

1 QBOi El Niño Southern Oscillation experiments: Assessing 2 relationships between ENSO, MJO, and QBO

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52 **Abstract.** This study uses an ensemble of climate model experiments coordinated by the Quasi-Biennial Oscillation initiative
53 (QBOi) to analyze the Madden-Julian Oscillation (MJO) in the presence of either perpetual El Niño or La Niña sea surface
54 temperatures during boreal winter. In addition to the prescribed El Niño Southern Oscillation (ENSO) conditions, the nine
55 models internally generate QBOs, meaning each may influence the MJO. Objectives of our analyses are to assess the response
56 of the MJO to strong idealized ENSO forcing and look for evidence of a QBO influence on the MJO in a multi-model context.
57 The diagnostics used include wavenumber-frequency spectra of tropical convective and dynamical fields, measures of MJO
58 lifetime, an evaluation of MJO diversity and visualization of MJO vertical structure, as well as an assessment of QBO
59 morphology and the QBO’s impact on tropical convection. Kelvin wave spectral power increases in the El Niño simulations
60 whereas equatorial Rossby waves power is stronger in the La Niña simulations. All models simulate faster MJO propagation
61 under El Niño conditions. This change in speed is corroborated by the MJO diversity analysis, which reveals that models better
62 reproduce the observed “fast propagating” and “standing” MJO archetypes given perpetual El Niño and La Niña, respectively.
63 Regardless of ENSO, QBO descent into the lower stratosphere is underestimated and we detect little QBO influence on tropical
64 tropopause stability and MJO activity. With little influence from the QBO on the MJO activity in these runs, we can be
65 confident that the aforementioned changes in the MJO indeed arise from the different ENSO boundary conditions.

66 **1 Introduction**

67 The tropical circulation is influenced by various forms of internal variability, each operating at different timescales, yet still
68 influencing each other. The Madden-Julian Oscillation (MJO) is dominant at intraseasonal timescales (Madden and Julian
69 1994; Lin, 2022). It consists of large-scale eastward propagating fluctuations in tropical precipitation and circulation that
70 traverse the Indian Ocean and Maritime Continent through to the Pacific over roughly 30 to 60 days (Hendon and Salby 1994).
71 MJO variability fluctuates a lot year to year as does other variability in the climate system.

72
73 At interannual timescales, the El Niño Southern Oscillation (ENSO; Philander, 1990) is one of the most consequential sources
74 of tropical tropospheric variability. It is characterized by shifting patterns of sea surface temperatures (SSTs) and associated
75 changes in ocean and atmospheric circulations in the tropical Pacific. ENSO varies on timescales between two to seven years,
76 and consists of three phases, the warm El Niño, the cold La Niña and a “neutral” phase where neither polarity dominates. Also

77 operating at interannual timescales is the Quasi Biennial Oscillation (QBO), which is the dominant mode of variability in the
78 lower tropical stratosphere, defined by alternating easterly and westerly zonal wind shear zones descending from 5 to 100 hPa
79 with an average periodicity of 28 months (Baldwin et al. 2001).
80

81 The MJO, ENSO, and QBO are known to influence each other in various ways. ENSO's La Niña and El Niño phases are
82 associated with shifts of intraseasonal tropical atmospheric variability like the MJO towards the Indo-Pacific Warm Pool and
83 Date Line, respectively (Kessler 2001; Tam and Lau 2005). In addition, ENSO can influence the amount of time the MJO
84 spends in particular Wheeler and Hendon (2004) MJO phases and the duration of MJO events overall, which are shorter during
85 El Niño and longer during La Niña (Pohl and Matthews 2007; Pang et al. 2016; Wei and Ren 2019; Wang et al. 2019; Dasgupta
86 et al. 2021; Fernandes and Grimm 2023).

