

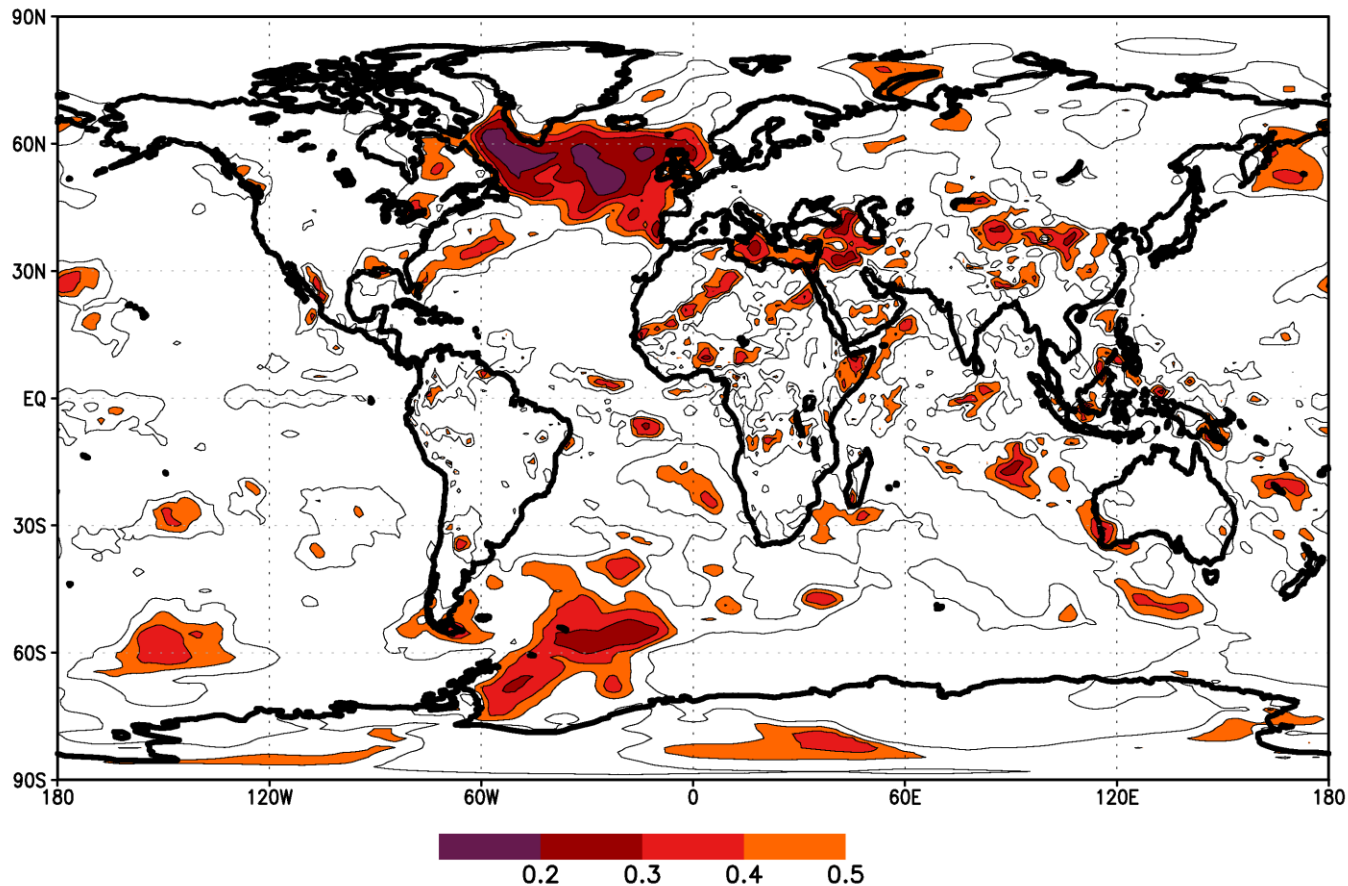
GEOPHYSICAL RESEARCH LETTERS, VOL. 40, 1–5, doi:10.1002/grl.50355, 2013

## **Multiyear climate predictions using two initialization strategies**

W. Hazeleger,<sup>1,5</sup> V. Guemas,<sup>2,8</sup> B. Wouters,<sup>1</sup> S. Corti,<sup>4,6</sup> I. Andreu-Burillo,<sup>2</sup> F. J. Doblas-Reyes,<sup>2,7</sup> K. Wyser,<sup>3</sup> and M. Caian<sup>3</sup> and R. Haarsma, D. Volpi

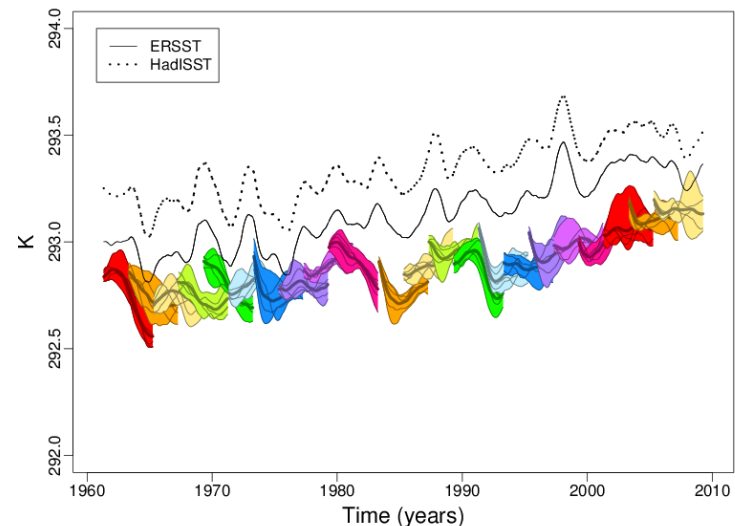
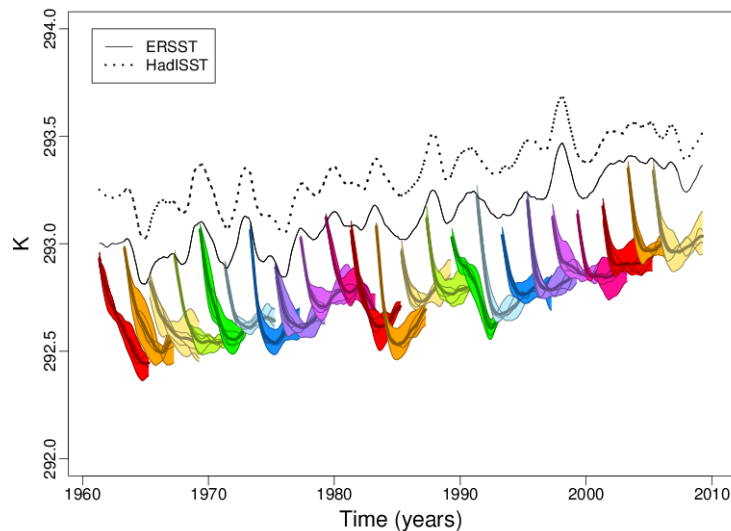


## Prognostic potential predictability in EC-Earth (T2m, yr 1-10; without trend)





# Questions for initialized decadal predictions



See poster *D. Volpi et al* for more details

- What is the impact of different initialization strategies on drift and skill within one coupled model
- What is the impact of sampling (1 yr vs 5 yr start dates) on skill within one coupled model



## experiments

- EC-Earth V2.3 (CMIP5 version, T159L62 IFS, 1 degree NEMO)
- Initial conditions: ERA40/interim atmosphere land, ORAS4 ocean, ice from Drakkar V4.3 forced ocean/ice model

