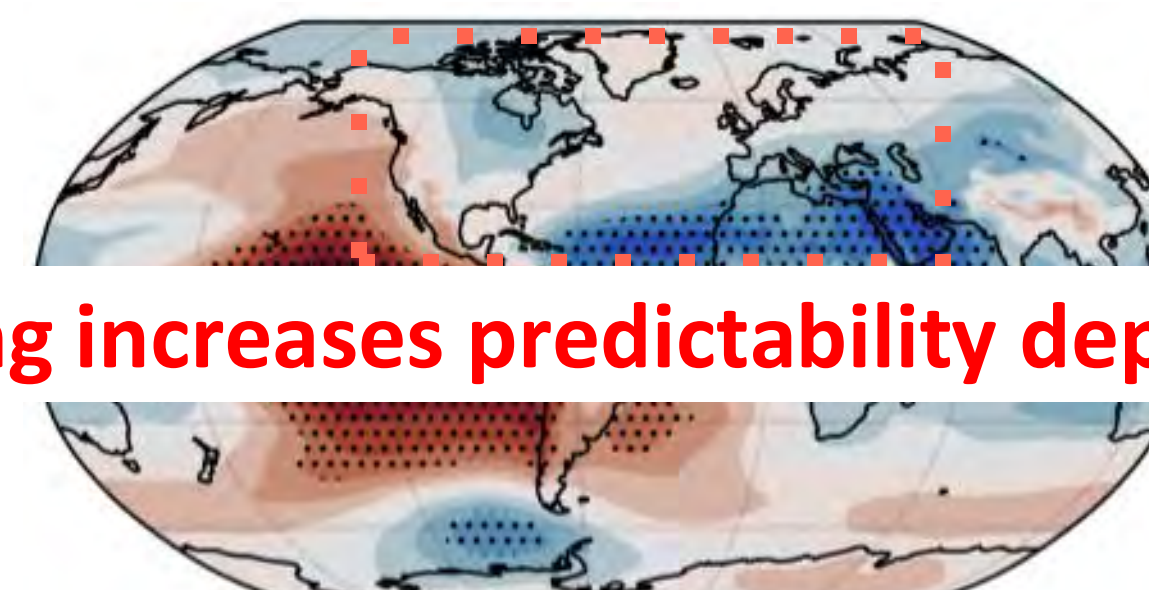
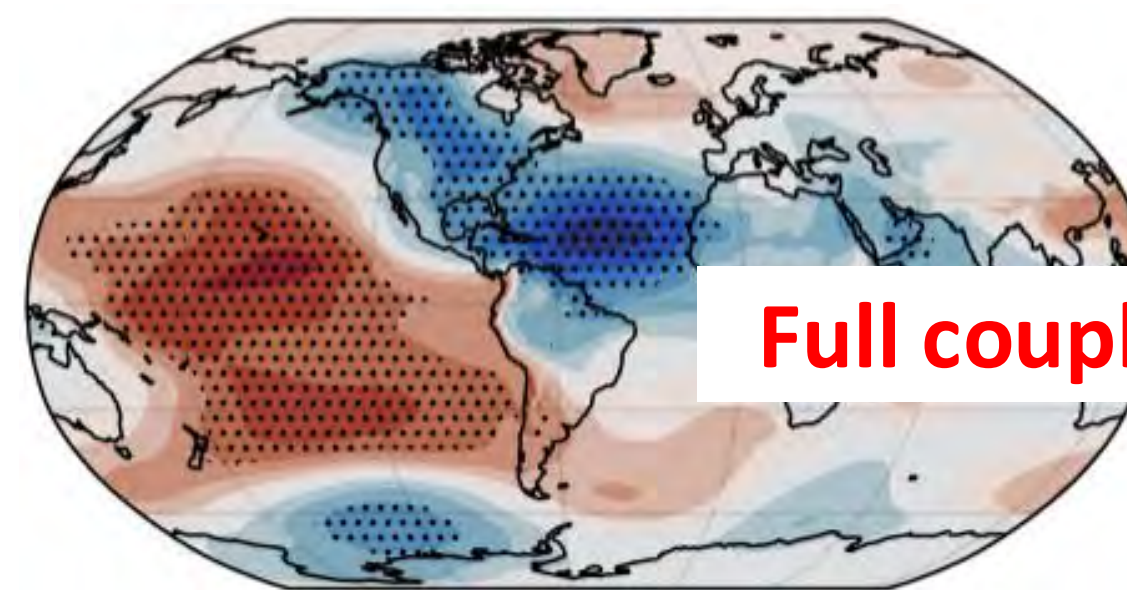
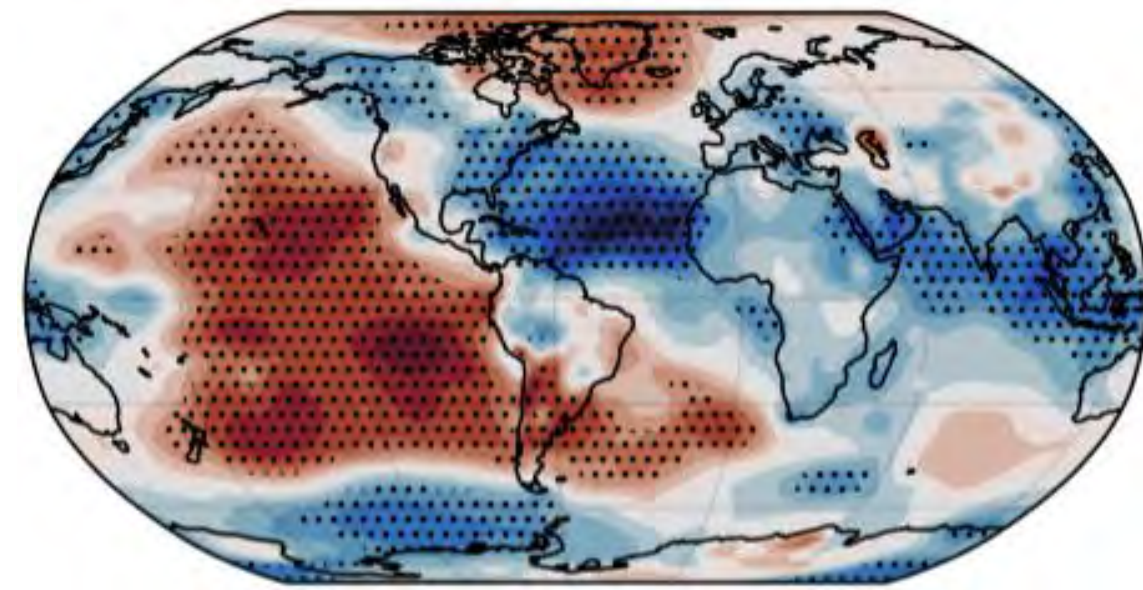
ODA

