

# The Central Role of Ocean Dynamics in Connecting the North Atlantic Oscillation to the Extratropical Component of the Atlantic Multidecadal Oscillation

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(Manuscript received 5 May 2016, in final form 27 January 2017)

## ABSTRACT

The relationship between the North Atlantic Oscillation (NAO) and Atlantic sea surface temperature (SST) variability is investigated using models and observations. Coupled climate models are used in which the ocean component is either a fully dynamic ocean or a slab ocean with no resolved ocean heat transport. On time scales less than 10 yr, NAO variations drive a tripole pattern of SST anomalies in both observations and models. This SST pattern is a direct response of the ocean mixed layer to turbulent surface heat flux anomalies associated with the NAO. On time scales longer than 10 yr, a similar relationship exists between the NAO and the tripole pattern of SST anomalies in models with a slab ocean. A different relationship exists both for the observations and for models with a dynamic ocean. In these models, a positive (negative) NAO anomaly leads, after a decadal-scale lag, to a monopole pattern of warming (cooling) that resembles the Atlantic multidecadal oscillation (AMO), although with smaller-than-observed amplitudes of tropical SST anomalies. Ocean dynamics are critical to this decadal-scale response in the models. The simulated Atlantic meridional overturning circulation (AMOC) strengthens (weakens) in response to a prolonged positive (negative) phase of the NAO, thereby enhancing (decreasing) poleward heat transport, leading to broad-scale warming (cooling). Additional simulations are used in which heat flux anomalies derived from observed NAO variations from 1901 to 2014 are applied to the ocean component of coupled models. It is shown that ocean dynamics allow models to reproduce important aspects of the observed AMO, mainly in the Subpolar Gyre.

## 1. Introduction

Numerous studies have examined sea surface temperature (SST) variability in the Atlantic and associated climatic impacts. These studies have shown that on short (interannual) time scales there exists a tripole pattern of SST anomalies that arises as a response of the oceanic mixed layer to turbulent surface heat flux anomalies driven by variations of the North Atlantic Oscillation (NAO) (Bjerknes 1964; Daly 1978; Cayan 1992; Battisti et al. 1995; Gulev et al. 2013). A more intriguing question, with greater relevance for larger-scale climate, is what drives SST variations in the Atlantic on decadal-to-multidecadal time scales. Analyses of both instrumental data and proxy climate reconstructions have shown the existence of a

basinwide monopole SST anomaly pattern that varies on multidecadal time scales. This has been called (Kerr 2000) the Atlantic multidecadal oscillation (AMO), or Atlantic multidecadal variability (AMV), and has been linked to numerous important climate phenomena around the world. These include droughts in Africa, changing tropical storm activity in the Atlantic, and drought over interior North America. (Folland et al. 1986; Enfield et al. 2001; Sutton and Hodson 2005; Knight et al. 2006; Zhang and Delworth 2006; Chylek et al. 2009, 2014; Mahajan et al. 2011; Nigam et al. 2011; Sutton and Dong 2012; Hu and Veres 2016).

Many previous studies have hypothesized that this AMO pattern is a result of variations in ocean circulation (Bjerknes 1964) involving the Atlantic meridional overturning circulation (AMOC). Extensive prior work has shown that AMOC variations occur in models on decadal-to-multidecadal time scales and that these AMOC variations produce monopole SST patterns in the Atlantic that resemble the observed AMO (Delworth et al. 1993; Knight 2005; Danabasoglu et al. 2012). However, models vary widely in the time scale of their AMOC and SST variability

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and in the underlying mechanisms producing that AMOC variability (Grossmann and Klotzbach 2009; Keenlyside et al. 2013). Further, while the models typically produce SST patterns that resemble observations in the Subpolar Gyre, the simulated anomalies are typically smaller than observed in the tropical North Atlantic. Recent work (Clement et al. 2015) has provided a different perspective, suggesting that ocean circulation changes do not play a substantial role in the observed AMO in models with a dynamic ocean. They suggest that the AMO is a direct response of the ocean to atmospherically generated NAO variations and associated turbulent surface heat flux variations, combined with wind–evaporation–SST feedback in the tropics. However, this new perspective is challenged by recent analyses showing that ocean dynamics plays a central role in the AMO (Zhang et al. 2016; O'Reilly et al. 2016).

In this study, we reexamine the connection between the NAO and Atlantic SST variability across a range of time scales using a combination of observational analyses and specifically designed climate model experiments. We show that ocean dynamics is critical to understanding the processes driving key aspects of the AMO in models and observations, especially in the Subpolar Gyre. AMO variability in the subtropics likely involves additional atmospheric processes, such as cloud or dust feedback (Bellomo et al. 2015; Martin et al. 2014; Evan et al. 2009; Wang et al. 2012; Yuan et al. 2016; Brown et al. 2016).

## 2. Observational data and model simulations

We use a combination of observational analyses and specifically designed numerical experiments.

### a. Observational data

We use sea surface temperatures (SST) from ERSST data (Smith et al. 2008; available online at <https://www.esrl.noaa.gov/psd/data/gridded/tables/sst.html>) and an index of the NAO based on normalized station data (Hurrell et al. 2016) over the December–March (DJFM) period. We use both annual mean and winter (DJFM) mean SST and find comparable results, so we only show results using annual mean SST.

### b. Models

We use two climate models, with two variants of each model. The first model is GFDL CM2.1 (Delworth et al. 2006), consisting of an atmospheric general circulation model coupled to an ocean general circulation model. The horizontal resolution of the atmosphere is approximately 200 km, with 24 vertical levels. The horizontal resolution of the ocean is approximately 1°, with meridional refinement to 0.33° in the deep tropics. The model ocean has 50 levels in the vertical. We also use the

Forecast-Oriented Low Ocean Resolution (FLOR; Vecchi et al. 2014) version of GFDL CM2.5 (Delworth et al. 2012). FLOR uses similar atmospheric physics as CM2.1, but with considerably higher spatial resolution in the atmosphere and land (approximately 50 km), with 32 vertical levels in the atmosphere. The ocean component of FLOR has the same spatial resolution as CM2.1, with similar physics. For each model (CM2.1 and FLOR) a variant is constructed in which the dynamic ocean component is replaced by a slab of fixed 50-m depth (CM2.1\_SLAB and FLOR\_SLAB, respectively), with no interannual variations of ocean heat transport. In order for the slab models to have a realistic mean state and seasonal cycle of SST in the absence of resolved ocean heat transport, we add to the slab ocean model an additional heat flux adjustment term. The flux adjustments are calculated using preliminary separate simulations of the CM2.1\_SLAB and FLOR\_SLAB models in which model SST are restored (with a 5-day restoring time scale) to observed monthly SST over the period 1971–2012. The observed monthly SSTs are interpolated to daily values for the restoring runs. The time mean of the restoring heat flux used in these simulations is then defined as the heat flux adjustment term (separate adjustment fields for the CM2.1\_SLAB and FLOR\_SLAB models). The heat flux adjustment term varies as a function of space and the seasonal cycle but is constant from one year to the next. The various simulations of the CM2.1\_SLAB and FLOR\_SLAB models are then conducted in which the heat flux adjustment terms are applied. Using these flux adjustments, the slab models have considerably reduced SST biases relative to the models with dynamic oceans.

We conduct several types of experiments with various sets of the available models (CM2.1, CM2.1\_SLAB, FLOR, and FLOR\_SLAB). The various experiments are listed in Table 1 and described below. With each model we conduct multicentury control simulations in which the atmospheric composition and radiative forcing is held fixed at either preindustrial (1860) or “modern” (1990) conditions. Using CM2.1, CM2.1\_SLAB, and FLOR, we conduct experiments (CM2.1\_HIST, CM2.1\_SLAB\_HIST, and FLOR\_HIST) in which estimates of the time-varying radiative forcing over the period 1861–2015 are applied to the model. We use ensembles to better define the response to radiative forcing. For CM2.1 and CM2.1\_SLAB\_HIST, we use 10-member ensembles, whereas for FLOR\_HIST we use 5-member ensembles. The ensemble members are started from widely separated points in their respective control simulations. The response to radiative forcing is defined as the ensemble mean from the simulation with time-varying radiative forcing minus the ensemble mean from the corresponding sections of the control simulations.

TABLE 1. List of experiments.

