Nov 2017, PAG meeting



JNU-KOPRI seasonal prediction models for Arctic sea-ice based on statistical and dynamical method

Jee-Hoon Jeong, Yoo-Geun Ham Chonnam National University

Baek-Min Kim KOPRI

Jun 14 20

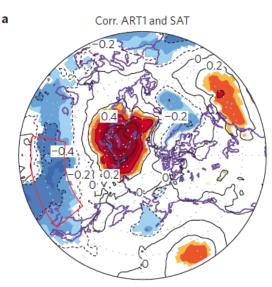
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Practical needs for the Arctic sea ice prediction

- Seasonal climate prediction
 - Anomalous pattern of Artic sea ice concentration is a useful predictor of East Asia winter climate
 - Low September SIC over the Barents-Kara Sea is known to lead cold winters, cold waves in East Asia

- Expectation for the Arctic route
 - Korean Government asks for seasonal prediction of Arctic sea ice
 - Needs from shipping companies



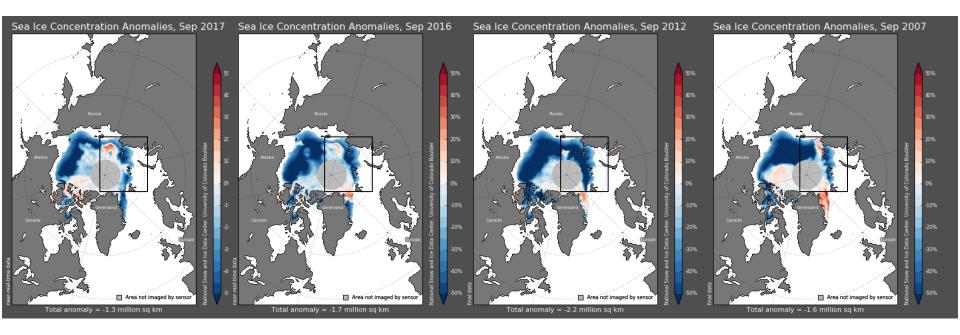
Kug et al. 2015 (Nat. Geo)



Large year-to-year variation of regional sea-ice conditions



• The Arctic sea ice extent is decreasing continuously, and relatively well predicted a season beyond. However, it is a difficult task to predict regional sea ice conditions, which varies greatly year-to-year.

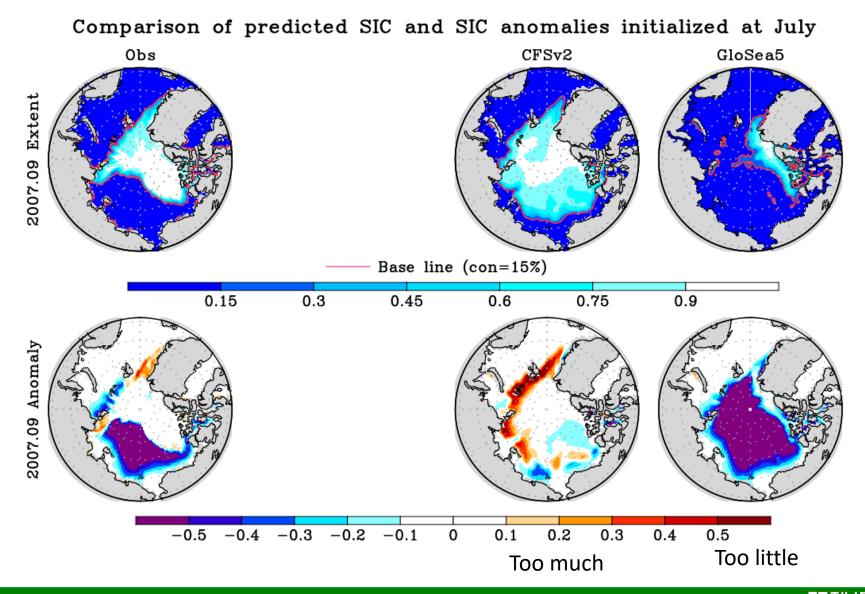


• Sea ice over a different region has a different atmospheric responses



Still, dynamical models, GCMs, have large bias: forecast example Sep. 2007

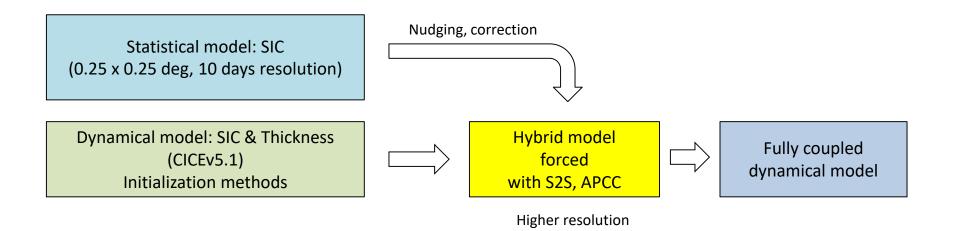




KOPRI-JNU's roadmap for Artic sea ice prediction



- At the moment, dynamical predictions have limitations for practical use, so we developed a statistical model only for forecasting sea ice concentration (SIC) (~'17), but
- are developing a dynamical prediction model as well for SIC and thickness based on offline sea ice model (~'18).
- A hybrid (dynamical model combined with statistical predictions) model will be developed (~'20).