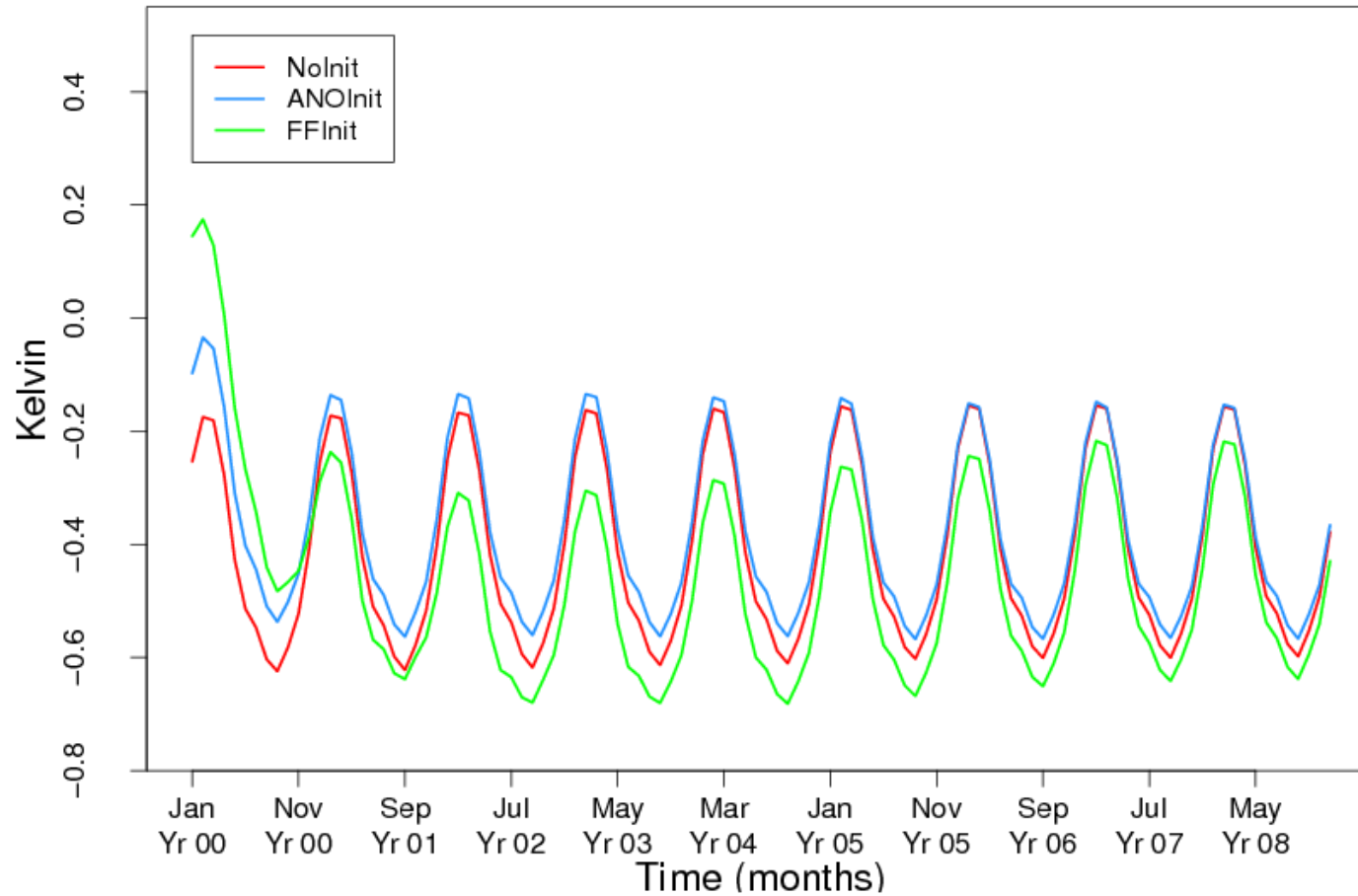
FULL state initialisation for atmosphere and ocean

ANOM initialisation (anomaly on top of climatology from historical runs) for ocean and sea ice.

- 10 members at each starting date for FULL, 5 for ANOM
- New compared to GRL paper: **1) annual start dates 2) three methods for anomaly initialisation**



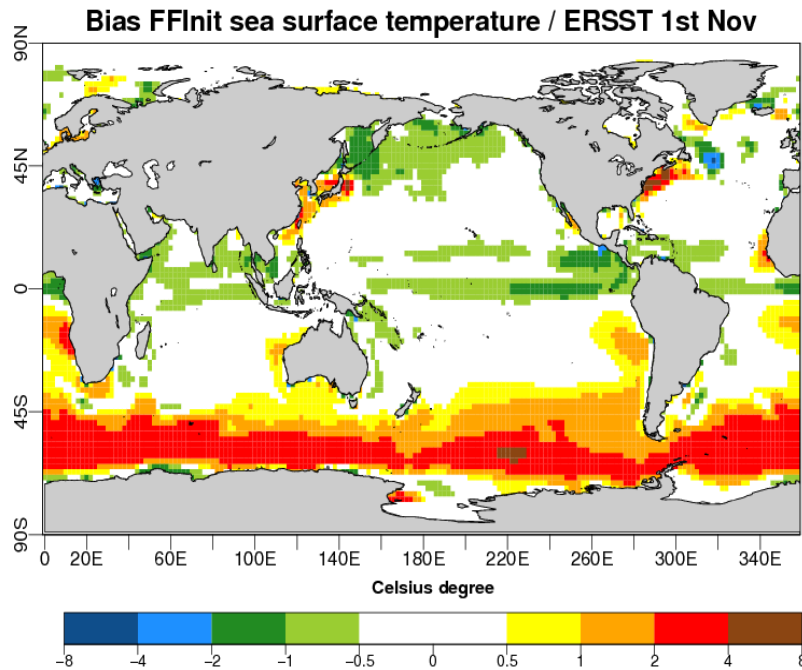
### Drift in global Sea Surface Temperature (60S–60N)



