

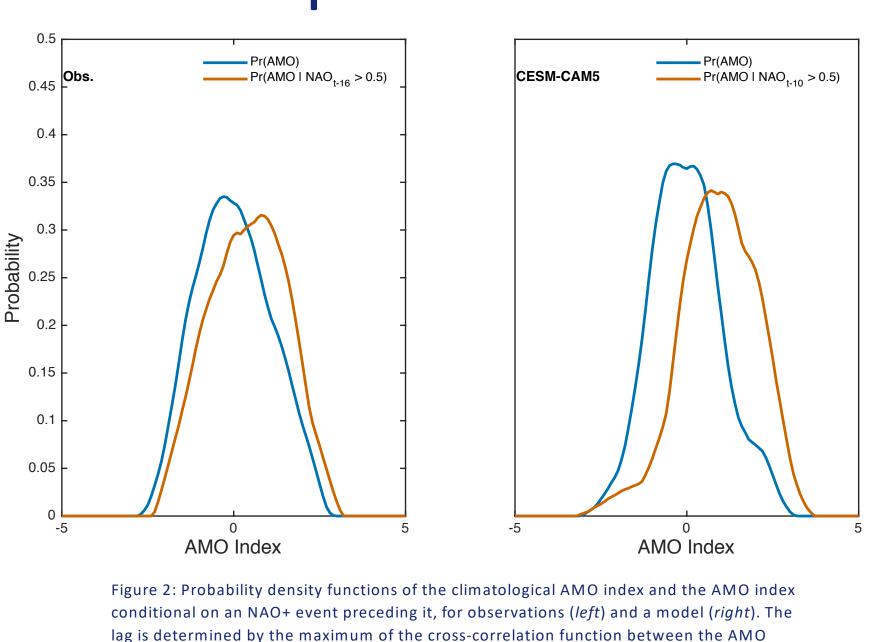
-20 through 20 (following the methodology of Delworth et al. 2017) correlation of the filter are colored white. Note the variability in the lag of maximum correlation across models as well as the discrepancy between models and observation

We cannot reject the null hypothesis that positive lagged correlations in the subtropics are an artifact of filtering. The mechanism responsible for the lagged response to the NAO may be limited to the subpolar gyre.

and NAO.

### Results: NAO adds little predictive skill

The distribution of the AMO index conditional on a prior NAO+ event is statistically significantly different from climatology. However, the climatological probability of an AMO warm event ( $\pm$  0.15) °C anomaly) is 23% and the probability conditional on a prior NAO+ event is 26%, in observations.



# Variable external forcing obscures the weak relationship between the NAO and north Atlantic multi-decadal SST variability

## Results: NAO explains a small portion of AMO variance

The linear lagged SST response to the NAO only accounts for between 1% and 12% of the variance in the AMO index in control runs of CMIP5 models (4% on average), and 19% in observations.

On average, the unlagged, low-pass filtered NAO index explains only ~5% and ~7% of the variability in the AMO index in control runs of CMIP5 models and observations, respectively.

