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Impact of tropical ocean SSTs on the late winter signal over the North Atlantic-European region

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North Atlantic-European (NAE) region

- under the dominant influence of the North Atlantic Oscillation (NAO)
- large internal variability of the atmosphere \rightarrow affects seasonal predictability



https://climatedataguide.ucar.edu/climate-data/hurrellnorth-atlantic-oscillation-nao-index-station-based

North Atlantic-European (NAE) region

- under the dominant influence of the North Atlantic Oscillation (NAO)
- large internal variability of the atmosphere \rightarrow affects seasonal predictability
- ➢ idea: improve seasonal predictability by using strong sources of signal outside the region of interest



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El Niño-Southern Oscillation (ENSO)

- quasiperiodic phenomenon in the atmosphere-ocean system
- global impacts on climate variability
- affects remote regions through teleconnections









Mezzina, B 2022, *Dynamics of the late-winter ENSO teleconnection to the North Atlantic-European region*, PhD thesis, University of Barcelona, Barcelona, <u>http://hdl.handle.net/2445/182562</u>

- goal \rightarrow distinguish internal from boundary-forced variability
- ensembles of numerical simulations used to estimate signal and noise

ICTP AGCM (SPEEDY; T30-L8)

- intermediately complex AGCM \rightarrow computationaly inexpensive
- parametrizations: SW and LW radiation, large-scale condensation, convection, surface fluxes of momentum, heat and moisture, and vetical diffusion
- successful in simulating the main features of ENSO-related global teleconnections (Abid et al. 2000; Herceg-Bulić et al. 2012, 2017; Kucharski et al. 2006, 2013)

ICTP AGCM experiments

- six ensembles of numerical simulations
- monthly varying NOAA ERSST V3 SST anomalies set as lower-boundary forcing only in limited ocean areas
- climatological exp \rightarrow no SST anomalies
- period: 1854-2010
- JFM geopotential height at 200 hPa
- output filtered with a high-pass filter with a cut off period of 11 years



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25

0.75

0.5

25

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0.25

0.5

0.75

25

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(a) Ctrl, (b) TroAtl, (c) TroPac, (d) TroInd, and (e) Tropics

Empirical Orthogonal Functions analysis

- linear decomposition tehnique
- result: orthogonal EOF patterns and corresponding time series (PCs) which explain the largest part of variability
- EOFs based on all ensemble members vs. enseble mean

Signal-to-noise optimal patterns method

- method described in the paper by Straus and Shukla et al. (2003)
- all ensemble members used to find a hierarchy of modes sorted by their signal-to-noise ratio
- result: OPT patterns which have the maximum signal-to-noise ratio
- not suitable for observations



Fig. 2 EOF1 pattern of JFM geopotential heights at 200 hPa (GH200) [m] based on all ensemble members and based on the ensemble mean in the period 1855-2010 in ICTP AGCM experiments: Clim, Ctrl, TroAtl, TroPac, TroInd, and Tropics. The percentage of explained variance (ExpVar) associated with each of the EOF patterns is indicated by the number in brackets.

Following the definitions from Branković and Molteni (2004):

Signal estimate

$$\sigma_s^2 = \frac{1}{N} \sum_{j=1}^N (\overline{x}_j - \overline{x})^2$$

Ensemble mean in the *j*-th year:



N – number of years (156) M – number of ensemble members (35)

Noise estimate

$$\sigma_n^2 = \frac{1}{N} \sum_{j=1}^{N} \left[\frac{1}{M} \sum_{i=1}^{M} (x_{ij} - \bar{x}_j)^2 \right]$$

Climatological mean of the ensemble mean:





Fig. 3 JFM signal variance of geopotential heights at 200 hPa (GH200) [m²] in ICTP AGCM experiments



Fig. 4 JFM noise variance of geopotential heights at 200 hPa (GH200) [m²] in ICTP AGCM experiments





зо́พ

-1.5 1.5

-3

3ÖE

4.5

6ÖW

-6 -4.5

40N

20N + 90W



Fig. 5 First spatial pattern (EOFOPT1) of JFM GH200 [m] in the signal-to-noise optimal patterns method in the period 1855-2010 for ICTP AGCM experiments:(a) Clim, (b) Ctrl, (c) TroAtl, (d) TroInd, (e) TroPac, and (f) Tropics. Panel (g) shows the corresponding signal-to-noise ratio (SNR) of the EOFOPT1 mode. All SNR values are considered statistically significant according to the F-test for the ratio of variances on the 95% confidence level.



12⁰W

120W

12'0W

6ÓW

60W

6ÓW

0.9

0

Correlation of GH200 with observed SSTAs

Fig. 6 Correlation of global NOAA ERSST V3 SST anomalies and the time series associated with the first optimal pattern (PCAVG1) of JFM GH200 in the period 1855-2010 for ICTP AGCM experiments: (a) Clim, (b) Ctrl, (c) TroAtl, (d) TroPac, (e) TroInd, and (f) Tropics. All statistically significant values based on the two-tailed Student's t-test on the 95% confidence level are encircled by dashed contours.

The EOF analysis was applied on the observed SST anomalies in the same areas as the SST-boundary forcing within ICTP AGCM experiments.