87
88 Nonetheless, not all research finds strong links between ENSO and seasonal-mean MJO activity. Slingo et al. (1999) found
89 that the observed intraseasonally filtered zonal mean 200 hPa zonal wind (their metric of "MJO activity") is weakly dependent
90 on ENSO phase. They affirmed this further by using an ensemble of AMIP simulations. Hendon et al. (1999) validated and
91 refined their definition of MJO activity, finding it to capture the salient features of the MJO and again that its variability is
92 mostly independent of ENSO. With these results in mind, it is less clear how the MJO should respond to the ENSO conditions
93 prescribed in our simulations. In fact, a common idea amongst the studies just mentioned is that the MJO's interannual
94 variability originates predominantly from internal atmospheric processes other than those associated with ENSO.

95
96 It is increasingly recognized that the easterly and westerly phases of the QBO exert an influence on the MJO (Yoo and Son,
97 2016; Son et al. 2017; Sakaeda et al. 2020; Martin et al. 2021; Jin et al. 2023; Huang et al. 2023). The MJO's amplitude is
98 stronger during easterly QBO boreal winters compared to westerly QBO winters over the observed record since 1979 (Yoo
99 and Son 2016; Densmore et al. 2019). However, despite improvement in the representation of simulated QBOs (Richter et al.
100 2020) and MJOs (Ahn et al. 2020) across model generations, current free-running Earth system models generally do not
101 simulate the QBO-MJO relationship (Kim et al. 2020; Lim and Son 2020; Martin et al. 2023), nor do they simulate a
102 sufficiently strong tropical tropopause response to the QBO (Serva et al. 2022). Further complicating interpretation is a
103 tendency for easterly QBO boreal winters to co-occur with La Niña events, introducing ambiguity about the source of MJO
104 modulation (Randall et al. 2023).

105
106 In this study, we document the influence of ENSO on the MJO using a multi-model ensemble of idealized experiments with
107 perpetual El Niño and La Niña forcings. These simulations were coordinated by the Atmospheric Processes And their Role in
108 Climate (APARC, previously "SPARC") Quasi-Biennial Oscillation initiative (QBOi, Butchart et al. 2018). In addition to
109 prescribing ENSO, the models internally generate QBOs, meaning both types of interannual internal variability may modulate
110 the MJO in these experiments. Our aim is to assess the extent to which model behavior is consistent with previously reported

111 ENSO-MJO relationships, recognizing that past studies have found different ENSO-MJO links and that the simulated QBOs
112 may also project onto the MJOs. The coordinated QBOi protocol allows us to revisit these relationships in a controlled
113 framework that isolates atmospheric responses to fixed, high-amplitude ENSO SSTs across many models. These perpetual
114 ENSO conditions represent the strongest observed El Niños and La Niñas and offer an upper bound on ENSO’s effect on MJO
115 characteristics, providing a high signal-to-noise database for studying this connection. We also selectively incorporate Phase
116 1 QBOi experiments, performed as 1979-2009 Atmospheric Model Intercomparison Project experiments, and hence, more
117 representative of typical ENSO amplitudes (Butchart et al. 2018, Experiment 1). Unlike Coupled or Atmospheric Model
118 Intercomparison Project experiments, or historical reanalysis datasets, these simulations avoid complications associated with
119 time-evolving forcings, event-to-event ENSO variability, and background SST biases. On the other hand, while perpetual
120 ENSO conditions help isolate atmospheric responses, they also preclude direct comparison with observation-based datasets,
121 which include event-to-event variability, for example. The perpetual-ENSO experiments used here are a continuation of the
122 QBOi Phase 1 experiments and have companion studies that examine ENSO’s effect on the QBO (Kawatani et al., 2025) and
123 the combined influence of ENSO and the QBO on global teleconnections (Naoe et al., 2025).