## Results: Variable external forcing obscures NAO-AMO relationship

— CanESM2 — CCSM4

CMCC-CE

- CMCC-CMS

- CSIRO-Mk3-6-

— GFDL-CM3 — GFDL-ESM2G

-GISS-E2-H-CC GISS-E2-H

-GISS-E2-R-CC

-HadGEM2-AO HadGEM2-CC

– IPSL-CM5A-LF

- IPSL-CM5A-MF

IPSL-CM5B-LR

MIROC4h

MIROC-ESN

MPI-ESM-MR

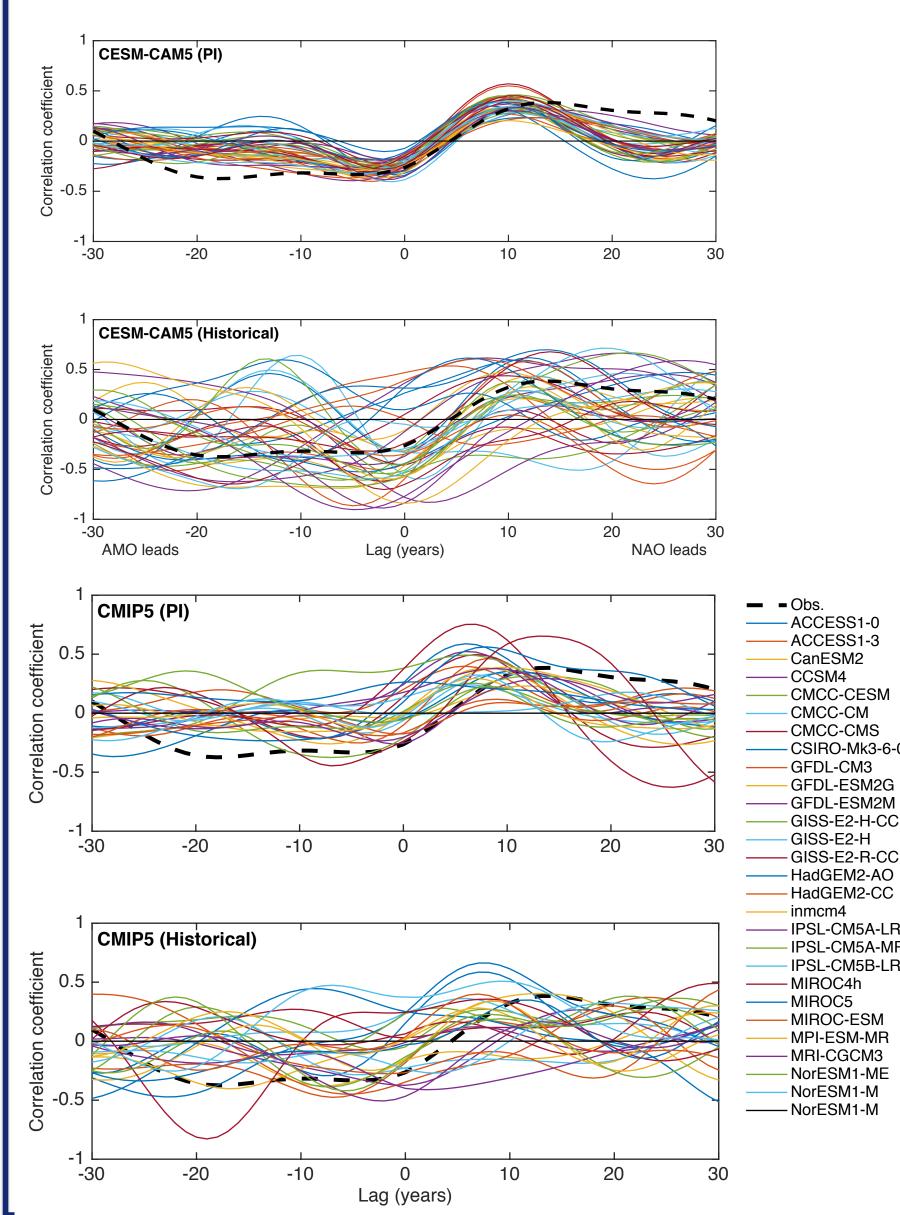
MRI-CGCM3

NorESM1-ME

NorESM1-M

MIROC5

inmcm4



	Years	Lag of maximum correlation (years)	% AIVIO <sub>MID</sub> variance explained by NAOt=0	% AMO variance explained by NAOt=0	% AIVIOMID variance explained by lagged	% AMO variance explained by lagged
					response	response
Observations						
ERSST/NAO	153	15	20%	7%	31%	19%
Models						
CESM1-CAM5	899	10	10%	0%	6%	4%
CESM1-SOM	899	-	44%	20%	-	-
GFDL-CM2.1	499	7	3%	0%	7%	8%
CIMIP5 models						
ACCESS1-0	499	6	0%	3%	7%	7%
ACCESS1-3	499	6	0%	0%	5%	3%
CanESM2	995	12	8%	2%	2%	3%
CCSM4	1050	12	3%	2%	4%	3%
CMCC-CESM	276	13	24%	16%	4%	8%
CMCC-CM	329	9	13%	5%	0%	2%
CMCC-CMS	499	7	1%	1%	8%	12%
CSIRO-Mk3-6-0	499	11	3%	1%	1%	3%
GFDL-CM3	499	8	0%	0%	8%	9%
GFDL-ESM2G	499	8	2%	0%	1%	1%
GFDL-ESM2M	499	6	3%	2%	5%	5%
GISS-E2-H-CC	250	6	1%	0%	0%	1%
GISS-E2-H	779	6	18%	15%	4%	3%
GISS-E2-R-CC	250	7	8%	11%	11%	11%
HadGEM2-AO	699	8	1%	0%	7%	6%
HadGEM2-CC	239	9	36%	32%	3%	4%
inmcm4	499	10	7%	3%	1%	1%
IPSL-CM5A-LR	999	9	6%	1%	3%	4%
IPSL-CM5A-MR	299	7	12%	1%	1%	6%
IPSL-CM5B-LR	299	22	4%	0%	3%	1%
MIROC4h	99	13	11%	23%	1%	1%
MIROC5	999	1	0%	0%	1%	1%
MIROC-ESM	629	23	22%	8%	1%	1%
MPI-ESM-MR	999	7	3%	0%	3%	5%
MRI-CGCM3	499	14	5%	0%	3%	5%
NorESM1-ME	251	8	5% 6%	2%	2%	4%
NorESM1-M	500	7	10%	0%	1%	1%
CMIP5	500	9.4	8%	5%	3%	4%
Average		5.4	070	370	<b>3</b> 70	