Average of
20 samples

1-yr HCST

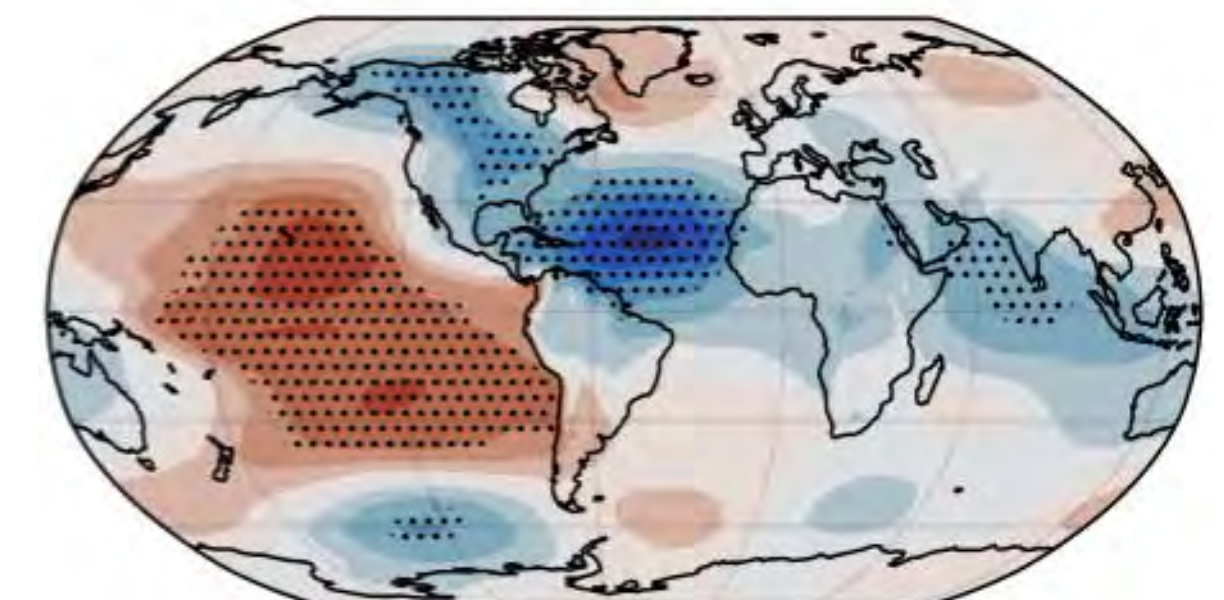
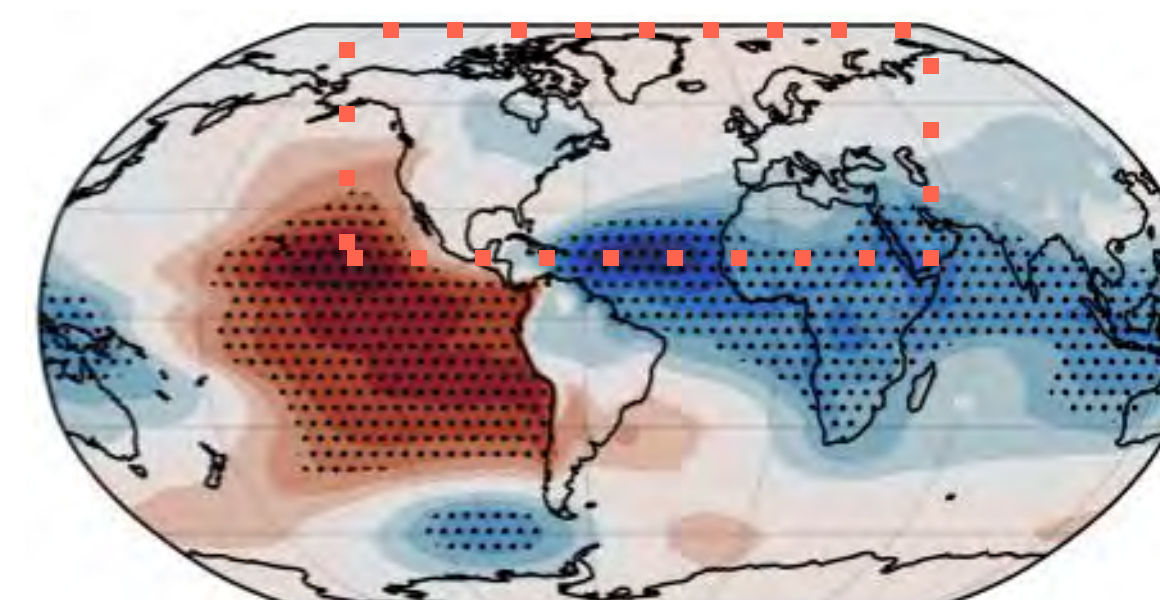
