

# Ocean dynamic persistence provides substantial predictability for El Niño-Southern Oscillation (ENSO)

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# Motivation

ENSO prediction skill  
from coupled models



Accuracy of the  
initial ocean state

+

Fidelity of coupled  
models in depicting  
air-sea coupling

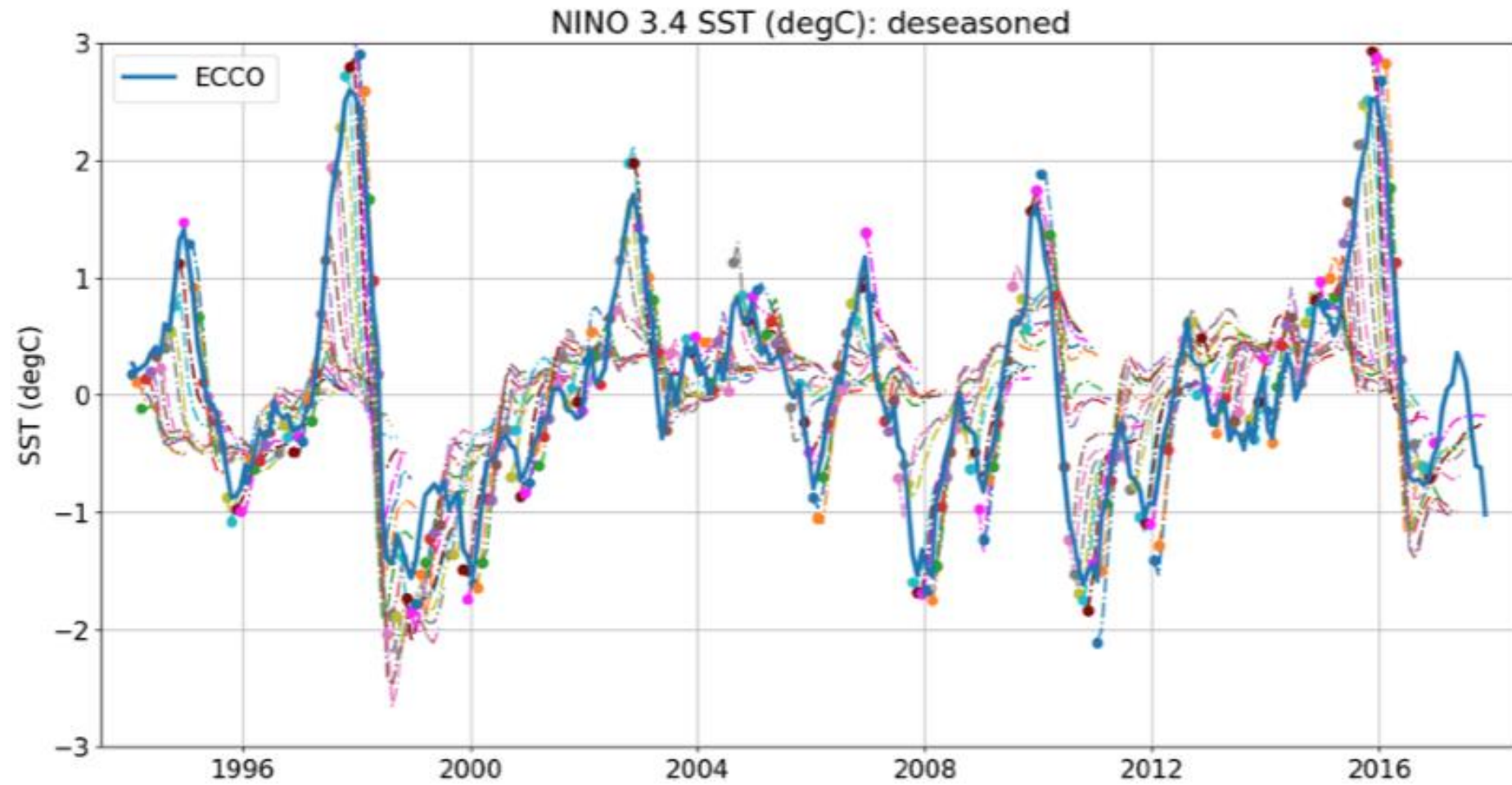
How much predictability does the initial ocean state provide w/o considering subsequent air-sea coupling?

Coupled-model initialization shock or drift can limit the contribution of an accurate initial ocean state to ENSO prediction skill.

# Method

- Perform a set of 12-month forward integration of a global ocean model forced by climatological seasonal atmospheric forcings, initialized from **ECCO** ocean state estimate monthly from 1994 to 2017.
- Interannual ocean anomalies in the 12-month hindcasts are solely due to the evolution of the initial ocean state without air-sea coupling - referred to as **ocean dynamic persistence**.
- It **contrasts statistical persistence**, a typical baseline metrics for evaluating prediction skill, where the initial anomaly is damped statistically as a function of prediction lead time.

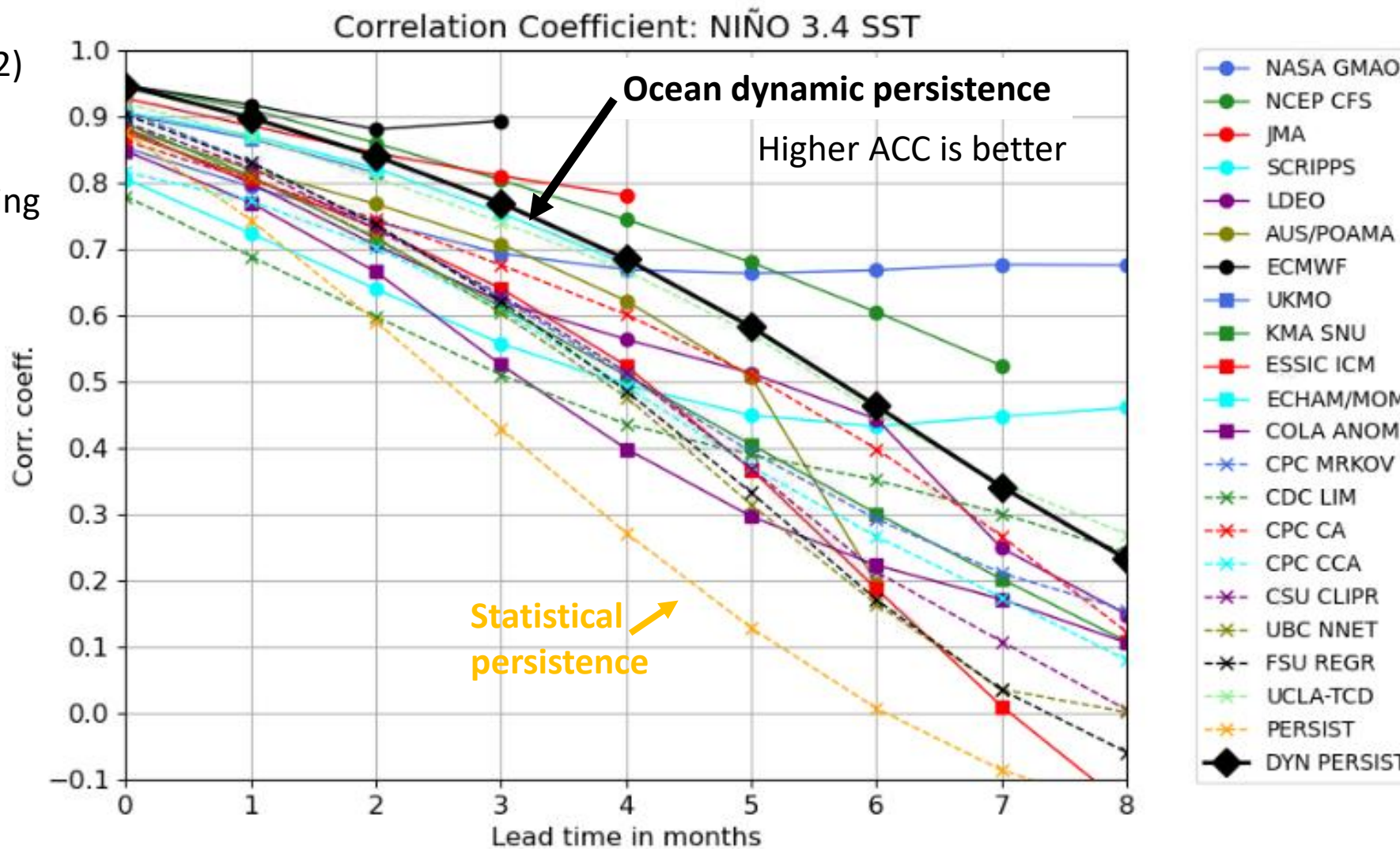
# Ocean dynamic persistence hindcasts of Niño3.4 SST index



