Underpredicted ENSO Teleconnections in Seasonal Forecasts

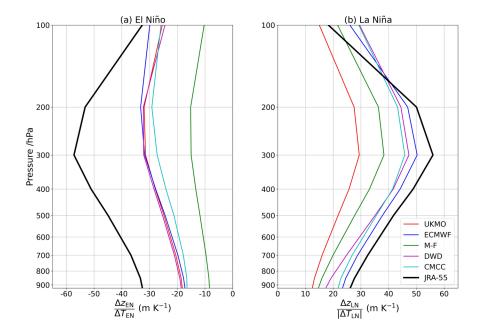
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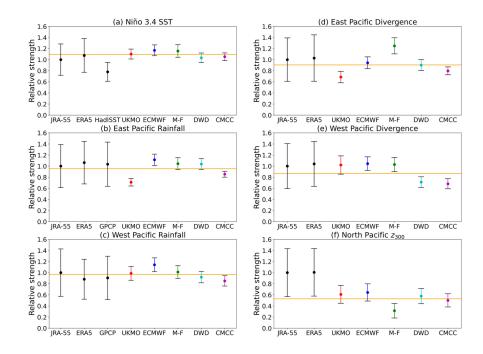
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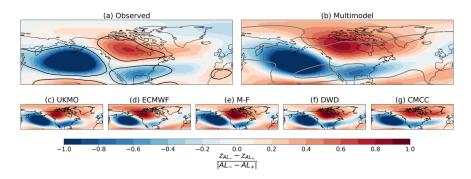
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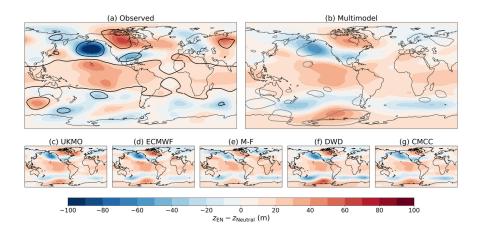
Abstract

The El Niño-Southern Oscillation (ENSO) influences climate variability across the globe. ENSO is highly predictable on seasonal timescales and therefore its teleconnections are a source of extratropical forecast skill. To fully harness this predictability, teleconnections must be represented accurately in seasonal forecasts. We find that a multimodel ensemble from five seasonal forecast systems can successfully capture the spatial structure of the late winter (JFM) El Niño teleconnection to the North Atlantic via North America, but the simulated amplitude is half of that observed. We find that weak amplitude teleconnections exist in all five models and throughout the troposphere, and that the La Niña teleconnection is also weak. We find evidence that the tropical forcing of the teleconnection is not underestimated and instead, deficiencies are likely to emerge in the extratropics. We investigate the impact of underestimated teleconnection strength on North Atlantic winter predictability, including its relevance to the signal-to-noise paradox.









Underpredicted ENSO Teleconnections in Seasonal Forecasts

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Key Points:

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7	• Seasonal forecasts severely underestimate the response to ENSO in the extratrop-
8	ical North Pacific
9	• The underestimated model response exists throughout the troposphere and is present
10	for both El Niño and La Niña
11	• Tropical processes which generate the teleconnection are well predicted and so the

problem does not appear to originate in the deep tropics

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13 Abstract

The El Niño-Southern Oscillation (ENSO) influences climate variability across the globe. 14 ENSO is highly predictable on seasonal timescales and therefore its teleconnections are 15 a source of extratropical forecast skill. To fully harness this predictability, teleconnec-16 tions must be represented accurately in seasonal forecasts. We find that a multimodel 17 ensemble from five seasonal forecast systems can successfully capture the spatial struc-18 ture of the late winter (JFM) El Niño teleconnection to the North Atlantic via North 19 America, but the simulated amplitude is half of that observed. We find that weak am-20 plitude teleconnections exist in all five models and throughout the troposphere, and that 21 the La Niña teleconnection is also weak. We find evidence that the tropical forcing of 22 the teleconnection is not underestimated and instead, deficiencies are likely to emerge 23 in the extratropics. We investigate the impact of underestimated teleconnection strength 24 on North Atlantic winter predictability, including its relevance to the signal-to-noise para-25 dox. 26

27 Plain Language Summary

The El Niño-Southern Oscillation (ENSO) describes the cycle of warmer and cooler 28 sea surface temperatures in the tropical Pacific Ocean, which influences climate around 29 the globe. The high heat capacity of the ocean means that ENSO changes relatively slowly 30 and so the ENSO phase — known as El Niño or La Niña — can be predicted with high 31 accuracy several months ahead. Far-flung influences — known as teleconnections — of 32 ENSO can provide predictability away from the tropics in seasonal forecasts if they are 33 accurately modelled. In this work, the late winter (Jan–Mar) ENSO teleconnection to 34 the North Atlantic, which travels via the North Pacific and North America, is investi-35 gated in five forecast models. We find that in all five models, the pattern of the telecon-36 nection is accurately captured, but the strength of the modelled teleconnection is half 37 of that in the real world. We find that the strength of processes in the tropics which cause 38 the teleconnection — including changes in sea surface temperatures and rainfall — are 39 not underestimated by models, meaning that the problem arises further along the path-40 way to the extratropics. This error likely contributes to the currently unresolved 'sig-41 nal to noise paradox' in climate forecasts. 42