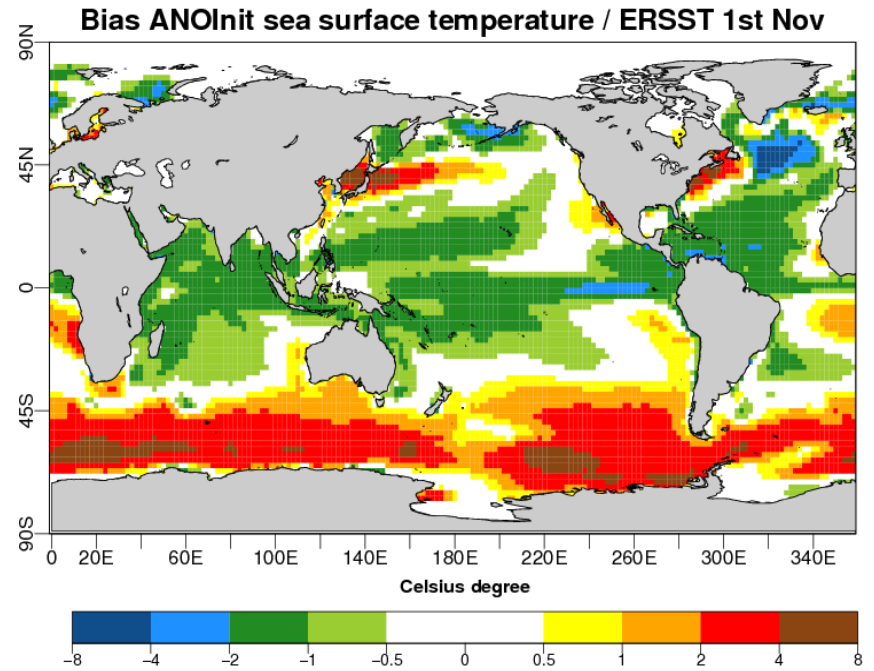


# Bias after 3 months

Full



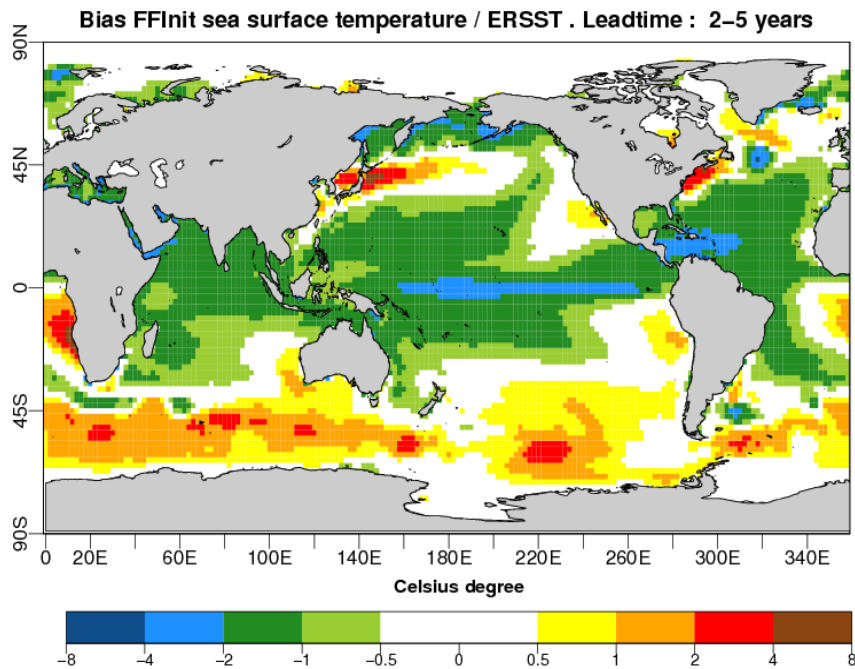
Anomaly



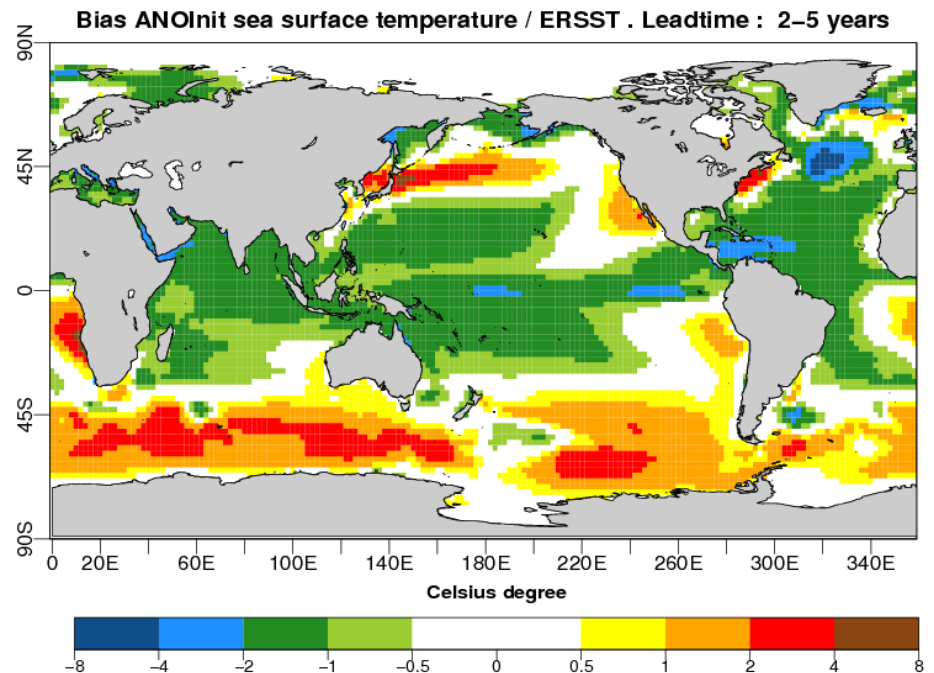


# Bias after 2-5 years

## Full



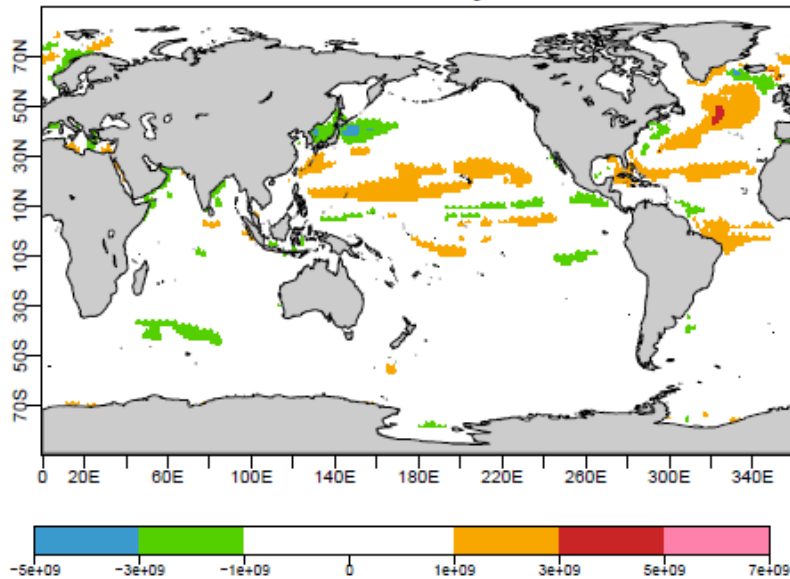
## Anomaly



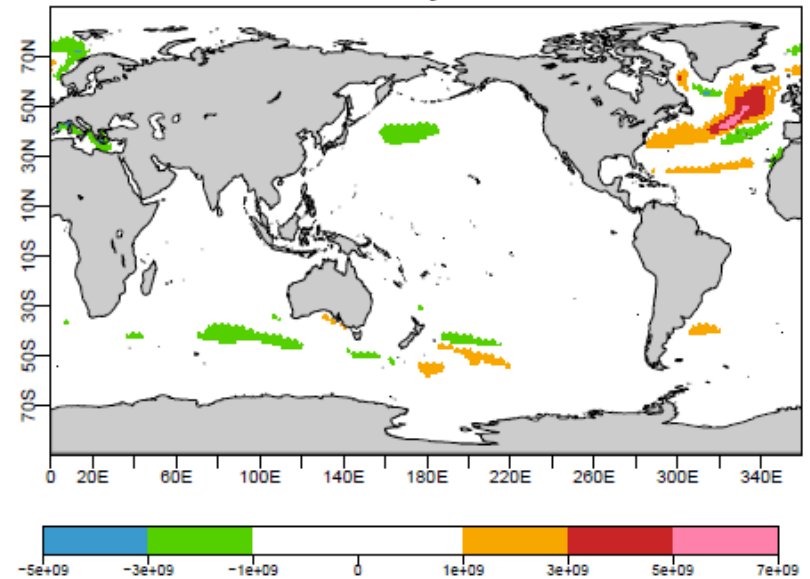


