

1 Supplementary data of

2 **Causal interactions between ENSO and the North Tropical Atlantic**

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10 **Contents of this Supplementary data**

11

12 Text S1

13 Table S1

14 Figures S1

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16 The North Tropical Atlantic index (NTA) was given as the average sea surface temperature
17 (SST) anomalies in the boreal spring (March–April–May, MAM) over the North Tropical
18 Atlantic region (90°W–20°E; 0°N–15°N). The ENSO index was computed as the average SST
19 anomalies in the Niño 3.4 area (120–170°W; 5°N–5°S) in boreal winter (December–January–
20 February, DJF). Confounding factors (i.e., the Indian Ocean Dipole (IOD) and the Indian Ocean
21 Basin mode (IOB)) may have effects on the connections between NTA and ENSO. The IOD
22 index (Saji *et al* 1999) was given as the difference in boreal fall (September–October–November,
23 SON) sea surface temperature (SST) anomalies between two Indian Ocean regions of the
24 western pole (50–70°E; 10°N–10°S) and southeastern pole (90–110°E; 0°N–10°S). The IOB
25 index is computed as the first empirical orthogonal function (EOF) of boreal spring (MAM) SST
26 anomalies in the Indian Ocean basin (40°E–120°E, 20°N–20°S).

27 We use the following multivariate predictive model (e.g., Stern and Kaufmann 2013, Mosedale
28 *et al* 2006) to estimate the causal links between the NTA and ENSO:

$$29 \quad X_t = \sum_{i=1}^p \alpha_i X_{t-i} + \sum_{i=1}^p \beta_i Y_{t-i} + \sum_{j=1}^m \sum_{i=1}^p \delta_{j,i} Z_{j,t-i} + \varepsilon_t \quad (1)$$

30 where X_t is the NTA (or ENSO index) for year t , Y_t is the ENSO index (or NTA), and $Z_{j,t}$ is the
31 confounding factor j for year t . For example, the NTA for year 1 [i.e., MAM(1)] may have
32 predictive power on ENSO for year 1 [i.e., D(1)JF(2)] and ENSO for year 2 [i.e., D(2)JF(3)].
33 Conversely, ENSO for year 0 [i.e., D(0)JF(1)] may have predictive power on the NTA for year 1
34 [i.e., MAM(1)] and the NTA for year 2 [i.e., MAM(2)]. In the predictive model shown in
35 equation 1, while estimating the influence of Y on X (i.e., the contribution of the term
36 $\sum_{i=1}^p \beta_i Y_{t-i}$ in predicting X), the contribution of past X events are already taken into account by
37 adding the term $\sum_{i=1}^p \alpha_i X_{t-i}$. Thus, the causal influence of Y on X , if detected, is robust and the
38 contribution of past X events are already considered in our analyses.

39 Here, m is number of confounding factors and $p \geq 1$ is the order of the multivariate predictive
40 model. The optimal order p is computed by minimizing the Schwarz criterion or the Bayesian
41 information criterion (Schwarz 1978). The optimal orders might be different for each model.
42 Here we consider the simultaneous impacts of confounding factors and thus provide more
43 information of the real-world teleconnections. Our analysis uses two different confounding

44 factors; thus, m is equal to 2. The noise residuals ε_t and the regression coefficients α_i , β_i and $\delta_{j,i}$
45 are the results of the multiple linear regression analysis using the least squares method. We
46 detrend and normalize all the climate indices.

47 We estimate the probability of no Granger causality by applying a test of Granger causality
48 (Mosedale *et al* 2006, Stern and Kaufmann 2013, Le and Bae 2019) for the multivariate
49 predictive model shown in equation 1.

50 For computing the degree of uncertainty, we followed recent guidance (Stocker *et al* 2013,
51 Masson-Delmotte, V., P. Zhai, A. Pirani, S.L. Connors, C. Péan, S. Berger, N. Caud, Y. Chen, L.
52 Goldfarb, M.I. Gomis, M. Huang, K. Leitzell, E. Lonnoy, J.B.R. Matthews, T.K. Maycock, T.
53 Waterfield, O. Yelekçi, R. Yu, and B. Zhou Masson-Delmotte, V., P. Zha 2021) and utilized the
54 terms ‘very unlikely’, ‘unlikely’, ‘likely’ for the 0–10%, 0–33%, and 66–100% probability of the
55 likelihood of the outcome, respectively. For example, if the p -value is less than 0.33, the result
56 indicates that ENSO is unlikely to display no Granger causality on NTA. In this instance, we
57 conclude that ENSO has ‘causal effect’ on NTA.