STEP1:

Extracting major spatio-temporal variation patterns from historical SIC observations of SIC using Season-reliant EOF (SEOF)

STEP2:

Estimating the current state of SIC with observed SIC anomalies for last 12 months

STEP3:

Projecting the future evolution of SIC using S-EOFS and the current

state

STEP4:

Corrections with atmospheric circulation, surface radiation



Sea ice concentration anomalies over the Arctic, Monthly average forecasts up to 12 months, half degree

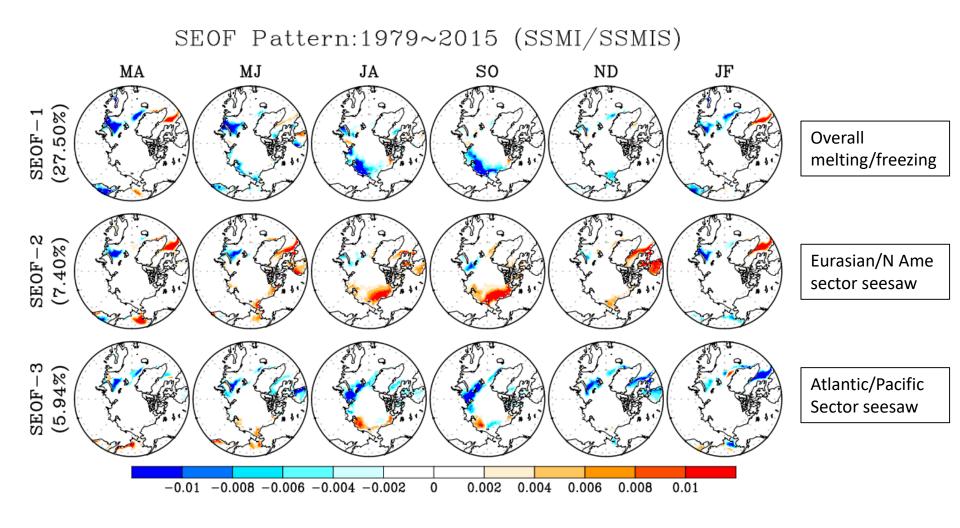




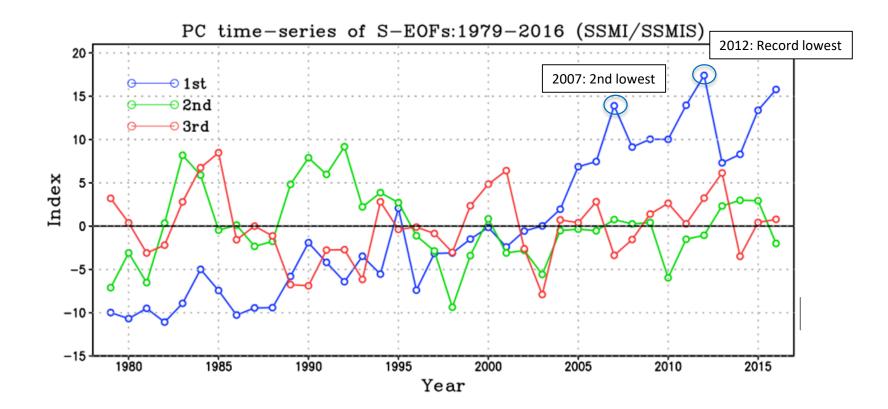
Extracting the spatio-temporal variation: Leading 3 Season-reliant EOFs of Arctic SIC



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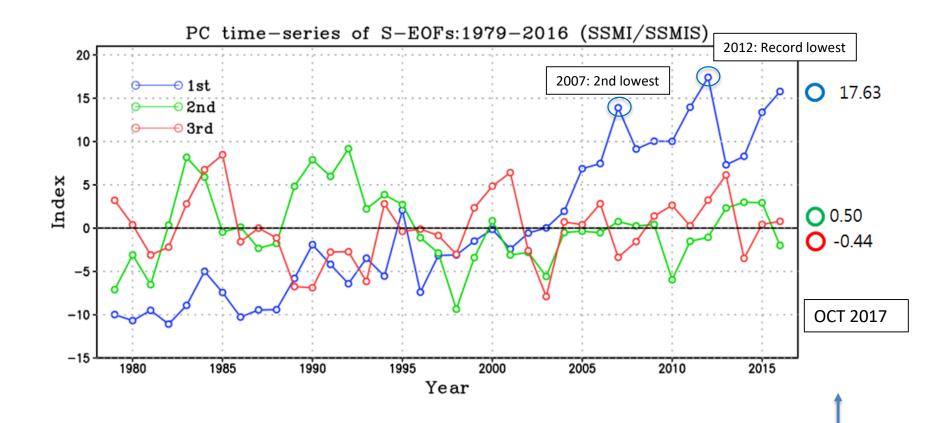










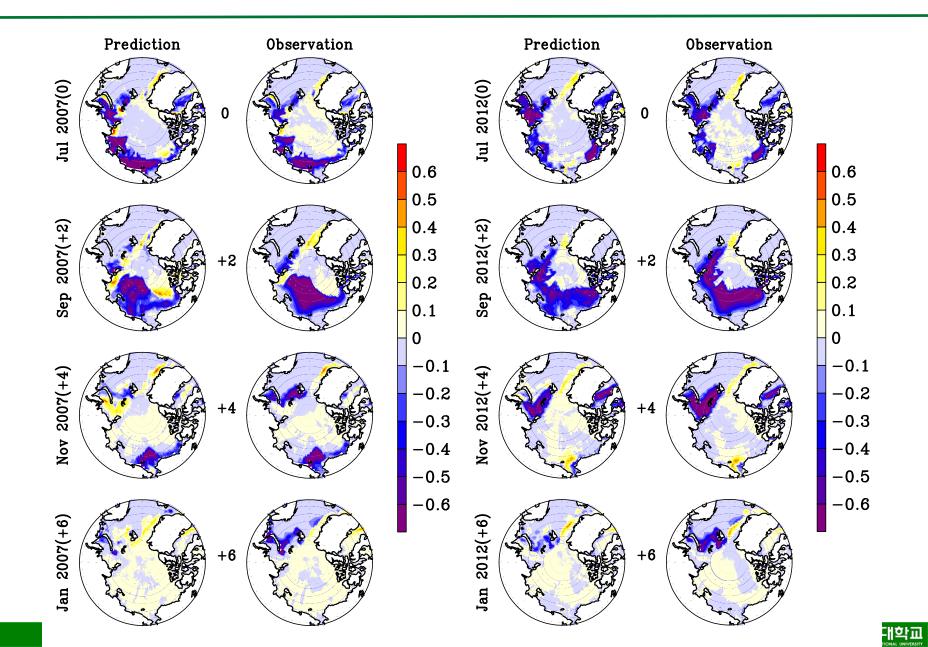


Estimating current state from the latest observations, and assuming it will persist for the next 12 months