No.	Experiment name	Model	Atm resolution	Ocean type	Radiative forcing	Extra NAO forcing
1	CM2.1_Ctrl	CM2.1	~200 km	Dynamic	Const 1860	None
2	CM2.1_Ctrl_NAO_50yr	CM2.1	~200 km	Dynamic	Const 1860	50-yr periodic
3	CM2.1_HIST	CM2.1	~200 km	Dynamic	Historical 1951–2014	None
4	CM2.1_HIST_NAO	CM2.1	~200 km	Dynamic	Historical 1951–2014	Obs 1951–2014
5	CM2.1_SLAB_Ctrl	CM2.1_SLAB	~200 km	50-m slab	Const 1860	None
6	CM2.1_SLAB_HIST	CM2.1_SLAB	~200 km	50-m slab	Historical 1901–2014	None
7	CM2.1_SLAB_HIST_NAO	CM2.1_SLAB	~200 km	50-m slab	Historical 1901–2014	Obs 1901–2014
8	CM2.1_SLAB_Ctrl_NAO_50yr	CM2.1_SLAB	~200 km	50-m slab	Const 1860	50-yr periodic
9	FLOR_Ctrl	FLOR	~50 km	Dynamic	Const 1860	None
10	FLOR_Ctrl_NAO_50yr	FLOR	~50 km	Dynamic	Const 1860	50-yr periodic
11	FLOR_HIST	FLOR	~50 km	Dynamic	Historical 1951–2014	None
12	FLOR_HIST_NAO	FLOR	~50 km	Dynamic	Historical 1951–2014	Obs 1951–2014
13	FLOR_SLAB_Ctrl	FLOR_SLAB	~50 km	50-m slab	Control 1990	None

We also conduct simulations to explore the response of the system to an imposed NAO forcing. In these simulations we impose on the model ocean an additional pattern of surface heat flux anomalies that has the spatial pattern of the NAO. The pattern of the imposed heat flux anomalies is shown in Fig. 1a of [Delworth and Zeng \(2016\)](#). Specifically, after the atmosphere–ocean heat flux is calculated in the model, we add an extra term to the heat flux going into the model ocean. This extra heat flux has the spatial pattern of the NAO. We derive this pattern by computing a linear regression at each grid point between the time series of surface heat fluxes from the ECMWF interim reanalysis (ECMWF-Interim; [Dee et al. 2011](#)) and a time series of the observed NAO index. Both the NAO index and the surface heat fluxes are time means over the DJFM period. The surface heat flux includes the latent and sensible terms, as well as shortwave and longwave radiative terms. The spatial pattern of the NAO-related heat flux anomalies is fixed, but we modulate the amplitude of the flux forcing in time in various ways as described below. We constrain the heat flux anomaly so that its spatial integral is zero. Therefore, the imposition of this anomaly pattern does not directly add heat to (or subtract heat from) the climate system. This technique is similar to that employed in an earlier pioneering study using an ocean-only model ([Eden and Jung 2001](#)), in contrast to the fully coupled model used here.

### 1) IDEALIZED FORCING EXPERIMENTS

In a first set of experiments we add the specified pattern of NAO-related heat flux forcing to the ocean component of the control simulations for CM2.1, CM2.1\_SLAB, and FLOR. The NAO flux forcing is modulated in time by a sine wave with a 50-yr period, the amplitude of which corresponds to one standard deviation of the observed NAO time series (identified in [Table 1](#) as CM2.1\_Ctrl\_NAO\_50yr, CM2.1\_SLAB\_Ctrl\_NAO\_50yr,

and FLOR\_Ctrl\_NAO\_50yr). This 50-yr time scale is idealized but is loosely based on the observed NAO variations over the twentieth century, which have substantial variability on multidecadal time scales (see, e.g., [Hurrell et al. 2016](#)). The simulations are 100-yr long, with 10-member ensembles for CM2.1 and CM2.1\_SLAB and 5-member ensembles for FLOR.

### 2) REALISTIC FORCING EXPERIMENTS

In an additional set of experiments we add to the historical simulations (CM2.1\_HIST, CM2.1\_SLAB\_HIST, and FLOR\_HIST) the spatial pattern of the NAO heat flux, but multiplied each year by the observed value of the NAO index over the period 1901–2014 for the CM2.1 and CM2.1\_SLAB models, and over the period 1951–2014 for the higher-resolution (and more computationally expensive) FLOR model. These experiments are identified in [Table 1](#) as CM2.1\_HIST\_NAO, CM2.1\_SLAB\_HIST\_NAO, and FLOR\_HIST\_NAO. Differences between the experiments with NAO forcing and their counterparts without NAO forcing (e.g., CM2.1\_HIST\_NAO minus CM2.1\_HIST) show the climatic impact of the NAO forcing added to the model ocean. This yields some assessment of the contribution of NAO variations to observed climate variations in terms of the impact of the NAO on the ocean and subsequent feedback to the atmosphere. In these experiments there is no anomalous flux forcing applied to the atmosphere; the anomalous NAO fluxes are only directly felt by the ocean. The observed NAO index is taken from the Climate Data Guide ([Hurrell et al. 2016](#)). The heat flux anomalies are only applied over the December–March period. We use ensembles to better estimate the response to the NAO (10 members for CM2.1\_HIST\_NAO and CM2.1\_SLAB\_HIST\_NAO and 5 members for FLOR\_HIST\_NAO).

The simulations with dynamic oceans have been previously analyzed and published in [Delworth and Zeng \(2016\)](#)

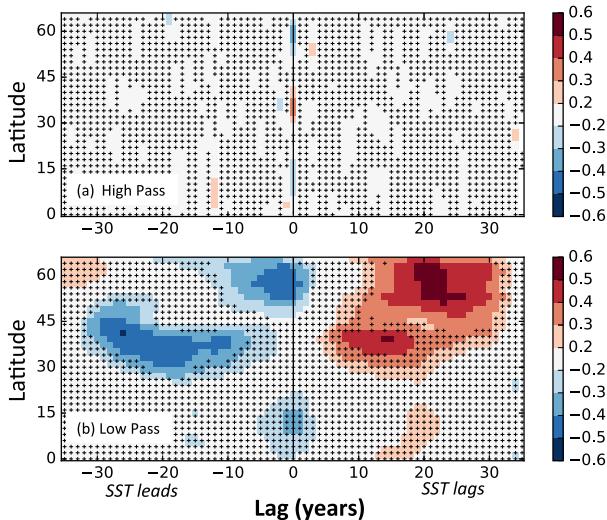


FIG. 1. Zonal averages of the lagged correlations between annual mean SST and the NAO (for DJFM) based on observations. Correlations were first computed at each grid point and then zonally averaged from  $60^{\circ}$  to  $20^{\circ}$ W. Negative (positive) lags denote years prior to (after) a maximum in the NAO. Lag 0 indicates a correlation between the NAO (December of year 0 through March of year 1) with annual mean SST (January of year 1 through December of year 1). (a) Correlations calculated using HP data (filtered to retain time scales shorter than 10 yr). (b) Correlations calculated using LP data (filtered to retain time scales longer than 10 yr). For both (a) and (b), the stippling denotes points that do not pass a statistical significance test (described in detail in the appendix). The analyses shown here were based on time series that were not detrended. Analyses using detrended time series produce similar results, with somewhat larger correlations.

and Delworth et al. (2016), whereas the simulations with slab oceans have not previously been analyzed.

We note that the model computes its own internal NAO variability, in addition to the imposed NAO forcing. This creates spread among the ensemble members in simulating the model response to the NAO forcing, since the model ocean responds both to the imposed NAO forcing and the internally generated NAO forcing. We evaluated the spread of the NAO by resampling the control simulations of each model. Based on these analyses [see also the methods section of Delworth et al. (2016)] we conclude that the NAO computed in each model is an important source of noise and ensemble spread but that the NAO-forced signal is still able to emerge from this noise when the NAO forcing is of sufficiently large amplitude, as is the case for the primary multidecadal swings of the NAO in the observational record.

### 3. Observed relationship between NAO and North Atlantic SST

We first examine the relationship between the NAO and SST in observations as a function of time scale. We

filter the observed time series for the NAO and SST to retain either time scales shorter than 10 yr [high-pass filtered (HP)] or longer than 10 yr [low-pass filtered (LP)], using a finite impulse response filter with 10 weights (Bloomfield 1976). We calculate the lagged correlations between the time series of annual mean SST at each grid point and the NAO time series for both the HP and LP data over the period 1861–2014. More specifically, a correlation at lag 0 refers to the correlation of the NAO index for the period of December (year 0) through March (year 1) with annual mean SST for January through December of year 1. As discussed in the appendix, statistical significance was estimated using a Monte Carlo resampling technique. We show results in which the time series were not detrended prior to the analyses, we find results that are generally similar to those shown below, with somewhat larger amplitudes.

At short time scales (Fig. 1a) the largest-amplitude correlations occur in the year immediately after the NAO maximum (defined as lag 0) and correspond to a tripole pattern, with negative values in the Subpolar Gyre and tropical North Atlantic and positive values in midlatitudes. This is consistent with atmospheric surface flux forcing of the ocean mixed layer, as shown by many past studies (Cayan 1992; Battisti et al. 1995). Correlation coefficients at other lags are considerably smaller.