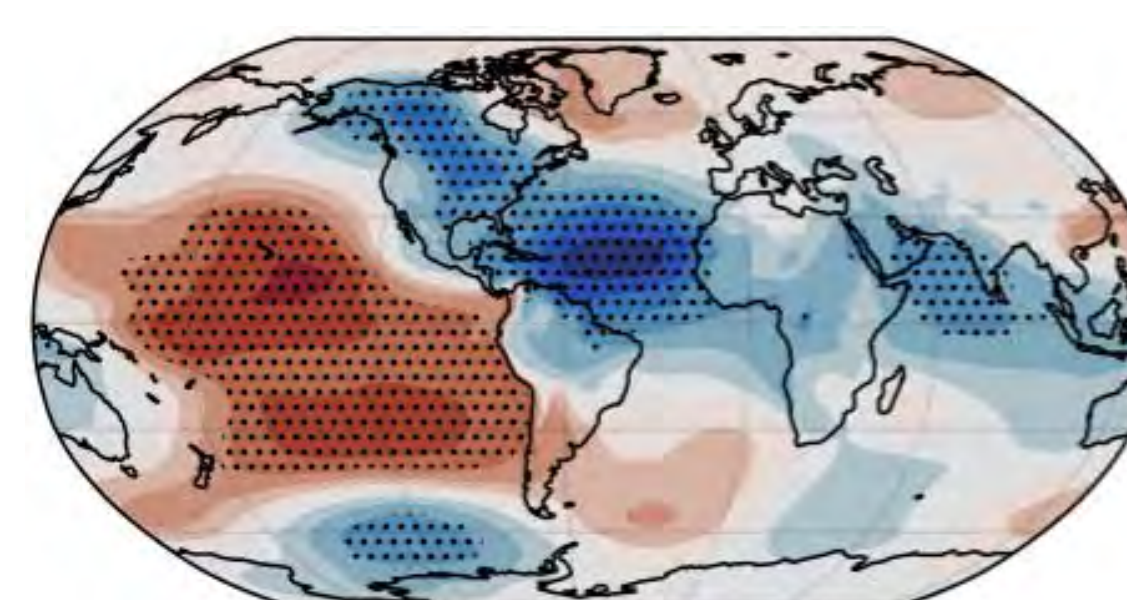
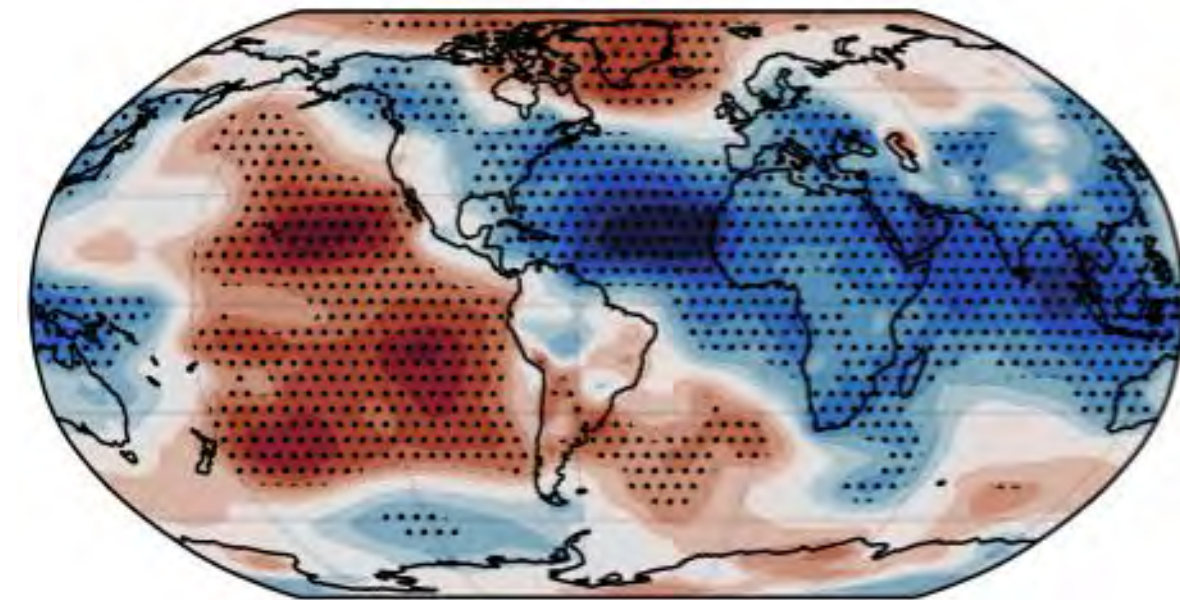
Average of
20 samples