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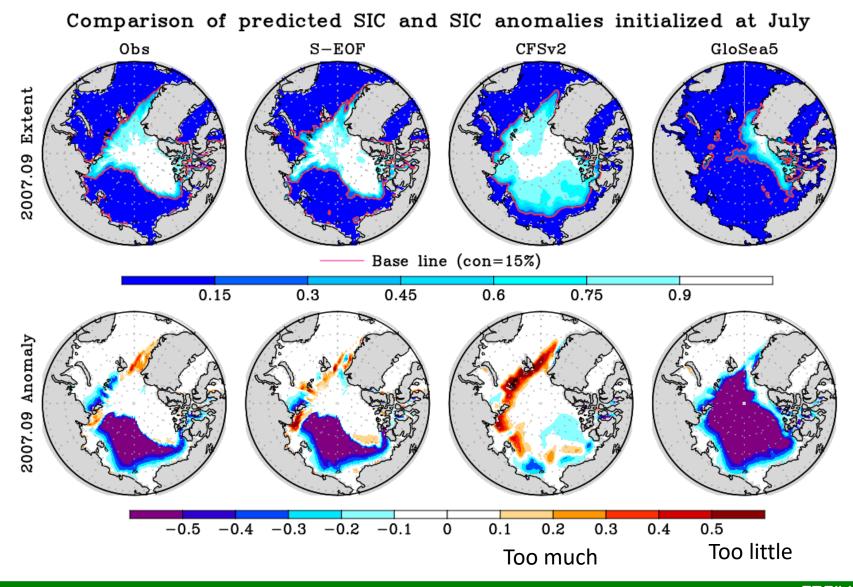
Forecast example: 2007, 2012





Still, dynamical models, GCMs, have large bias: forecast example Sep. 2007

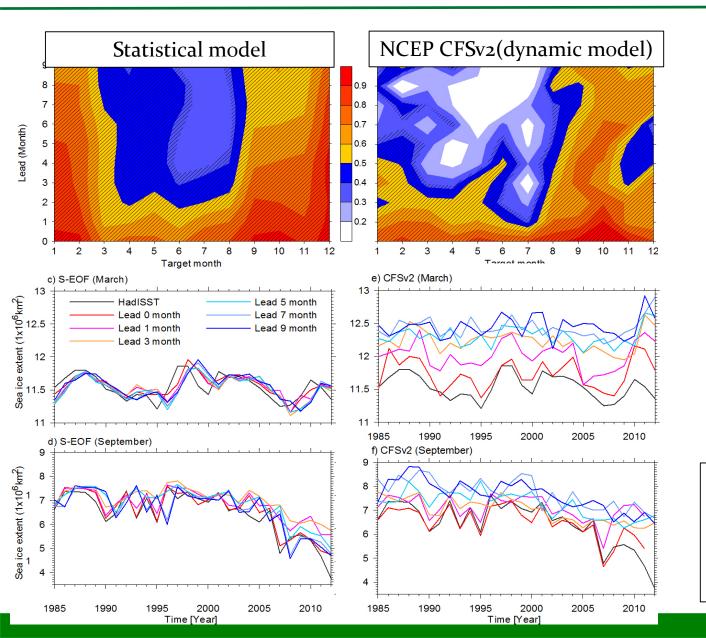




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Arctic SIE prediction skill (in correlation)