43 **1** Introduction

The El Niño-Southern Oscillation (ENSO) is a major driver of global climate on 44 interannual timescales. Sea surface temperature anomalies associated with ENSO lead 45 to shifts in tropical convection, which generate poleward propagating Rossby waves (Sardeshmukh 46 & Hoskins, 1988). This leads to a quasi-stationary wave response in the extratropics, which 47 is modulated by the annual cycle. During the warm El Niño phase, a negative pressure 48 anomaly persists in the North Pacific throughout the winter, with a positive anomaly 49 over western Canada (Horel & Wallace, 1981). The response in the North Atlantic is typ-50 ically strongest in late winter and resembles the negative phase of the North Atlantic Os-51 cillation (Moron & Gouirand, 2003; Ayarzagüena et al., 2018). The Aleutian Low, which 52 is strongly coupled to ENSO, has been shown to have an inverse relationship with the 53 North Atlantic Oscillation in late winter (Honda et al., 2001). Several causes of the intraseasonal change in the teleconnection to the North Atlantic have been suggested, in-55 cluding modulation by the stratosphere (Ineson & Scaife, 2009; Cagnazzo & Manzini, 56 2009), and by differences in the tropical convective response in the west Pacific (Bladé 57 et al., 2008) and in the Indian Ocean (Abid et al., 2021). 58

In seasonal forecasts, ENSO is a potential source of predictability as it varies on longer than seasonal timescales. In order to translate this potential predictability into true predictability, models need to accurately capture teleconnections from ENSO to the extratropics. Not all variability in the climate system is predictable, but ideally models should capture the fraction of variance which is predictable. However, seasonal forecasts have been found to underestimate this fraction in predictions of the North Atlantic Oscillation (Eade et al., 2014; Scaife et al., 2014; Baker et al., 2018). This phenomenon is known as the signal-to-noise paradox (Scaife & Smith, 2018), as it is associated with an underestimated signal-to-noise ratio, and counter-intuitively implies that models are better at predicting the real world than they are at predicting their own ensemble members.

Whilst there is extensive literature on the winter extratropical response to ENSO in observations and free-running GCMs, existing literature on modelled amplitude of the extratropical response, or the performance of seasonal forecasts in capturing the response is relatively limited. Very few studies have used multimodel forecasts to examine their representation of ENSO-extratropical teleconnections (e.g. L'Heureux et al. (2017)).

We investigate the performance of five seasonal forecast systems in capturing teleconnections from ENSO to the North Pacific, North America and the the North Atlantic. We then seek to establish the region in which teleconnection errors originate. We then further examine the North Pacific–North Atlantic teleconnection pathway (cf. Honda et al. (2001)) in order to understand the effect of errors in modelled teleconnections on North Atlantic predictability in the context of the signal-to-noise paradox.

⁸¹ 2 Data and Methods

We use hindcasts for the winters 1993/1994 to 2016/2017 from five seasonal fore-82 cast systems. Met Office GloSea5 (MacLachlan et al. (2015); hereafter UKMO) hind-83 casts have 21 members, with 7 each initialized on the 25th October, 1st November and 84 8th November. Météo-France System 8 (Voldoire et al. (2019); M-F) hindcasts have 25 85 members, with 1 initialized on the 1st November, 12 initialized on the last Thursday of 86 October, and 12 initialized on the penultimate Thursday of October. CMCC-SPS3 (Sanna 87 et al. (2016); CMCC), DWD GCFS 2 (Fröhlich et al. (2021); DWD) and ECMWF SEAS5 88 (Johnson et al. (2019); ECMWF) hindcasts respectively have 40, 30 and 25 members ini-89 tialized on the 1st November. Where multimodel means have been computed, ensem-90 ble members are weighted equally, although results are robust to equal weighting of mod-91 els (not shown). The JRA-55 (Kobayashi et al., 2015) reanalysis with data from 1979/80-92 2016/17 is used for observations, and the ERA5 (Hersbach et al., 2020) reanalysis with 93 the same period is used for comparison. Additionally, the Global Precipitation Clima-94 tology Project v2.3 (Adler et al. (2003); hereafter GPCP) observational dataset from 1979/1980-95 2016/2017 is also used for precipitation, and the HadISST 2.2 dataset (Kennedy et al. 96 (2017); hereafter HadISST) from 1950/51-2014/15 is also used for sea surface temper-97 atures. For sea surface temperatures, December–March (DJFM) means are used. For 98 all other fields, January–March (JFM) means are used. Late winter is the focus of this work as the North Atlantic response to ENSO is most robust during this period (Moron 100 & Gouirand, 2003). 101

The Niño 3.4 index (Trenberth, 1997) — defined as the mean SST anomaly ΔT between 190–240°E and 5°S–5°N — is computed for the JRA-55 reanalysis and is used to split winters into El Niño ($\Delta T > 0.5$ K), La Niña ($\Delta T < -0.5$ K) and neutral ($|\Delta T| < 0.5$ K) phases. Of the 24 hindcast winters, 6 are classified as El Niño, 9 as La Niña, and 9 as neutral. The 38 winters in the reanalysis period consist of 10 El Niño, 11 La Niña and 17 neutral seasons. Hindcast Niño 3.4 indices correlate strongly with observations (r is between 0.96 and 0.99 for the ensemble mean of each model).

