

CL5.3.4 – Predictions of climate from seasonal to (multi) decadal timescales (S2D) and their applications

Predictable Patterns of Winter Temperature and Their Predictability Sources

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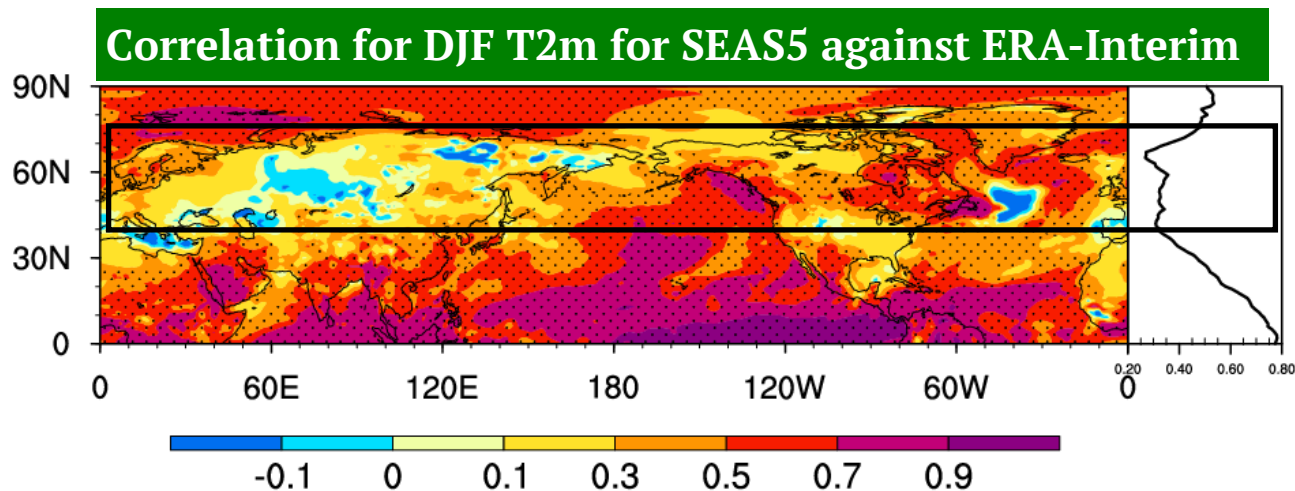
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1. Low Predictability over Mid-latitude Eurasia



➤ Atmospheric external forcing has been suggested to play a role: **ENSO**, **snow cover** and **sea ice** (*Ineson and Scaife 2008; Cohen 2014; Zuo 2011*)

➤ **Question**

Can we improve the prediction of winter temperature (T2m) via predictable patterns?

(*Fan 2020*)

2. Maximize the Signal-to-noise Ratio to Extract Predictable Patterns

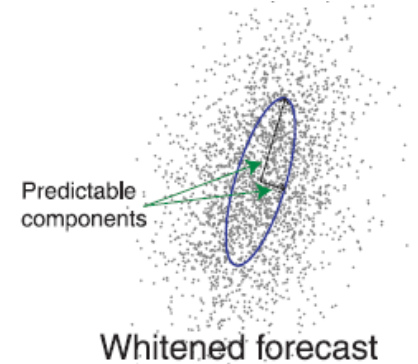
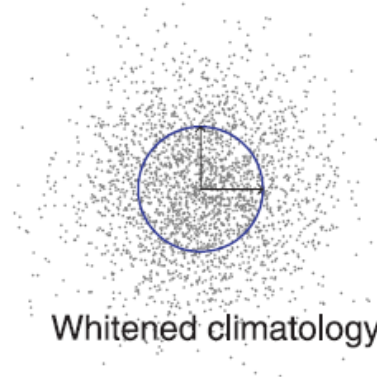
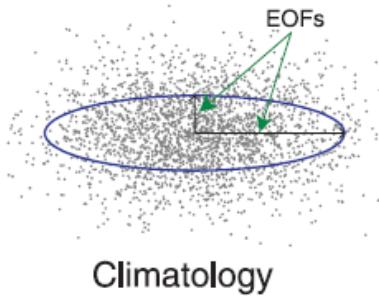
Internal variation



White noise



Predictable patterns

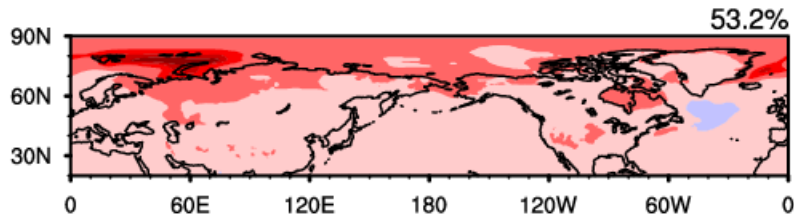


Maximum signal-to-noise EOF analysis (Allen and Smith 1997; Schneider and Griffies 1999; Timothy 2006)