The AMV Index

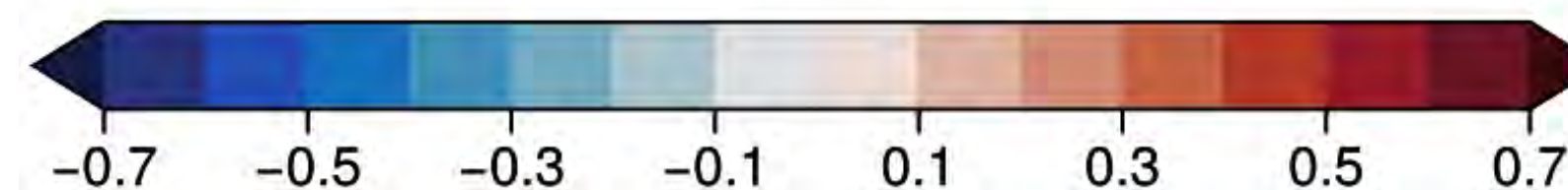


Full coupling increases predictability depending on region and variable

The iAMV Index



Analysis Period: 1961-2020



Correlation Coefficient

Modes of Variability as Important Sources for Wildfire Occurrences

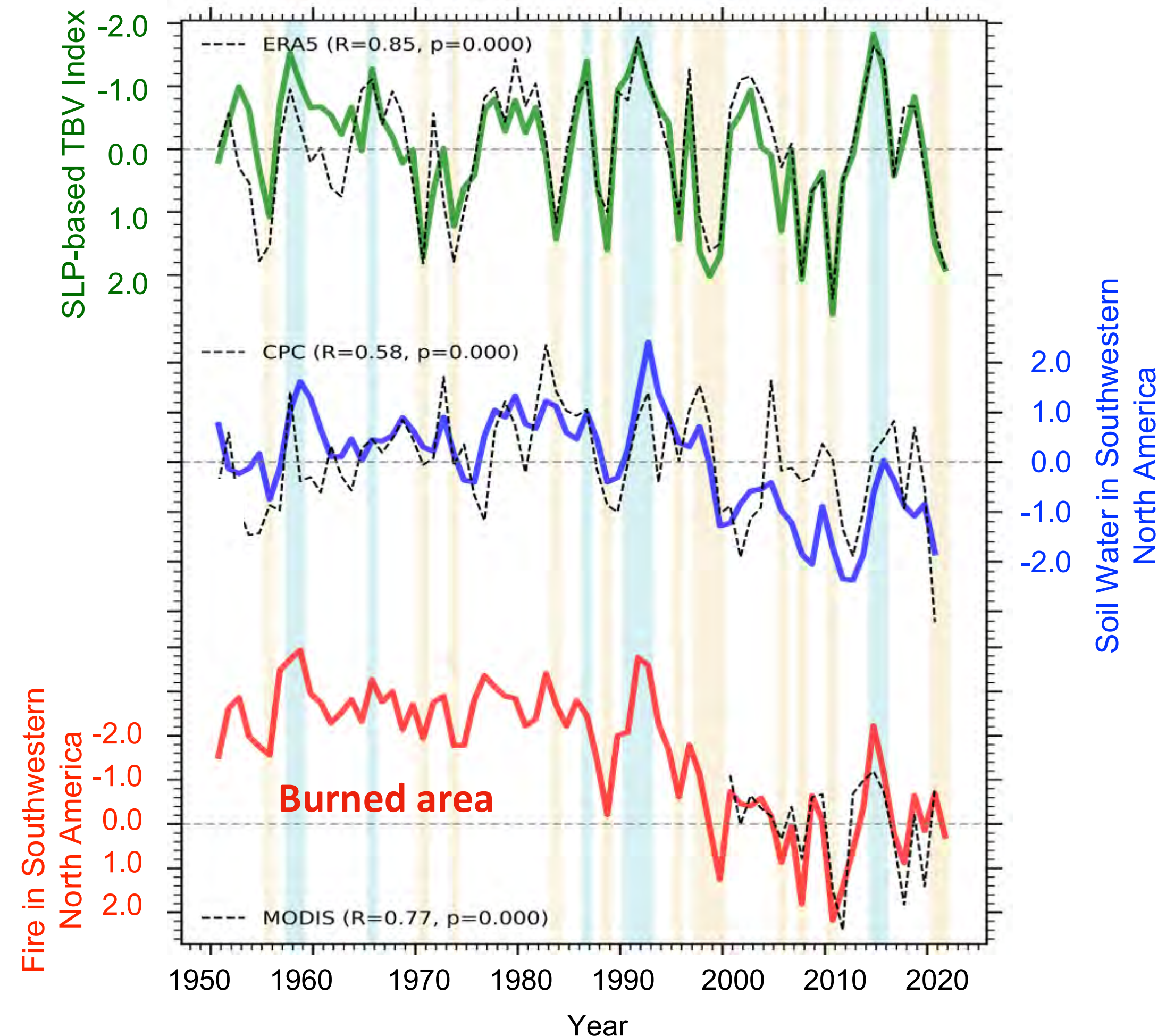
Trans-basin variability (TBV) between the Pacific and Atlantic sea surface temperature as one of key sources for multi-year predictability of drought and wildfire in Southwestern North America.