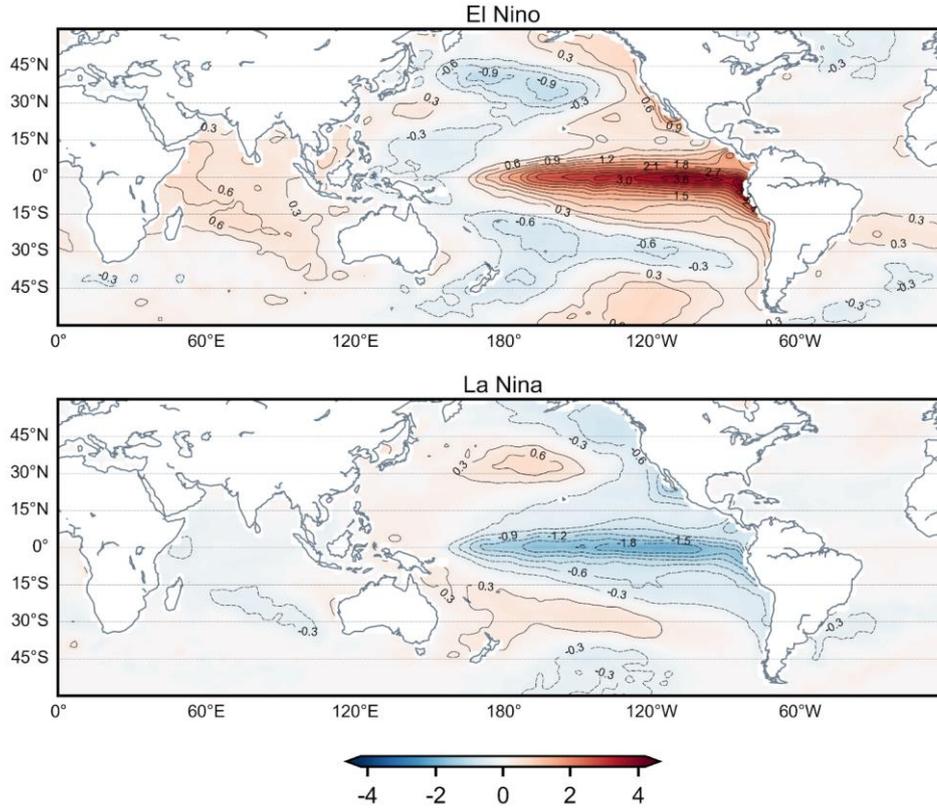
124 **2 Methods**

125 **2.1 Experimental setup**

126 Butchart et al. (2018) established a set of simplified modeling experiments for Phase 1 of the QBOi. Their Experiment 2, the
127 “present-day time slice” simulation, forms the basis for these perpetual ENSO simulations. It was designed to allow for an
128 evaluation of the accuracy of modeled QBOs under present-day conditions, that is, how the model QBOs operate in a climate
129 forced with fixed repeating annual cycles of global sea surface temperature (SSTs), sea ice concentration (SIC), and external
130 forcings representative of the time averaged 1988-2007 state.

131
132 The perpetual ENSO runs analyzed here are equivalent to Experiment 2, but with global El Niño or La Niña SST anomalies
133 superimposed on top of the climatological SST forcing. An assessment of the MJO is not conducted for Experiment 2 because
134 essential variables such as daily horizontal winds, outgoing longwave radiation (OLR), and precipitation, were not archived.
135 In creating the composite El Niño or La Niña forcings, the characterization of ENSO follows the Japan Meteorological Agency
136 (JMA) convention, where ENSO is defined by the spatially averaged NINO.3 (5°S-5°N,150°W-90°W) monthly SST anomalies
137 from 1950-2016. Anomalies are defined as deviations from the climatological seasonal cycle and computed relative to the
138 most recent sliding 30-year period of JMA COBE-SST version 1 data (JMA, 2006). The anomalies are smoothed using a five-
139 month running mean and the periods during which the anomalies exceed 0.5°C (-0.5°C) for at least six consecutive months
140 are labeled as El Niño (La Niña) periods. However, after averaging the SST anomalies for all El Niño Januarys, Februarys,
141 etc., and doing the same for La Niña, the composite average annual cycles of El Niño and La Niña SSTs show only modest
142 amplitudes (e.g., 1.92 °C for El Niño Januarys). To amplify the atmospheric response to ENSO in the simulations and increase