# Mixed layer heat content (ANOM minus FULL)

ML OHC , Nov, year 1



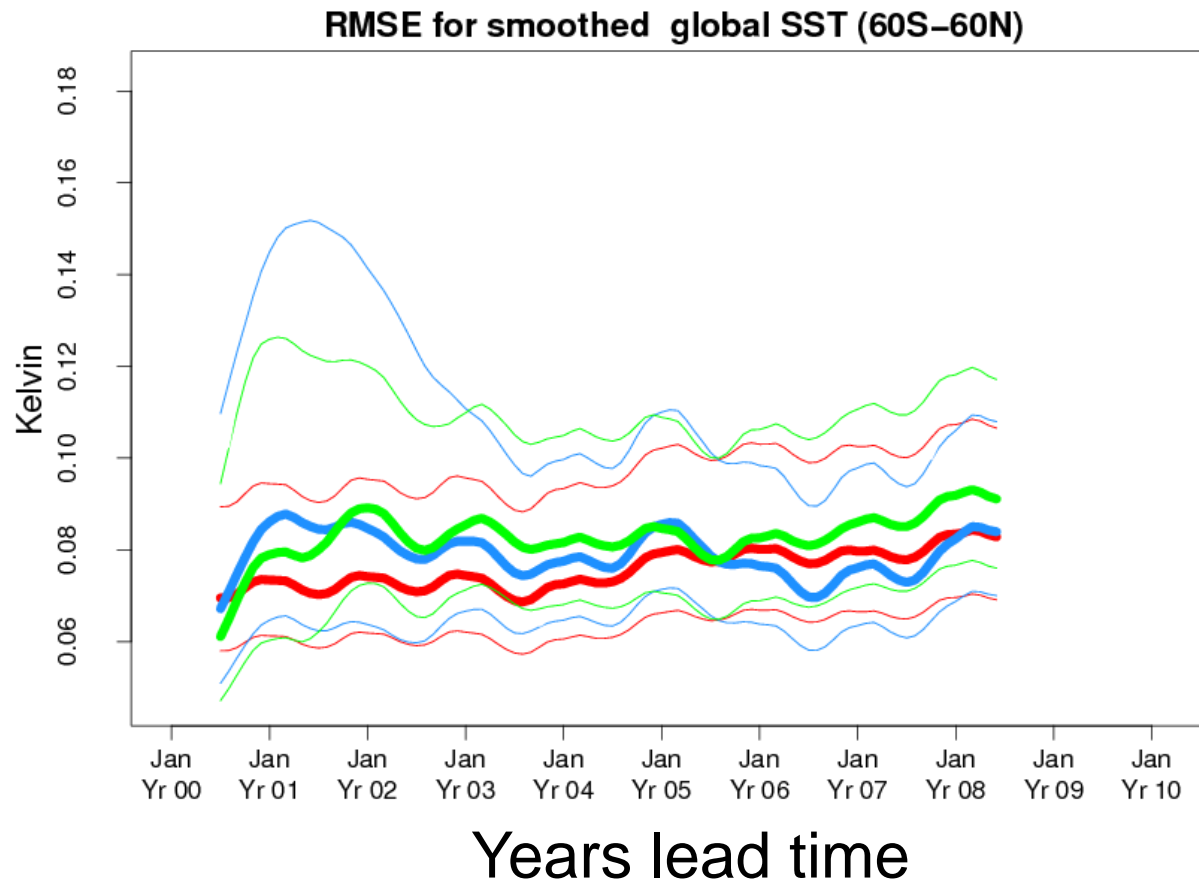
ML OHC , years 2-5







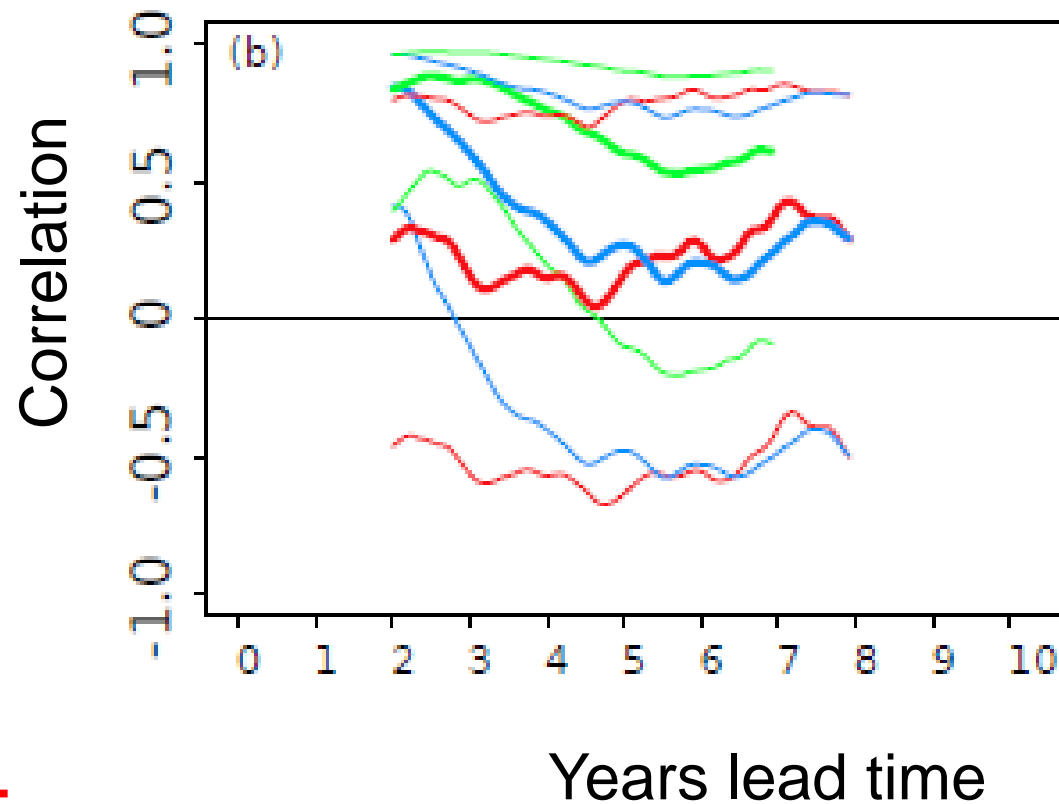
## RMSE of reforecasts (global)



**FULL**  
**ANOM**  
**NOINIT**



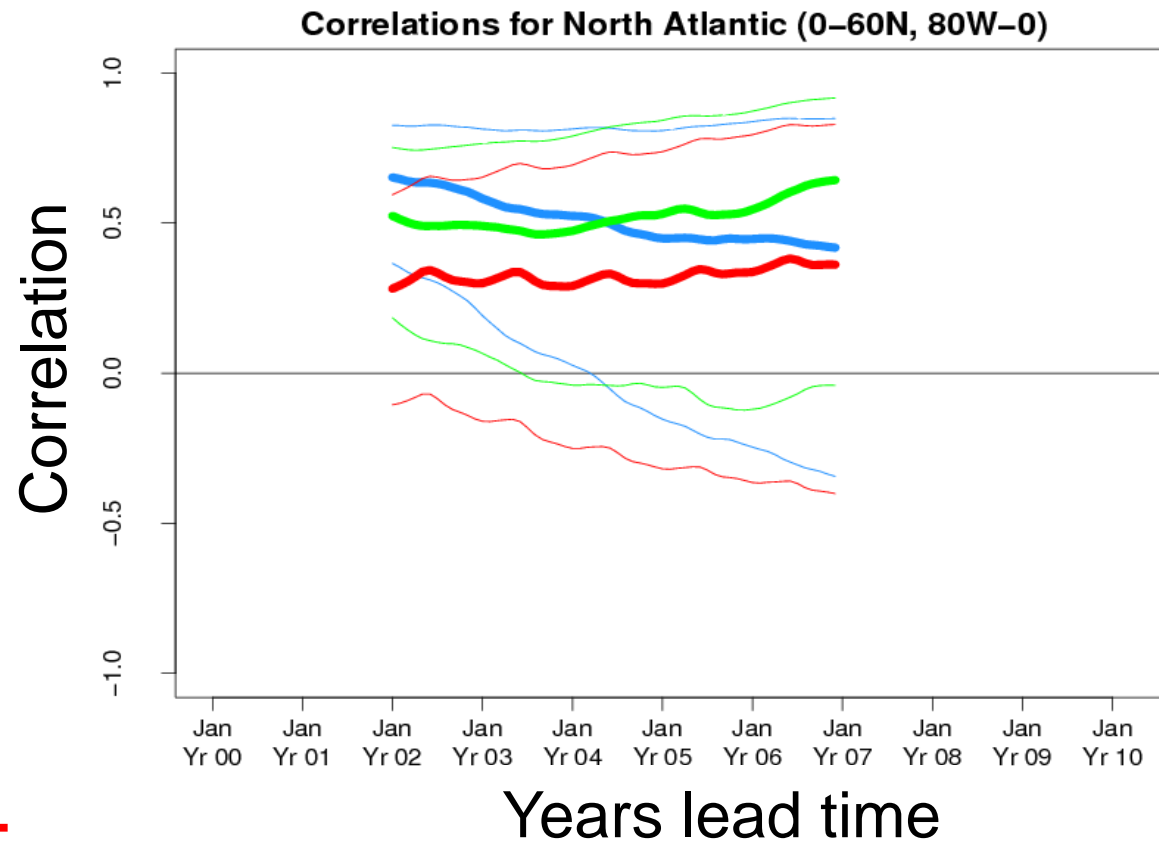
## North Atlantic scores with 5 yearly start dates