3. Predictable Patterns and Their Possible Atmospheric External Forcing

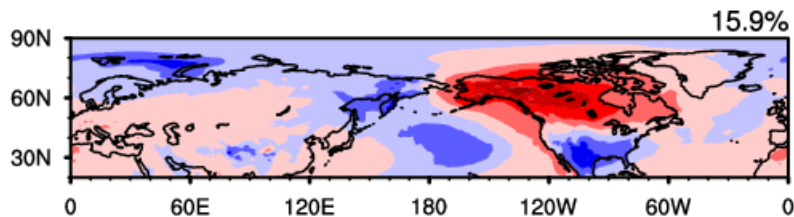
Predictable patterns

MEOF1



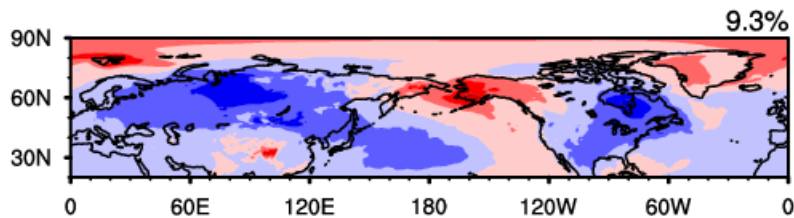
Trend

MEOF2



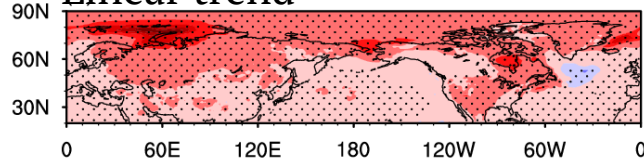
EP ENSO

MEOF3

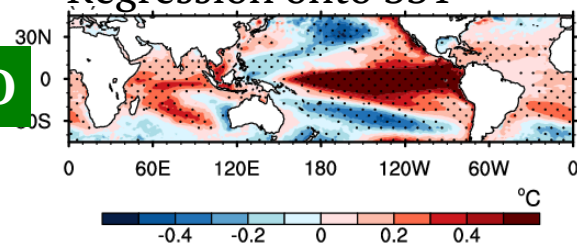


Predictability sources

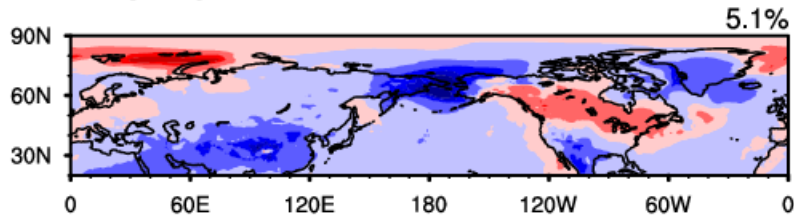
Linear trend



Regression onto SST