# Prediction skill: Anomaly Correlation Coefficient (ACC) of predicted vs. observed Niño3.4 SSTA (for the 2002-2011 period as in Barnston et al. 2012, BAMS)

**Color curves:** from Barnston et al. (2012)  
 Solid – coupled models.  
 Dashed – statistical models.  
 (Thanks to Dr. Tony Barnston for providing the model forecast data)

**Findings:**  
 Ocean dynamic persistence skill  
 > skill of most coupled and  
 statistical forecast models;  
 >> skill of statistical persistence.

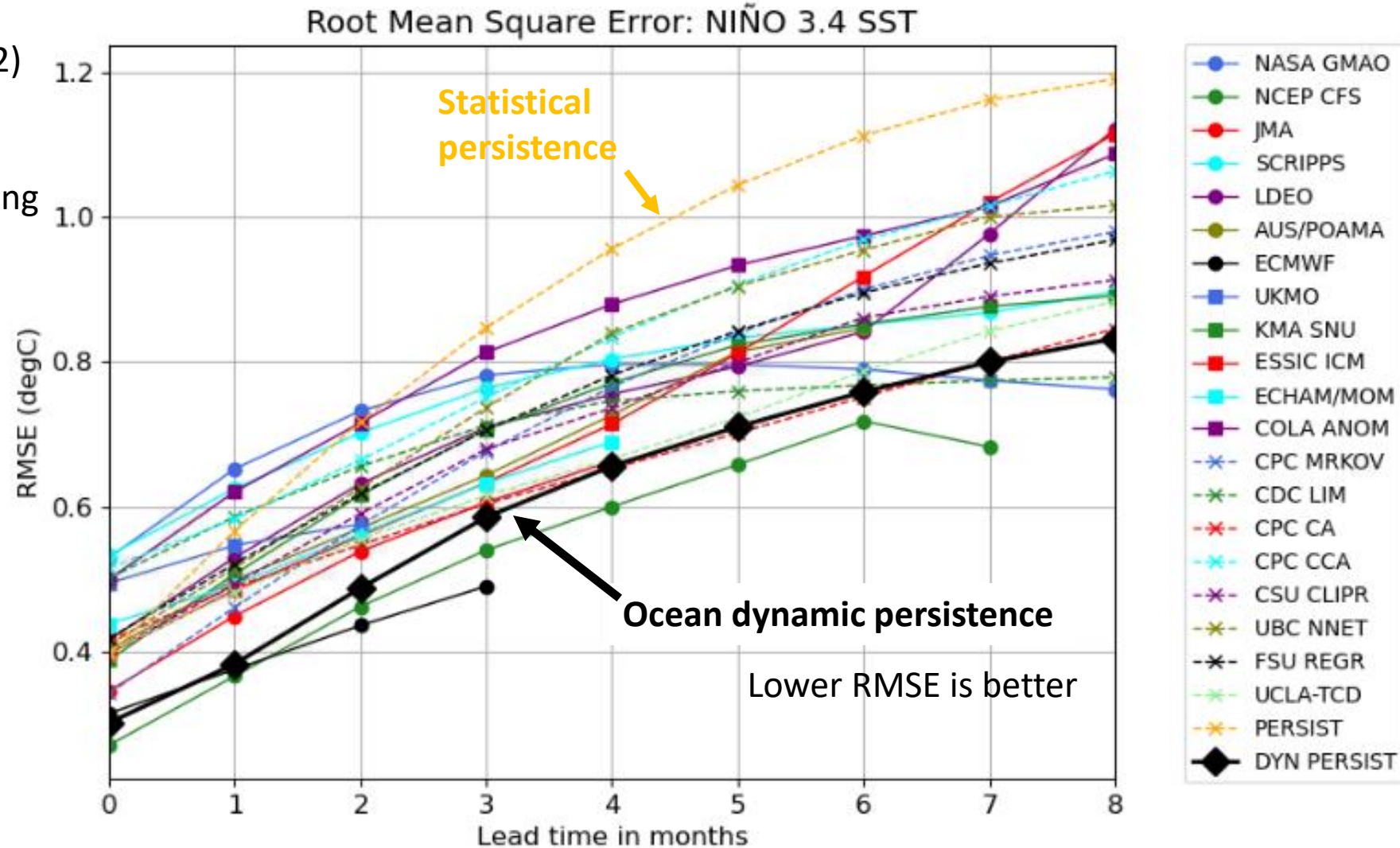


Lead time month 0 represents the average for predictions in the first 3 months  
 (Barnston et al. 2012)

# Prediction skill: Root Mean Squared Error (RMSE) of predicted vs. observed Niño3.4 SSTA (for the 2002-2011 period as in Barnston et al. 2012, BAMS)