**FULL**  
**ANOM**  
**NOINIT**



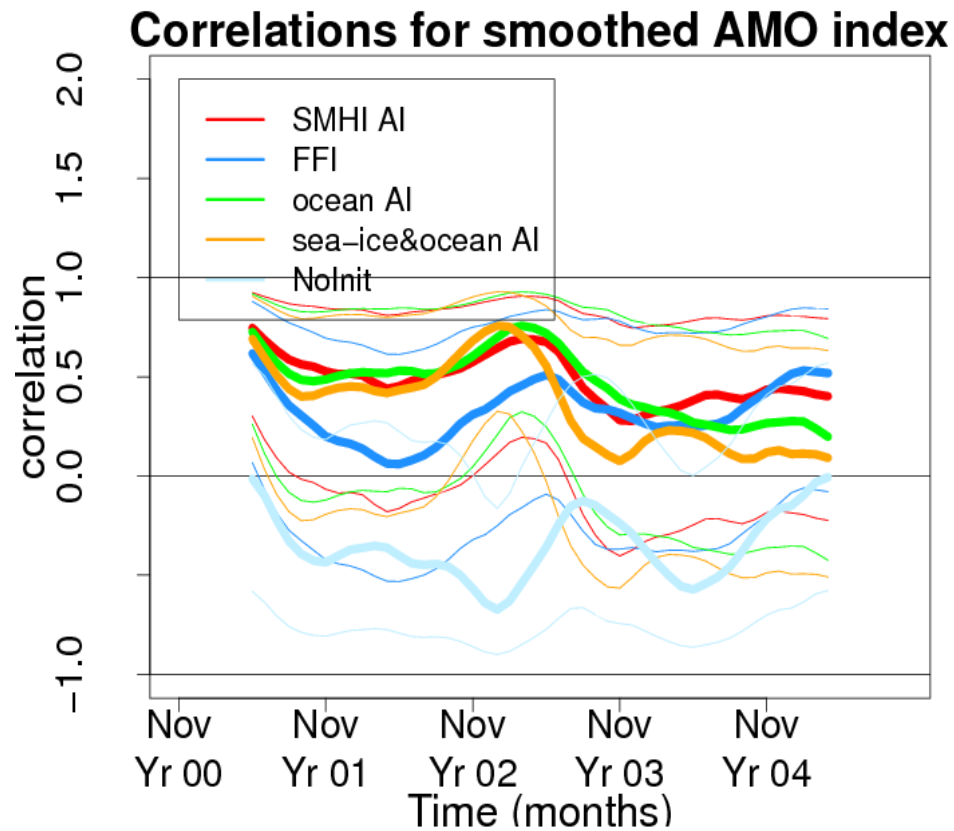
## North Atlantic scores with annual start dates



**FULL**  
**ANOM**  
**NOINIT**



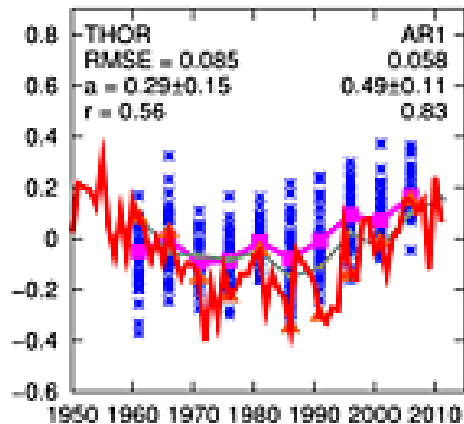
## North Atlantic scores with alternative anomaly initialisation





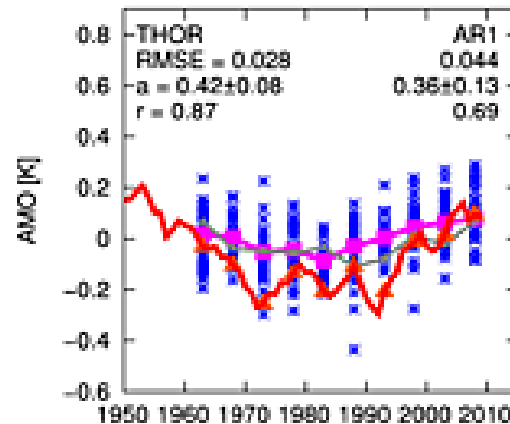
## But...hard to beat damped persistence

1 yr



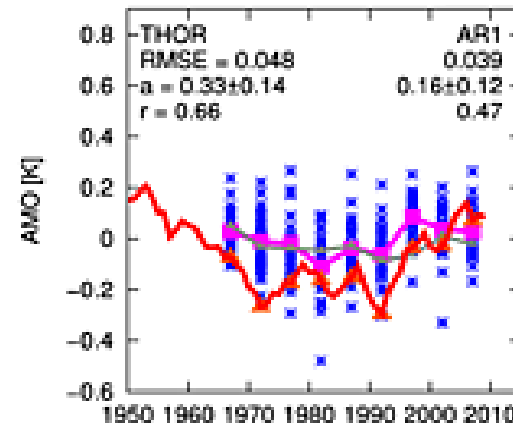
$$r_{\text{model}} = 0.56$$
$$r_{\text{ar1}} = 0.83$$

2-5 yr

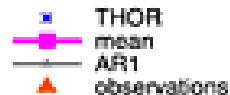


$$r_{\text{model}} = 0.87$$
$$r_{\text{ar1}} = 0.69$$

6-9 yr



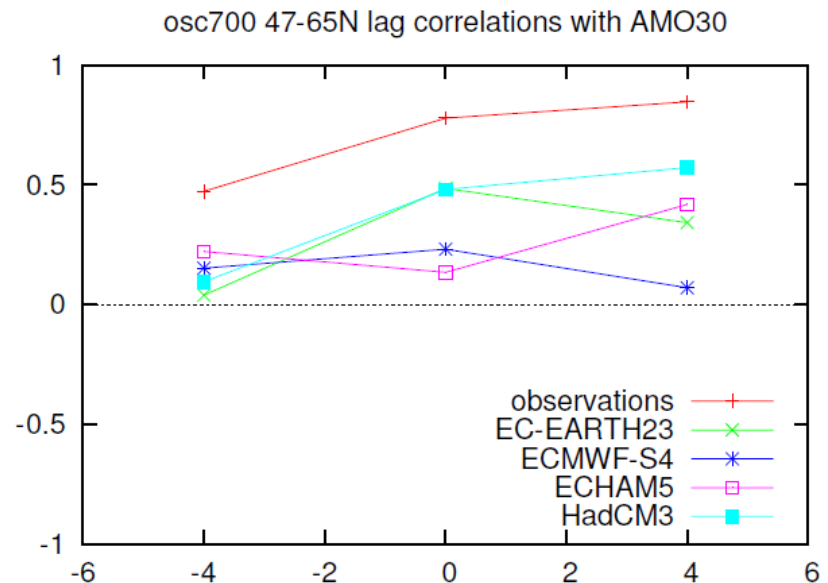
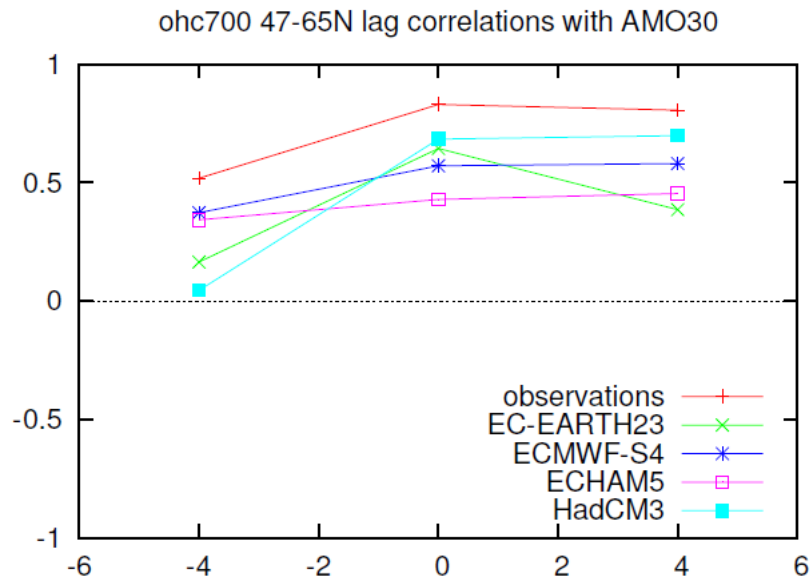
$$r_{\text{model}} = 0.66$$
$$r_{\text{ar1}} = 0.47$$



Hazeleger et al. Predicting multiyear North Atlantic Ocean variability  
JGR Oceans 2013