For a given field, the response to El Niño is defined as the composite mean of El Niño years for that field with the neutral composite subtracted. Similarly, the response to La Niña is the La Niña composite with neutral subtracted. For the purpose of quantifying uncertainty, standard deviations for observed El Niño responses are computed as the standard deviation of all El Niño years for the relevant dataset. In order to compare model standard deviations with observations, a sampling technique is used. For each sample, a neutral mean is computed using a random member from each neutral year. This is then subtracted from a random ensemble member from an El Niño year. 10,000 samples are taken, for which the standard deviation is computed.

Tropical East Pacific precipitation (TEP) is defined as the mean precipitation be-118 tween 160–240°E and 5°S–5°N, and is strongly coupled to ENSO. Tropical West Pacific 119 precipitation (TWP) is defined as the mean precipitation between 110–150°E and 0–20°N. 120 121 These boxes were chosen to capture the strongest precipitation responses during both El Niño and La Niña years, and are similar to the TEP and TWP boxes used in Scaife 122 et al. (2017). TWP is considered as it has a known impact on the extratropical North-123 ern Hemisphere, including the North Atlantic Oscillation (Kucharski et al., 2006; Scaife 124 et al., 2017; Scaife, Ferranti, et al., 2019). 125

Due to the observed asymmetry of the North Pacific response to El Niño and La Niña, boxed regions of 180–220°E and 40–60°N for El Niño, and 170–210°E and 30–50°N for La Niña, are used when comparing the strength of the extratropical responses. These boxes are chosen in order to capture the strongest response in each phase.

The Aleutian Low is defined as the mean 300 hPa geopotential height z between 130 $170-220^{\circ}$ E and $30-60^{\circ}$ N. This is a region of climatological low geopotential height with 131 high interannual variability, and the boundaries are chosen to capture the regions with 132 the strongest responses during El Niño and La Niña years. Aleutian Low composites are 133 taken by defining negative and positive phases, where the Aleutian Low is deeper (more 134 negative) or shallower (less negative) than the mean respectively. As the correlation be-135 tween models and observation is less strong for the Aleutian Low than it is for the Niño 136 3.4 index (0.5 < r < 0.7 for the ensemble mean of models), model Aleutian Low in-137 dices are used to create model composites instead of the observed index. The 38 reanal-138 ysis winters consist of 17 negative seasons and 21 positive seasons. The multimodel en-139 semble mean evenly splits into 12 negative and 12 positive winters. DWD has 11 neg-140 ative winters out of 24, UKMO and M-F both have 12, and ECMWF and CMCC both 141 have 14. 142

¹⁴³ **3** Weak Teleconnections

Figure 1 (a) shows the response of 300 hPa geopotential height to El Niño computed using JRA-55 data. A significant increase in geopotential height occurs throughout most of the tropics due to tropical warming and the canonical Pacific-North American pattern (e.g. Horel and Wallace (1981)) is present. In the North Atlantic a tripole pattern emerges, with a large positive anomaly across the basin just south of Greenland, and negative anomalies to the north east and south west.

Figure 1 (b) shows the multimodel response to El Niño. The simulated pattern is 150 very similar to that observed but the strength of the response in the extratropical North-151 ern Hemisphere is severely underestimated. This is most evident over the North Pacific, 152 where the magnitude of the composite response is greater than 90 m over a large region 153 in observations but does not exceed 60 m anywhere in the multimodel mean. Figure 1 154 (c)–(g) shows the individual model responses. Each model captures the structure of the 155 response over the North Pacific and North America, and all except the Météo-France model 156 show a North Atlantic pattern which resembles the observed pattern. The underestimated 157 amplitude of the teleconnection is present in all five models. Additionally, in all mod-158 els the North Pacific response is tilted on a north west-south east axis relative to ob-159 servations. Whilst the response over North America appears to be weak, significance is 160 lower than in the North Pacific. 161

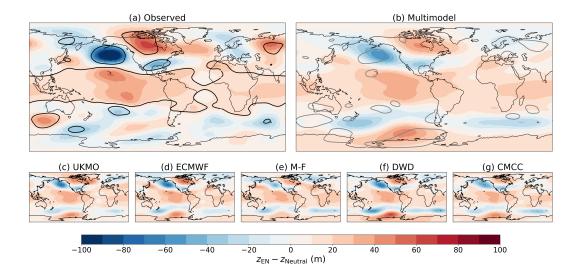


Figure 1. El Niño teleconnections in observations and models. Difference between El Niño and neutral composites of 300 hPa geopotential height for JRA-55 (a), the mean of all five models (b) and each of the five models ((c)–(g), labelled). Black line contours on (a) bound regions where the El Niño and neutral years are significantly different above the 10% level in observations according to a two-tailed t test. Gray line contours on (b) bound regions where the modelled response is significantly different to the observed response at the 10% level according to a two-tailed t test.