143 the signal-to-noise ratio, the annual cycles are multiplied by 1.8 and 1.4, respectively, making their amplitudes comparable to
144 the strongest observed ENSO events. A similar scaling is applied to the corresponding global signatures in NINO.3 SST
145 anomalies (Fig. 1), which are superimposed on 1988–2007 climatological SSTs and prescribed in the models. Note that this
146 procedure does not completely capture the development, mature phase, and decay of all observed El Niño events, due to
147 diversity in the evolutions of events. We refer the reader to Kawatani et al. (2025) for more details on the experimental design
148 and sensitivity of the model QBOs.



149
150 **Figure 1: November-April composites of the El Niño (top) and La Niña (bottom) JMA COBE SST anomalies (units K) that are**
151 **prescribed in the perpetual ENSO simulations.**

152 In addition to the prominent El Niño and La Niña signals, the November-April (NDJFMA) SSTs shown in Figure 1 include
153 the signatures of the basin-scale Interdecadal Pacific Oscillation (IPO) (Henley et al. 2015) and Indian Ocean SSTs that are in
154 phase with ENSO. In some regions like the tropical Pacific and Indian Oceans, the amplitude of the global SSTs associated
155 with the El Niño are roughly double that of La Niña (Fig. 1).

156

157 2.2 Models

158

159 *Table 1: The models used in this study, the horizontal (given as spectral triangular truncation, T, or longitude-latitude increments) and*
 160 *vertical configuration (number of levels and model top, in hPa), the number of years per simulation, and relevant literature for the convective*
 161 *schemes. Only one realization is used from each model.*

Model	Horizontal resolution	Vertical levels (top, hPa)	Number of years	Convective parameterizations
EC-EARTH3.3	T255	91 (0.01)	101	Bechtold (2014)
ECHAM5sh	T63	95 (0.01)	40	Tiedtke (1989), Nordeng (1994)
EMAC	T42	90(0.01)	106	Tiedtke (1989)
LMDz6	2°-1.25°	79 (0.015)	80	Emanuel (1991), Hourdin et al. (2013)
GISS-E2-2-G	2°-2.5°	102 (0.002)	30	Rind et al. (2020); Kelley et al. (2020)
MIROC-AGCM-LL	T106	72 (1.2)	100	Pan and Randall (1998); Emori et al. (2001)
MIROC-ESM	T42	80 (0.0036)	100	Pan and Randall (1998); Emori et al. (2001)
MRI-ESM2.0	T159	80 (0.01)	50	Yukimoto et al. (2019)
CESM1(WACCM5-110L)	1.25°-0.94°	110 (6.1E-6)	101	Zhang and McFarlane (1995)

162
 163 The models considered are listed in Table 1 along with the number of years analyzed for each model and references on each
 164 model’s convective parameterization. These parameterizations impact the representation of tropical phenomena (Holt et al.
 165 2020; Kawatani et al., 2025), including the simulation of intraseasonal oscillations (Ham and Hong, 2013). For example, past
 166 sensitivity tests with the version of MIROC-ESM that we use here have shown that its cumulus parameterization struggles to
 167 simulate an MJO of realistic amplitude with capability to propagate over the Maritime Continent (Miura et al. 2012). The
 168 updated scheme (Chikira and Sugiyama 2010) in use in newer versions of the model, MIROC6, has helped ameliorate these
 169 issues (Ahn et al. 2017; 2020). The importance of simulated convection-circulation coupling has been identified for other
 170 models (Kim et al. 2014; Zhu et al. 2020; Wang et al. 2022).