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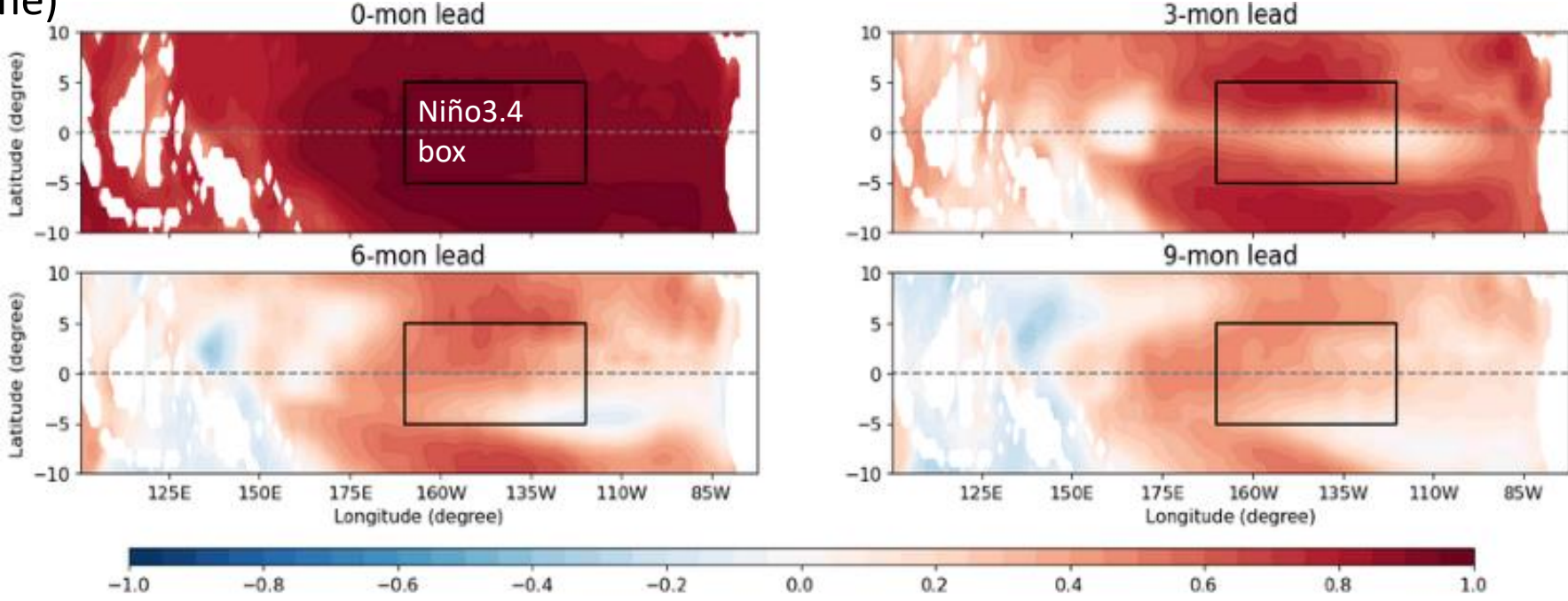


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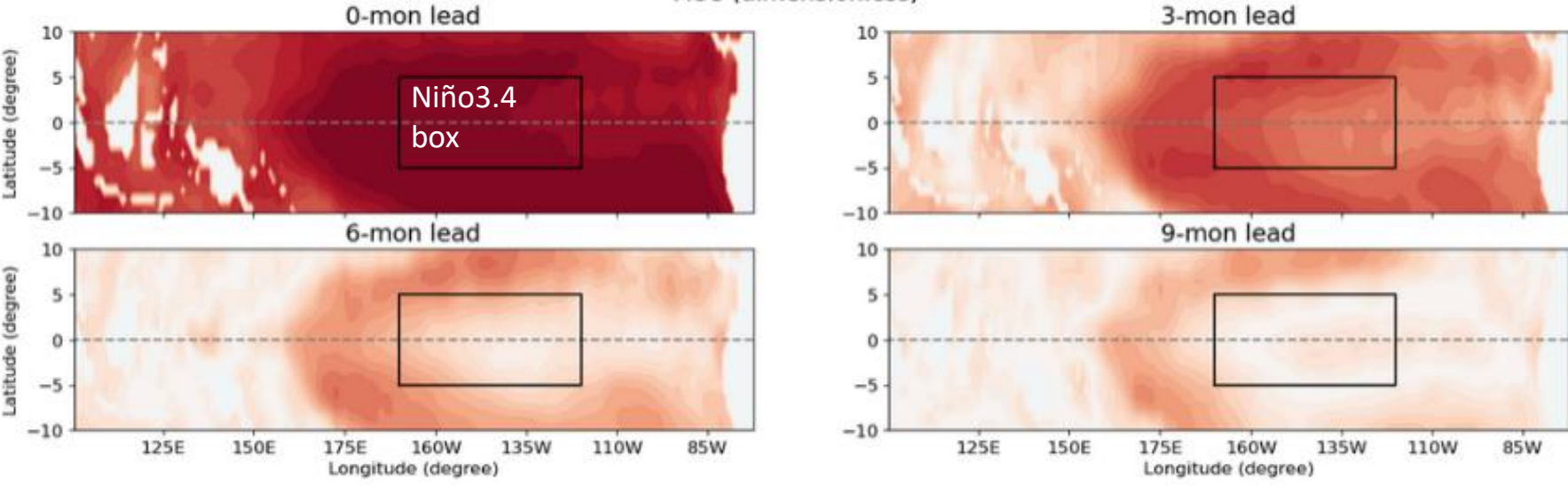
# Contrasting prediction skills by ocean dynamic persistence & statistical persistence

(as a function of space and lead time)

Prediction skill (ACC) from ocean dynamic persistence



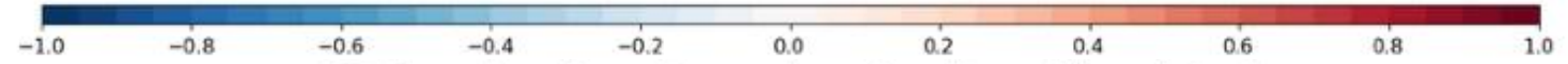
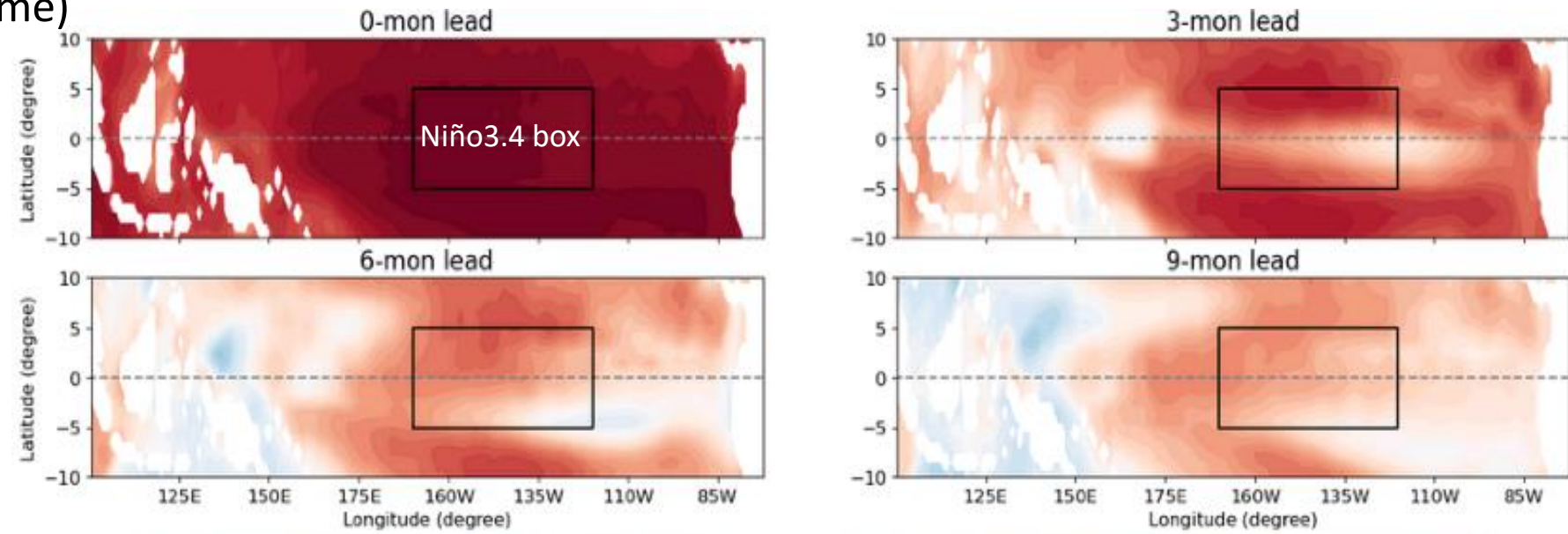
Prediction skill (ACC) from statistical persistence