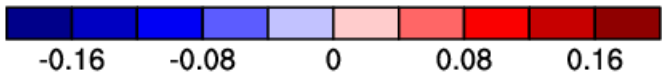
MEOF4



CP ENSO

Eurasia Snow

Arctic Sea Ice



4. Scheme to Improve the Predictive Skill

Predictable patterns



Projecting

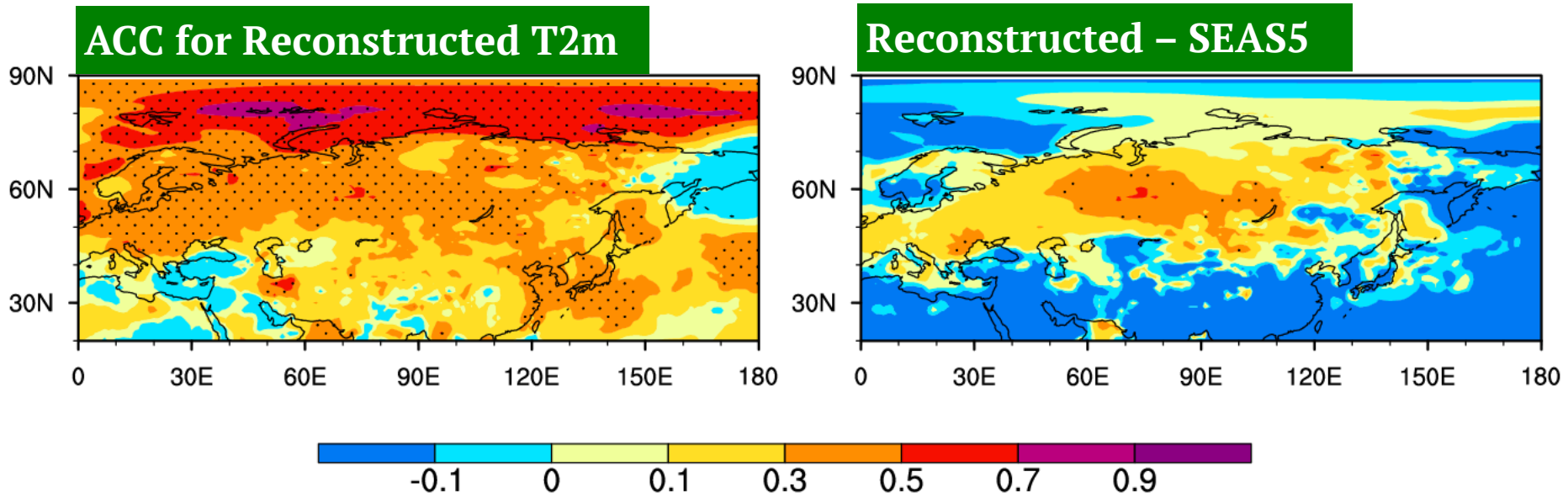
Time series of predictability sources

>> 4. Scheme to Improve the Predictive Skill

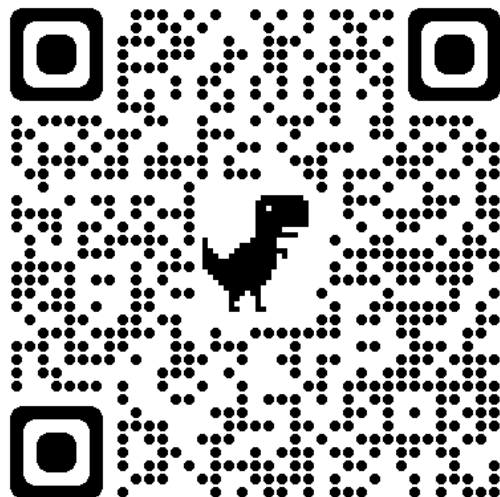
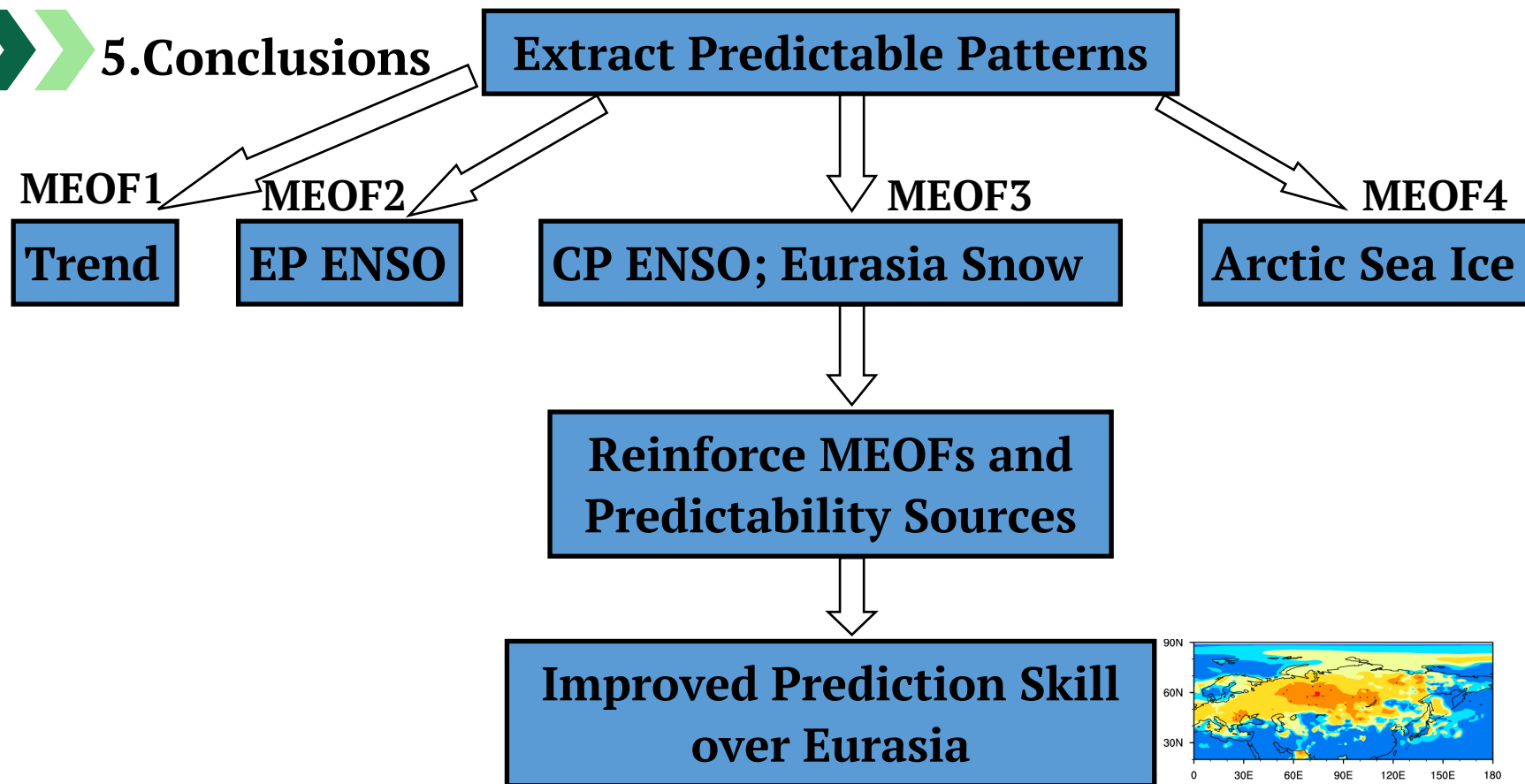
Predictable patterns