While there is reasonable agreement between observed and modelled teleconnec-162 tions in the North Atlantic, the observed North Atlantic pattern for this period closely 163 resembles the Atlantic wavetrain response to strong El Niño events found in Toniazzo 164 and Scaife (2006). Removing the four El Niño events (1983, 1992, 1998, 2016) classed 165 as strong by their definition ($\Delta T > 1.5$ K) from the observed composite leads to a pat-166 tern which more closely resembles the pattern from the multimodel mean (not shown). 167 This may mean that either the models do not capture the non-linearity in the North At-168 lantic response, or that part of the strong El Niño pathway via the tropical Atlantic found 169 in Toniazzo and Scaife (2006) is also too weak. However, the limited availability of ob-170 served El Niño years means that caution should be taken when comparing strength-based 171 subsamples of the available data. 172

Figure 2 (a) shows that the extratropical Pacific response to El Niño is underes-173 timated throughout the troposphere and in all five models, by about half in four mod-174 els and even more than this in M-F. Figure 2 (b) shows that the response to La Niña 175 is also underestimated in all models, at all tropospheric levels except for 100 hPa. The 176 modelled El Niño and La Niña responses are not equally weak, with a less weak response 177 to La Niña in most models. For this reason, we focus on the response to El Niño in the 178 next section. Figure S1 shows maps of the 300 hPa geopotential height response to La 179 Niña, calculated using JRA-55 (a) and the multimodel ensemble mean (b). The peak strength 180 of the response to La Niña in the North Pacific is strongly underestimated in the mul-181 timodel mean, but the spatial extent of the positive geopotential height anomaly is broader 182 than observed. 183

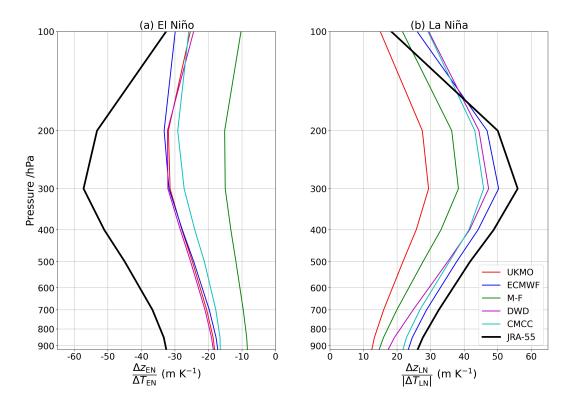


Figure 2. Vertical structure of ENSO teleconnections to the North Pacific. Different boxes are used for the two phases — (180–220°E, 40–60°N) for El Niño and (170–210°E, 30–50°N) for La Niña — to account for asymmetry in the teleconnection. The geopotential height responses are divided by the Niño 3.4 index response from the same source to account for SST biases.

¹⁸⁴ 4 Errors in the Teleconnection Pathway

We next consider where the model errors emerge in the chain of processes involved 185 in the teleconnection. Figure 3 shows the strength of the response of six indices to El 186 Niño in observational datasets and in hindcasts, with \pm two standard errors from the 187 mean shown to represent uncertainty. Figure 3 (a) shows that all five models actually 188 overestimate changes in Niño 3.4 sea surface temperatures relative to JRA-55, although 189 the model means are within the uncertainty range. This suggests that if sea surface tem-190 peratures were perfectly predicted, the underestimation of the modelled teleconnection 191 would be even worse. When comparing to ERA5, there is almost no error in the mul-192 timodel mean response. All five models are above the uncertainty range of HadISST, al-193 though it should be noted that the HadISST time period (1950/51-2014/15) is differ-194 ent to the reanalysis period. Modelled responses of East Pacific (TEP) rainfall to El Niño 195 (Figure 3 (b)) are similar to those in JRA-55, with two models underestimating the re-196 sponse and three models overestimating, and the difference between the JRA-55 and mul-197 timodel mean is much smaller than their respective uncertainties. ERA5 and GPCP are 198 in strong agreement with JRA-55 for the TEP response to El Niño. For West Pacific (TWP) 199 rainfall (Figure 3 (c)) all three observational datasets and all five models show similar 200 responses, with model responses well within the uncertainty range of each observational 201 dataset. 202

The generation of Rossby waves requires a Rossby wave source S (Sardeshmukh & Hoskins, 1988)

$$S = -\zeta \nabla \cdot \boldsymbol{u}_{\chi} - \boldsymbol{u}_{\chi} \cdot \nabla \zeta, \tag{1}$$

where ζ is the absolute vorticity and u_{χ} is the divergent component of the hori-205 zontal wind. Increased tropical precipitation due to surface heating is associated with 206 a baroclinic divergence response, with convergence near the surface and divergence at 207 the level of convective outflow. Tropical divergence anomalies lead to divergent winds 208 over a broader region, extending into areas with higher background vorticity and more 209 favourable conditions for Rossby wave propagation. We found that the divergence re-210 sponse to ENSO in the tropical Pacific is not robust at 300 hPa, but it is highly robust 211 and strongly correlated with precipitation at 200 hPa. This level is consistent with pre-212 vious studies (e.g. Sardeshmukh and Hoskins (1988), Scaife et al. (2017)). Outside of the 213 tropics, Rossby wave sources are reinforced by the Rossby waves themselves, as they re-214 sult in anomalous vorticity and divergence. For this reason, we do not include any cal-215 culations of subtropical or extratropical Rossby wave source, as underestimated values 216 in models may be caused by, rather than the cause of, underestimated teleconnection am-217 plitude. 218