# Contrasting prediction skills by ocean dynamic persistence & statistical persistence

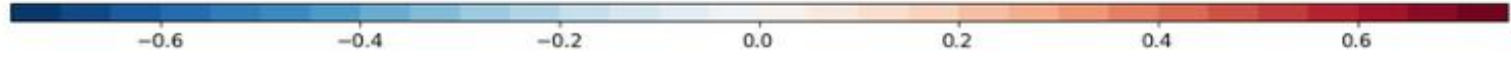
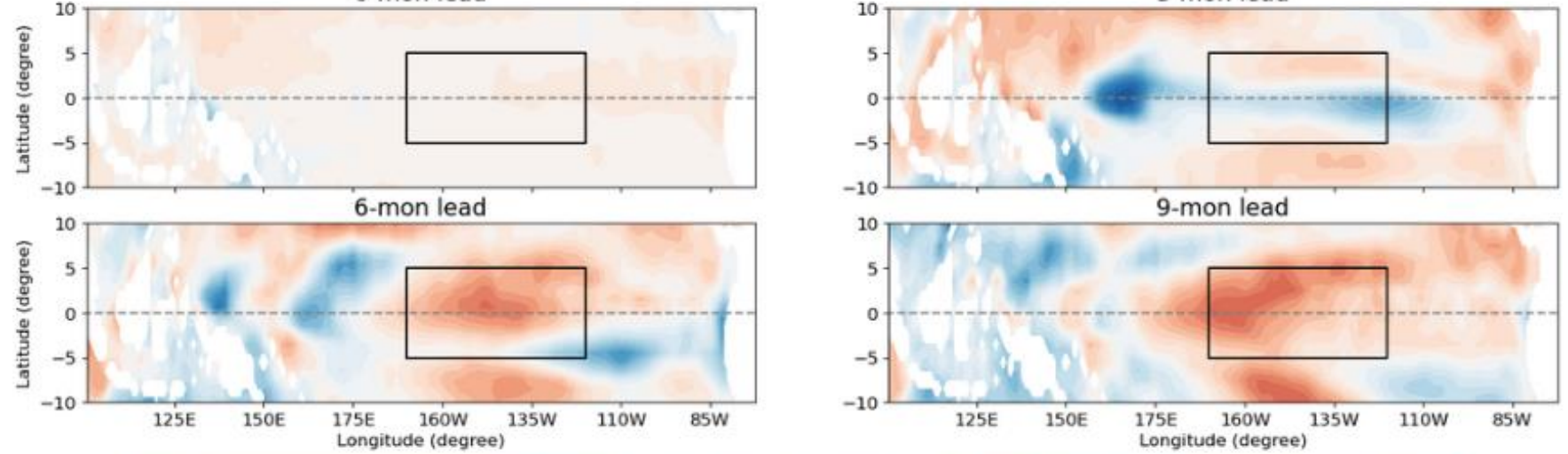
(as a function of space and lead time)

Prediction skill (ACC) from ocean dynamic persistence



ACC: Ocean dynamic persistence - damped persistence (dimensionless)

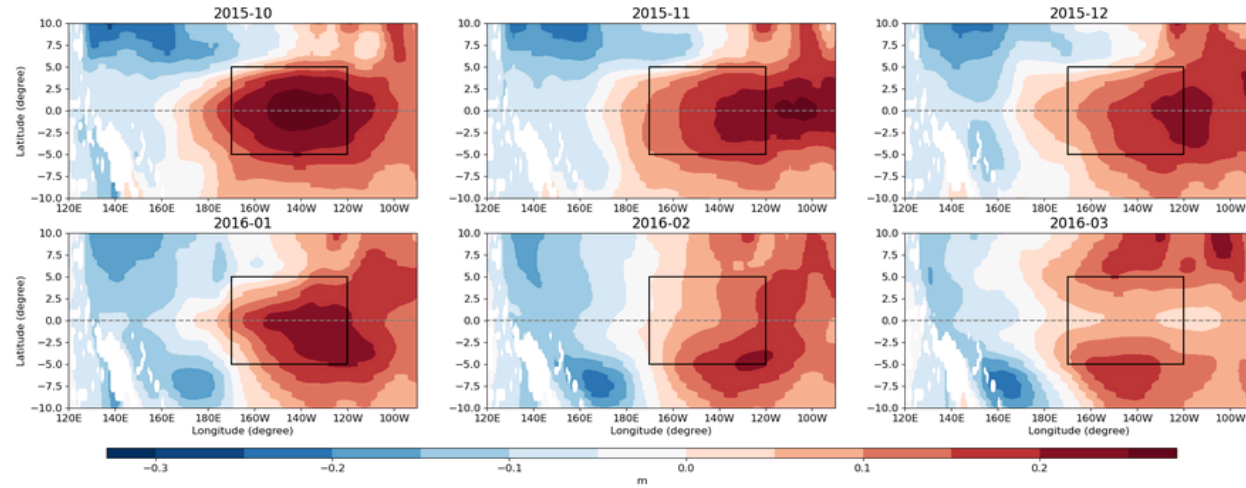
Prediction skill (ACC) difference: dynamic-statistical persistence (red values indicate ocean dynamic persistence having higher skill)



# Equatorial wave dynamics is important to ocean dynamic persistence: example for equatorial & off-equatorial SSHA & SSTA anomalies during the 2015 El Niño