We show the LP results in Fig. 1b. There is a tripole-like pattern for small lags, but the largest correlations occur at much larger lags. Positive correlations, covering most of the North Atlantic, reach their maximum 15–30 yr after the NAO maximum. This decadal lagged relationship is distinctly different than the direct surface flux forcing of the mixed layer shown for the HP analyses.

We also note that there are significant negative correlations when SST leads the NAO by a decade or two. We present analyses in sections 4b and 4d to suggest that these negative correlations in fact represent a response to a preceding negative phase of the NAO, just as the positive correlations at a lag of 1–2 decades represent a response to a positive phase of the NAO. Underlying these relationships is the fact that the NAO has substantial multidecadal variability in the (relatively short) observed record.

The spatial patterns of the correlation coefficients are shown in Fig. 2. The pattern with maximum correlations using HP data is shown in Fig. 2a, occurring for annual mean SST immediately following the DJFM NAO, and is a clear tripole. Shown in Fig. 2b are the correlations using LP data when SST lags the NAO by 15 yr. The pattern is of largely uniform sign across the basin, resembling the observed AMO (Sutton and Hodson 2005), and is distinctly different than the tripole structure. We show in Figs. 2c–h the temporal evolution of the LP SST

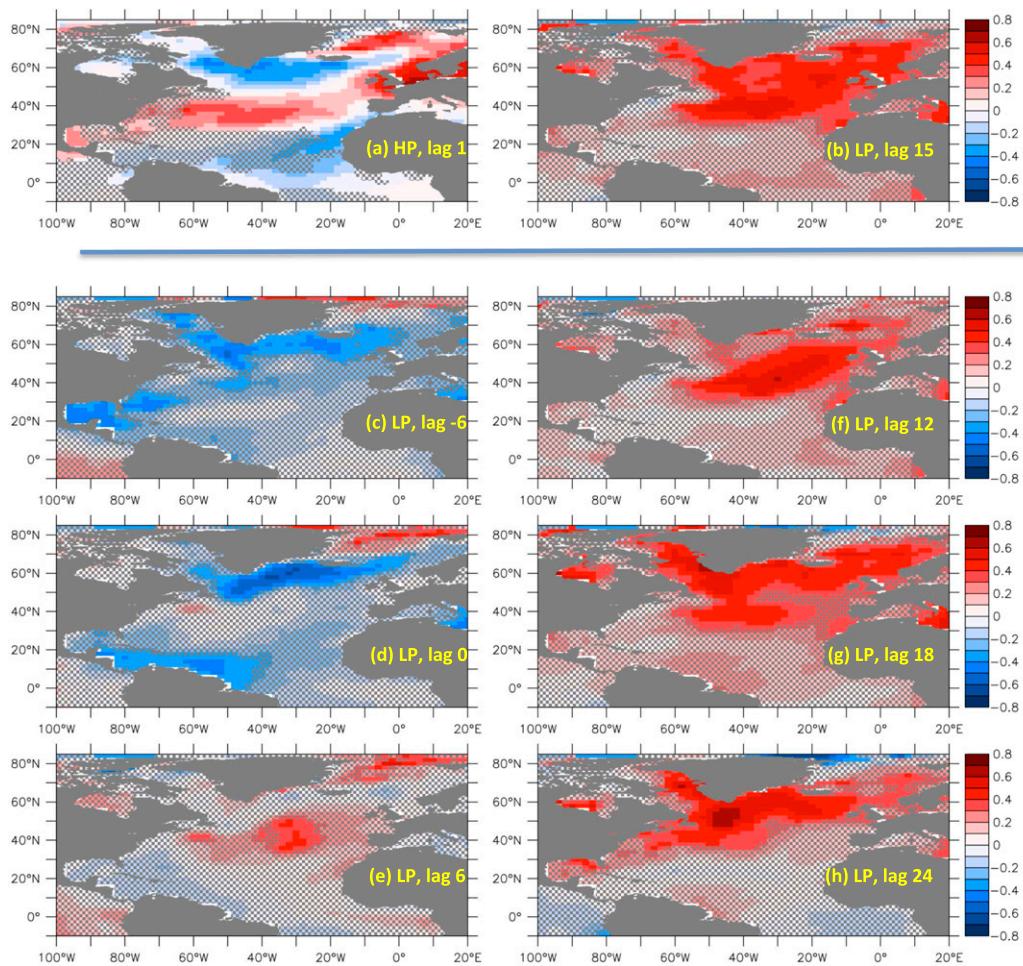


FIG. 2. Maps of the correlation coefficient between observed annual mean SST at each grid point and the NAO index. (a) Correlation coefficients between the DJFM NAO index and annual mean SST in the immediately following year (January–December) using the HP data. (b) Correlation coefficients when annual mean SST lags the DJFM NAO index by 15 yr using the LP data. (c) Correlation coefficients between annual mean SST and the NAO using low-frequency data when the SST leads the NAO by 6 yr. (d)–(h) As in (c), but for a lag of (d) 0, (e) 6, (f) 12, (g) 18, and (h) 24 yr. Correlations were computed without detrending (analyses using detrending produced similar results, with somewhat larger correlations). Regions without stippling are significant at the 80% confidence level using the two-sided Student's  $t$  test.

signal with respect to the NAO. For periods before the NAO maximum, SST anomalies are generally negative over the North Atlantic. A few years after the NAO reaches its maximum, positive SST anomalies develop in the central North Atlantic. This region expands northward and equatorward, encompassing most of the North Atlantic by 10–15 yr after the NAO maximum, with the largest signal in the subpolar North Atlantic. The initial appearance of the positive SST anomaly in the central North Atlantic and its subsequent apparent propagation appear consistent with a significant role for ocean dynamics, but this is difficult to explore solely from observational analyses. We therefore turn to modeling experiments to explore further

the relationship between the NAO and Atlantic decadal SST variability.

#### 4. Simulated relationship between NAO and North Atlantic SST

In this section we explore the relationship between the NAO and North Atlantic interannual-to-multidecadal SST variability in a suite of climate model simulations.

##### a. Control simulations

We first explore this relationship in a set of control simulations. We show in Fig. 3 the zonal means of the