↓ Projecting

Time series of predictability sources



5. Conclusions



This presentation participates in OSPP



Outstanding Student & PhD
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Thanks for listening!

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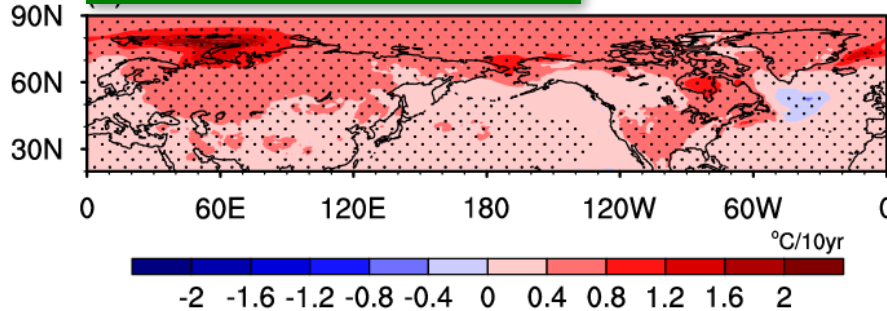
Fan, H. et al., 2020. *J. Clim.*, 33, 10743-10754.

Supplement

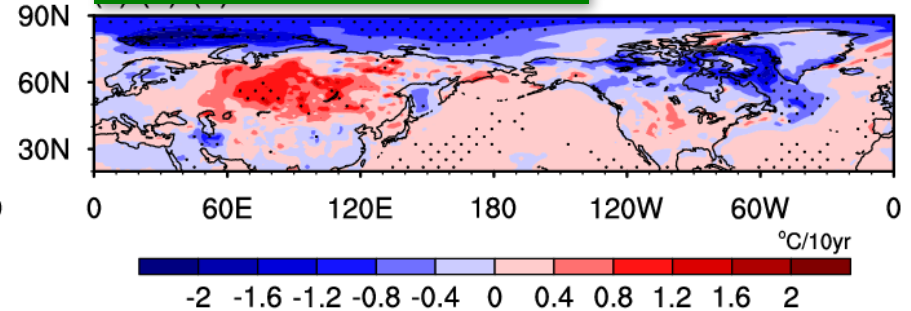
3. Predictability sources of the leading predictable patterns