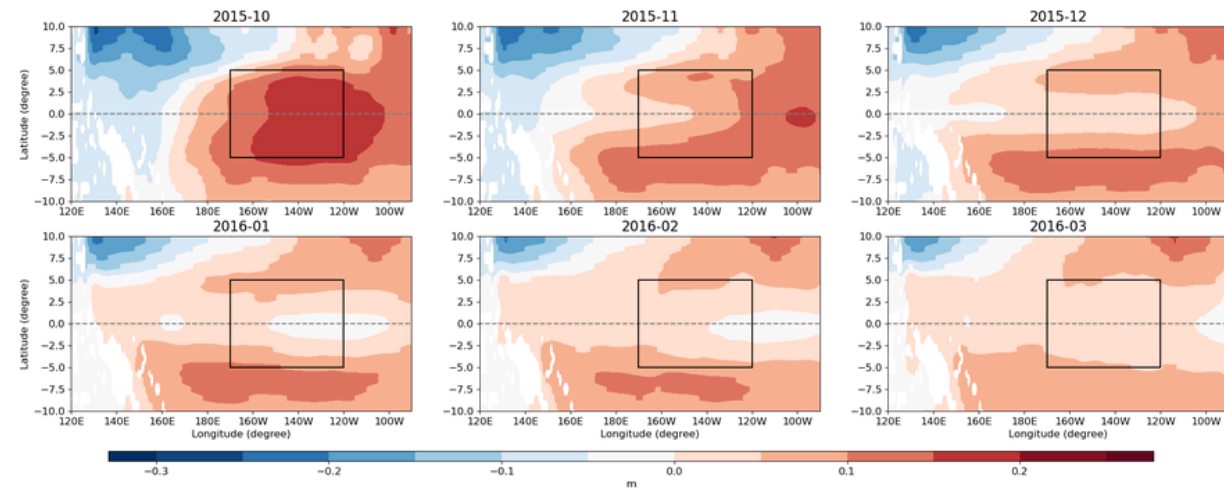
## SSHA from ECCO

Sea Surface Height (m): ECCO



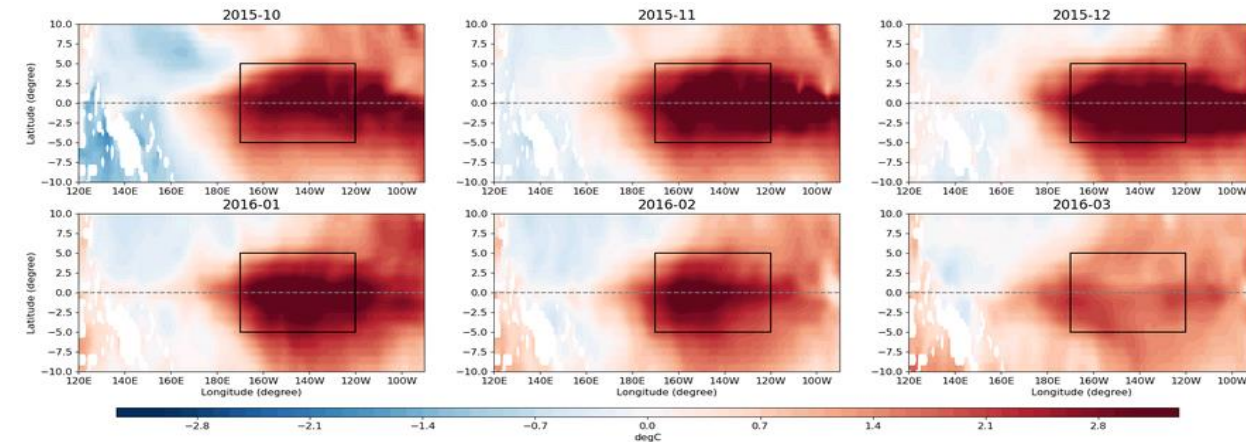
## SSHA dynamic persistence, initialized 2015-09-30, 24Z