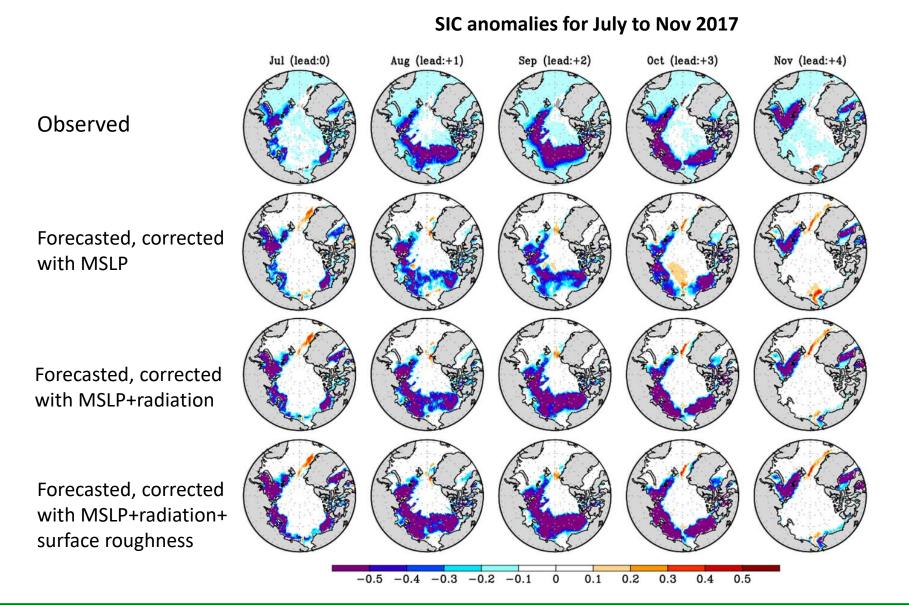
Compared with NCEP CFS,

- High obs-fcst correlation
- No bias
- Spring barrier exist

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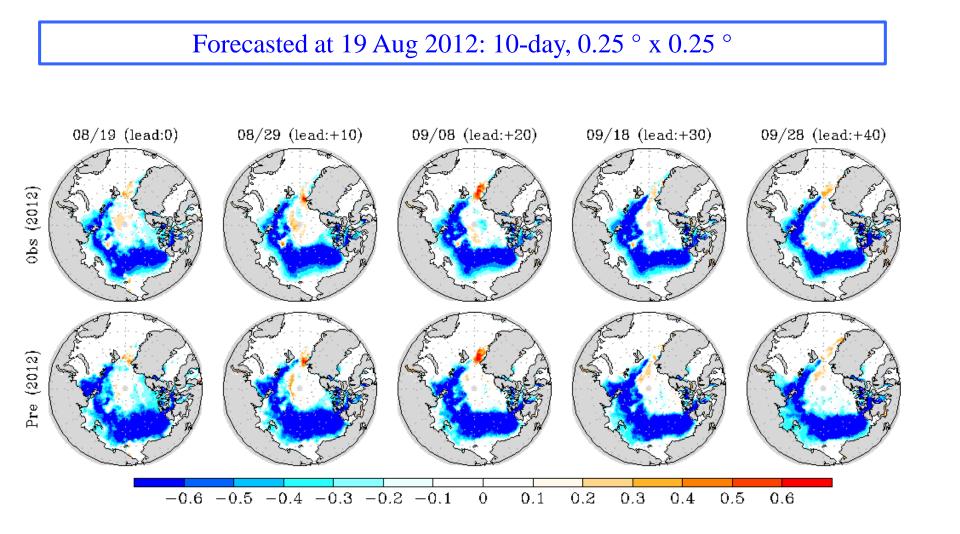
Corrections using atmospheric circulation, radiations..





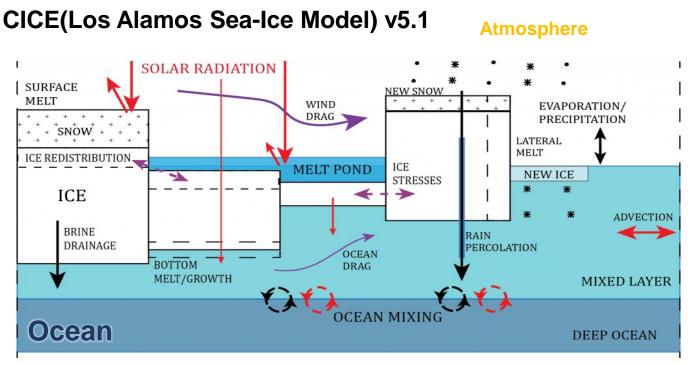








Dynamical model under development

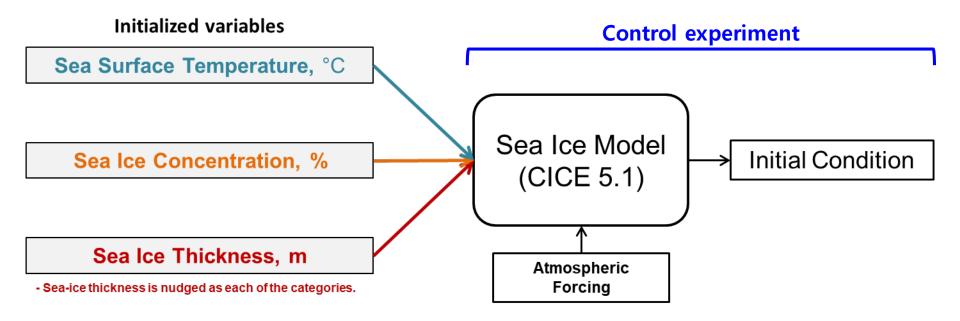


Some of the various physics now included in state-of-the-art sea ice models (e.g. CICE). Red arrows indicate heat fluxes, black arrows indicate salt/freshwater fluxes, and purple arrows indicate dynamic forces

http://www.climate-lab-book.ac.uk/2015/the-sea-ice-orchestra

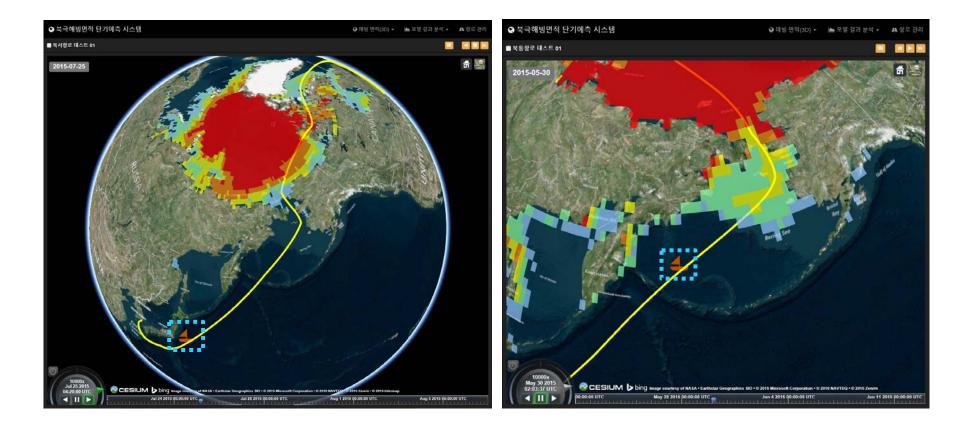
- The stand-alone ice model that predict sea ice fields by simulating the dynamic/thermodynamic processes.
- The CICE5 is the latest version of the Los Alamos Sea Ice Model.
- It can be run coupled in a global climate model(CICE4) or uncoupled as a stand-alone ice model.

<Initialization system for SST/SIC/thickness>



- Initialization scheme: Nudging(Newtonian relaxation)
- Experiment period: 1990.01.01 ~ 2008.12.31(19-years)
- Control experiment: Only atmospheric forcing
- Initialization system: Initialization + Atmospheric forcing



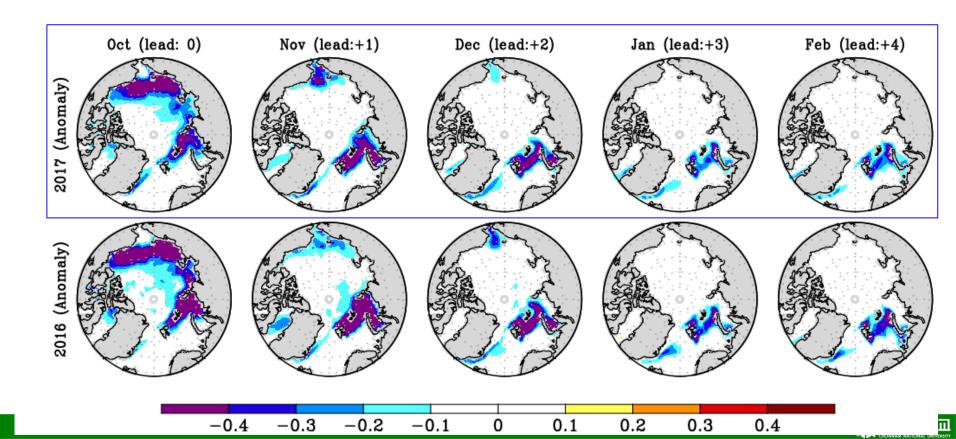




Summary



- JNU-KOPRI developed a prototype of Arctic sea ice statistical prediction model for ~12 months SIC predictions. Dynamical and hybrid model is being developed.
- Forecasted for this winter (initialized at Oct2017) as below





Thanks for your attention!