Figure 3 (d) shows the response of 200 hPa divergence in the tropical east Pacific 219 (using the same box as TEP rainfall) to El Niño calculated using JRA-55, ERA5 and 220 each model. The observational estimates have very similar mean responses. The mul-221 timodel mean is slightly weaker than observations, but it is well within the margin of un-222 certainty and all five individual models are within the 2 standard error range of both re-223 analyses. Figure 3 (e) shows the 200 hPa divergence response in the tropical west Pa-224 cific (using the box for TWP rainfall). The multimodel mean response is weaker than 225 that of TEP divergence (around 85% of the JRA-55 response), but it is still well within 226 the margin of error of both reanalyses. Finally, the extratropical anomalies in Figure 3 227 (f) demonstrate the significance of the weak response of North Pacific geopotential height 228 to El Niño, as the strength of the multimodel mean is below the two standard error range 229 from the JRA-55 mean. JRA-55 and ERA5 are in strong agreement on both the mean 230 response and its error. 231

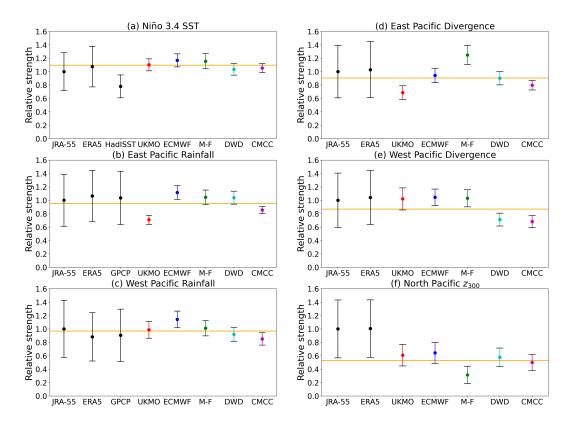


Figure 3. Strength of responses to El Niño for six indices: (a) Niño 3.4 sea surface temperature; (b) Tropical East Pacific precipitation; (c) Tropical West Pacific precipitation; (d) Tropical East Pacific divergence at 200 hPa; (e) Tropical West Pacific divergence at 300 hPa; (f) North Pacific geopotential height at 300 hPa, all normalized relative to JRA-55. Circles denote means from each data source. Error bars denote two standard errors from the mean. Orange horizontal lines denote the multimodel mean.

If it is assumed that the multimodel tropical divergence response is systematically 232 weak, the extent to which it is weak (a factor of around 0.85 relative to JRA-55 for both 233 TEP and TWP) is not sufficient to explain the weak North Pacific response (a factor of 234 around 0.55) without the influence of highly non-linear mechanisms. Therefore, the trop-235 ical forcing of the teleconnection does not appear to be the source of the weak telecon-236 nection. For La Niña we found that the TWP precipitation and divergence responses are 237 underestimated in all models, with high significance for the multimodel mean and three 238 of the models (ECMWF, M-F, CMCC; not shown), making it more difficult to separate 239 tropical and extratropical effects which contribute to the underestimated teleconnection 240 strength. 241

Finally, we consider the link from the Pacific to the Atlantic. Figure 4 (a) shows 242 the geopotential height response to a unit deepening of the Aleutian Low, for JRA-55 243 data. Figure 4 (b) shows the same, but for the multimodel ensemble. At mid-latitudes, 244 the geopotential height response to deepening of the Aleutian Low is relatively accurate 245 in the multimodel ensemble — a negative NAO-like pattern (cf. Honda et al. (2001)) ex-246 ists in both observations and models, and a similar dipole anomaly occurs over North 247 America. The amplitude of the pattern is much closer to the observed link than is the 248 case for the response to ENSO. This suggests that improving the strength of the Aleu-249 tian Low response to ENSO would improve prediction over North America and in the 250 North Atlantic sector, as well as in the North Pacific sector. Figure 4 (c)-(g) shows the 251

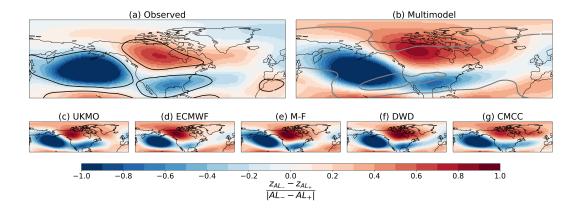


Figure 4. Teleconnections between the Pacific and Atlantic basins. 300 hPa geopotential height response to a unit deepening of the Aleutian Low. Defined as the difference between composites of geopotential height during years with a deeper-than-average Aleutian Low and those with a less deep Aleutian Low, divided by the absolute difference in the mean Aleutian Low for the same sets of years. For JRA-55 (a), the multimodel mean (b) and individual models ((c)–(g), labelled). Black line contours on (a) bound regions where the difference is significant at the 10% level in observations according to a two-tailed t test. Gray line contours on (b) bound regions where the observed and multimodel results are significantly different at the 10% level according to a two-tailed t test.

response for individual models. Each model has a similar response to the multimodel mean
and observations, although they vary in how far the dipole extends eastwards into the
North Atlantic, and the North Atlantic response in the ECMWF model is weaker than
others.