Table 1 Correlation between the first principal component (PC1) of NOAA sea surface temperature anomalies (SSTA) and the time series associated with the first optimal pattern (PCAVG1) of geopotential heights at 200 hPa (GH200) in JFM season in ICTP AGCM experiments. All values are statistically significant based on the two-tailed Student's t-test on the 95% confidence level.

Table 2 Correlation between the first principal component (PC1)
of NOAA sea surface temperature anomalies (SSTA) and PC1 of
GH200 in JFM season calculated for the ensemble mean of each
ICTP AGCM experiment. All values are statistically significant
based on the two-tailed Student's t-test on the 95% confidence
level

	Ctrl GH200 PCAVG1	TroAtl GH200 PCAVG1	TroPac GH200 PCAVG1	TroInd GH200 PCAVG1	Tropics GH200 PCAVG1
TroAtl SSTA PC1	0.81	0.42	0.77	0.47	0.81
TroPac SSTA PC1	0.88	0.34	0.95	0.60	0.87
TroInd SSTA PC1	0.81	0.51	0.71	0.86	0.78
Tropics SSTA PC1	0.91	0.41	0.95	0.66	0.90

	Ctrl	TroAtl	TroPac	TroInd	Tropics
	GH200	GH200	GH200	GH200	GH200
	PC1	PC1	PC1	PC1	PC1
	EnsMean	EnsMean	EnsMean	EnsMean	EnsMean
TroAtl	0 55	0.33	0.69	-0.25	0.69
SSTA PC1	TA PC1				
TroPac	0 70	0.27	0.97	0 22	0.91
SSTA PC1	0.70	0.27	0.87	-0.55	0.01
TroInd	0 47	0.27	0.64	0 5 2	0 50
SSTA PC1	0.47	0.57	0.04	-0.52	0.33
Tropics	0 60	0.21	0 96	0 27	0.01
SSTA PC1	0.09	0.31	0.00	-0.57	0.01

ICTP AGCM experiments

- SST boundary forcing, especially in the tropical Pacific, enhances potential predictability of the late-winter atmospheric circulation over the North Atlantic-European region
- boundary-forced signal can be detected in the first EOF based on the ensemble mean and the first optimal pattern, which has the highest signal-to-noise ratio
- time series connected to the optimal pattern is spatially and temporally connected to the variability of the observed SSTs

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Results were first published in the paper:

Ivasić, S., & Herceg-Bulić, I. (2022). A modelling study of the impact of tropical SSTs on the variability and predictable components of seasonal atmospheric circulation in the North Atlantic-European region. Climate Dynamics, https://doi.org/10.1007/s00382-022-06357-3

References

- Abid MA, Kucharski F, Molteni F, et al (2020) Separating the Indian and Pacific Ocean impacts on the Euro-Atlantic response to ENSO and its transition from early to late winter. J Clim 1–57. <u>https://doi.org/10.1175/jcli-d-20-0075.1</u>
- Branković, Č., Molteni, F. Seasonal climate and variability of the ECMWF ERA-40 model. Climate Dynamics 22, 139–155 (2004).
- Herceg-Bulić I, Branković Č, Kucharski F (2012) Winter ENSO teleconnections in a warmer climate. Clim Dyn 38:1593–1613. <u>https://doi.org/10.1007/s00382-010-0987-8</u>
- Herceg-Bulić I, Mezzina B, Kucharski F, et al (2017) Wintertime ENSO influence on late spring European climate: the stratospheric response and the role of North Atlantic SST. Int J Climatol 37:87–108. <u>https://doi.org/10.1002/joc.4980</u>
- Horel, J. D., and J. M. Wallace. 1981. "Planetary-Scale Atmospheric Phenomena Associated with the Southern Oscillation." Monthly Weather Review 109 (4): 813–29. <u>https://doi.org/10.1175/1520-0493(1981)109<0813:PSAPAW>2.0.C0;2</u>
- Kucharski F, Molteni F, Yoo JH (2006) SST forcing of decadal Indian Monsoon rainfall variability. Geophys Res Lett 33:. <u>https://doi.org/10.1029/2005GL025371</u>
- Kucharski F, Molteni F, King MP, et al (2013) On the need of intermediate complexity general circulation models: A "sPEEDY" example." Bull Am Meteorol Soc 94:25–30. <u>https://doi.org/10.1175/BAMS-D-11-00238.1</u>
- Mezzina, B 2022, Dynamics of the late-winter ENSO teleconnection to the North Atlantic-European region, PhD thesis, University of Barcelona, Barcelona, http://hdl.handle.net/2445/182562
- Smith, Thomas M., Richard W. Reynolds, Thomas C. Peterson, and Jay Lawrimore. 2008. "Improvements to NOAA's Historical Merged Land-Ocean Surface Temperature Analysis (1880-2006)." Journal of Climate 21 (10): 2283–96. <u>https://doi.org/10.1175/2007JCLI2100.1</u>.
- Straus D, Shukla J, Paolino D, et al (2003) Predictability of the seasonal mean atmospheric circulation during autumn, winter and spring. J Clim.