Backup slides





Dynamical model

by coupled GCM of interactive atmosphere, ocean, and sea ice with proper initialization

VS.

Statistical model

based on statistical relationship between the Arctic sea ice and preceding SST/OLR/circulation etc.



GCMs have large-bias of sea ice



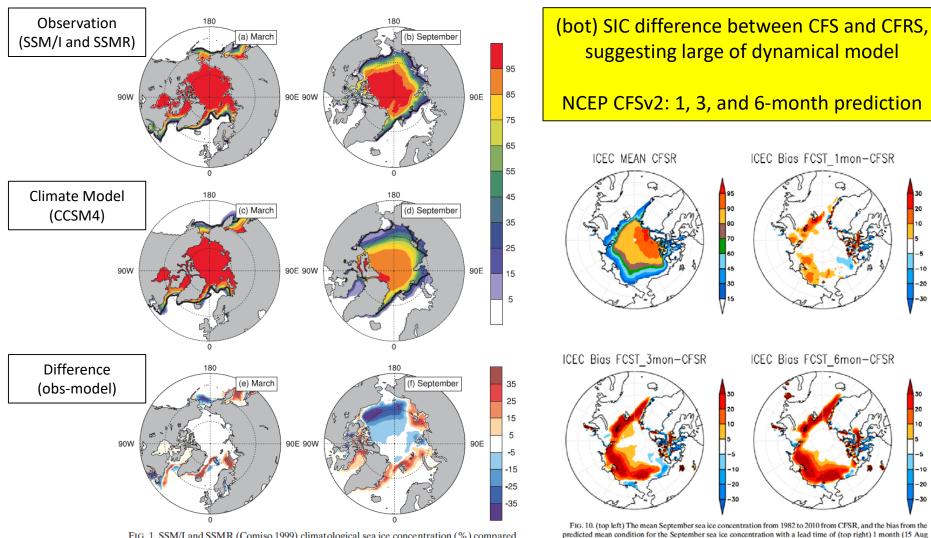


FIG. 1. SSM/I and SSMR (Comiso 1999) climatological sea ice concentration (%) compared to the six member CCSM4 ensemble average sea ice concentration for 1981–2005 for (a),(c) March and (b),(d) September. The ice edge (taken as 15% sea ice concentration contour) from the SSM/I and SSMR data are shown as black line. (e),(f) The difference between the CCSM4 and the SSM/I and SSMR ice concentration, showing regions with too much ice in the simulation (red colors) and regions with not enough ice in the simulation (blue).

Saha et al 2014, JC

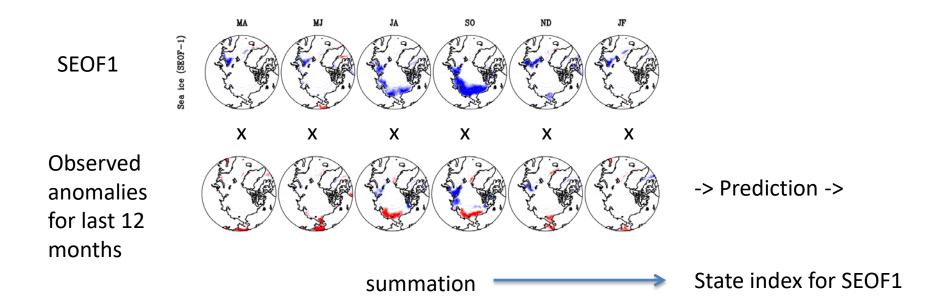
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IC), (bottom left) 3 months (15 Jun IC), and (bottom right) 6 months (15 Mar IC).

Step 2: SIC state index



- 1. 현재(0~ -11 month) 해빙 상태지수의 결정
 - SIC anomalies for the last 12 months
 - Projection onto the identified S-EOFs