a) MSN EOF1

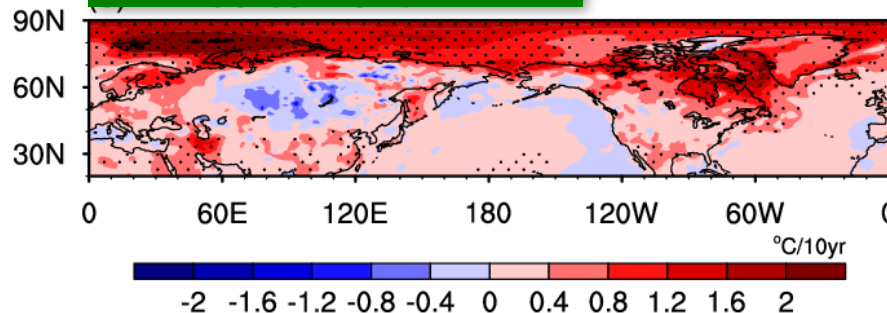
T2m trends in SEAS5



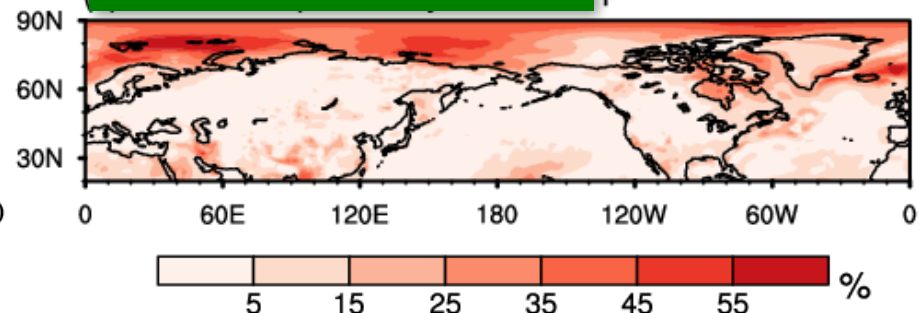
SEAS5 - OBS



T2m trends in OBS



Explained Variance



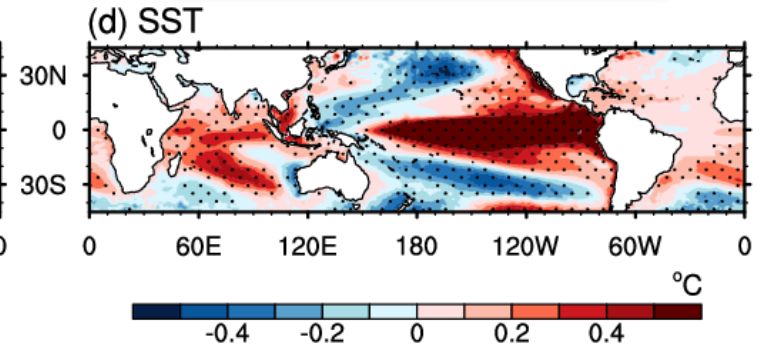
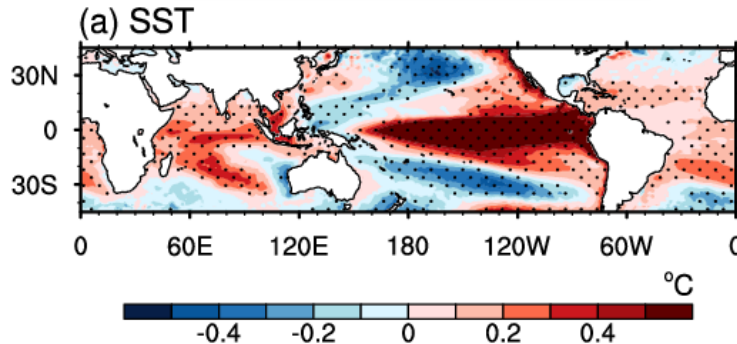
SEAS5 underestimates the observed magnitude of the warming trend and misses the observed warm Arctic-old continent pattern.

b) MSN EOF2

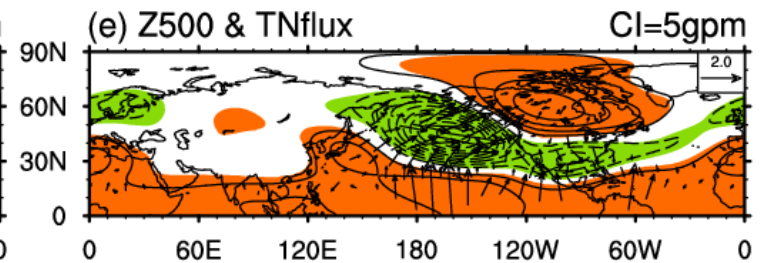
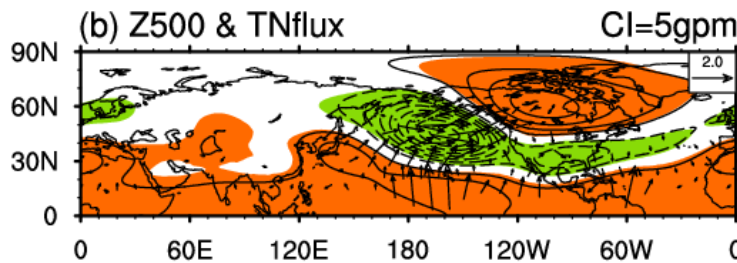
Regression onto MPC2