## But...models get skill for different reasons



← AMO lags    AMO leads →

← AMO lags    AMO leads →

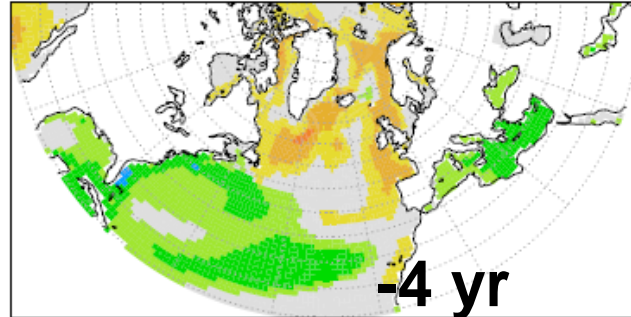
Lagged correlations between ocean heat content (0-700m)  
in subpolar gyre and the AMO



# Salt content – MOC relation

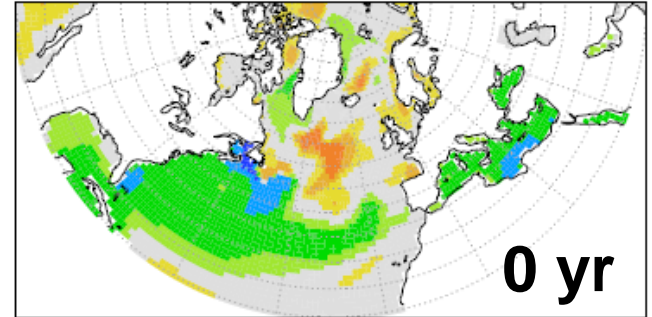
EC-Earth V2.3

corr EC-EARTH23 osc700 with MOC at 40N lag -48m



-4 yr

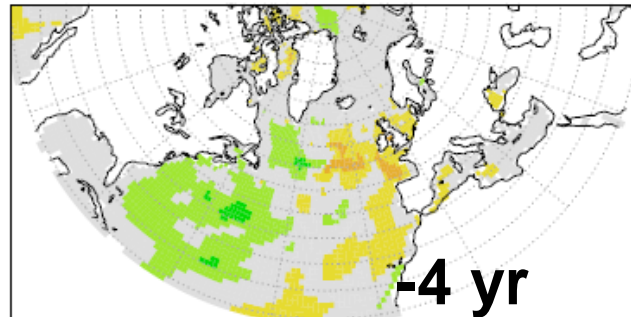
corr EC-EARTH23 osc700 with MOC at 40N lag 0m



0 yr

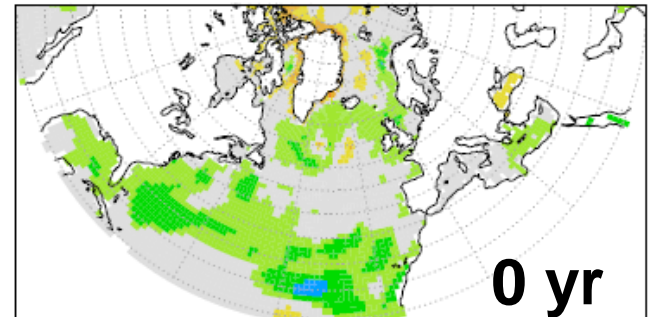
ECHAM5-OM

corr ECHAM5 osc700 with MOC at 40N lag -48m



-4 yr

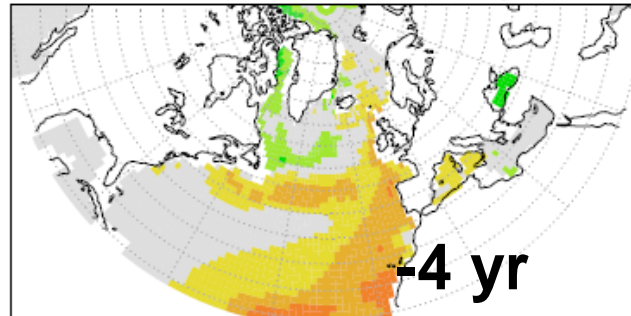
corr ECHAM5 osc700 with MOC at 40N lag 0m



0 yr

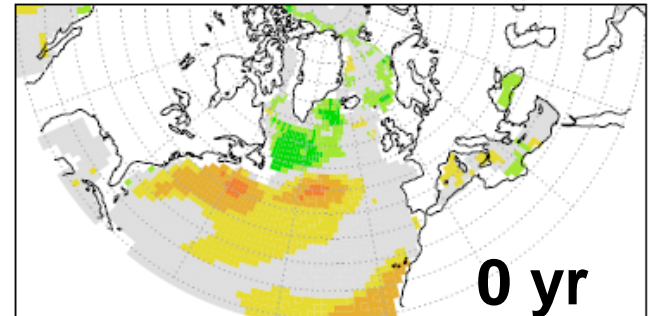
HADCM3

corr HadCM3 osc700 with MOC at 40N lag -48m



-4 yr

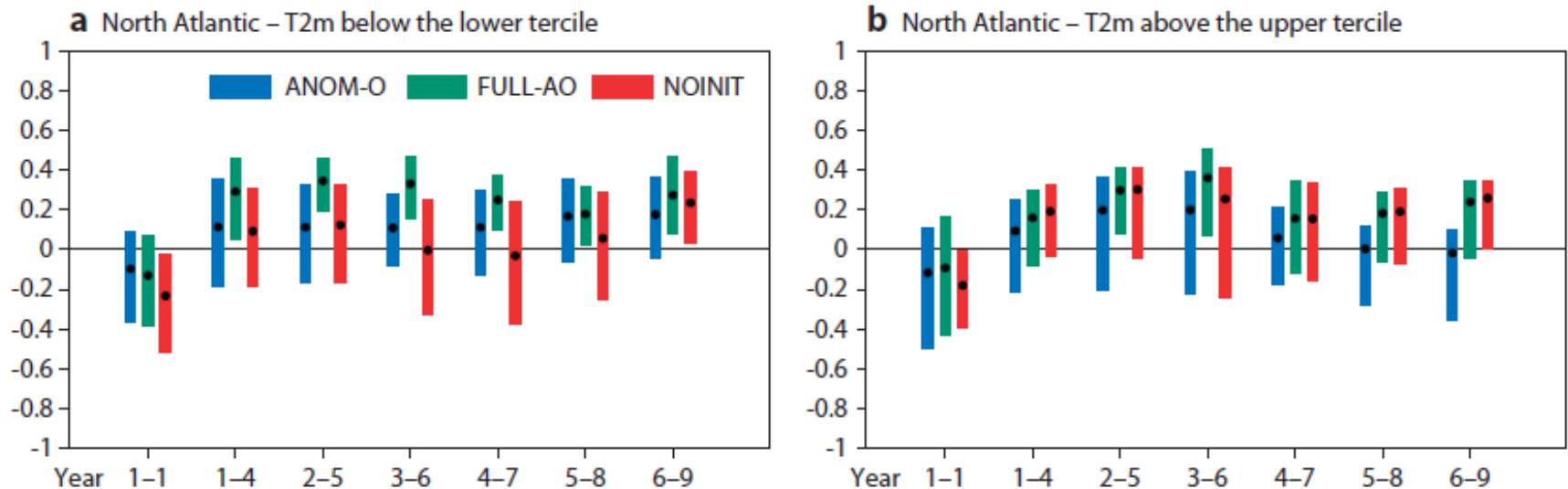
corr HadCM3 osc700 with MOC at 40N lag 0m



