



Supplement of

Future wintertime meridional wind trends through the lens of subseasonal teleconnections

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Figure S1. Individual CMIP5 models' first two leading V EOFs, based on Historical DJF data - 300hPa monthly subseasonal anomalies for the entire NH. The λ value for each EOF is denoted in parentheses in the titles. Positive (negative) values are displayed in red (blue), with an interval of 1 ms^{-1} . Contours greater than 2 ms^{-1} were omitted for clarity.



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Figure S2. Same as Fig. S1, but for EOFs calculated with RCP8.5 runs.



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Figure S3. The first two leading Historical MMM V EOFs, calculated for the entire NH (shading) and for the NA and AS sectors (contours).



Figure S4. Model skill for EOF representation, as expressed by the Pearson correlation coefficient between the cosine latitude weighted global functions and the NCEP-I patterns.

S1 Definition and filtering of CTP Events

We define "CTP events" in daily mean data. These are essentially Rossby Wave Packets which are nearly in phase with the 500 hPa equivalent of the preferably-phased pattern that was found in the 300 hPa monthly projections. First, we apply a

3-day running mean on the 500 hPa daily V field. After calculating the projection index and excluding low values (as in the

- 5 monthly case), we detect all sequences of 3 or more consecutive daily timesteps in which $|\gamma_d \gamma_m| \le \pi/8$, where γ_d and γ_m are the daily and preferred-monthly phases respectively. Lag 0 of a CTP event is defined as the first day of the sequence. Two consecutive sequences that are not separated by at least 48 hours are considered one CTP event. Our results were not found to be sensitive to the choice of these parameters. This method of filtering results in some false positive matches. We therefore further filter out all sequences whose composite does not display a wavy signature (alternating mean negative-positive-negative)
- 10 anomalies in the boxes marked in Fig. 8c). In order to determine the statistical significance of the resulting event composites, we use a 1000 member bootstrap method.

S2 Lagged linear regression of tropical proxies

We used lagged linear regression in an attempt to establish a causal relationship between tropical convective forcing (expressed by OLR and upper tropospheric divergence) and the excitation of CTP events. After choosing a base daily time series as an
independent variable (X_i), we regress onto it the 7-day lagged time series of a chosen dependent field (Y_i). Multiple regression equations are then derived, one for every gridpoint of Y. We map the resulting Y pattern by plugging an identical arbitrary x value in every equation. Statistical significance is assessed through the p-value of the correlation coefficient.

Following Livezey and Chen (1983), we assume that the coefficient's distribution is normal with a standard deviation of $1/\sqrt{n-3}$. *n* is the number of degrees of freedom, estimated by:

$$n = N/[1 + 2\sum_{i=1}^{N} C_{XX}(i\Delta t)C_{YY}(i\Delta t)]$$

Where N is the number of samples, Δt is the sampling time (1 day) and $C_{ZZ}(i\Delta t)$ is the autocorrelation of Z for lag $i\Delta t$.

Table S1. CMIP5 models used in this study. All data was taken from monthly-resolution Historical and RCP8.5 runs. Models denoted by (*) were studied as test cases using daily data.

Model Name	Institution
ACCESS1.0	Commonwealth Scientific and Industrial Research Organization
ACCESS1.3	(CSIRO) and Bureau of Meteorology
BCC-CSM1.1	Beijing Climate Center, China Meteorological Administration
BCC-CSM1.1(m)	
CCSM4	National Center for Atmospheric Research, United States
CESM1(BGC)	National Center for Atmospheric Research United States: Community
CESM1(CAM5)	Earth System Model Contributors
CESM1(WACCM)	
CMCC-CESM	
CMCC-CM	Centro Euro-Mediterraneo per I Cambiamenti Climatici
CMCC-CMS	-
CNRM-CM5	Centre National de Recherches Météorologiques / Centre Européen de
	Recherche et Formation Avancée en Calcul Scientifique
CSIRO-Mk3.6.0	Commonwealth Scientific and Industrial Research Organization in
	collaboration with Queensland Climate Change Centre of Excellence
FGOALS-g2	LASG, Institute of Atmospheric Physics, Chinese Academy of
	Sciences and CESS, Tsinghua University
FIO-ESM	The First Institute of Oceanography, SOA, China
GFDL-CM3	NOAA Geophysical Fluid Dynamics Laboratory, United States
GFDL-ESM2G	
GFDL-ESM2M	
GISS-E2-H	
GISS-E2-H-CC	NASA Goddard Institute for Space Studies, United States
GISS-E2-R-CC	• *
HadGEM2-AO	National Institute of Meteorological Research / Korea Meteorological
	Administration
HadGEM2-CC	Met Office Hadley Centre, United Kingdom
HadGEM2-ES	
INM-CM4	Institute for Numerical Mathematics, Russia
IPSL-CM5A-LR	
IPSL-CM5A-MR(*)	Institut Pierre-Simon Laplace, France
IPSL-CM5B-LR	
MIROC-ESM	Japan Agency for Marine-Earth Science and Technology, Atmosphere
MIROC-ESM-CHEM(*)	and Ocean Research Institute and National Institute for Environmental
MIROC5	Studies
MPI-ESM-MR	Max Planck Institute for Meteorology, Germany
MPI-ESM-LR	
MRI-CGCM3	Meteorological Research Institute, Japan
NorESM1-M	Norwegian Climate Centre
NorESM1-ME	