171 **2.3 Observation-based reference data**

172 To be consistent with previous studies (e.g., Wei and Ren 2019), the ENSO-MJO relationship is considered during
173 November-April. The six historical La Niña years, where the year is associated with November, are 1970, 1984, 1988, 2017,
174 2020, and 2021 and the eight El Niño years are 1968, 1982, 1986, 1991, 1997, 2009, 2015, 2018. Each corresponds to an
175 instance when the smoothed Niño3 anomalies exceed $\pm 0.5^{\circ}\text{C}$ for at least six consecutive months. For comparison with the
176 models, the subsequent analyses include “observed” El Niño and La Niña composites, formed by averaged deseasonalized
177 1959-2022 ERA5 reanalysis (Hersbach et al. 2020) over the years. Unlike the model simulations, which prescribe ENSO
178 SST anomalies scaled by factors of 1.8 (El Niño) and 1.4 (La Niña), the ERA5 composites reflect unscaled, event-based
179 variability. While this distinction makes direct comparison between models and reanalysis imperfect, ERA5 nonetheless
180 provides a useful benchmark for expected atmospheric responses. The ERA5 El Niño and La Niña composites are best
181 interpreted in relation to each other, rather than as strict analogs to the simulations.

182 Satellite-based daily precipitation estimates from a 1996-2023 Global Precipitation Climatology Project (GPCP, v1.3) record
183 (Huffman et al. 2001; Adler et al. 2017) are used to assess ENSO-related precipitation anomalies, as is NOAA Interpolated
184 Outgoing Longwave Radiation (Liebmann and Smith 1996). For these analyses, El Niño and La Niña years are defined as
185 ± 1 variations in standardized November-April Niño3 anomalies, yielding 1997, 2009, and 2015, and La Niña years as
186 1999, 2007, and 2010. Unlike ERA5, where cloud fields and precipitation are parameterized without direct assimilation,
187 satellite-based datasets provide a more direct observational reference, particularly valuable in the tropics where reanalysis
188 precipitation is known to be less reliable (Gehne et al. 2016).

189 **2.4 MJO analyses**

190 We implement a number of widely-used methods to evaluate the MJO in the perpetual ENSO simulations. In the interest of
191 exploring changes to MJO lifetime by ENSO phase as well as visualizing the MJO’s vertical structure, we compute Real-time
192 Multivariate MJO indices (RMMs) for each perpetual ENSO simulation using a similar methodology to Wheeler and Hendon
193 (2004, WH04). The RMMs are derived from a combined empirical orthogonal function (EOF) analysis of tropically averaged
194 (15°S - 15°N) anomalous daily outgoing longwave radiation (OLR), 200-hPa zonal wind (U200), and 850-hPa zonal wind
195 (U850). As in WH04, we deseasonalize, remove interannual variability, and normalize the anomalies by their global variance.
196 To enable a fairer comparison between the models and reanalysis, we project the anomalous model fields onto 1959-2022
197 ERA5 WH04 EOFs, which are computed by using ERA5 winds and ERA5 OLR; projecting onto each model’s respective
198 ENSO simulation instead does not change the conclusions and using NOAA OLR in the EOFs has minimal impact on model
199 the RMMs (not shown). Daily OLR and U200 were not available for two models (GISS-E2-2-G and LMDz6) so their RMMs
200 are computed using U250 and U850.

201

202 The number of MJO events within a given data set is tallied like in Pohl and Matthews (2007) by counting the number of times
203 the MJO makes a complete rotation through its RMM1 and RMM2 phase space. Average lifetime and MJO amplitudes
204 ($\sqrt{(RMM1^2) + (RMM2^2)}$) are computed across events. To visualize the MJO’s vertical structure, latitudinally averaged
205 10°S-10°N longitude-pressure cross-section of zonal wind and temperature are projected onto the RMMs using the same steps
206 as Hendon and Abhik (2018), but applied across ENSO years in the present study rather than QBO years.