0 yr



## North Atlantic probabilistic skill scores

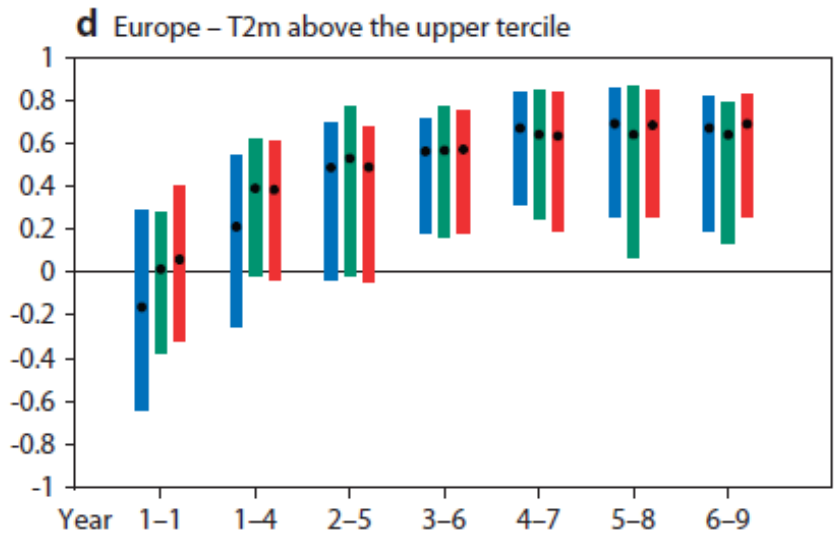
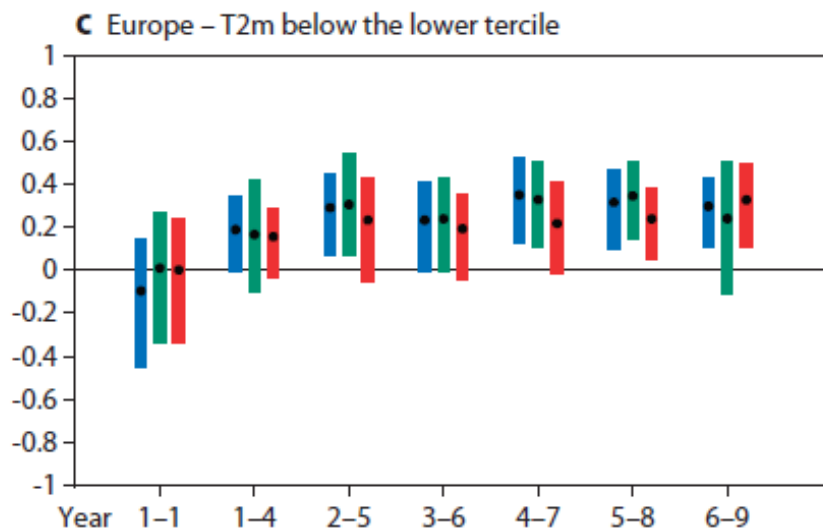


Brier skill score: compares predicted probability of events to a climatological reference forecast (based on 5 yr restarts).





# Europe probabilistic skill scores



**FULL**  
**ANOM**  
**NOINIT**



## Conclusions full state vs anomaly initialization

Drift is highest in FULL and can be traced back to known biases in EC-Earth

Warm Southern Ocean bias develops fast → atmospheric processes  
Cold bias develops slower → atmospheric & oceanic processes

FULL and ANOM have similar skill scores. Most skill from external forcing except for North Atlantic, where ANOM outperforms FULL

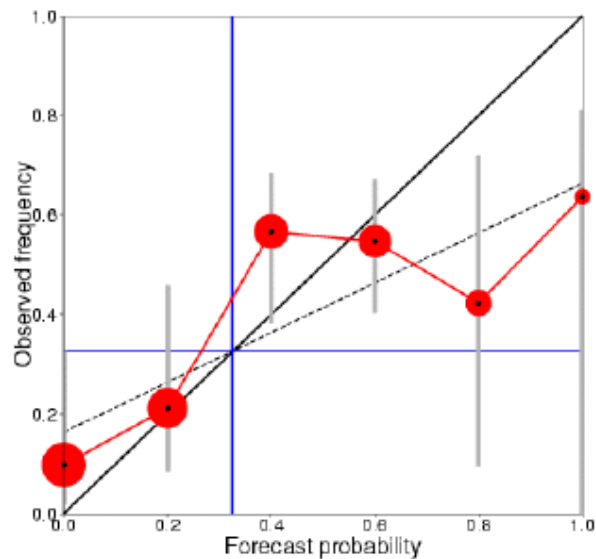
Indication for probabilistic skill over Europe on multiyear time scales

But ... Sampling has substantial impact on skill scores

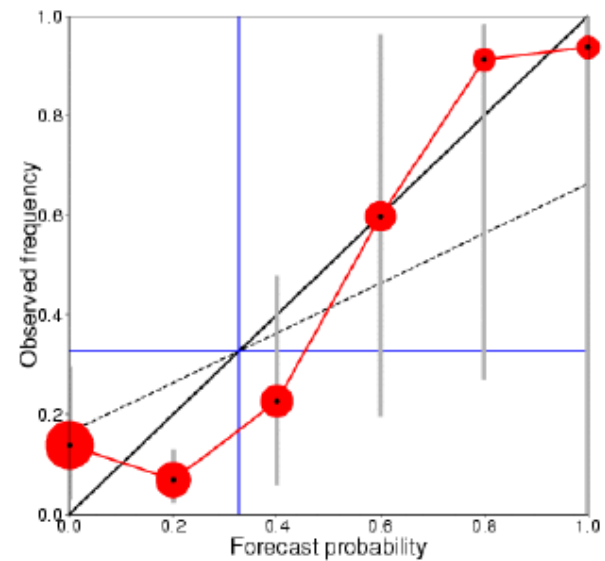
But ... This is a single model study using statistical scores, but mechanisms differ among models and should be understood and verified against observations!