²⁵⁶ 5 Discussion and Conclusions

Underprediction of teleconnections has serious ramifications for seasonal forecasts. 257 If total ensemble variance of forecasts is realistic, halving the strength of teleconnections 258 will degrade model predictability and reduce skill. These results also have a bearing on 259 the signal-to-noise paradox on interannual timescales. However, this problem is also known 260 to exist in longer term predictions (Eade et al., 2014; Smith et al., 2020) where ENSO 261 is not the main driver of predictable signals. Therefore, whilst ENSO is unlikely to be 262 the direct cause of the paradox, its weak teleconnections may have the same origin, as 263 has also been found for the Quasi-Biennial Oscillation (O'Reilly et al., 2019). It is there-264 fore necessary to establish the cause of weak teleconnections in general, in order to not 265 only improve modelled ENSO-associated interannual variability but potentially to also 266 improve modelled variability due to other drivers and at different timescales. Leading 267 hypotheses for this more general problem include transient eddy feedback (Scaife, Camp, 268 et al., 2019; Hardiman et al., 2022) and ocean-atmosphere interaction (Zhang & Kirt-269 man, 2019; Ossó et al., 2020). ENSO is a global driver of climate variability, but its in-270 fluence on North American and European winters is underestimated in current seasonal 271 forecasts. Solving this problem would reduce uncertainty and increase skill in winter pre-272 diction in these regions. 273

The underestimation of the El Niño teleconnection amplitude has strong statistical significance (cf. Figure 3 (f)) despite the uncertainty of the observed amplitude due to the limited number of observed events (cf. Deser et al. (2017)). This level of significance was found to remain when including pre-satellite era JRA-55 and ERA5 reanalysis data (starting from 1958/59 and 1950/51 respectively; not shown).

This study demonstrates that current seasonal forecasts are unable to capture the strength of the atmospheric response to ENSO in the North Pacific, which in turn affects ENSO teleconnections to the North Atlantic. This has an impact on seasonal prediction of wind speed, temperature and precipitation during winter in North America and Europe.

²⁸⁴ 6 Open Research

Hindcast data from the five seasonal forecast systems used in this work are freely 285 available online (https://doi.org/10.24381/cds.0b79e7c5 for geopotential and wind com-286 ponent data, and https://doi.org/10.24381/cds.68dd14c3 for all other fields), with the 287 following originating centre and system labels: UK Met Office, 15 for UKMO; ECMWF, 288 5 for ECMWF; Météo France, 8 for M-F; DWD, 21 for DWD; CMCC, 35 for CMCC. 289 Hindcast data was produced by the institutes that developed each forecast system: The 290 UK Met Office for UKMO (MacLachlan et al., 2015); the European Centre for Medium-291 Range Weather Forecasts for ECMWF (Johnson et al., 2019); Météo-France for M-F (Voldoire 292 et al., 2019); The Deutscher Wetterdienst for DWD (Fröhlich et al., 2021); The Euro-293 Mediterranean Center on Climate Change for CMCC (Sanna et al., 2016). 294

The JRA-55 reanalysis data (Kobayashi et al., 2015) used in this work are freely 295 available online (https://doi.org/10.5065/D60G3H5B). The GPCP data (Adler et al., 296 2003) used in this work are freely available online (https://psl.noaa.gov/data/gridded/data.gpcp.html). 297 The ERA5 reanalysis data (Hersbach et al., 2020) used in this work are freely available 298 online (https://doi.org/10.24381/cds.6860a573 for geopotential and wind component data, 299 and https://doi.org/10.24381/cds.f17050d7 for all other data). The HadISST (Kennedy 300 et al., 2017) sea surface temperature data used in this work is freely available online at 301 https://esgf-node.llnl.gov/search/input4mips/ under the HighResMIP (Haarsma et al., 302 2016) Target MIP. 303