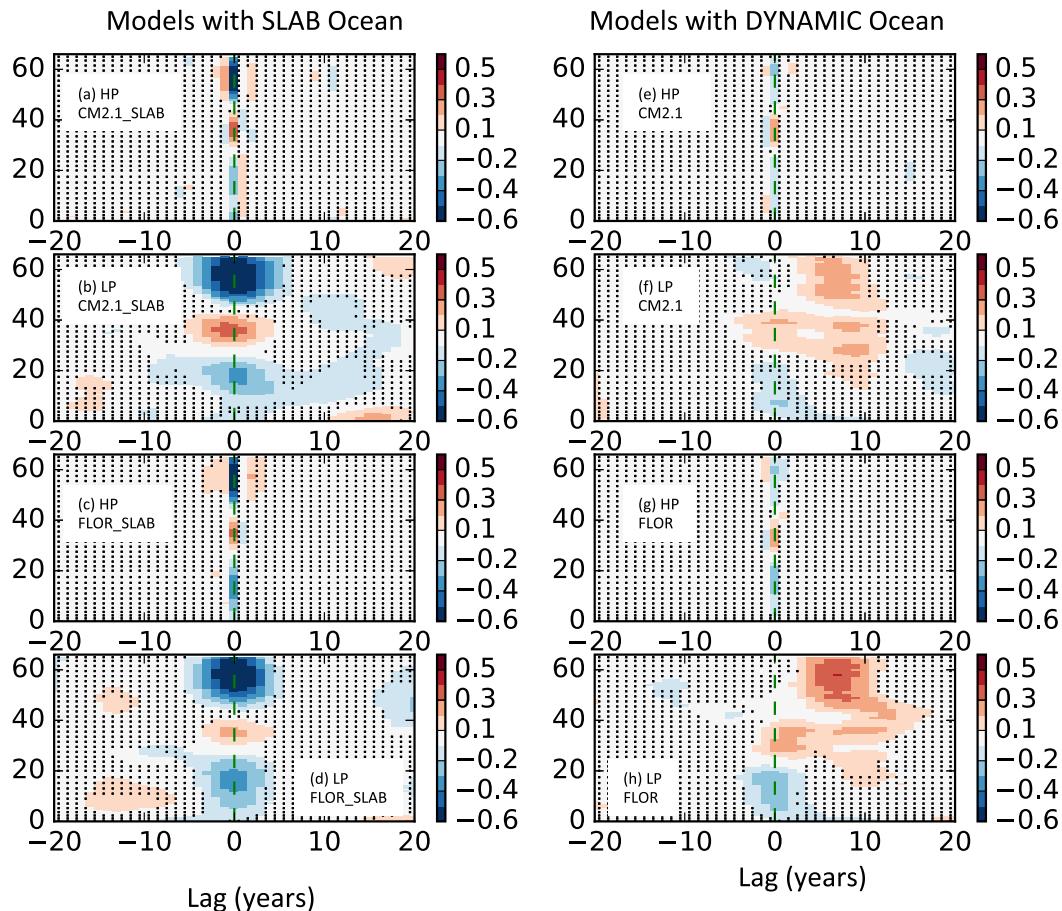


FIG. 3. Zonal mean ( $60^{\circ}$ – $20^{\circ}$ W) of the correlation coefficient between simulated annual mean SST and the model's DJFM NAO index, calculated from various control simulations: (a) HP output, CM2.1\_SLAB; (b) LP output, CM2.1\_SLAB; (c) HP output, FLOR\_SLAB; (d) LP output, FLOR\_SLAB; (e) HP output, CM2.1; (f) LP output, CM2.1; (g) HP output, FLOR; and (h) LP output, FLOR. Data were time filtered prior to analysis. HP data retain only time scales shorter than 10 yr, while LP data contain only time scales longer than 10 yr. Negative (positive) lags along the  $x$  axis indicate years prior to (following) a maximum NAO value. A lag of 0 yr (dashed line) denotes a correlation coefficient calculated between the December–March NAO index and the mean SST for January–December, where the January–March period is the same for the NAO and SST. The stippling denotes points that do not pass a statistical significance test (described in detail in the [appendix](#)).

correlation coefficients between annual mean SST anomalies and the DJFM NAO as a function of lag. We show results for HP and LP data. We also show results for models using either a slab ocean (Figs. 3a–d) or a dynamic ocean (Figs. 3e–h). For all results with models using a slab ocean model (i.e., no ocean dynamics), the largest correlations are at lag 0 (HP data) or are centered around lag 0 (LP data), with no substantial correlations at lags greater than  $\pm 5$  yr. This is consistent with observational analyses at short time scales (cf. Figs. 3a,c to Fig. 1a) but inconsistent with observations at long time scales (cf. Figs. 3b,d to Fig. 1b). When using a dynamic ocean (Figs. 3e–h), the models are similar to observations at short time scales (Figs. 3e,g), with maximum correlations at lag 1 yr, and the spatial structure is consistent with the

tripole pattern. The model correlations are somewhat weaker than those from observational analyses. This difference could be related to sampling uncertainty with far fewer points in the observational record or to incorrectly represented processes in the models.

At long time scales the models with dynamic oceans show the maximum correlation lagged a decade or so after the NAO maximum (Figs. 3f,h). This lag is shorter in the models than in observations (Fig. 1b). This phase lag between the NAO and SST anomalies at long time scales, similar to observations, only appears when ocean dynamics are considered.

We show in Fig. 4 the spatial pattern of the correlations between SST and the NAO at the lag corresponding to the largest correlations based on Fig. 3. For

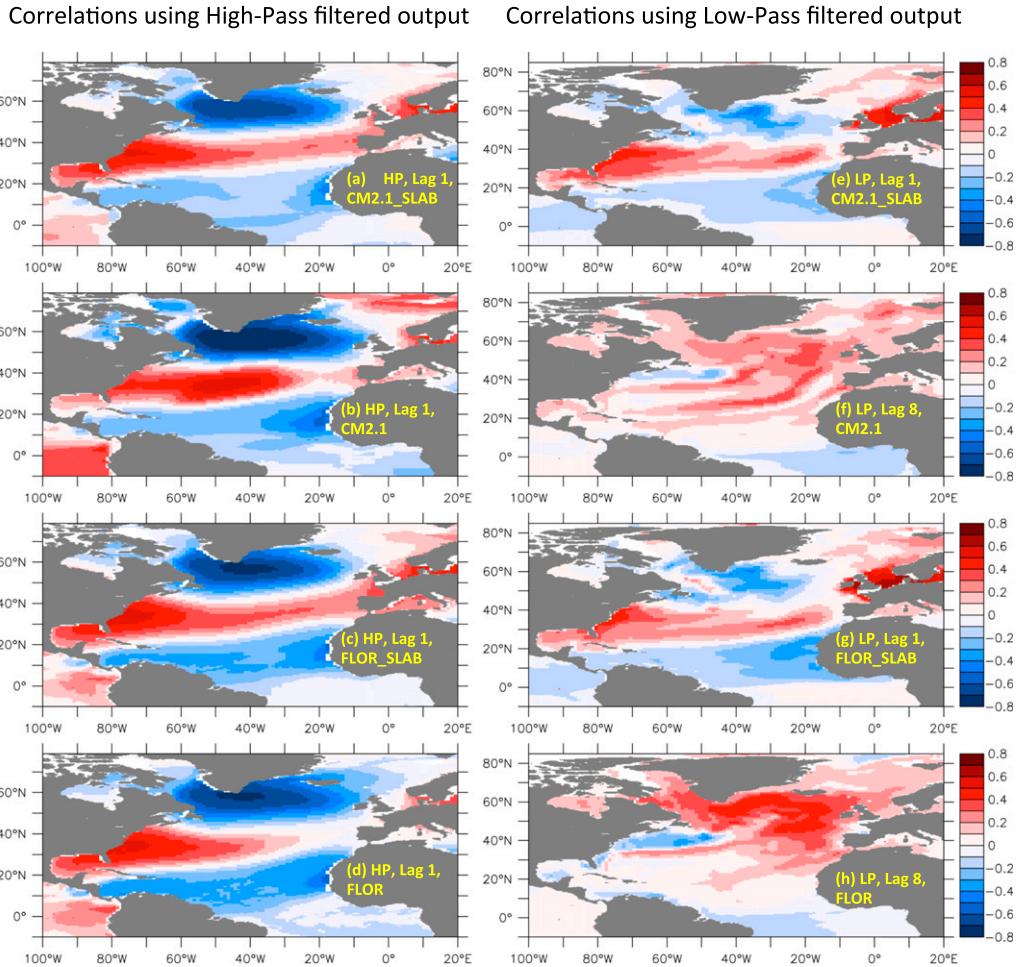


FIG. 4. Maps of the correlation coefficients between annual mean SST and the NAO index in various models. Each map is at a lag (yr) where the overall field of correlations is at a maximum. Positive lags indicate the NAO leading SST. LP (HP) indicates data have been filtered prior to the correlation analysis to retain time scales longer than (shorter than) 10 yr.

all of the cases using HP data (Figs. 4a–d), the maximum correlation occurs for the year immediately after the NAO maximum and resembles the tripole pattern. This was true regardless of whether the model used a slab ocean (Figs. 4a,c) or a dynamic ocean (Figs. 4b,d). In contrast, there are differing behaviors using LP data (Figs. 4e–h). For the models with slab oceans (Figs. 4e,g), the maximum correlation is also at the year immediately following the NAO maximum, and the spatial correlation is the familiar tripole pattern. In contrast, for models with dynamic oceans (Figs. 4f,h), the maximum correlation occurs approximately 7–10 yr after the NAO maximum and is characterized by a monopole, AMO-like pattern over the North Atlantic, somewhat resembling the observations (Fig. 1b), although with a smaller-than-observed amplitude in the tropical North Atlantic. Again, this points to a fundamental role of

ocean dynamics for decadal-scale Atlantic SST variability, especially in the Subpolar Gyre region.

#### b. Simulations with periodic NAO forcing

We further explore the relationship between the NAO and Atlantic SST anomalies by analyzing simulations using a 50-yr idealized NAO forcing using the CM2.1, CM2.1\_SLAB, and FLOR models. Normalized time series of the idealized NAO variability and associated heat flux forcing in the Subpolar Gyre are shown in Fig. 5a. Positive NAO values are associated with strengthened westerly winds over the subpolar North Atlantic. These in turn lead to negative heat flux anomalies, where a negative heat flux anomaly denotes enhanced ocean to atmosphere heat flux.