58 **References**

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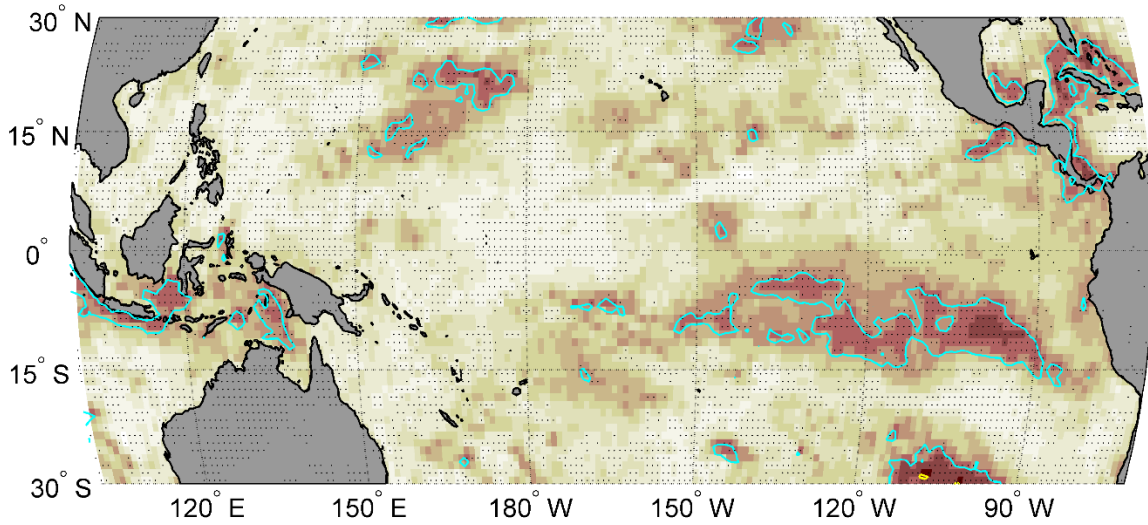
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84 [art](https://www.cambridge.org/core/product/identifier/CBO9781107415324A011/type/book_part)
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86 **Table S1.** List of CMIP6 HighResMIP models used in this study