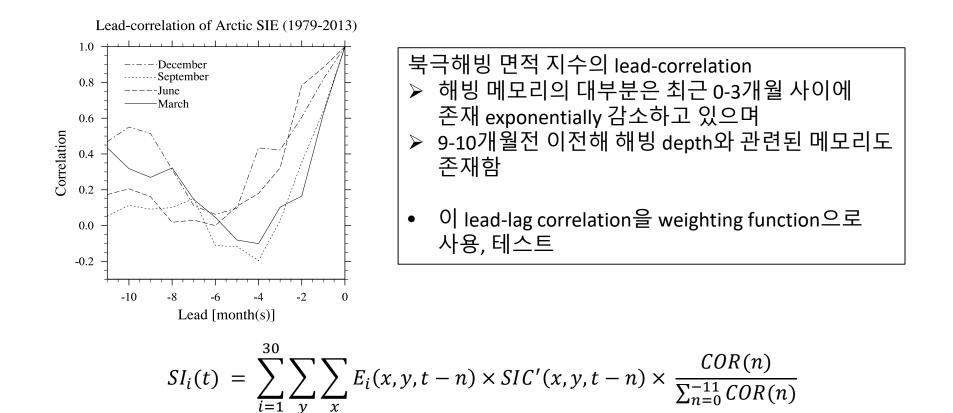
Step 2: SIC state index – weighting



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현재 해빙 상태지수 계산 기법

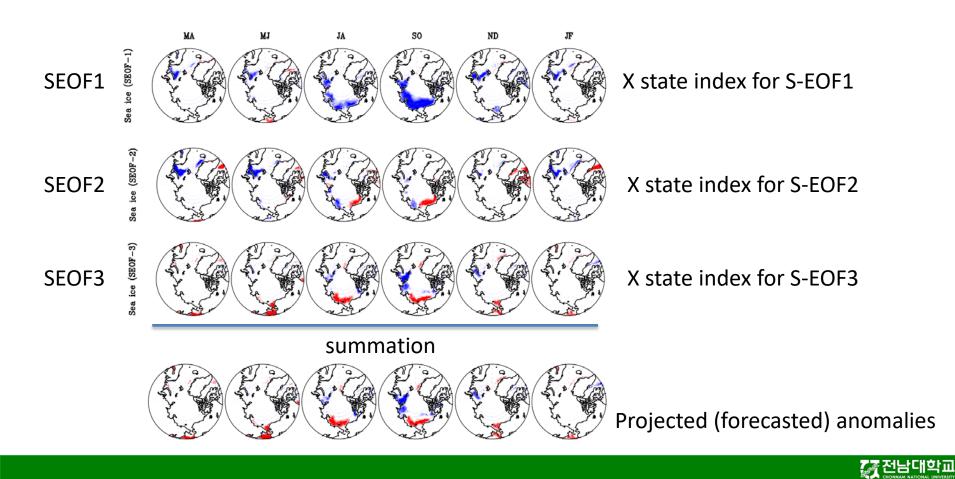


해빙면적 autocorrelation을 weighting function으로 이용

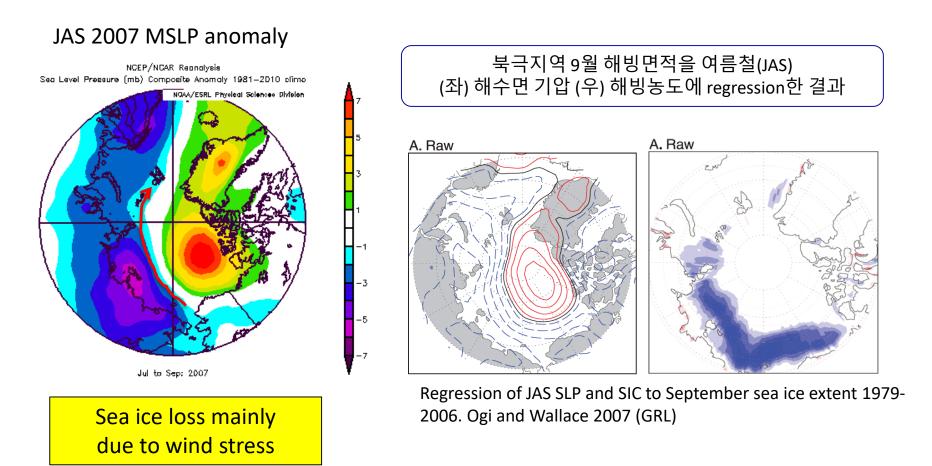
Step3: S-EOF를 이용한 해빙 통계예측 모델



- 2. 상태지수가 예측기간 동안에 유지(혹은 trend가 유지) 된다고 가정
- 3. Pre-identifie된 S-EOF 패턴과 상태지수를 곱하여 각 mode에 해당하는 미래 해빙 anomalies를 reconstruction