Regression onto PC1_{SST}

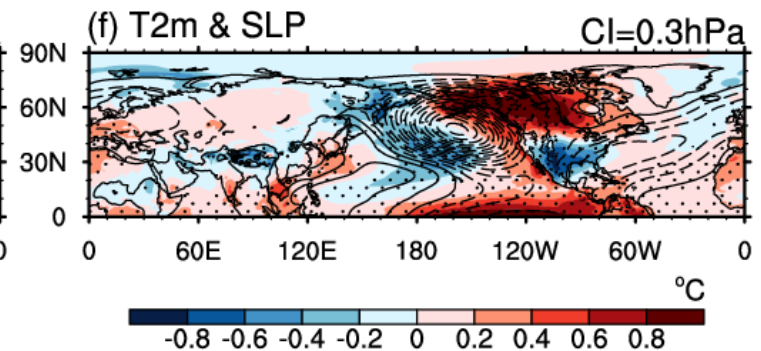
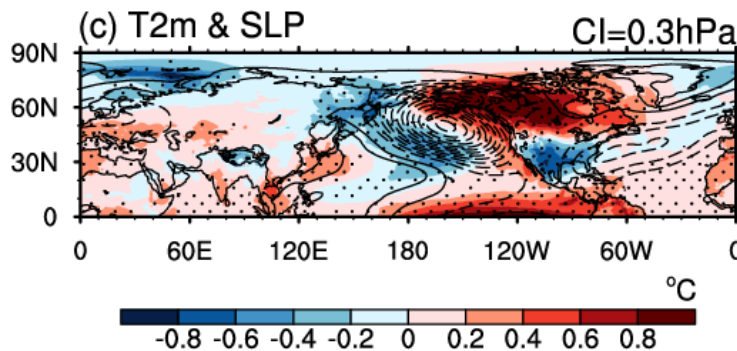
SST



Z500 & TNflux



T2m & SLP



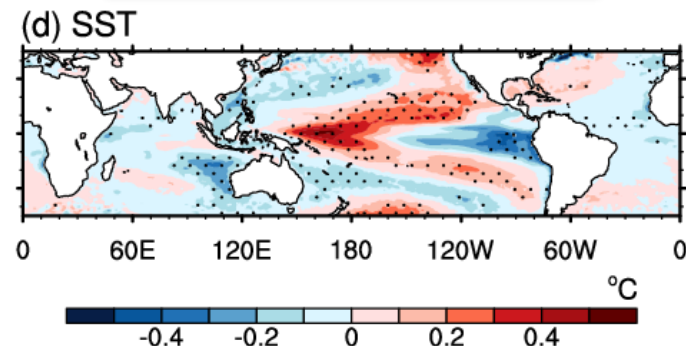
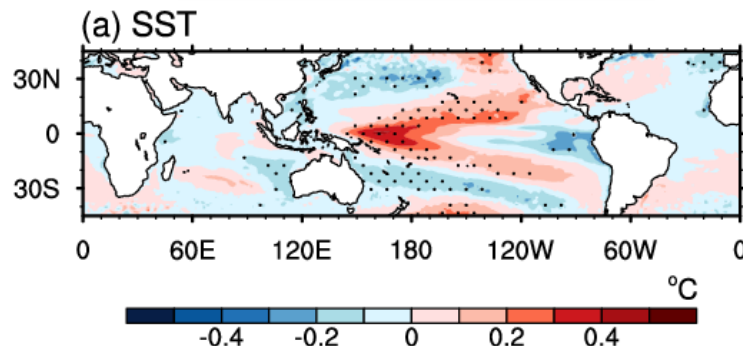
The MSN EOF2 is rooted in the tropical forcing for the eastern Pacific-type ENSO

c) MSN EOF3

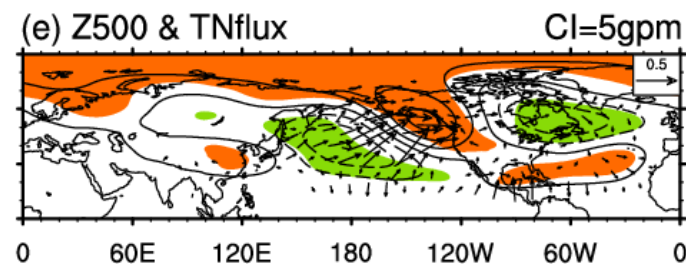
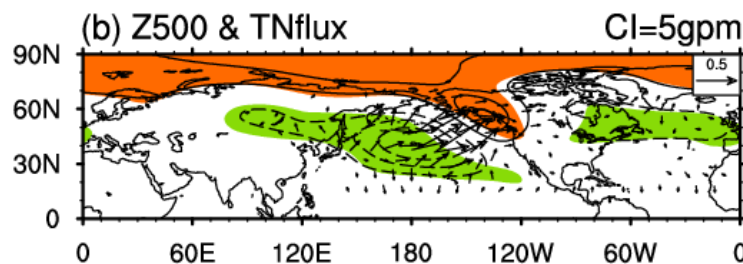
Regression onto MPC3