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309 References

- Abid, M. A., Kucharski, F., Molteni, F., Kang, I.-S., Tompkins, A. M., & Almazroui,
 M. (2021). Separating the Indian and Pacific Ocean impacts on the Euro Atlantic response to ENSO and its transition from early to late winter. Jour nal of Climate, 34 (4), 1531–1548.
- 314Adler, R. F., Huffman, G. J., Chang, A., Ferraro, R., Xie, P.-P., Janowiak, J., ...315Nelkin, E. (2003). The version-2 global precipitation climatology project316(GPCP) monthly precipitation analysis (1979–present). Journal of Hydrom-317eteorology, 4(6), 1147–1167. Retrieved from https://doi.org/10.1175/3181525–7541(2003)004<1147:TVGPCP>2.0.C0;2
- Ayarzagüena, B., Ineson, S., Dunstone, N. J., Baldwin, M. P., & Scaife, A. A.
 (2018). Intraseasonal effects of El Niño–Southern Oscillation on North Atlantic Climate. *Journal of Climate*, *31*(21), 8861–8873.
- Baker, L., Shaffrey, L., Sutton, R., Weisheimer, A., & Scaife, A. A. (2018). An
 intercomparison of skill and overconfidence/underconfidence of the winter time North Atlantic Oscillation in multimodel seasonal forecasts. *Geophysical Research Letters*, 45 (15), 7808–7817.
- Bladé, I., Newman, M., Alexander, M. A., & Scott, J. D. (2008). The late fall extratropical response to ENSO: sensitivity to coupling and convection in the tropical West Pacific. *Journal of Climate*, 21 (23), 6101–6118.
- Cagnazzo, C., & Manzini, E. (2009). Impact of the stratosphere on the winter tropospheric teleconnections between ENSO and the North Atlantic and European region. *Journal of climate*, 22(5), 1223–1238.
- Deser, C., Simpson, I. R., McKinnon, K. A., & Phillips, A. S. (2017). The North ern Hemisphere extratropical atmospheric circulation response to ENSO: How
 well do we know it and how do we evaluate models accordingly? *Journal of Climate*, 30(13), 5059–5082.
- Eade, R., Smith, D. M., Scaife, A. A., Wallace, E., Dunstone, N. J., Hermanson, L.,
 & Robinson, N. (2014). Do seasonal-to-decadal climate predictions underestimate the predictability of the real world? *Geophysical Research Letters*, 41(15), 5620–5628.
- Fröhlich, K., Dobrynin, M., Isensee, K., Gessner, C., Paxian, A., Pohlmann, H., ...
 Baehr, J. (2021). The German climate forecast system: GCFS. Journal of Advances in Modeling Earth Systems, 13(2), e2020MS002101. Retrieved from https://doi.org/10.1029/2020MS002101
- Haarsma, R. J., Roberts, M. J., Vidale, P. L., Senior, C. A., Bellucci, A., Bao, Q.,
 ... von Storch, J.-S. (2016). High resolution model intercomparison project (HighResMIP v1. 0) for CMIP6. *Geoscientific Model Development*, 9(11), 4185-4208.
- Hardiman, S. C., Dunstone, N. J., Scaife, A. A., Smith, D. M., Comer, R., Nie, Y.,
 & Ren, H.-L. (2022). Missing eddy feedback may explain weak signal-to-noise
 ratios in climate predictions. *npj Climate and Atmospheric Science*, 5(57).
- Hersbach, H., Bell, B., Berrisford, P., Hirahara, S., Horányi, A., Muñoz-Sabater, J.,
 ... Thépaut, J.-N. (2020). The ERA5 global reanalysis. *Quarterly Journal* of the Royal Meteorological Society, 146(730), 1999–2049. Retrieved from
 https://doi.org/10.1002/qj.3803
- Honda, M., Nakamura, H., Ukita, J., Kousaka, I., & Takeuchi, K. (2001). Interannual seesaw between the Aleutian and Icelandic lows. Part I: Seasonal
 dependence and life cycle. *Journal of Climate*, 14(6), 1029–1042.
- Horel, J. D., & Wallace, J. M. (1981). Planetary-scale atmospheric phenomena as sociated with the Southern Oscillation. Monthly Weather Review, 109(4), 813–
 829.
- Ineson, S., & Scaife, A. A. (2009). The role of the stratosphere in the European climate response to El Niño. *Nature Geoscience*, 2(1), 32–36.