correlation for observations is lower than other estimates in the literature.

In pre-industrial control runs, most models exhibit a lagged warm response to the NAO. However, when model runs account for variations in external forcing, the NAO-AMO relationship is obscured. The influence of external forcing is noted in both an ensemble of a single model (CESM-LENS) and a multi-model ensemble (CMIP5).

> Figure 3: Cross-correlation functions between the low-pass filtered NAO index and the low-pass filtered AMO index. The dashed black line is observations and is the same in each panel. Note that the top panel is created by sub-sampling a single long PI run of CESM1 into 85-year segments to allow for direct comparison to CESM-LENS.

## Discussion: Is the observed NAO-AMO relationship due to chance alone?

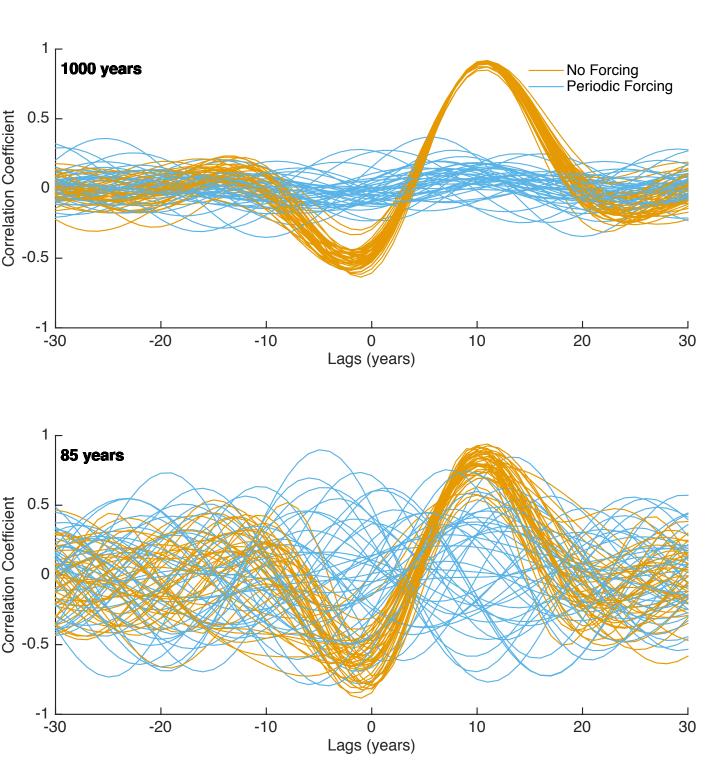
 $AMO_{LP} = \beta_1 NAO_{LP,t=0} + \beta_2 NAO_{LP,t=10} + \beta_3 \cos \omega t$ 

In the stochastic, statistical model above, we prescribe the NAO-AMO relationship as well as the influence of variable external forcing. Coefficients are calculated via independent linear regressions, yielding values of -0.05 C/std. dev., 0.1 C/std. dev., and -0.68 C/unit forcing. Inclusion of the  $\beta_3$  term interrupts or obscures the prescribed lagged relationship between the NAO and AMO.

We find evidence to support the hypothesis that ocean dynamics play a role in multidecadal SST variability; however, its contribution to overall variability and predictability in the region is small. When climate models include variable external forcing, the NAO-AMO relationship is obscured. Historical runs of climate models as well as a statistical model allow for the possibility that the observed relationship between the NAO and AMO is due to chance alone.

## Outstanding questions

- of AMO transitions?
- variance?
- variability
- AMO?



without the B<sub>3</sub> term

## Summary

What (if any) is the role of ocean heat transport in setting the timing

Through what mechanism does external forcing change AMO

What induces non-stationarity in the AMO index? When did variable external forcing become a key influence on AMO

What details of external forcing are valuable for prediction of the

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