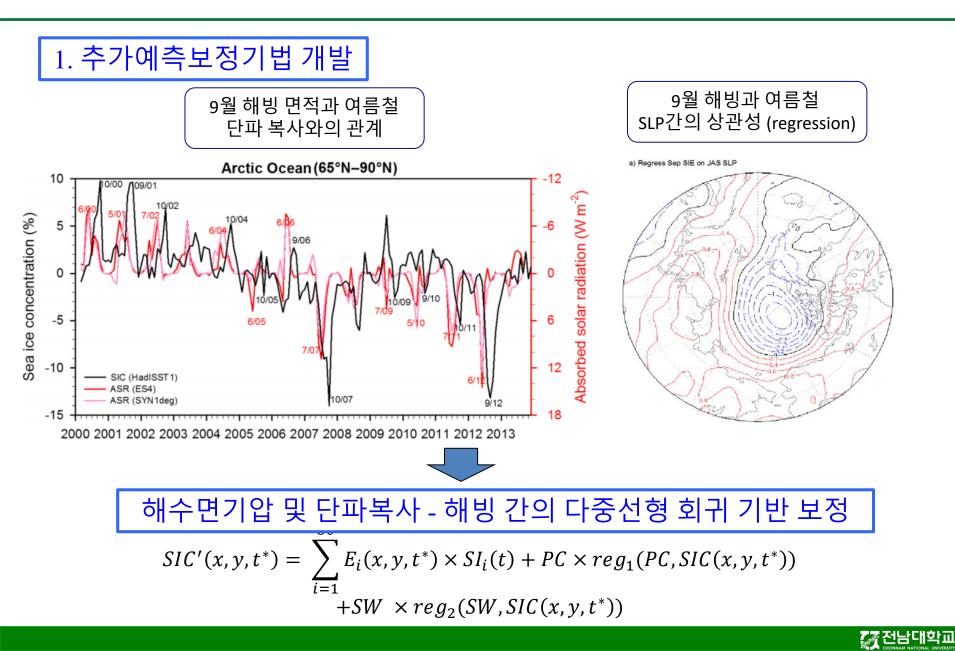




Melting phase의 대기정보-해빙농도 regression 모델 구축 통계모델에 보정 term으로 적용



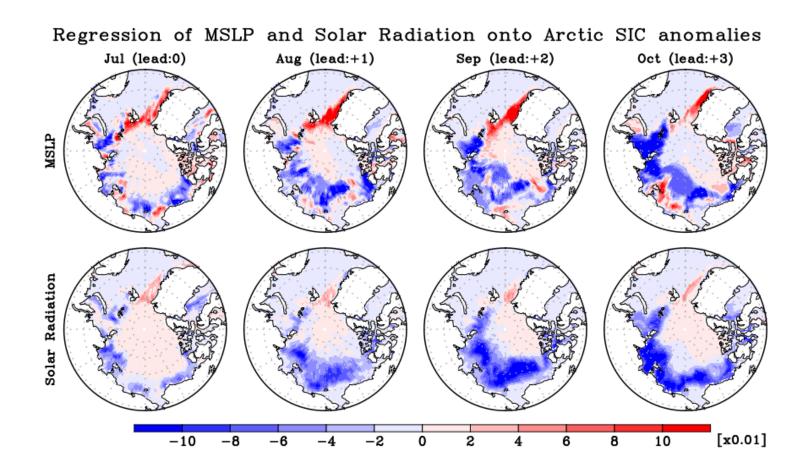




Step4: Correction with MSLP, radiation

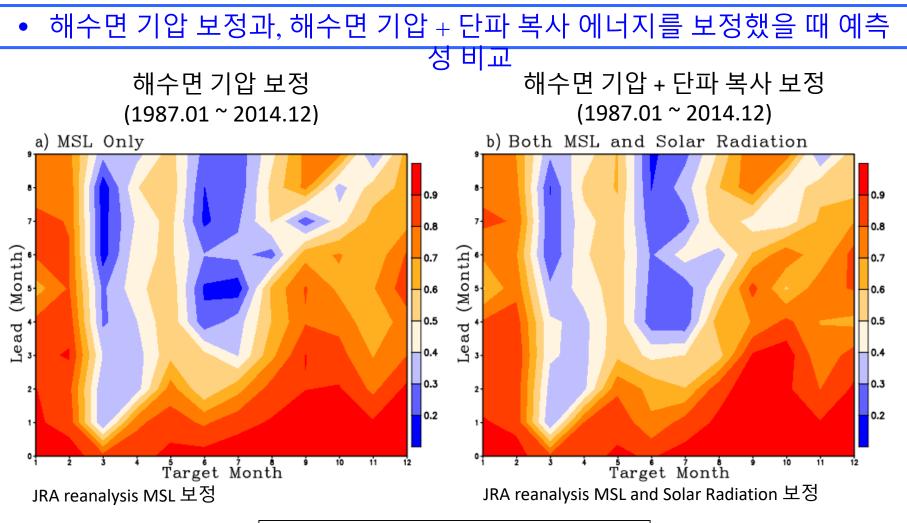


Spatial patterns of coefficient for multiple regression between MSLP and Solar Radiation onto Arctic SIC anomalies





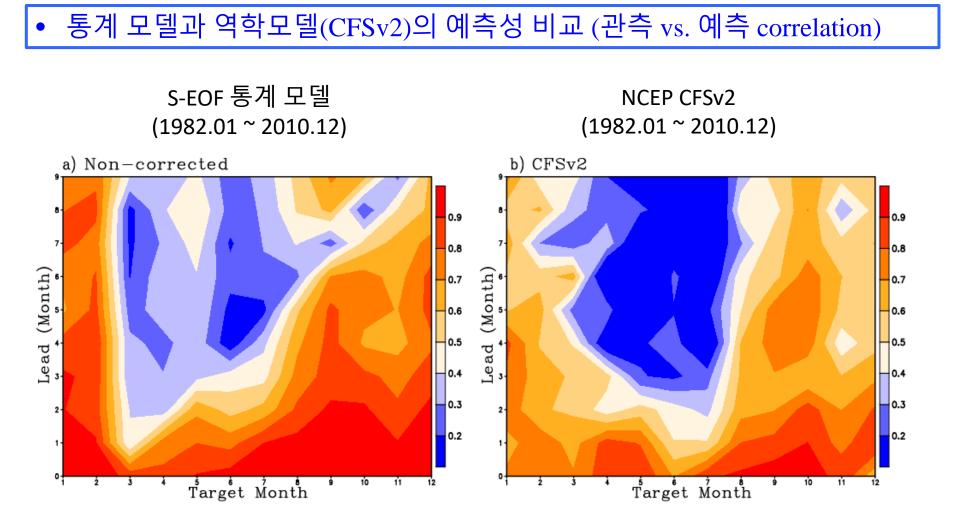




여름-가을철 예측성 향상이 나타남

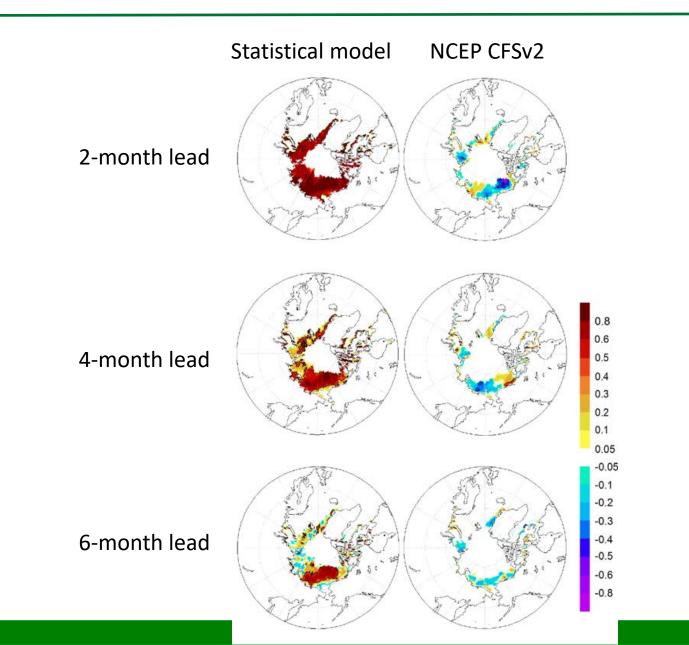
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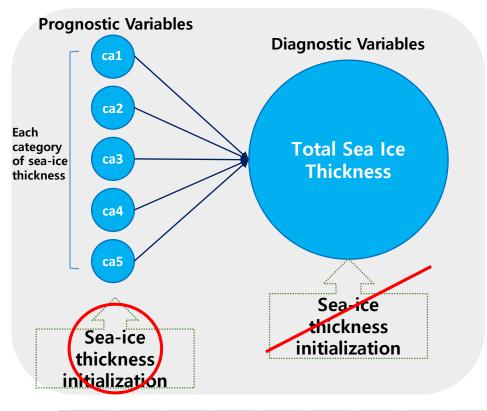
Sep. SIC observation-forecast correlation