For the model with a slab ocean (Fig. 5b), the zonal mean SST response to the 50-yr NAO forcing is largely

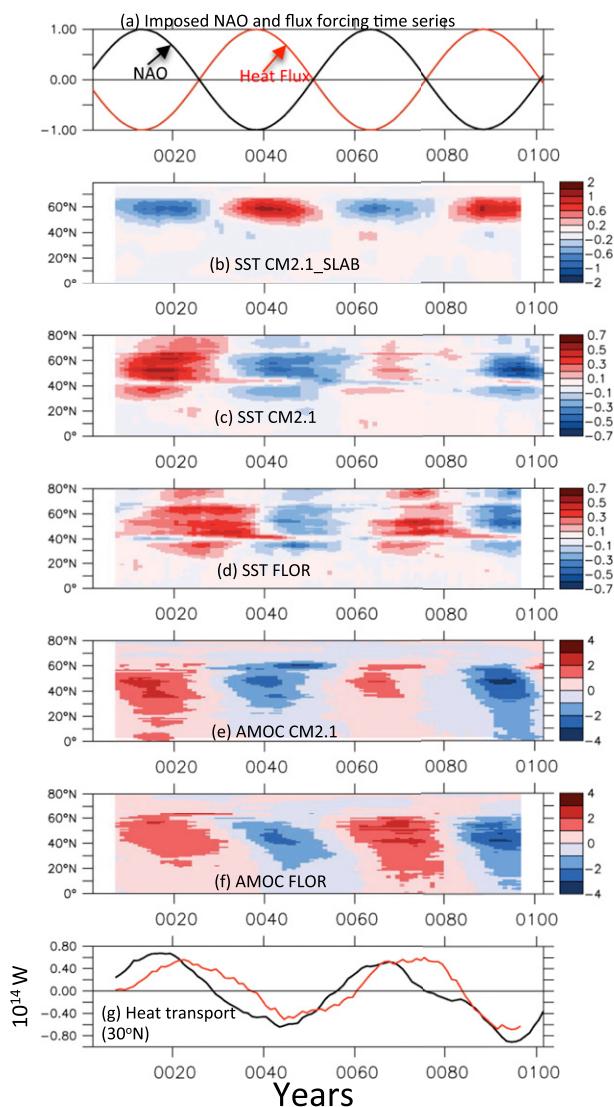


FIG. 5. (a) Normalized time series of the NAO (black) and associated heat flux time series (red) used as forcing in the idealized experiments. A positive (negative) phase of the NAO implies enhanced (reduced) ocean to atmosphere heat flux in the Subpolar Gyre. Negative values for heat flux indicate an enhanced ocean to atmosphere heat flux. (b) Zonal mean ( $60^{\circ}$ – $20^{\circ}$ W) of annual mean SST response (K) to NAO forcing in CM2.1\_SLAB, calculated as SST in simulations with NAO forcing minus SST in simulations without NAO forcing. (c) As in (b), but for CM2.1. (d) As in (b), but for FLOR. (e) AMOC response to NAO forcing in CM2.1, calculated as the AMOC in simulations with NAO forcing minus the AMOC in simulations without NAO forcing ( $\text{Sv}$ ;  $1 \text{ Sv} \equiv 10^6 \text{ m}^3 \text{ s}^{-1}$ ). (f) As in (e), but for the FLOR model. (g) North Atlantic meridional ocean heat transport response ( $30^{\circ}$ N) to NAO forcing in CM2.1 (black) and FLOR (red), calculated as the ocean heat transport in the simulations with the NAO forcing minus the ocean heat transport in the simulations without NAO forcing. An 11-yr running mean was applied to results in (e)–(g).

in phase with the NAO flux forcing (Fig. 5a), but with a small lag. Here, “in phase” means that negative SST anomalies in the Subpolar Gyre occur at the same time as a positive phase of the NAO (with enhanced ocean to atmosphere heat flux, which leads to cooling of the ocean mixed layer). The small phase lag between the maximum NAO and the coolest SSTs represents the finite heat capacity of the 50-m mixed layer ocean, such that the mixed layer temperature lags the forcing somewhat. The above results show that the response is dominated by the direct effects of the turbulent surface heat fluxes on the upper ocean.

In contrast, for the two models with dynamic oceans (Figs. 5c,d), positive SST anomalies over the Subpolar Gyre lag the NAO by 5–10 yr. This behavior reflects the role of ocean dynamics and shows a fundamentally different relationship between the NAO and Atlantic SSTs at middle and high latitudes in the presence of a dynamic ocean (Bjerknes 1964; Edén and Jung 2001). With a dynamic ocean the dominant response on long time scales (i.e., to the 50-yr forcing) is not the direct flux-forced tripole pattern, but a lagged monopole pattern that arises because of a lagged response of the AMOC (Delworth and Zeng 2016). In the models with dynamic oceans the NAO forcing extracts heat from the Subpolar Gyre, leading to an increase of upper-ocean density, oceanic convection, deep water formation, and the AMOC. The AMOC response to the 50-yr NAO forcing is shown in Figs. 5e and 5f. The periodic NAO forcing generates a periodic AMOC variability, with the AMOC lagging the forcing by a few years. The positive (negative) AMOC anomalies induce stronger (weaker) than normal poleward heat transport in the North Atlantic (Fig. 5g), leading to widespread warming (cooling) in the North Atlantic. The accumulating heat in the Subpolar Gyre associated with enhanced ocean heat transport leads to a lag of several years between the maximum AMOC anomaly and the maximum SST anomaly.

We further illustrate in Fig. 6 the relationships on decadal scales between the NAO and large-scale simulated SST by computing lead-lag correlations between the simulated SST response in the 50-yr NAO forcing experiments and the imposed NAO anomaly, similar to Figs. 1 and 3. For the slab ocean case (Fig. 6, top), the largest negative correlations occur at a small lag with respect to the NAO. This again shows that the slab ocean response is dominated by the direct impact of the anomalous surface heat flux forcing, with a small lag as a result of the thermal capacity of the ocean mixed layer.

For both the CM2.1 and FLOR simulations with dynamic ocean, the maximum positive correlations lag the imposed NAO forcing by approximately 10 yr, demonstrating that the NAO forcing induces a warming of the

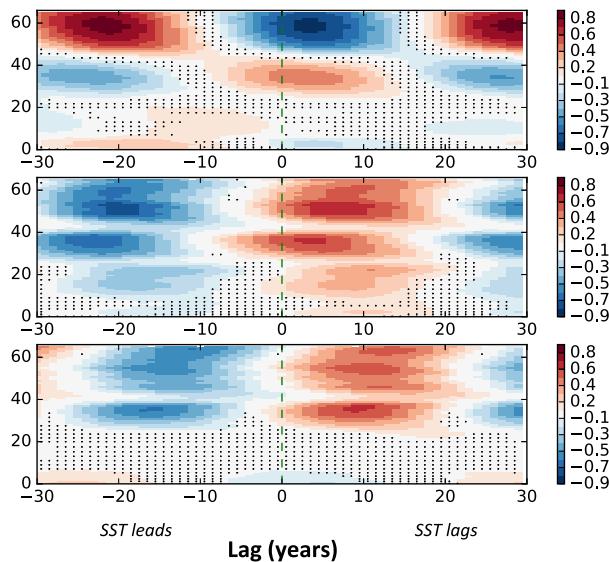


FIG. 6. Lead–lag correlations between the SST response and the imposed NAO forcing for the 50-yr idealized NAO forcing experiments. The SST response was first calculated as the SST in the simulations with NAO forcing minus the SST in corresponding sections of the control simulation. Linear correlations were computed at each grid point between the SST response and the imposed NAO forcing. The correlations were then zonally averaged over the domain 60°–20°W. Regions with stippling do not pass a statistical significance test (using a resampling technique as described in the appendix). (top) Results from CM2.1\_SLAB\_Ctrl\_NAO\_50yr, (middle) CM2.1\_Ctrl\_NAO\_50yr, and (bottom) FLOR\_Ctrl\_NAO\_50yr.

North Atlantic with a decadal-scale lag. We note that there are also negative correlations for periods preceding the NAO forcing, reminiscent of the observational analyses in Fig. 1b. The North Atlantic SST responds to a positive phase of the NAO with a delayed warming as a result of a decadal-scale adjustment of the AMOC to the NAO forcing. A similar process occurs in response to a negative NAO. The negative correlations for negative lags indicate a weakening of the AMOC in response to the negative phase of the NAO, which precedes the positive phase of the NAO by 25 yr in these idealized experiments. These results are of relevance for interpreting the observational analyses in Fig. 1b. During the instrumental record, the observed NAO has been characterized by substantial multidecadal variability (see Hurrell et al. 2016), with positive NAO phases in the early and late twentieth century and negative phases in the late nineteenth and midtwentieth centuries. The model results shown in Fig. 6 suggest that the observed negative correlations in Fig. 1b for negative lags may be related to multidecadal variability in the observed NAO. Just as for the model results in Fig. 6, the observed negative correlations at negative lags may

be the lagged response to a negative phase of the NAO, which precedes the positive phase of the NAO by several decades in both the observations and the idealized simulations. This is explored further in section 4d.