207
208 The MJO is visualized using two related wavenumber–frequency approaches. Both Fourier transform deseasonalized and
209 detrended, highpass filtered, tapered time-longitude data, reorder coefficients into eastward/westward components, and
210 compute power spectra to isolate intraseasonal variability. The CLIVAR/MJO metric (Waliser et al. 2009) applies this to
211 centered November-April segments of tropical U850, yielding single variable power for eastward and westward disturbances
212 (Figure 3). The Wheeler and Kiladis (1999) style analysis is similar, but designed to diagnose the broader family of
213 convectively coupled equatorial waves alongside the MJO: we linearly detrend multiyear daily fields, highpass filter at 96 days,
214 use overlapping 96-day segments (65-day overlap), perform successive longitude–time transforms, recover symmetric and
215 antisymmetric spectra about the equator, and normalize by a smoothed background; results are presented as the ratio of raw
216 symmetric daily mean precipitation power to background (Figure 2). Our Wheeler-Kiladis implementation also subsets data
217 to November-April. For model evaluation, spectra are divided by each model’s perpetual ENSO background, whereas GPCP
218 (ERA5) El Niño and La Niña spectra are compared to a common 1996-2022 (1959–2022) background.

219
220 We also implement an MJO diversity analysis in which MJO events are classified into distinct types based on their propagation
221 characteristics using k-means clustering (Wang et al. 2019). Each MJO event is binned as one of four archetypes, “standing”
222 or “jumping” MJOs, which propagate across the Indian Ocean, but are distinguished by reemergence of the MJO over the
223 western Pacific during jumping events, and “slow” or “fast” MJOs, which both continuously propagate across the Maritime
224 Continent, but at different speeds. An MJO event occurs when the 20-70 day bandpass-filtered OLR anomalies (from seasonal
225 cycle) averaged over the equatorial Indian Ocean (10°S-10°N, 75°E-95°E) are smaller than negative one standard deviation
226 for five successive days; the reference day (day 0) is the day of minimum OLR. The MJO events are categorized by a k-means
227 clustering of the enhanced convective signal (OLR anomalies under -5 Wm^{-2}) of the latitudinally averaged 10°S to 10°N time-
228 longitude OLR anomalies taken over 60°E to 180°E and over a 31-day period from day -10 to day 20. For brevity, we omit
229 further diversity analysis methodological details, for which we refer the reader to Back et al. (2024) for all steps. Unlike Wang
230 et al. (2019), in which initial centroids for clustering are randomly chosen, initial centroids for model MJO events are set to
231 those of the four observation-based clusters, allowing the present study to evaluate how well climate models reproduce
232 observed MJO diversity with minimal subjectivity.

233 2.5 QBO and analyses

234 The space-time form of the QBOs varies from model to model as each is generated with different amounts of forcing from
235 resolved waves and parameterized non-orographic gravity wave drag. Properties of these models that are particularly relevant
236 for simulating the QBOs are listed in Butchart et al. (2018), details on QBO morphology (e.g., its amplitude, latitudinal width)
237 given the observed SST record are presented in Bushell et al. (2022), and the relative contribution of resolved and
238 parameterized tropical waves to forcing the QBO is analyzed in detail in Holt et al. (2020). Of note, MIROC-AGCM-LL's
239 QBO is forced solely by resolved waves. As EC-EARTH and GISS-E2-2-G did not contribute to some of the earliest QBO
240 analyses, relevant details on their internal QBOs can be found in Serva et al. (2024) and Rind et al. (2014, 2020), respectively.
241 For a thorough analysis of how the QBO responds to the perpetual ENSO simulations, we refer the reader to Kawatani et al.
242 (2025).