Sea Surface Height (m): Dyn. Persistence Simulation



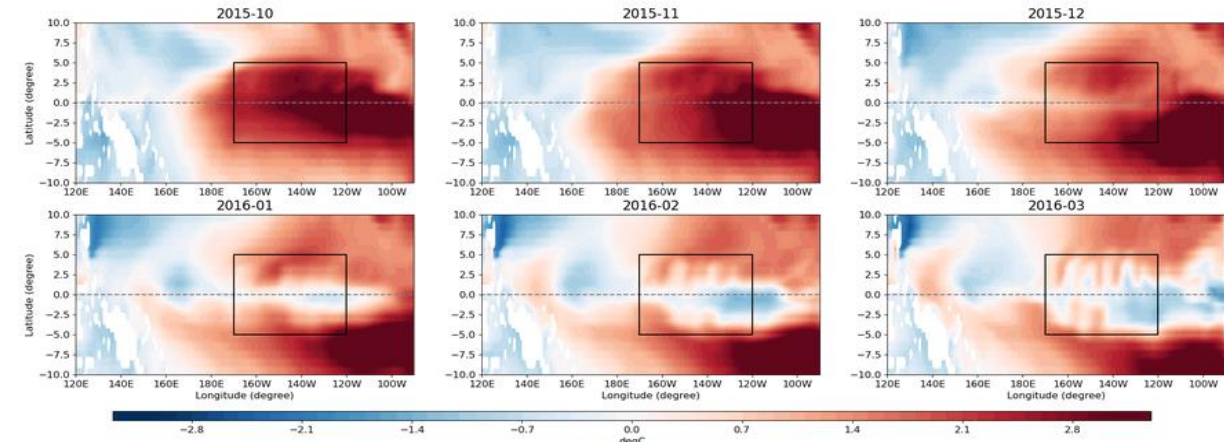
## SSTA from ECCO

Sea Surface Temperature (degC): ECCO



## SSTA dynamic persistence, initialized 2015-09-30, 24Z

Sea Surface Temperature (degC): Dyn. Persistence Simulation



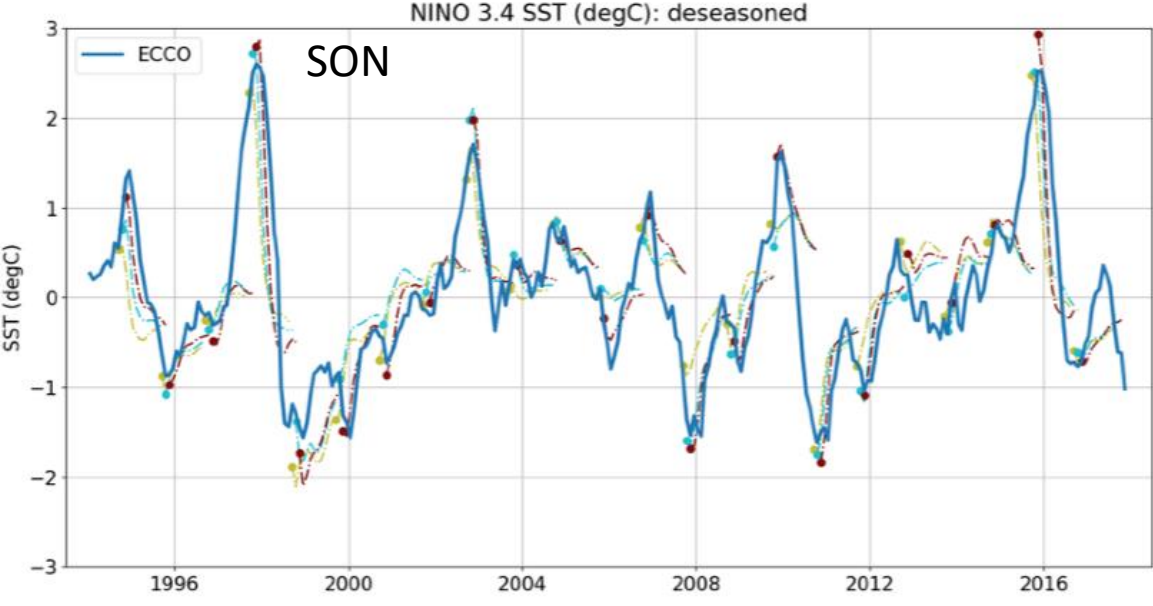
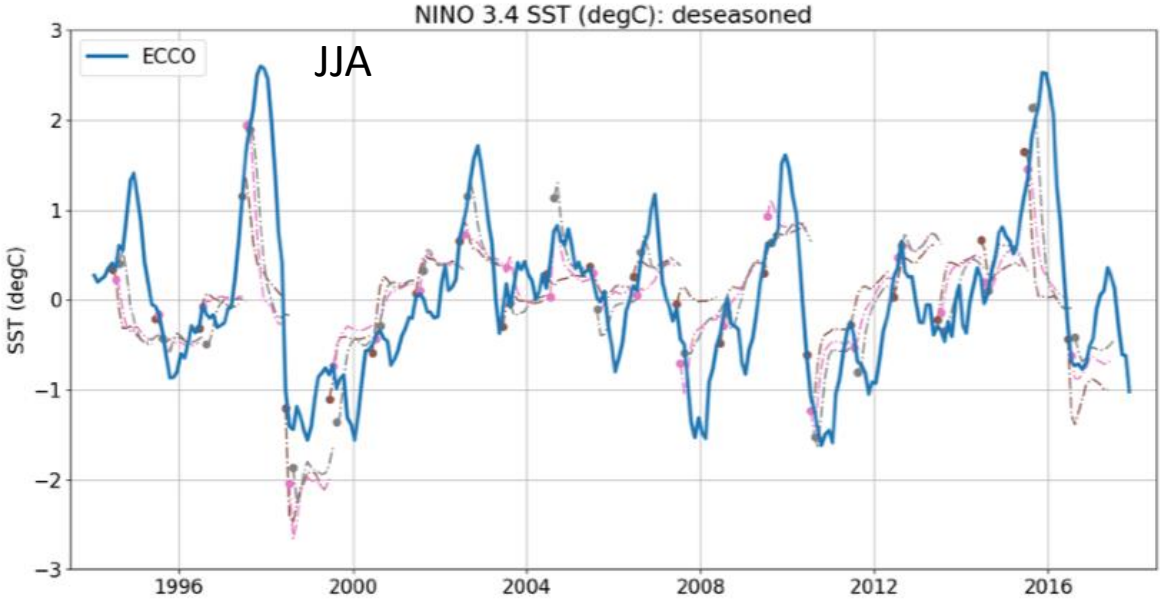
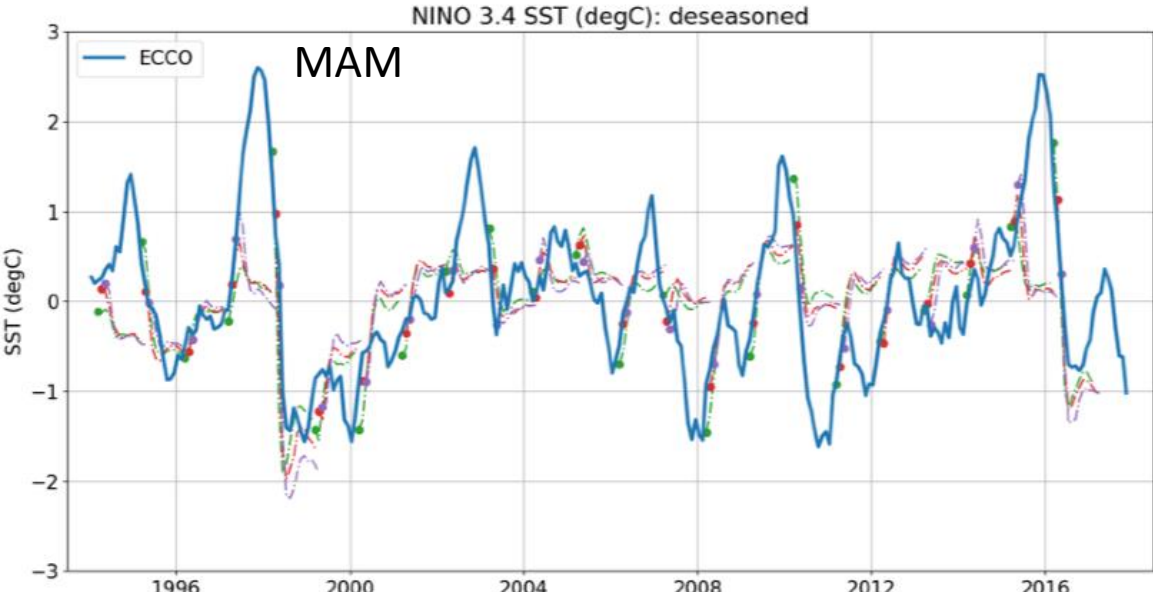
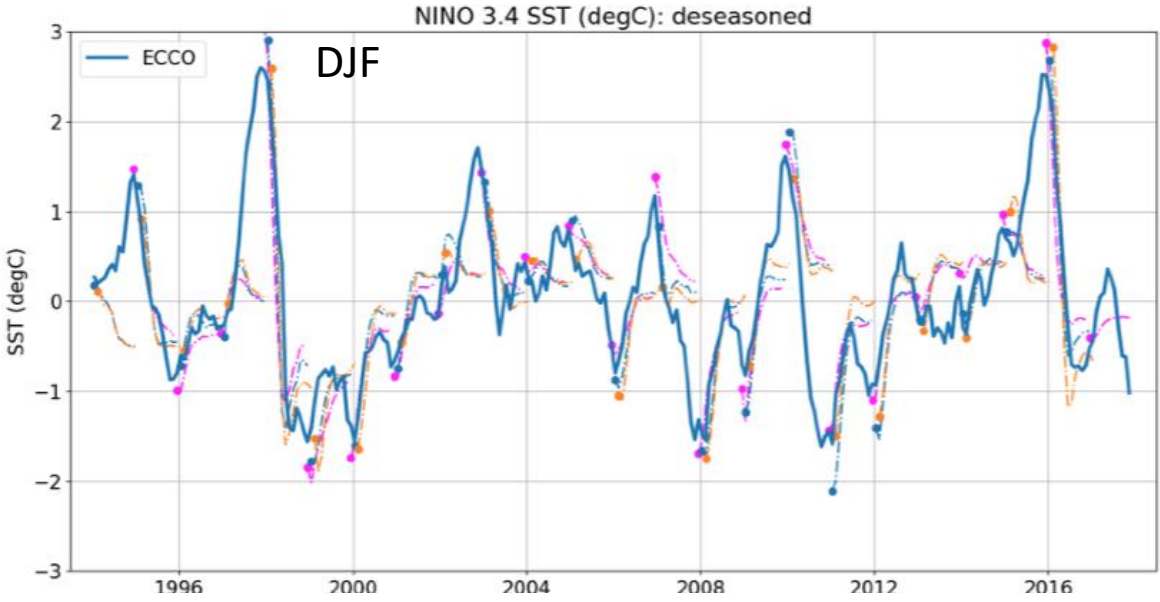