### c. Results from other models

To provide some assessment of the robustness of the NAO relationship with simulated North Atlantic decadal SST variability, we repeat the correlation analyses shown in Fig. 3 using a variety of control simulations from the CMIP5 archive. The results are shown in Fig. 7 and indicate that many (but not all) models have a similar relationship between the NAO and decadal-scale North Atlantic SST variability. Somewhat similar behavior is seen in ACCESS1.0, NorESM1-M, CCSM4, MPI-ESM-LR, MPI-ESM-MR, CSIRO Mk3.6.0, IPSL-CM5A-LR, and IPSL-CM5-MR. Rather different behavior, however, is seen in HadGEM2C-CC and MIROC-ESM-CHEM. (Expansions of acronyms are available online at <http://www.ametsoc.org/PubsAcronymList>.) The details can vary widely, however, and may be sensitive to a number of factors, including the spatial structure of the model-simulated NAO, the climatological ocean circulation, regions of simulated deep water formation, and the overall structure of biases in the model simulation. The overall results are sufficiently similar to suggest that the results shown in Fig. 3 are not unique to the GFDL models, but substantial uncertainty remains.

### d. Simulations of twentieth and twenty-first centuries

We next analyze the output from experiments using a realistic time history of NAO forcing to provide a perspective on the relationship between the NAO and Atlantic decadal SST variations in the observed record. For example, differences between CM2.1\_HIST\_NAO and CM2.1\_HIST reflect the impact of the observed NAO variations on the climate system over the period 1901–2014. For each model, we use ensembles to improve the signal-to-noise ratio (10-member ensembles for CM2.1 and CM2.1\_SLAB and 5-member ensembles for FLOR). We note that the FLOR simulations only go from 1951 to 2015 because of the greater computational expense of the FLOR model.

We show in Fig. 8a the time series of the observed NAO and in Fig. 8b the time series of the imposed surface heat flux anomalies over the Labrador Sea that are associated with the observed NAO anomalies (these are the extra fluxes added in the NAO experiments). The heat flux anomalies are only applied over December–March, with zero anomalies over the rest of the year; the time series shown in Fig. 8 are expressed here as annual means. The positive heat flux anomalies

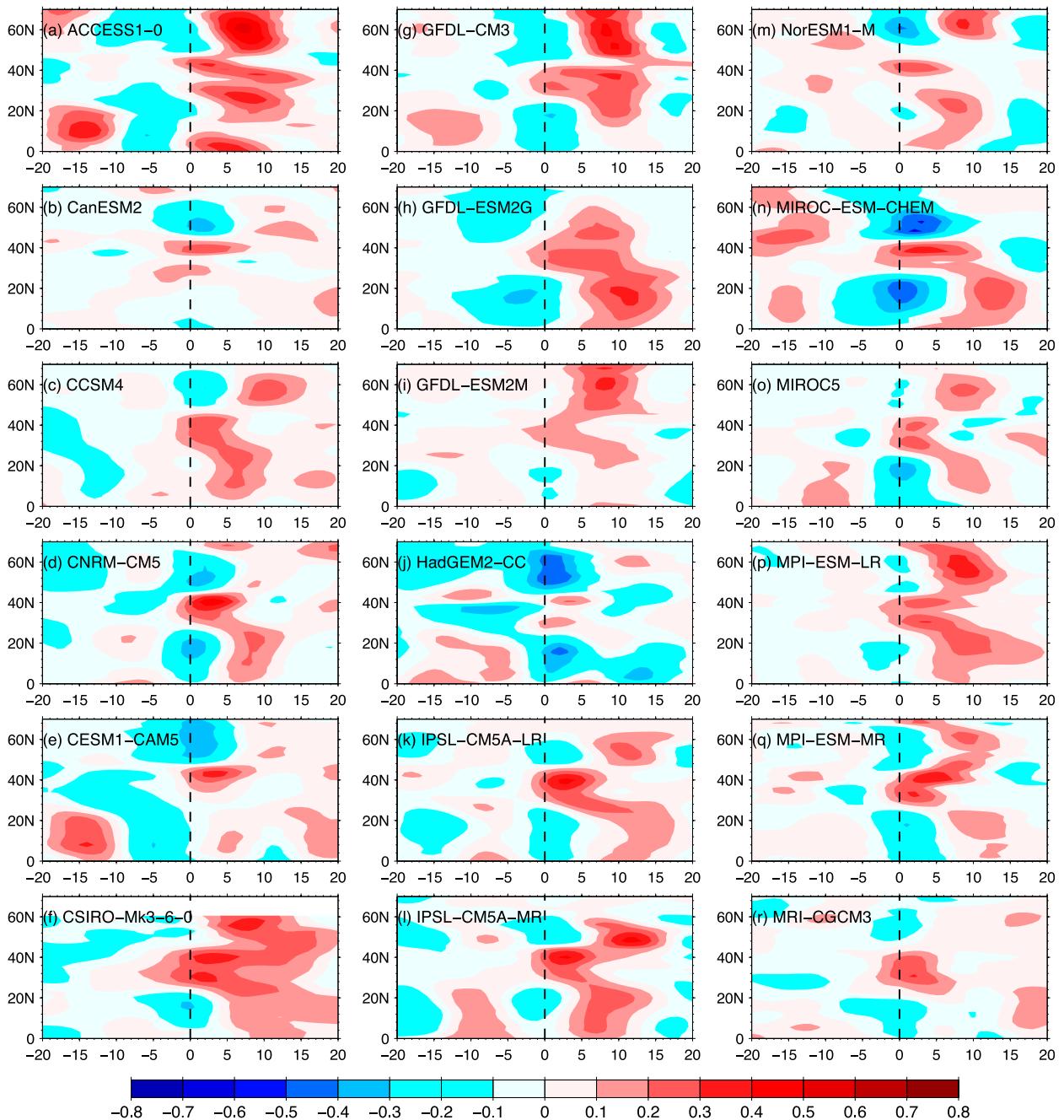


FIG. 7. Lead-lag correlation analyses, similar to Figs. 3f and 3h, using output from CMIP5 models (models were used that had control simulations at least 300 yr in length).

in the 1960s and 1970s are consistent with negative NAO anomalies in that time period, resulting in reduced ocean to atmosphere heat fluxes (appearing here as positive heat flux anomalies into the ocean). We show in Fig. 8c the time series of annual mean observed SST anomalies averaged over the domain  $30^{\circ}$ – $65^{\circ}$ N,  $60^{\circ}$ – $20^{\circ}$ W. This clearly shows multidecadal warming and cooling associated with the AMO. In Figs. 8d and 8e, we show the

simulated response to the observed NAO flux anomalies using CM2.1 and FLOR, respectively, averaged over the same domain as the observations. Both models capture the essence of the observed AMO signal in the extratropical North Atlantic, with negative (positive) SST anomalies appearing a decade or so after the NAO minimum (maximum). In both cases the SST response is associated with a lagged response of the AMOC to the

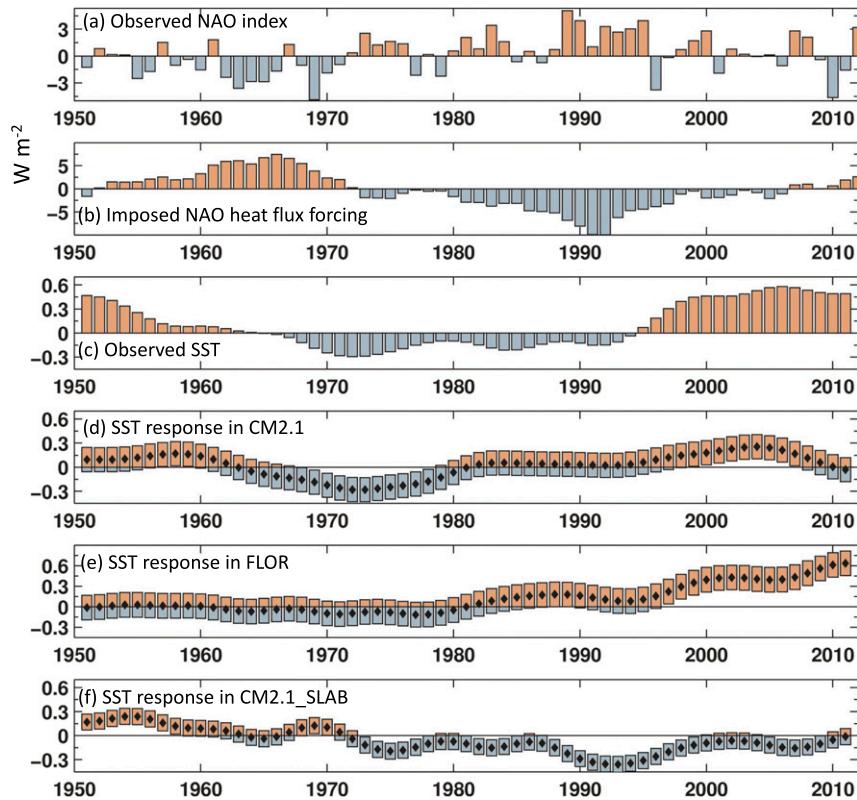


FIG. 8. (a) Time series of observed NAO index (DJFM; station-based index from [Hurrell et al. 2016](#)). (b) Time series of anomalous flux forcing associated with imposed NAO anomalies. Time series shows annual mean flux anomalies averaged over  $50^{\circ}$ – $60^{\circ}$ N,  $60^{\circ}$ – $30^{\circ}$ W, after applying a 7-yr running mean. (c) Observed annual mean SST anomalies averaged over  $30^{\circ}$ – $65^{\circ}$ N,  $60^{\circ}$ – $20^{\circ}$ W, after applying a 10-yr low-pass filter. Values plotted are anomalies with respect to time mean over 1951–80. (d) Time series of annual mean SST response to NAO flux anomalies using CM2.1, calculated as SST in CM2.1\_HIST\_NAO minus SST in CM2.1\_HIST (calculated as the differences in the ensemble means). A 10-yr low-pass filter was applied to the output. The black diamonds represent the difference of the ensemble means, while the colored bars represent the 90% uncertainty limits around that ensemble mean difference, calculated by resampling the control simulation. (e) As in (d), but using FLOR. (f) As in (d), but using CM2.1\_SLAB.