243
244 To help clarify the ability of the QBOs to interact with the MJOs, we use established metrics to characterize the morphology
245 of the ERA5 and model QBOs. The main field used to document QBO morphology is the monthly zonal-mean zonal wind.
246 "QBO cycles" (consecutive easterly/westerly phases) are identified by marking the first month when the deseasonalized and
247 smoothed (5-month running mean) 20 hPa 5°S-5°N wind changes from westerly to easterly, ending one month before the next
248 transition at 20 hPa (Kawatani et al. 2019). From these cycles, we calculate average QBO easterly, westerly, and total
249 amplitudes using the QBO "transition time" methodology of Richter et al. (2020). The easterly (westerly) amplitude is equal
250 to the average of the minimum (maximum) monthly QBO winds from each QBO cycle. The QBO cycles are used further to
251 calculate minimum, mean, and maximum QBO periodicity statistics. These statistics are a key result of Kawatani et al. (2025)
252 and are discussed thoroughly there. In short, the periodicity of the QBO decreases in all El Niño simulations and increases in
253 all La Niña simulations, which is attributed to ENSO modulating convection and the low-frequency circulation, thereby
254 influencing generation of tropical waves and their filtering by the large-scale circulation and the QBO. For the purposes of the
255 present study, the minimum and maximum periodicities are required to evaluate the QBO's spatial structure, defined as the
256 latitude-pressure cross sections of each data set's QBO Fourier amplitude. These are made by applying a discrete Fourier
257 transform in time to the multi-year monthly zonal-mean zonal wind at each pressure-latitude grid point and dividing the sum
258 of squares of the amplitudes of the harmonics corresponding to periods between the minimum and maximum QBO periods by
259 the sum of squares of the amplitudes of all harmonics. This ratio is subsequently multiplied by the standard deviation of the
260 zonal-mean zonal wind (Pascoe et al. 2005). Using this QBO Fourier amplitude, the lowest altitude the QBO reaches, its
261 vertical extent (i.e., how tall it is), and its latitudinal extent are defined as in Schenzinger et al. (2017), except that here the
262 QBO's maximum amplitude is assumed to be at 20 hPa for all models and ERA5.

263
264 QBO impact on the MJO is assessed using the techniques of Klotzbach et al. (2019) and Kim et al. (2020). Following the prior,
265 we make scatterplots of December-February warm-pool (10°S-10°N, 45°E-180°E) averaged tropopause stability (100 hPa

266 minus 200 hPa temperature) versus December-February MJO amplitude as a function of QBO phase (sign of DJF averaged
267 5°S-5°N 50 hPa zonal mean zonal-wind) for each of the simulations. MJO amplitude is expected to increase as tropopause
268 stability decreases, which happens during the easterly QBO phase. As in Kim et al. (2020), MJO activity is also computed as
269 a function of QBO phase. Specifically, MJO-filtered OLR is calculated following Wheeler and Kiladis (1999) with one
270 exception, the full time series is detrended rather than using 96-day overlapping segments. To minimize spectral leakage, 5%
271 of the data are tapered to zero at the ends of the timeseries. After tapering, a complex Fourier Transform is performed, and the
272 spectral wavenumber-frequency data are filtered to retain only the eastward propagating coefficients for 20-100 day periods
273 and wavenumbers 1-5. MJO activity is then defined as the standard deviation of the MJO-filtered OLR across all December-
274 February days that fall into a particular category, for instance all years, easterly QBO years or westerly QBO years. For this
275 analysis, easterly and westerly QBO years are defined as those which exceed +/- 0.5 standard deviation of the 50 hPa monthly
276 zonal wind anomalies, seasonally smoothed and averaged over 10°S to 10°N. We allow the QBO to be defined differently
277 between the Klotzbach and Kim et al. analyses; here we prioritize using the aforementioned metrics in their original form
278 rather than using customizing them.