Regression onto PC2_{SST}

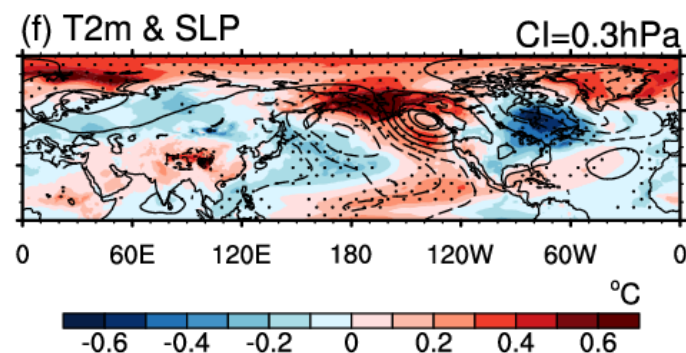
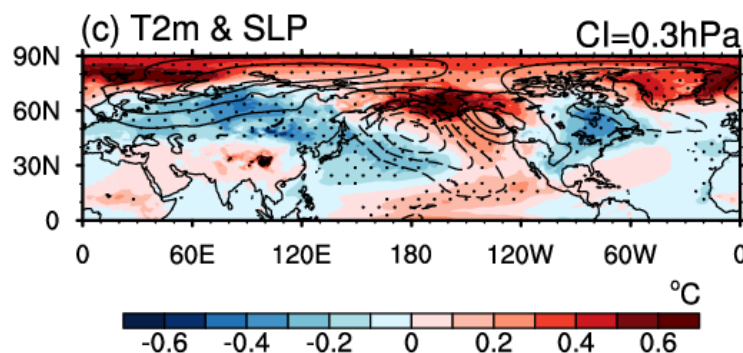
SST



Z500 & TNflux



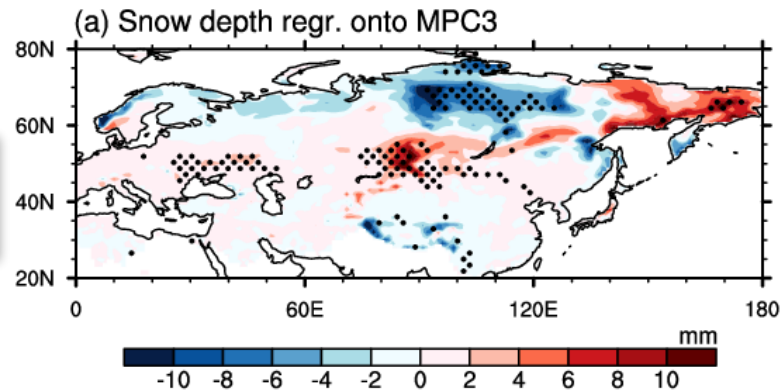
T2m & SLP



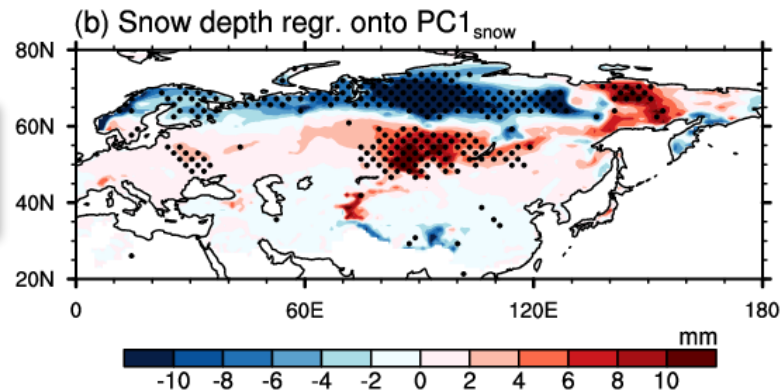
Source of predictability: central Pacific-type ENSO