**a) T2m Europe (land only) 2-5 yr NOINIT**  
**Below lower tercile**

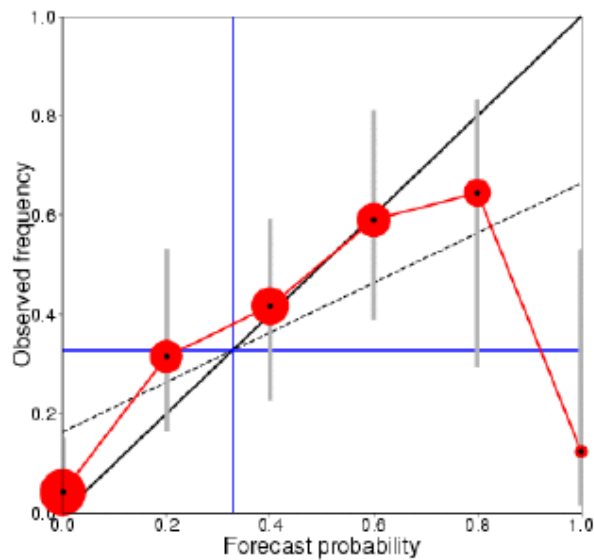


**b) T2m Europe (land only) 2-5 yr NOINIT**  
**Above upper tercile**

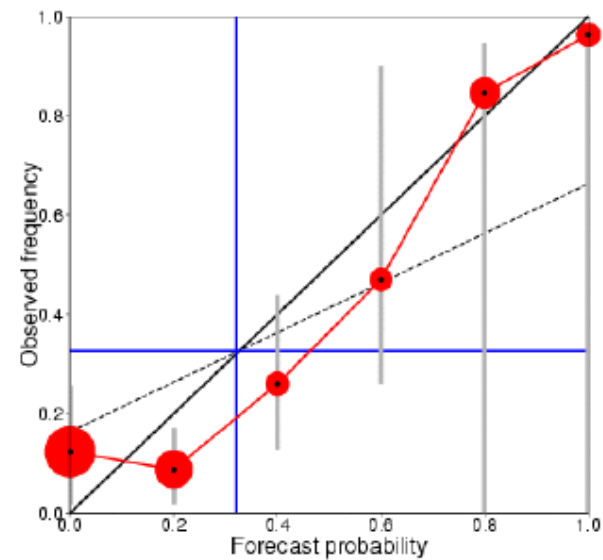




**c) T2m Europe (land only) 2-5 yr FULL**  
**Below lower tercile**

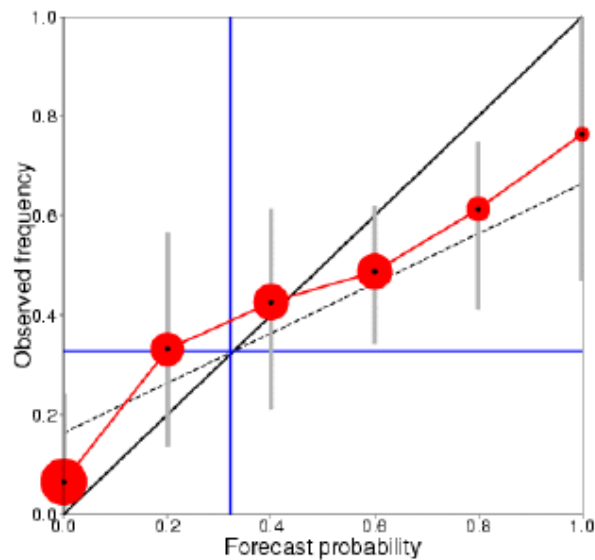


**d) T2m Europe (land only) 2-5 yr FULL**  
**Above upper tercile**

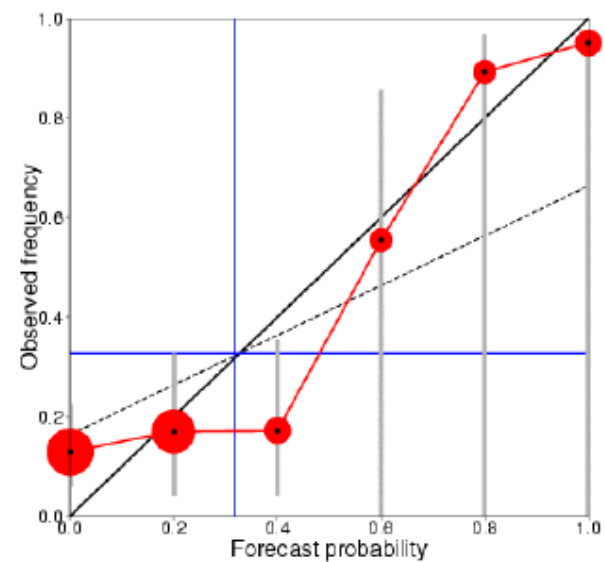




**e) T2m Europe (land only) 2-5 yr ANOM**  
**Below lower tercile**

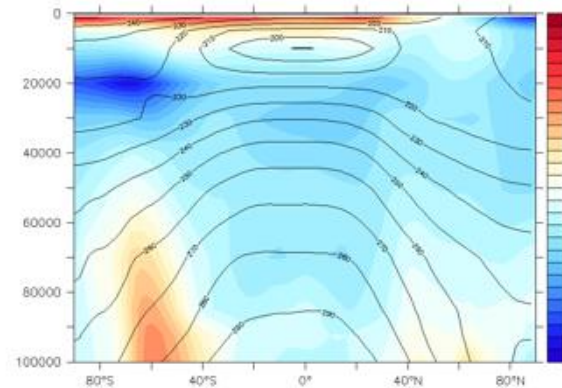


**f) T2m Europe (land only) 2-5 yr ANOM**  
**Above upper tercile**



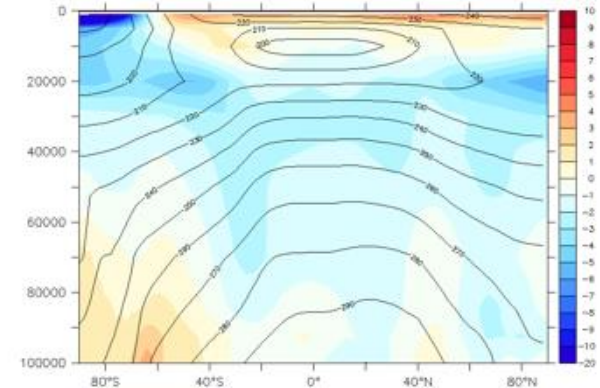


# Temperature bias



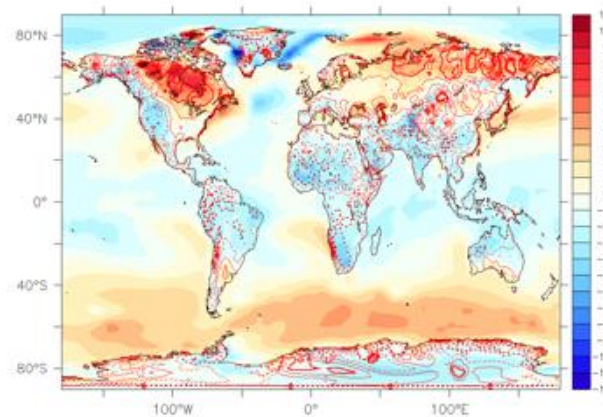
T PresentDay - ERAint(DJF)

DJF

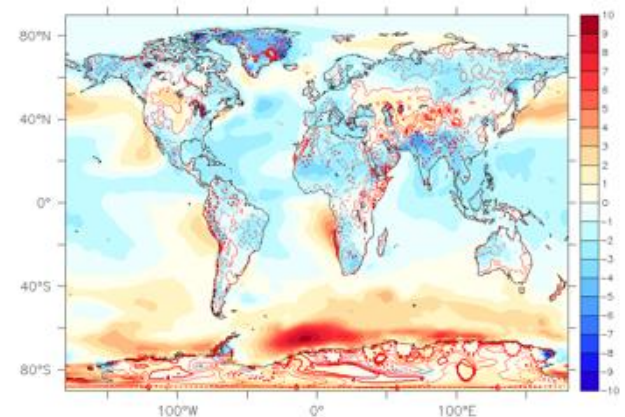


T PresentDay - ERAint (JJA)

JJA



T2M PresentDay - Reanalysis/Obs.(contour)(DJF)

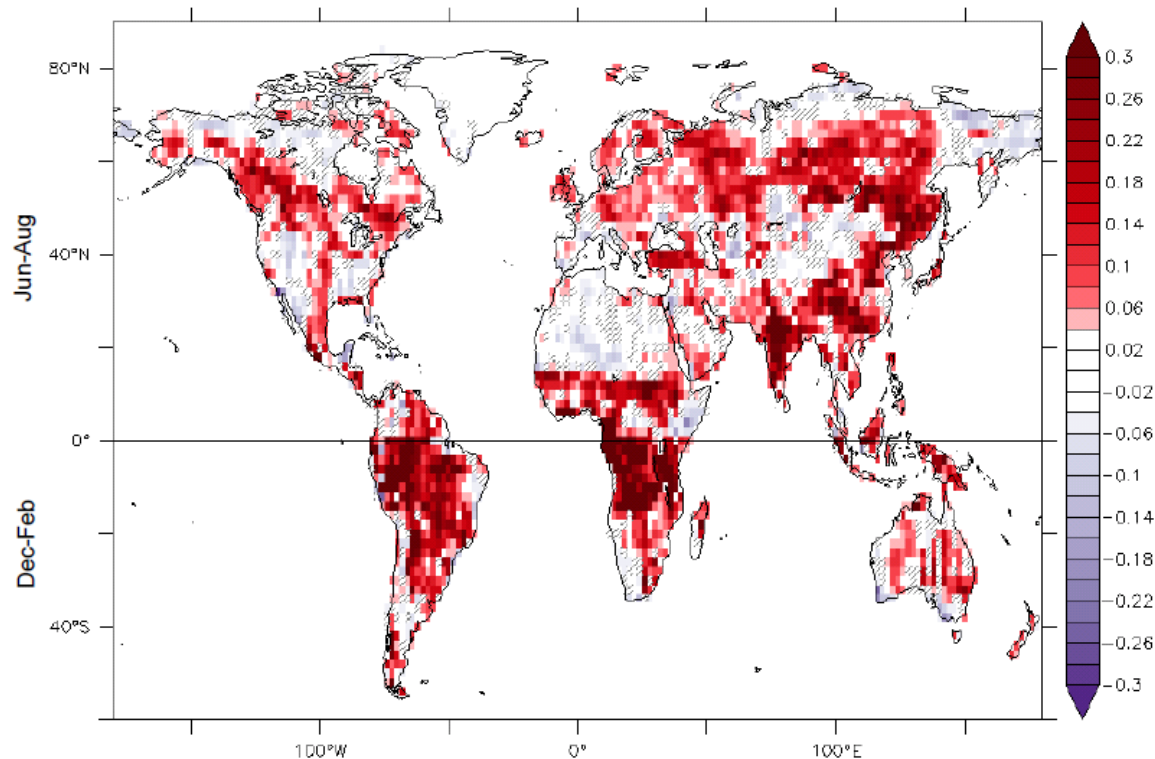


T2M PresentDay - Reanalysis/Obs.(contour) (JJA)





## Effect vegetation on predictability (Weiss et al 2012)



**Figure 3. Gain in potential predictability of evaporation during summer due to switching from climatological to time varying LAI values (PP of Exp E2 minus E1, 2000-2010). Summer is defined as in Fig 1. Hatched areas are not statistically significant at 95% level.**