# Summary

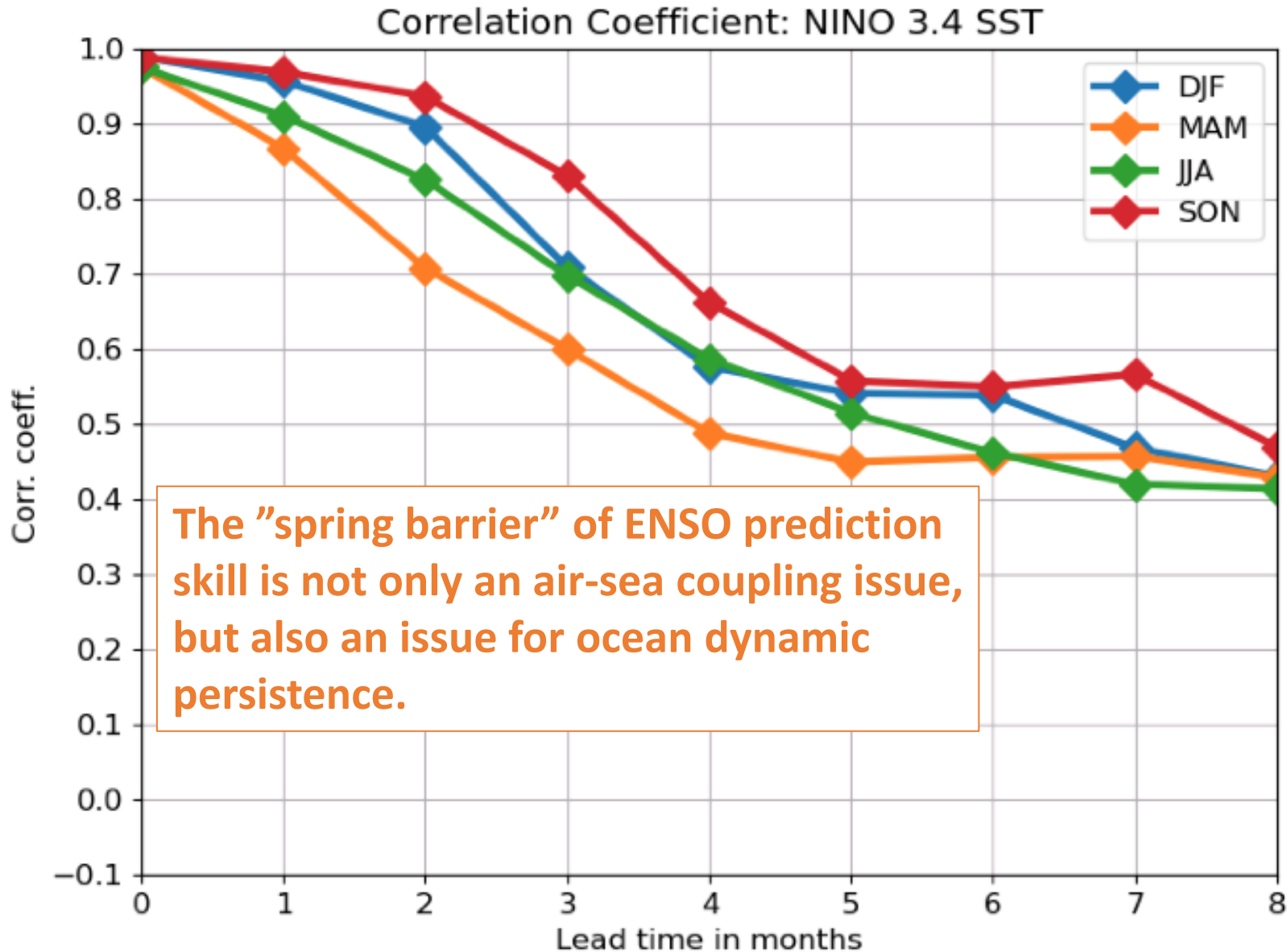
- The skill of **ocean dynamic persistence** in hindcasting Niño3.4 SST index
  - exceeds the skills of most statistical & coupled models;
  - is much better than that of statistical persistence;
  - raises the bar for baseline metrics of evaluating ENSO prediction skill.
- Equatorial wave dynamics explains much of the ocean dynamic persistence.
- The role of oceanic advection/diffusion at longer lead times needs to be investigated.

# Backup slides

# Ocean dynamic persistence hindcasts of Niño3.4 SST index: by initialization season



# The skill of ocean dynamic persistence in hindcasting Niño3.4 SST index by initialization season

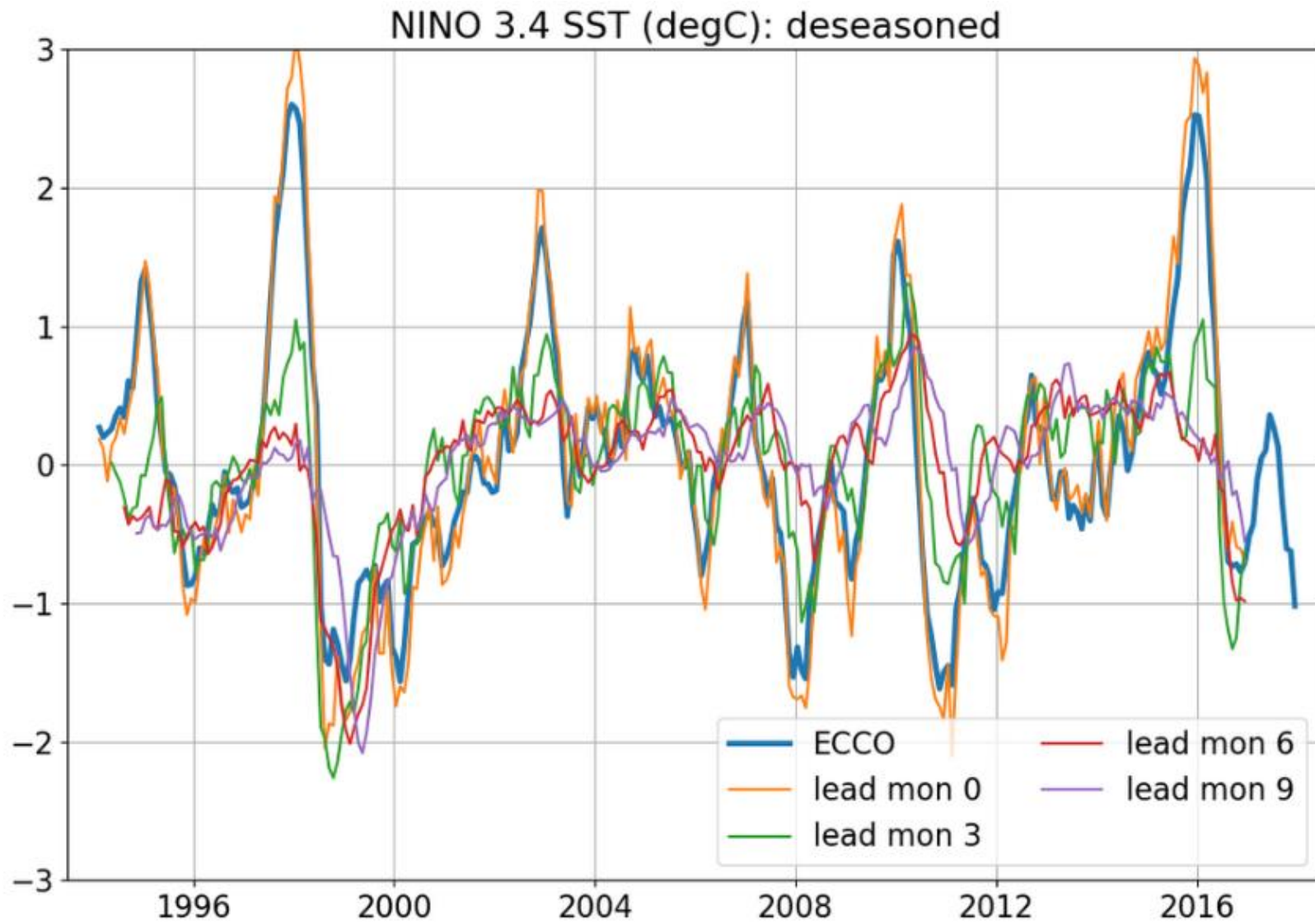


The "spring barrier" of ENSO prediction skill is not only an air-sea coupling issue, but also an issue for ocean dynamic persistence.