363	Johnson, S. J., Stockdale, T. N., Ferranti, L., Balmaseda, M. A., Molteni, F., Mag-
364	nusson, L., Monge-Sanz, B. M. (2019). SEAS5: the new ECMWF seasonal
365	forecast system. Geoscientific Model Development, 12(3), 1087–1117. Re-
366	trieved from https://doi.org/10.5194/gmd-12-1087-2019
367	Kennedy, J., Titchner, H., Rayner, N., & Roberts, M. (2017). input4MIPs.
368	MOHC. SSTsAndSeaIce. HighResMIP. MOHC-HadISST-2-2-0-0. Ver-
369	sion 20170505. Earth System Grid Federation, 10. Retrieved from
370	https://esgf-node.llnl.gov/search/input4mips
371	Kobayashi, S., Ota, Y., Harada, Y., Ebita, A., Moriya, M., Onoda, H., Taka-
372	hashi, K. (2015). The JRA-55 reanalysis: General specifications and basic
373	characteristics. Journal of the Meteorological Society of Japan. Ser. II, $93(1)$,
374	5-48. Retrieved from https://doi.org/10.2151/jmsj.2015-001
375	Kucharski, F., Molteni, F., & Bracco, A. (2006). Decadal interactions between the
376	western tropical Pacific and the North Atlantic Oscillation. <i>Climate Dynamics</i> ,
377	26(1), 79-91.
378	L'Heureux, M. L., Tippett, M. K., Kumar, A., Butler, A. H., Ciasto, L. M., Ding,
379	Q., Johnson, N. C. (2017). Strong relations between ENSO and the Arctic
380	Oscillation in the North American multimodel ensemble. <i>Geophysical Research</i>
381	Letters, $44(22)$, $11-654$.
382	MacLachlan, C., Arribas, A., Peterson, K. A., Maidens, A., Fereday, D., Scaife, A., Madec, G. (2015). Global Seasonal forecast system version 5
383	(GloSea5): A high-resolution seasonal forecast system. Quarterly Journal
384 385	of the Royal Meteorological Society, 141(689), 1072–1084. Retrieved from
386	https://doi.org/10.1002/qj.2396
387	Moron, V., & Gouirand, I. (2003). Seasonal modulation of the El Niño–Southern Os-
388	cillation relationship with sea level pressure anomalies over the North Atlantic
389	in October–March 1873–1996. International Journal of Climatology: A Journal
390	of the Royal Meteorological Society, 23(2), 143–155.
391	O'Reilly, C. H., Weisheimer, A., Woollings, T., Gray, L. J., & MacLeod, D. (2019).
392	The importance of stratospheric initial conditions for winter North Atlantic
393	Oscillation predictability and implications for the signal-to-noise paradox.
394	Quarterly Journal of the Royal Meteorological Society, 145(718), 131–146.
395	Ossó, A., Sutton, R. T., Shaffrey, L. C., & Dong, B. (2020). Development, ampli-
396	fication, and decay of Atlantic/European summer weather patterns linked to
397	spring North Atlantic sea surface temperatures. Journal of Climate, $33(14)$,
398	5939–5951.
399	Sanna, A., Borrelli, A., Athanasiadis, P., Materia, S., Storto, A., Navarra, A.,
400	Gualdi, S. (2016). CMCC-SPS3: the CMCC Seasonal Prediction System 3.
401	CMCC Research Paper (RP0285). Retrieved from https://www.cmcc.it/
402	<pre>publications/rp0285-cmcc-sps3-the-cmcc-seasonal-prediction-system -3</pre>
403	
404	Sardeshmukh, P. D., & Hoskins, B. J. (1988). The generation of global rotational flow by steady idealized tropical divergence. <i>Journal of the Atmospheric Sci</i> -
405 406	ences, 45(7), 1228–1251.
407	Scaife, A. A., Arribas, A., Blockley, E., Brookshaw, A., Clark, R., Dunstone, N. J.,
408	Williams, A. (2014). Skillful long-range prediction of European and North
409	American winters. Geophysical Research Letters, 41(7), 2514–2519.
410	Scaife, A. A., Camp, J., Comer, R. E., Davis, P., Dunstone, N. J., Gordon, M.,
411	Vidale, P. L. (2019). Does increased atmospheric resolution improve seasonal
412	climate predictions? Atmospheric Science Letters, 20(8), e922.
413	Scaife, A. A., Comer, R. E., Dunstone, N. J., Knight, J. R., Smith, D. M., MacLach-
414	lan, C., Slingo, J. (2017). Tropical rainfall, Rossby waves and regional
415	winter climate predictions. Quarterly Journal of the Royal Meteorological
416	Society, $143(702)$, 1–11.
417	Scaife, A. A., Ferranti, L., Alves, O., Athanasiadis, P., Baehr, J., Dequé, M.,

418	Yang, X. (2019). Tropical rainfall predictions from multiple seasonal forecast
419	systems. International Journal of Climatology, 39(2), 974–988.
420	Scaife, A. A., & Smith, D. M. (2018). A signal-to-noise paradox in climate science.
421	npj Climate and Atmospheric Science, 1(1), 1–8.
422	Smith, D. M., Scaife, A. A., Eade, R., Athanasiadis, P., Bellucci, A., Bethke, I.,
423	Zhang, L. (2020). North Atlantic climate far more predictable than models
424	imply. <i>Nature</i> , 583(7818), 796–800.
425	Toniazzo, T., & Scaife, A. A. (2006). The influence of ENSO on winter North At-
426	lantic climate. Geophysical Research Letters, $33(24)$.
427	Trenberth, K. E. (1997). The definition of El Niño. Bulletin of the American Meteo-
428	$rological \ Society, \ 78(12), \ 2771-2778.$
429	Voldoire, A., Saint-Martin, D., Sénési, S., Decharme, B., Alias, A., Chevallier, M.,
430	Waldman, R. (2019). Evaluation of CMIP6 deck experiments with CNRM-
431	CM6-1. Journal of Advances in Modeling Earth Systems, 11(7), 2177–2213.
432	Retrieved from https://doi.org/10.1029/2019MS001683
433	Zhang, W., & Kirtman, B. (2019). Understanding the signal-to-noise paradox with a
434	simple Markov model. Geophysical Research Letters, $46(22)$, 13308–13317.

Supporting Information for "Underpredicted ENSO Teleconnections in Seasonal Forecasts"

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Contents of this file

1. Figure S1

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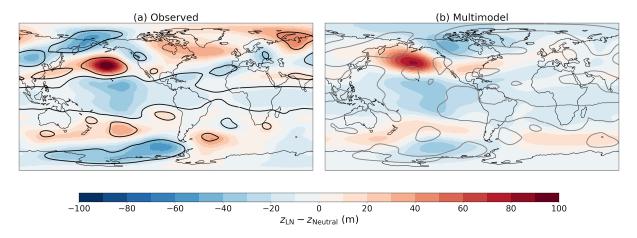


Figure S1. Geopotential height response to La Niña. Difference between La Niña and neutral composites of 300 hPa geopotential height for JRA-55 (a) and the mean of all five models (b). Black line contours on (a) bound regions where the La Niña and neutral years are significantly different above the 10% level in observations according to a two-tailed t test. Gray line contours on (b) bound regions where the modelled response is significantly different to the observed response at the 10% level according to a two-tailed t test.