c) MSN EOF3

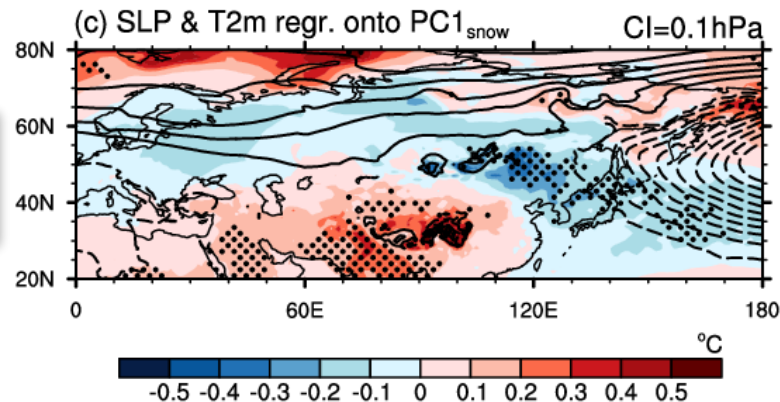
Snow regression
onto MPC3



Snow regression
onto PC1_{snow}



SLP & T2m regression
onto PC1_{snow}



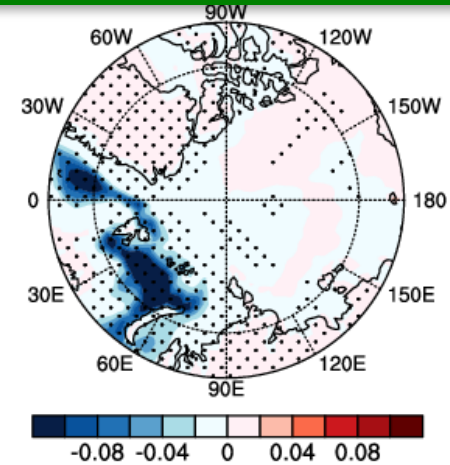
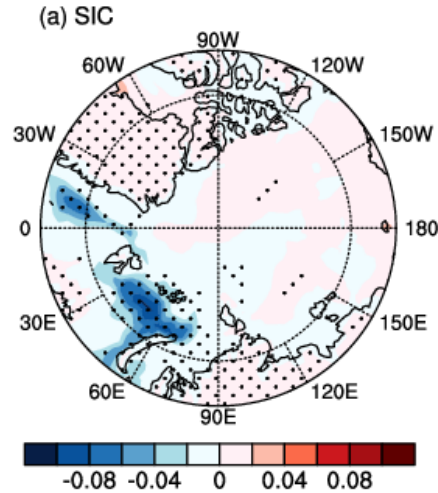
Source of predictability: Eurasian snow depth

>>> d) MSN EOF4

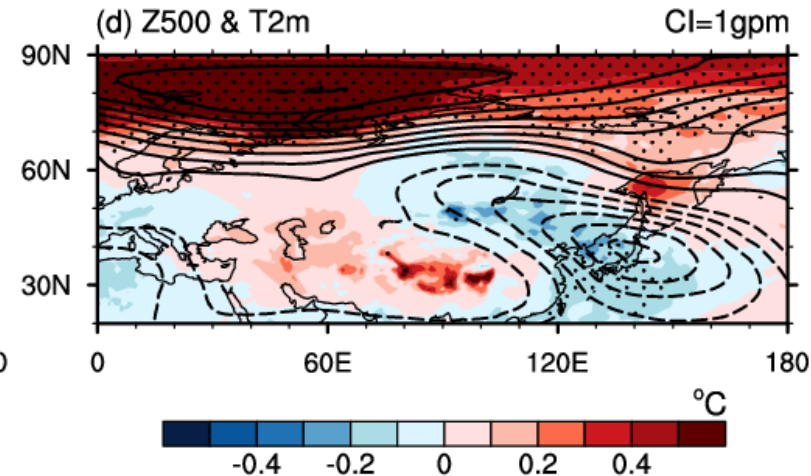
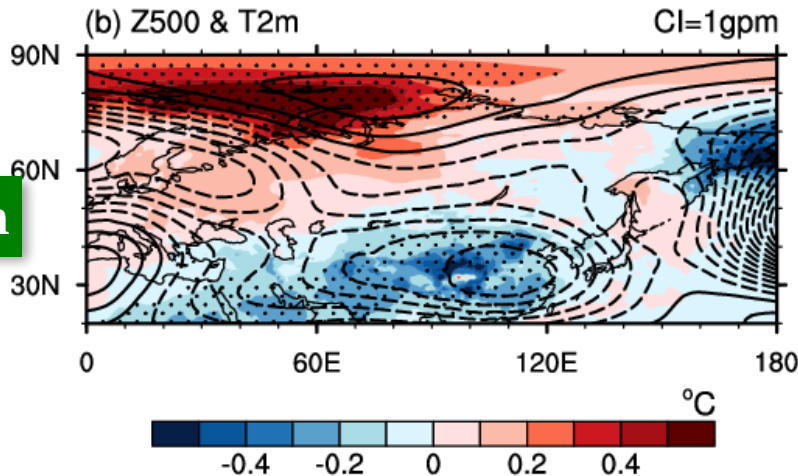
Regression onto MPC4

Regression onto PC1_{SIC}

SIC



Z500 & T2m



Partially traced to the Arctic sea ice in the Barents and Norwegian Seas with some uncertainty

3. Predictability sources of the leading predictable patterns

5% and 1% significance levels based on two-tailed Student's t -test are denoted by * and **, respectively.

	Trend	PC1 _{SIC}
MPC1	0.98**	-0.40*

	PC1 _{SST}	PC1 _{SIC}
MPC2	0.87**	0.33*

	PC2 _{SST}	PC1 _{Snow}	PC1 _{SIC}
MPC3	0.75**	0.46**	-0.40*

	PC1 _{SIC}
MPC4	-0.51**

Tropical Pacific sst : 30°S-30°N, 110°E-70°W
Arctic sea sic : 70°N-89°N
Eurasian snow depth: 20°E-130°E, 45°N-75°N

4. Improved predictions over Eurasia by incorporating the predictable patterns