NAO fluxes [see Fig. 1 in [Delworth et al. \(2016\)](#)], thereby altering oceanic heat transport and creating basin-scale monopole SST anomaly patterns. We show in [Fig. 8f](#) the results using CM2.1\_SLAB. When we replace the dynamic ocean of CM2.1 with the 50-m slab ocean, the response to exactly the same set of NAO surface heat flux anomalies is substantially different. In the slab model, negative SST anomalies are a direct response to a positive phase of the NAO. Therefore, while both observations and the models with the dynamic ocean show widespread warming in the 1990s and 2000s (a delayed response to the positive phase of the NAO), the slab model shows pronounced negative SST anomalies in the 1990s and 2000s, opposite to the observed signal. Therefore, we conclude that ocean dynamics are an essential part of the processes

governing AMO-like SST variations over the Subpolar Gyre in the models and also in observations.

The observations ([Fig. 8c](#)) show a rapid warming in the 1990s in the North Atlantic. In the top panel of [Fig. 9](#) we show the spatial pattern of that warming, relative to the cool period of the 1970s and early 1980s. The observations show warming extending from the Subpolar Gyre to the tropical North Atlantic. We show the degree to which each of the models is able to simulate that warming in response to historical radiative forcing and/or NAO flux forcing. [Figure 9](#), left, shows the SST response due to radiative forcing alone (HIST simulations), [Fig. 9](#), center, shows the SST response due to both radiative forcing and the imposed NAO flux anomalies (HIST\_NAO simulations), and [Fig. 9](#), right, shows the impact of only the NAO flux forcing, estimated as the



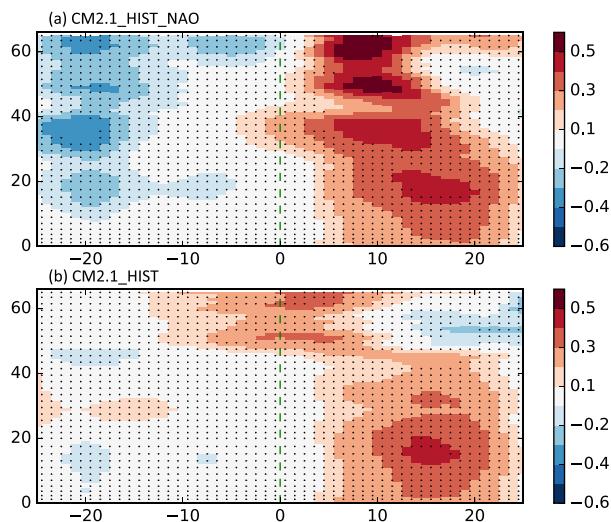


FIG. 10. Zonal mean ( $60^{\circ}$ – $20^{\circ}$ W) of the correlation coefficient between simulated annual mean SST and the observed DJFM NAO index. The data were subject to a 10-yr low-pass filter prior to analysis. (a) Results using 10-member ensemble of CM2.1\_HIST\_NAO simulations. (b) Results using 10-member ensemble of CM2.1\_HIST simulations. Stippled areas are not significant, using a similar technique as described in the [appendix](#).

model with a slab ocean, the warming for CM2.1 and FLOR extends much farther north, into the Nordic Seas. The impact of the NAO is isolated in [Fig. 9](#), right, by subtracting the HIST results from the HIST\_NAO results. This shows that, for the models with dynamic oceans, the extratropical warming (from the 1970s to the 1990s) has a very significant contribution from NAO-forced ocean circulation changes. The subpolar warming in the models with a dynamic ocean (two middle panels of [Fig. 9](#), center) is thus a combination of two effects: cooling that would occur from a radiatively induced AMOC weakening (top and middle panels of [Fig. 9](#), left), and warming that occurs in response to a NAO-induced AMOC strengthening (top and middle panels of [Fig. 9](#), right). The NAO-induced AMOC strengthening dominates, leading to the subpolar warming in the models with a dynamic ocean (two middle panels of [Fig. 9](#), center). In contrast, the slab ocean shows little additional impact from the NAO forcing.

The CM2.1\_HIST\_NAO simulation extends from 1901 to 2014, and the CM2.1\_HIST covers the period 1861–2014. We can therefore compute lead–lag correlations between the observed NAO and SST in each of these two simulations, similar to [Fig. 1](#). These are shown in [Fig. 10](#) for the CM2.1\_HIST\_NAO simulations ([Fig. 10a](#)) and the CM2.1\_HIST simulations ([Fig. 10b](#)). A number of features are present. First, we note that the CM2.1\_HIST\_NAO simulations have weak negative correlations when SST leads the NAO by approximately

20 yr, somewhat similar to the observations in [Fig. 1b](#). As discussed with regard to [Fig. 6](#), one interpretation of this is that these negative correlations reflect a weakening of the AMOC (and hence negative SST anomalies) in response to a previous negative phase of the NAO. We note that the observed NAO in the twentieth century is characterized by clear multidecadal variability. When we impose this NAO forcing on the model ocean, as in CM2.1\_HIST\_NAO, we reproduce some aspects of the negative correlations between SST and the NAO when SST leads the NAO. In contrast, this negative correlation is absent in [Fig. 10b](#), in which there are no imposed NAO anomalies. This provides support to the idea that the negative correlations that occur when SST leads the NAO by a decade or two are a result of the weakened AMOC in response to a prior negative phase of the NAO.

We also note that there are positive correlations when SST lags the NAO by a decade or so in the CM2.1\_HIST\_NAO simulations, as we have seen previously. These have largest values in the Subpolar Gyre, consistent with our previous analyses. We also note modest positive correlations in the NAO\_HIST simulations. This relationship is weak and likely insignificant and could result from some similar aspects of long term trends in the observed NAO and the response of the North Atlantic Ocean to radiative forcing changes.

In summary, the above results suggest that the multidecadal variations of North Atlantic SST are a combination of the response to changing radiative forcing and to multidecadal ocean circulation changes induced by multidecadal NAO variations. For these models with a dynamic ocean, the NAO-driven changes have largest amplitude in the middle and higher latitudes of the North Atlantic, whereas at lower latitudes the relative impact of radiative forcing is larger. This decomposition occurs in models for which NAO-driven AMOC variability results in a much smaller response in the tropical North Atlantic than in the subpolar North Atlantic. It is possible that this small tropical response is related to deficiencies in model processes, such as cloud or dust feedbacks. If that were the case, the potential importance of NAO-driven AMOC and ocean heat transport variability for tropical North Atlantic climate variability could be larger.

We note that our results provide both contrast and complementarity to a recent study on this topic ([Clement et al. 2015](#)), in which it was concluded that the AMO in models with a dynamic ocean is mainly due to the direct influence of NAO variations on the heat budget of the ocean mixed layer, combined with wind–evaporation–SST feedback in the tropics. The pattern of surface flux forcing employed in the current study is

similar in character to the dominant pattern of coupled ocean–atmosphere variability in their study that resembled the NAO. Nevertheless, our results point to the importance of ocean dynamics in the extratropical North Atlantic through a delayed response of the AMOC and resultant changes to basin-scale meridional oceanic heat transport. Our results also suggest a smaller direct role for ocean dynamics in the tropical North Atlantic than in the extratropical North Atlantic, highlighting the potential role of additional atmospheric processes and feedbacks.

## 5. Summary and discussion

Using observations and models, we have examined the relationship between the North Atlantic Oscillation (NAO) and Atlantic decadal SST variations. Consistent with many previous studies, on short time scales NAO-related surface heat flux anomalies drive a tripole pattern of SST anomalies in the Atlantic. On decadal and longer time scales, there is a lagged response of the ocean to the NAO fluxes, with the AMOC playing a prime role in modulating meridional oceanic heat transport and generating an AMO-like SST response. A prolonged positive phase of the NAO enhances the AMOC after a decadal-scale delay.