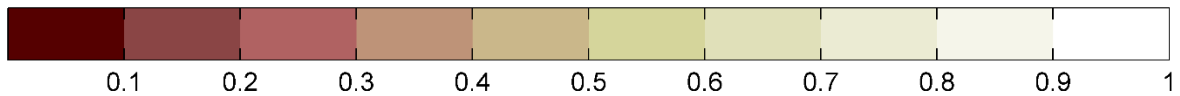
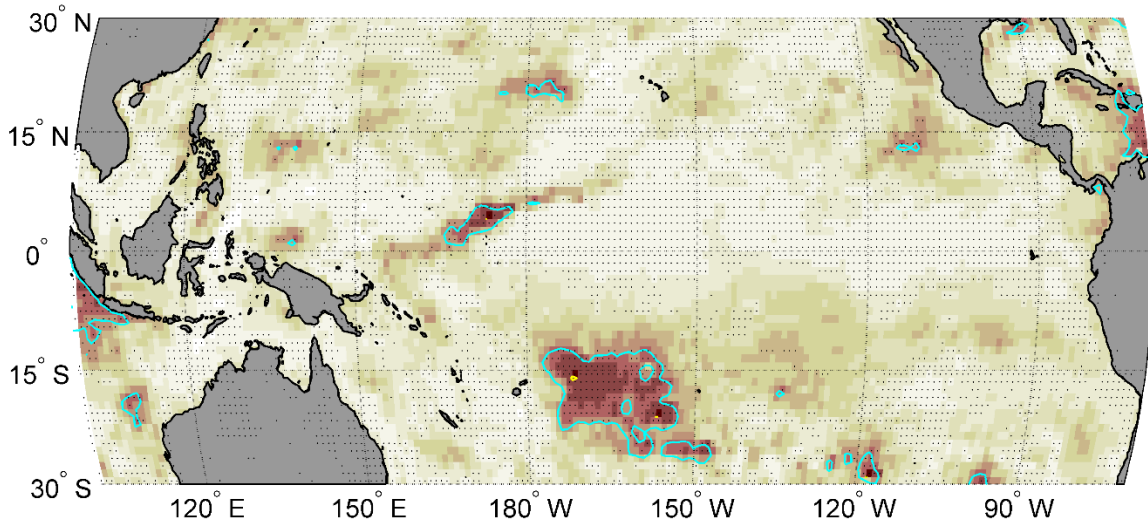
No.	Model name	Modelling center, country	Atmosphere resolution	Ocean resolution
1	BCC-CSM2-HR	BCC, China	$0.5^\circ \times 0.5^\circ$ (~ 50 km \times 50 km)	$0.5^\circ \times 0.5^\circ$
2	CAMS-CSM1-0	CAMS, China	$1^\circ \times 1^\circ$	$1^\circ \times 1^\circ$
3	CMCC-CM2-HR4	CMCC, Italy	$1^\circ \times 1^\circ$	$0.25^\circ \times 0.25^\circ$
4	CMCC-CM2-VHR4	CMCC, Italy	$0.25^\circ \times 0.25^\circ$	$0.25^\circ \times 0.25^\circ$
5	CNRM-CM6-1	CNRM-CERFACS, France	$2.5^\circ \times 2.5^\circ$	$1^\circ \times 1^\circ$
6	CNRM-CM6-1-HR	CNRM-CERFACS, France	$1^\circ \times 1^\circ$	$0.25^\circ \times 0.25^\circ$
7	ECMWF-IFS-HR	ECMWF, Europe	$0.25^\circ \times 0.25^\circ$	$0.25^\circ \times 0.25^\circ$
8	ECMWF-IFS-LR	ECMWF, Europe	$0.5^\circ \times 0.5^\circ$	$1^\circ \times 1^\circ$
9	FGOALS-f3-H	CAS, China	$0.25^\circ \times 0.25^\circ$	$0.1^\circ \times 0.1^\circ$
10	FGOALS-f3-L	CAS, China	$1^\circ \times 1^\circ$	$1^\circ \times 1^\circ$
11	GFDL-CM4C192	NOAA-GFDL, United States	$0.5^\circ \times 0.5^\circ$	$0.25^\circ \times 0.25^\circ$
12	HadGEM3-GC31-HM	MOHC NERC, United Kingdom	$0.5^\circ \times 0.5^\circ$	$0.25^\circ \times 0.25^\circ$
13	HadGEM3-GC31-LM	MOHC NERC, United Kingdom	$2.5^\circ \times 2.5^\circ$	$0.25^\circ \times 0.25^\circ$
14	HadGEM3-GC31-MM	MOHC NERC, United Kingdom	$1^\circ \times 1^\circ$	$0.25^\circ \times 0.25^\circ$
15	INM-CM5-H	INM, Russia	$0.5^\circ \times 0.5^\circ$	$0.1^\circ \times 0.1^\circ$
16	IPSL-CM6A-ATM-HR	IPSL, France	$0.5^\circ \times 0.5^\circ$	none
17	IPSL-CM6A-LR	IPSL, France	$2.5^\circ \times 2.5^\circ$	$1^\circ \times 1^\circ$
18	MPI-ESM1-2-HR	MPI-M DWD DKRZ, Germany	$1^\circ \times 1^\circ$	$0.5^\circ \times 0.5^\circ$
19	MPI-ESM1-2-XR	MPI-M, Germany	$0.5^\circ \times 0.5^\circ$	$0.5^\circ \times 0.5^\circ$
20	MRI-AGCM3-2-H	MRI, Japan	$0.5^\circ \times 0.5^\circ$	none

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MODELS MEAN: NTA - summer SST



MODELS MEAN: NTA - fall SST



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89 **Figure S1.** NTA [MAM(t)] effects on boreal summer [JJA(t)] (a) and fall [SON(t)] (b) tropical Pacific
90 SST. The yellow and cyan contour lines specify p value = 0.1 and 0.33, respectively. Brown shades imply
91 low probability for the absence of Granger causality. Stippling designates that more than 70% of models
92 agree on the multi-model mean probability. For a given model, the agreement is defined when the

93 discrepancy between that model's probability and the multi-model mean probability is less than one
94 standard deviation of the multi-model mean probability. NTA: North Tropical Atlantic; SST: sea surface
95 temperature; MAM: March-April-May; JJA: June-July-August; SON: September-October-November.