Correlation values shown here are for 1994-2017, not 2022-2011 as in Barnston et al. (2012).

The value at each lead time represents the value for that lead-time month, not 3-month lead time average as in Barnston et al. (2012).

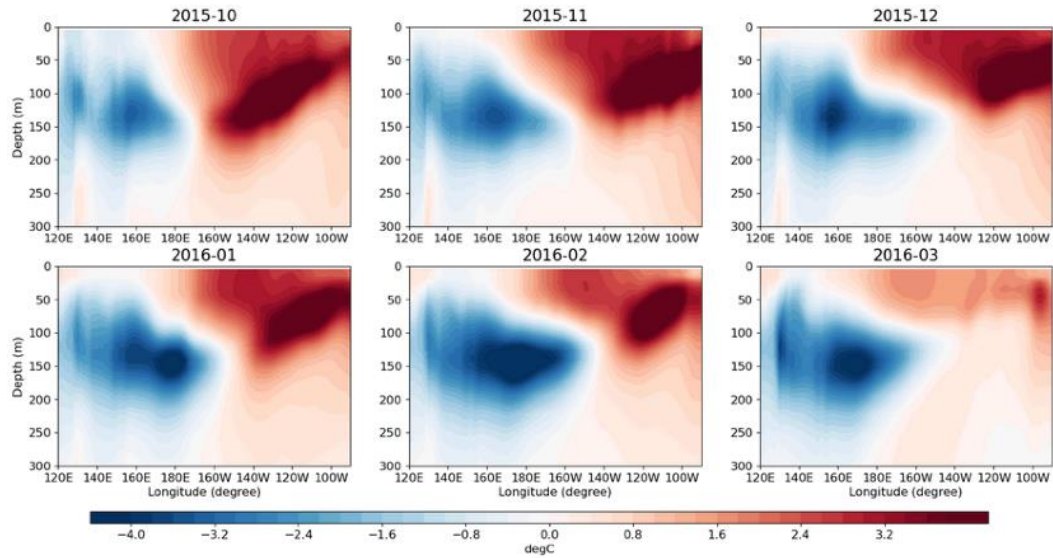
# NINO3.4 SST (degC): ECCO vs. Predictions



# Temperature Anomaly (degC): 5S-5N Average

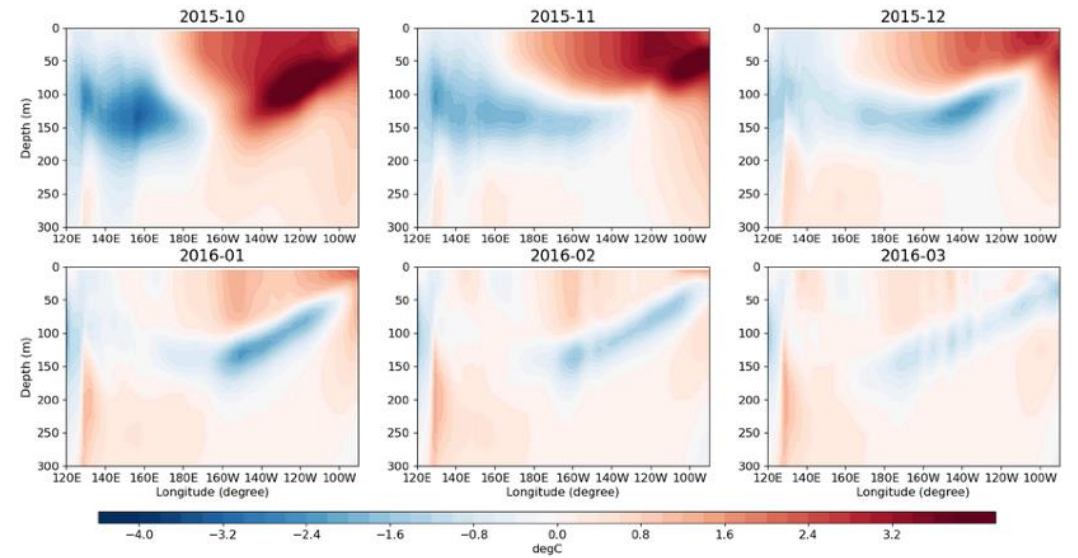
## ECCO

Ocean Temperature (degC): ECCO



## Prediction: Initialized on 09/30/2015, 24Z

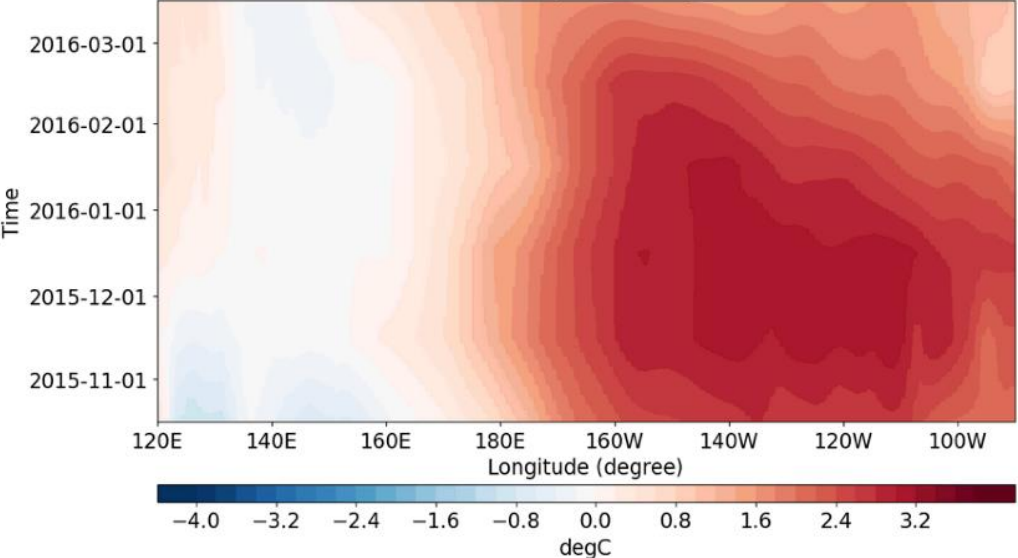
Ocean Temperature (degC): Dyn. Persistence Simulation



# Sea Surface Temperature Anomaly (degC): 5S-5N Average

## ECCO

Sea Surface Temperature (degC): ECCO



## Prediction: Initialized on 09/30/2015, 24Z

Sea Surface Temperature (degC): Dyn. Persistence Simulation