We further show that ocean dynamics are crucial to the relationship on decadal scales in the extratropical North Atlantic. We use simulations with slab oceans to show that, without ocean dynamics, the relationship of Atlantic decadal SST anomalies to the NAO is nearly opposite to that in models with ocean dynamics, primarily over the Subpolar Gyre, and different from observations. Further, the NAO–SST relationship in models with ocean dynamics bears a considerable resemblance in middle and higher latitudes of the North Atlantic to the relationship diagnosed from observations.

In the models used for the present study, the phase lag between the NAO and Atlantic decadal SST anomalies is shorter than that seen in observations. It is difficult to properly characterize this relationship, given the short observational record. However, this difference is consistent with the relatively short time scale of AMOC variability in these models (20–30 yr) relative to the time scale of the observed AMO. The response of models to NAO variability is a key aspect of Atlantic decadal variability and is an important component underlying the physical basis for decadal prediction. It is important to stress that other types of atmospheric circulation variability in addition to the NAO may be very important for driving Atlantic Ocean variability (Barrier et al. 2014).

The direct impact of NAO-induced ocean circulation changes is weaker over the tropical North Atlantic than

over the subpolar North Atlantic in our modeling study. This difference highlights the importance of additional atmospheric processes, some not well captured in many current models, which influence decadal-scale variability over the North Atlantic, especially over the tropical North Atlantic. The impact of time-varying anthropogenic aerosols (Booth et al. 2012), dust emissions from the African Continent (Evan et al. 2009), and cloud feedback and circulation linkages (Martin et al. 2014) may be important for tropical Atlantic variability. Variations of the North Atlantic Oscillation are one mechanism contributing to the observed decadal-scale variability of the North Atlantic through its impact on the AMOC, but a more complete understanding of observed Atlantic decadal variability needs to properly account for all of the important additional factors. This is particularly important since many of the large-scale tropical climatic impacts associated with Atlantic decadal SST variability are influenced most strongly by the SST signal in the tropical North Atlantic (Sutton and Hodson 2005). It is critical that models improve their ability to faithfully represent all of these important processes to allow us a better quantitative assessment of the processes governing observed Atlantic changes. Such improved understanding would then lend increased confidence to our predictions of future changes.

In the instrumental record, decadal-scale SST variability in the subpolar and tropical North Atlantic is well correlated. This spatial coherence is the underpinning for the commonly used definition of the AMO (areal-averaged SST anomalies in the North Atlantic from 0° to 70°N). However, our results suggest that one needs to think of multiple processes that influence Atlantic decadal SST variability in these regions. While ocean dynamics plays a crucial role for decadal-scale SST variability in the extratropical North Atlantic, the results of this study suggest that its direct influence in the tropical North Atlantic appears to be smaller, with air–sea fluxes playing a larger role. A key question is whether the observed spatial coherence in SST between the tropical and extratropical North Atlantic is an artifact of the short observational record or a robust aspect of Atlantic variability. If the latter, then what processes contribute to this coherence? What are the relative roles of natural and anthropogenic radiative forcing variations versus feedback processes (such as dust or cloud feedback) that may serve to connect the regions and that may not be well represented in models? Answering such questions poses a key challenge, as they will shed light on future predictions of the Atlantic and its climatic impact.

*Acknowledgments.* The authors are grateful to Dr. Yohan-Ruprich Robert and Dr. Honghai Zhang for

very valuable comments on an earlier version of this manuscript, as well as to three anonymous reviewers, who provided extremely insightful and valuable feedback and suggestions. The NOAA ERSST.v3 data were downloaded from NOAA/OAR/ESRL PSD, Boulder, Colorado, from their website. Observed NAO data were downloaded from the UCAR website (Hurrell et al. 2016). We acknowledge the World Climate Research Programme's Working Group on Coupled Modelling, which is responsible for CMIP, and we thank the climate modeling groups (models shown in Fig. 7) for producing and making available their model output. For CMIP, the U.S. Department of Energy's Program for Climate Model Diagnosis and Intercomparison provides coordinating support and led development of software infrastructure in partnership with the Global Organization for Earth System Science Portals. The work of T. Delworth, G. Vecchi, F. Zeng, and R. Zhang is supported as a base activity of NOAA's Geophysical Fluid Dynamics Laboratory. X. Yang is supported through UCAR under block funding from NOAA/GFDL. L. Zhang is supported through Princeton University under block funding from NOAA/GFDL.

## APPENDIX

### Statistical Testing for Figs. 1, 3, 6, and 10

Figure 1 is constructed by first computing the linear correlation coefficients between the observed NAO index and the time series of observed SST at each grid point in the North Atlantic. Prior to the analysis, the NAO and SST data were subject to either a low-pass (longer than 10 yr) or high-pass (shorter than 10 yr) filter. The correlation analysis is performed for various leads and lags. The correlation values are then zonally averaged over the longitudinal span  $60^{\circ}$ – $20^{\circ}$ W and then plotted as a function of lag and latitude. The analysis using high-pass-filtered data is shown in Fig. 1a, and using low-pass-filtered data in Fig. 1b.

We assess the significance of these plotted values using the following resampling strategy. A similar strategy is used for both Figs. 1a and 1b. We use observed time series of both the NAO and SST over the period 1864–2014 (151 yr total), using the November–February (NDJF) seasonal mean for the NAO and annual means for SST. In our resampling strategy we first choose a year at random between 1864 and 2014. We then create a new “shuffled” NAO time series. The first part of this new shuffled time series consists of the original NAO time series starting from the chosen random year, continuing to 2014; the second part of the time series uses the original NAO data from 1864 to the year

before the randomly chosen year. For example, if the random year were 1922, then the shuffled NAO time series would consist of the original NAO time series value from 1922 to 2014, followed immediately by the original NAO time series from 1864 to 1921. This shuffled time series has many of the same temporal properties as the original NAO time series, but the starting year is redefined. We then choose a second random year to do a similar reshuffling process to the time series of observed annual mean SST, using the same random year to perform the reshuffling for SST at each grid point. Thus, the NAO and SST time series have been independently reshuffled. We now compute the linear correlation coefficients between the reshuffled NAO and SST time series at each grid point. We then compute the zonal average of the correlations over the range  $60^{\circ}$ – $20^{\circ}$ W. We repeat the above process 10 000 times to form a distribution for each latitude and lag of the zonal mean correlations between randomly reshuffled NAO and SST time series. Threshold values in those distributions are found such that 10% of the distribution has values smaller than the first threshold, and 10% of the distribution has values larger than the second threshold. Hatching is placed in Fig. 1 on those grid points where the correlation values are between the two thresholds. We note that, when computing lagged correlations, the number of points in the correlation calculation is reduced by the lag in use (i.e., for lag 10 yr, there are 10 fewer points in the correlation calculations). We therefore compute a separate distribution for each length of time series ranging from 151 (original time series) to 131 (lag 20 yr) and use the appropriate distribution for each lag to calculate the threshold values.

By chance, we would expect 80% of the points to have hatching and 20% of the points to have no hatching. For the area poleward of  $30^{\circ}$ N (where we have an a priori expectation that on long time scales the NAO will lead SST), we calculate that 40% of the points have no hatching in Fig. 1b, which is substantially larger than the 20% one would expect from chance. We therefore conclude that there is a statistically meaningful correlation between the NAO and SSTs at long time scales, especially in the mid and higher latitudes of the North Atlantic.

Figure 3 is constructed in a similar fashion to Fig. 1, but using model output from three separate control simulations. At each lag, the correlations (calculated between the model's SST and NAO) are zonally averaged over  $60^{\circ}$ – $20^{\circ}$ W in the North Atlantic for each latitude band. We place stippling over latitudes and lags that are not statistically significant. This assessment was based on the following: at each grid point, and for each lag, we use the Student's *t* test to assess whether the

correlation coefficient was significantly different from zero at the 90% confidence level for that particular lag. We then evaluate, for each lag and latitude, how many points in the zonal band from 60° to 20°W are significant by that metric. Separately, we evaluate the number of grid points that would be expected to pass such a significance test by chance at various levels of confidence. The regions without (with) stippling indicate areas where this value of the zonal mean correlation has less than (greater than) a 10% chance of occurring by chance. A similar method was used to assess statistical significance in Figs. 6 and 10, with the exception that, for Fig. 10, the regions without (with) stippling indicate areas where this value of the zonal mean correlation has less than (greater than) a 20% chance of occurring by chance. We have performed these tests with varying confidence levels and estimates of degrees of freedom, but the essential results are largely robust.

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