Maxar shortwave infrared closer satellite image of burning buildings in Altadena, California, on Jan. 8, 2025. Credit: Satellite image (c) 2025 Maxar Technologies via Getty Images

Maxar shortwave infrared closer satellite image of burning buildings in Altadena, California, on Jan. 8, 2025
(Credit: Satellite image (c) 2025 Maxar Technologies via Getty Images)

CESM2-based prediction system



**Oral
Session A-2**



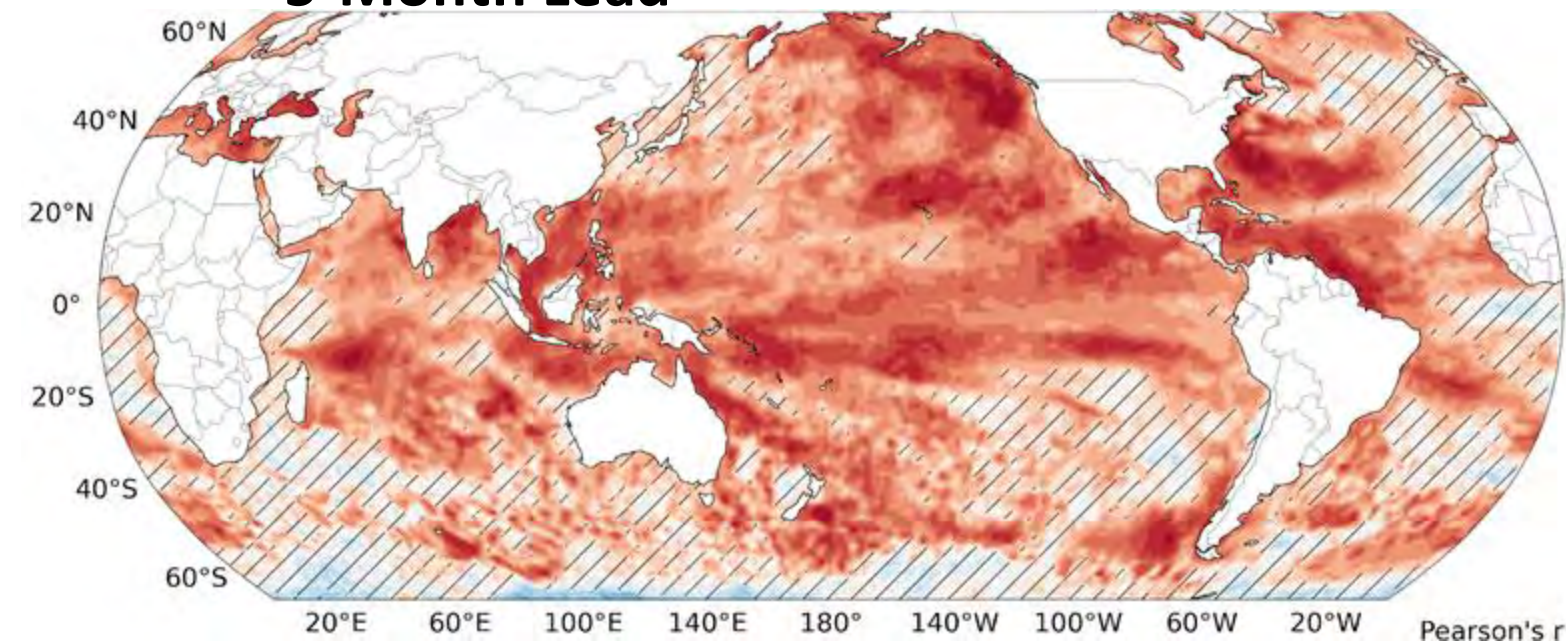
Modes of Variability as Important Sources for Marine Heat Wave Predictability