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Development of initialization scheme of sea-ice thickness

Method of initialization for each ice thickness category



There is no DATA for each ice thickness category But, Nudging term for them is required

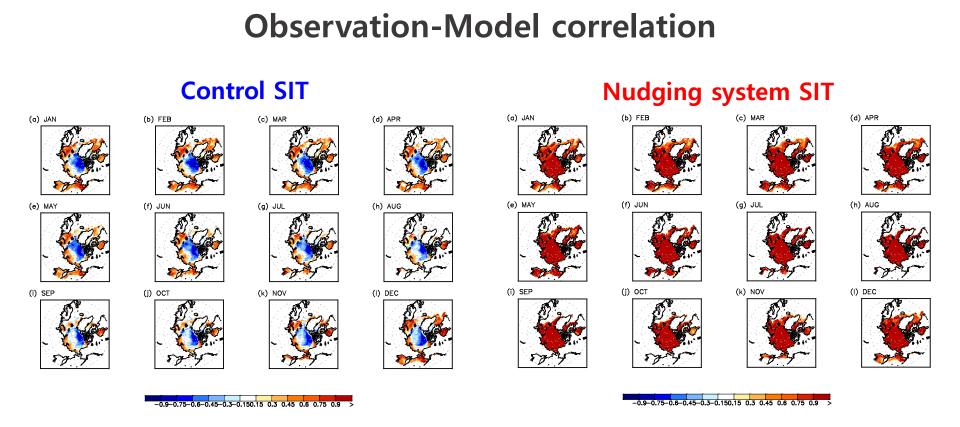
Control Run

 $H_{mod_tot} = H_{mod_ca1} + H_{mod_ca2} + H_{mod_ca3} + H_{mod_ca4} + H_{mod_ca5}$

Nudging Run $H_{nu_tot} = (H_{mod_ca1} + \Delta H_1) + (H_{mod_ca2} + \Delta H_2) + (H_{mod_ca3} + \Delta H_3) + (H_{mod_ca4} + \Delta H_4) + (H_{mod_ca5} + \Delta H_5)$ $\Delta H_1 = (H_{piomas_tot} - H_{mod_tot}) \times \left(\frac{H_{mod_ca1}}{H_{mod_tot}}\right)$ $\Delta H_2 = (H_{piomas_tot} - H_{mod_tot}) \times \left(\frac{H_{mod_ca2}}{H_{mod_tot}}\right)$

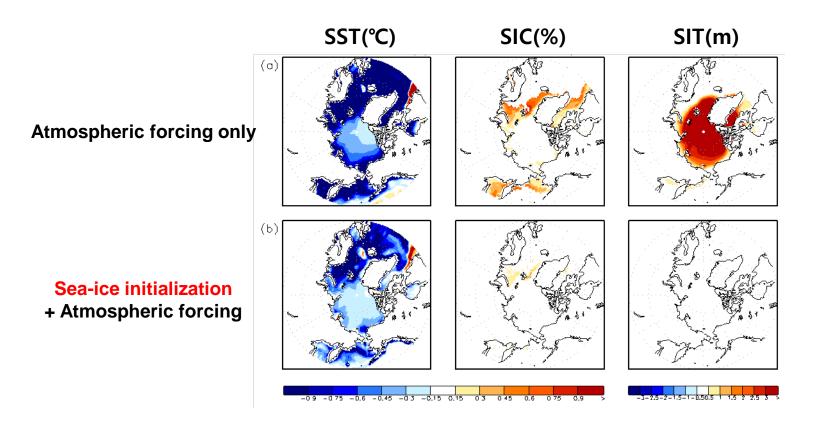
- The nudging term for prognostic variable each category of sea-ice thickness is required, because total thickness is diagnostic variable in model.
- The nudging terms for them are constructed by multiplying the ratio of each category in the model by the total sea-ice thickness difference.
- This method can directly initialize sea-ice thickness, maintaining the balance within the model.

Only each category of sea-ice thickness initialization



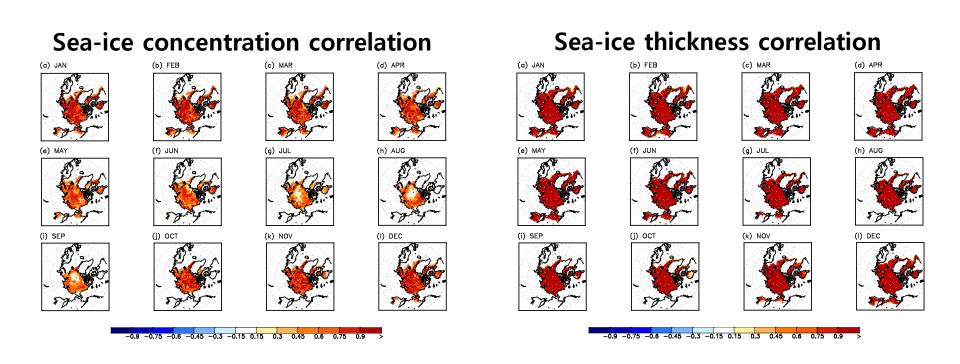
• SIT correlation becomes high in all seasons compared to control experiment.

 \rightarrow The each category of sea-ice thickness nudging initialization system is successfully setup and total SIT is well constrained by observation through each SIT category initialization method .



- The combined nudging initialization system is successfully setup.
- Bias of the sea-ice nudging experiments becomes significantly low
 → Well constrained by the observation.

Correlation(Model↔Observation) of initial condition



- The combined nudging initialization system is successfully setup
- Correlation between the sea-ice nudging experiments and the observations shows significantly high value for both SIC and SIT.
 - → Well constrained by the observation, especially sea-ice thickness