Prediction of Marine Heat Wave Days during JJA/ 1981-2020

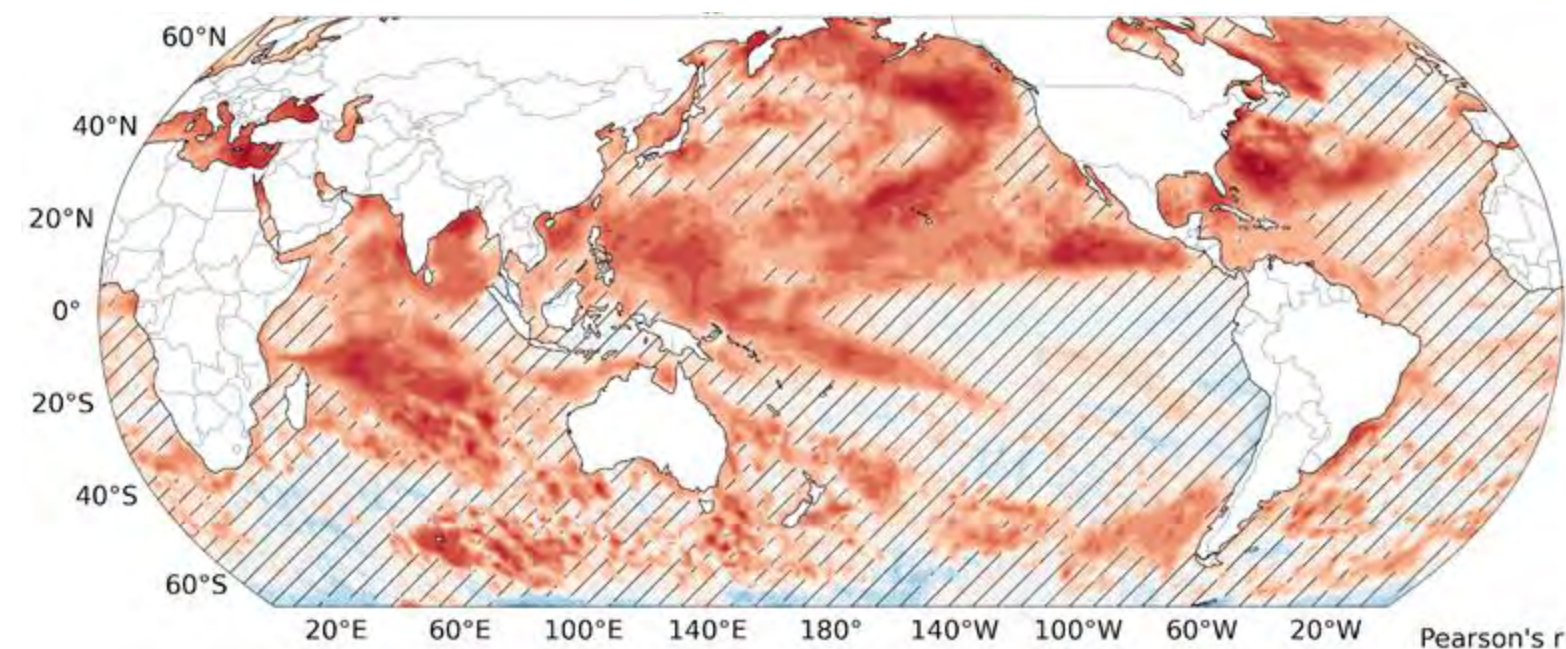
Total ACC

Skill

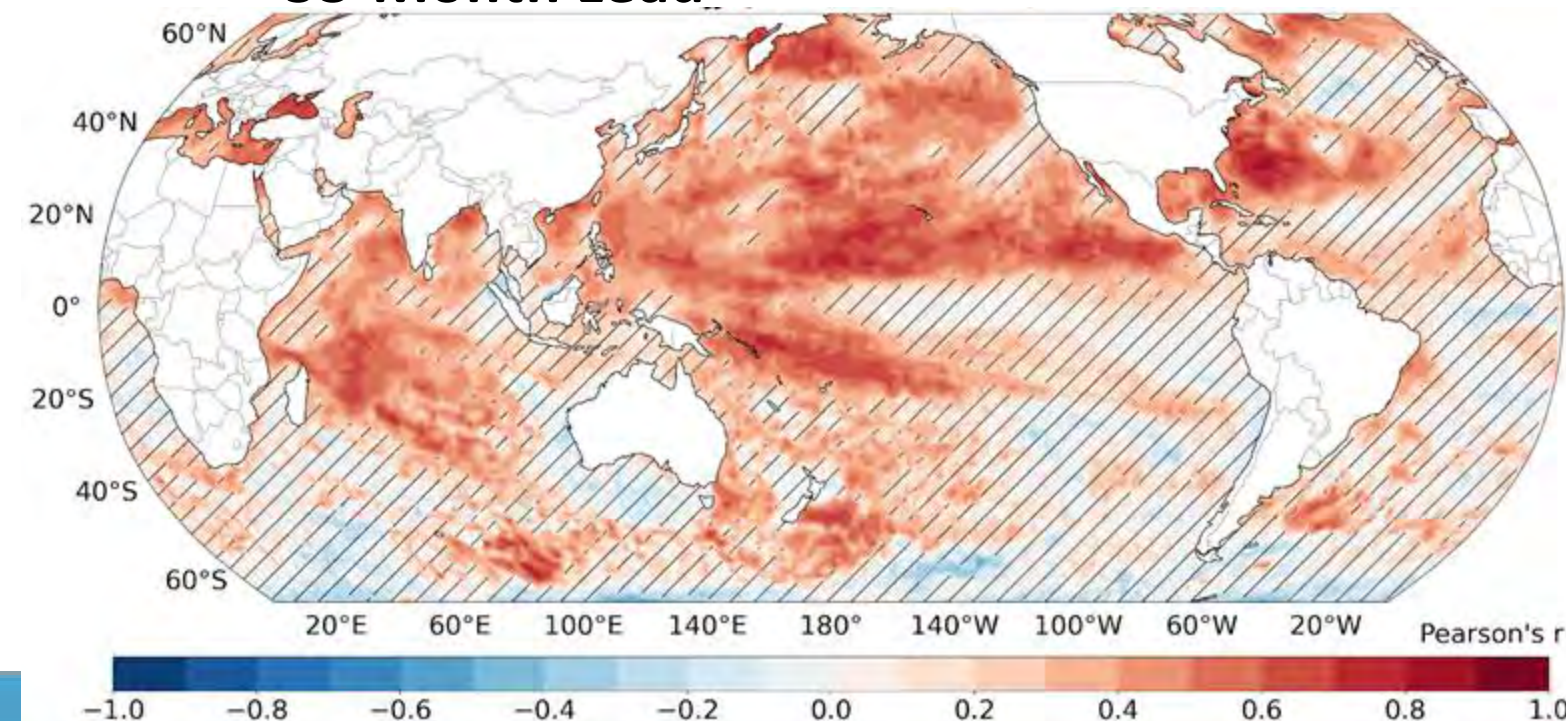
5-Month Lead



17-Month Lead

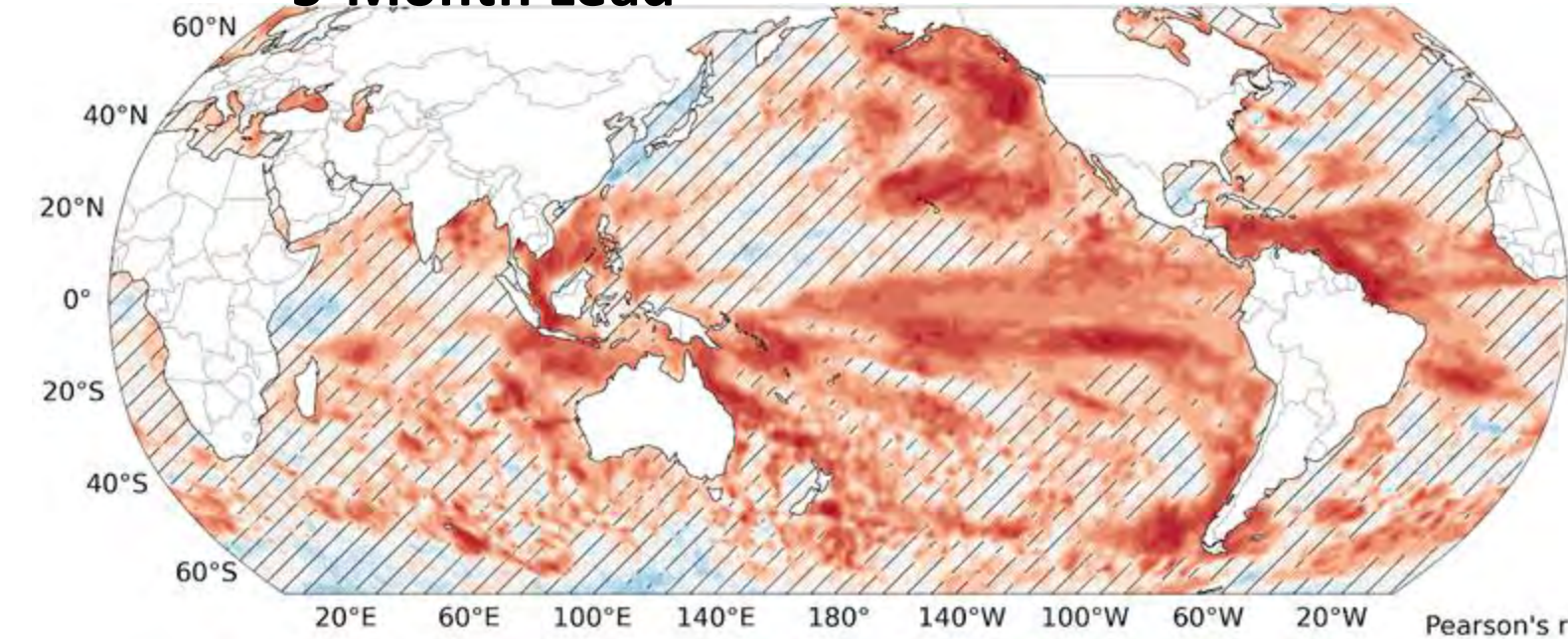


53-Month Lead

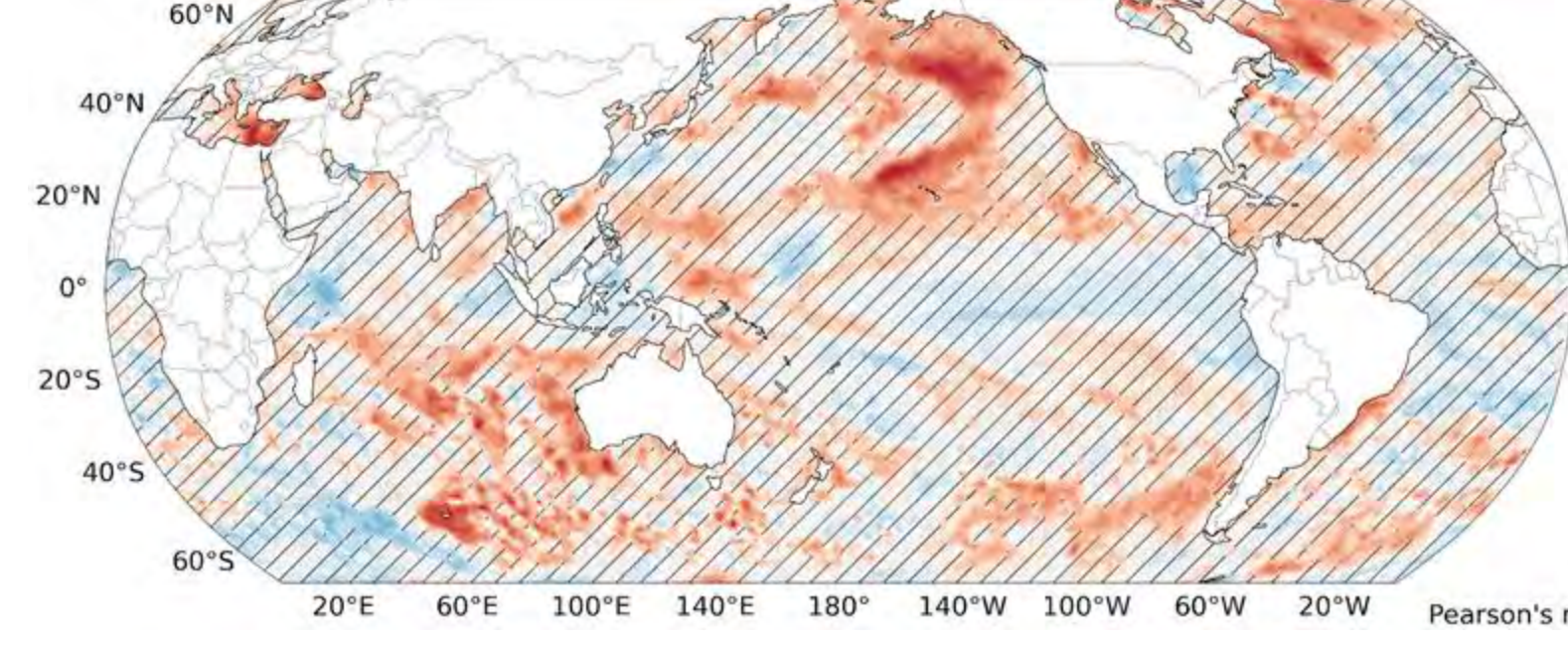


ACC Skill Removing External Forced Response

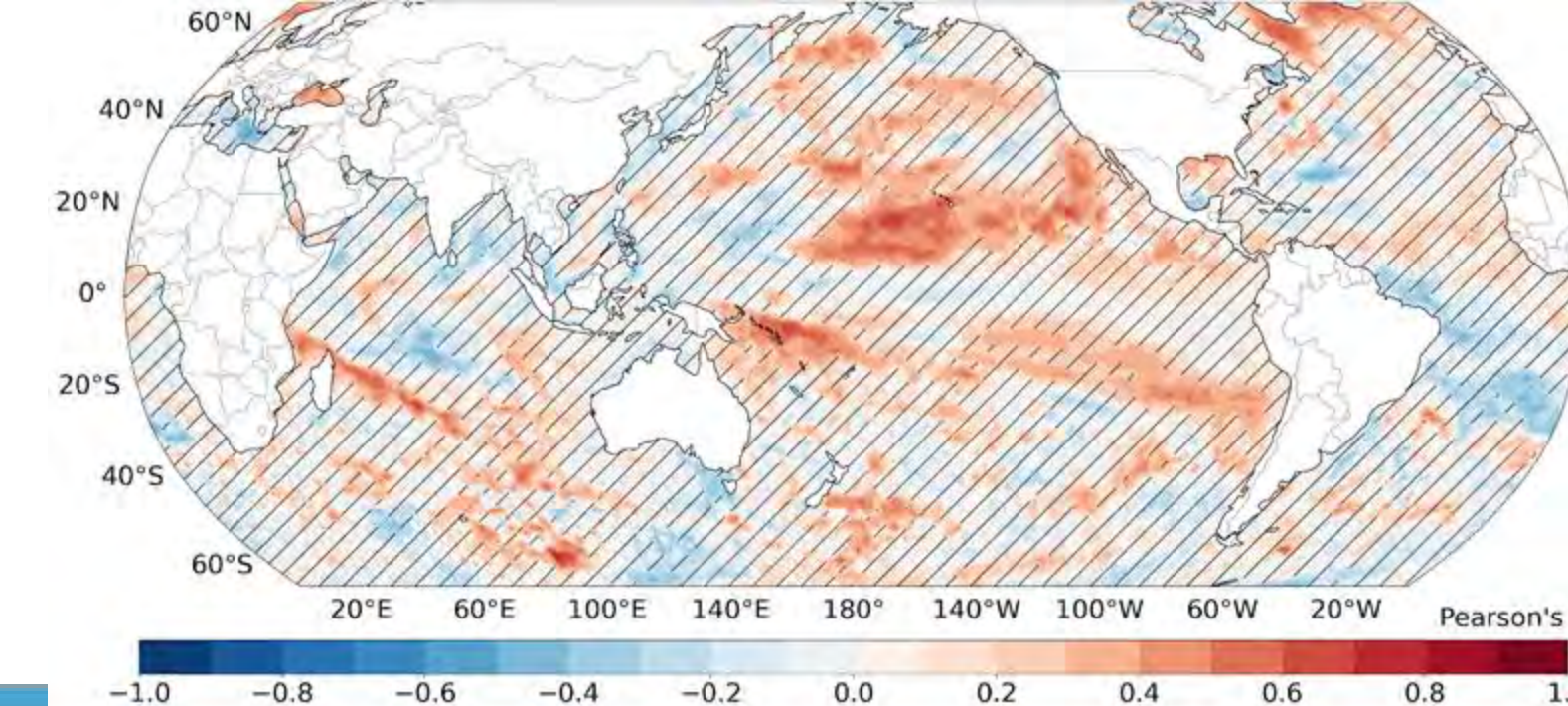
5-Month Lead



17-Month Lead



53-Month Lead



Oral
Session B-4
for terrestrial heat wave
predictability



Summary and Discussion

- A new seasonal-to-multiyear Earth prediction system has been developed based on the Community Earth System Model version 2 (CESM2). A 20-member ensemble which assimilates ocean temperature and salinity anomalies provides the initial conditions for 5-year prediction from 1960 to 2020.
- The CESM2-based Multi-year Prediction System has a useful skill for ENSO up to 1 year, for TBV up to several years, and AMV for more than 5 years.
- Modes of variability, including ENSO, TBV, PDO and AMV, are essential sources of climate variability and predictability, including statistics of extremes, over many parts of the globe. The model's systematic biases in the modes of variability and their teleconnection still hinder our prediction capability.
- The anomaly ocean data assimilation is an effective way to calibrate the pattern of modes of variability in the climate model.
- In many cases, the externally forced responses tend to reinforce the teleconnection of internal components of modes of variability.
- Full coupling of ocean, atmosphere, and land tends to increase predictability depending on region and variables.





Thank You!