279 **3 Results**

280 **3.1 Impact of perpetual ENSO conditions on the MJO**

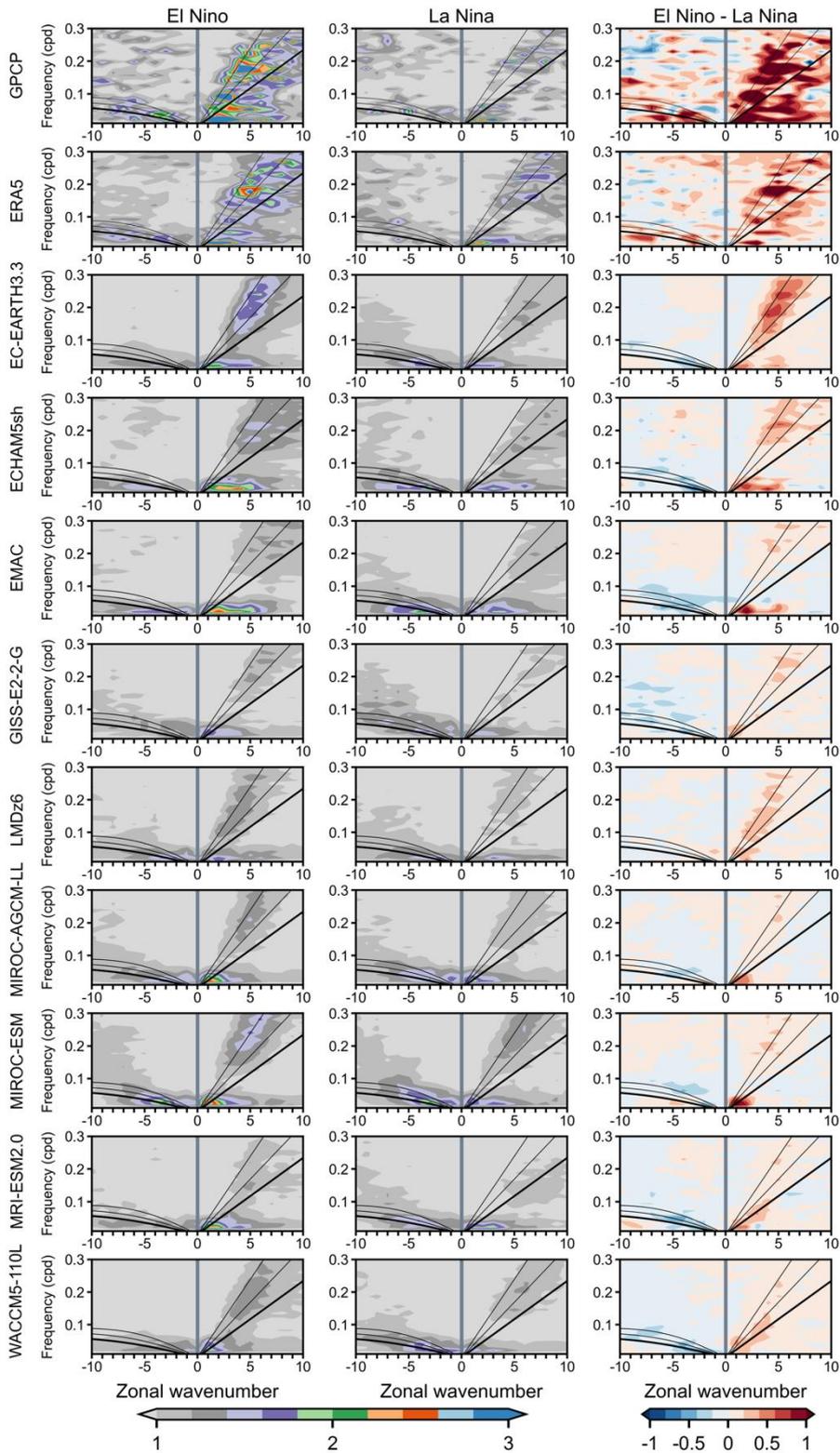
281 **3.1.1 Role of convectively coupled waves**

282 Before examining the influence of ENSO on MJO, we evaluate some of the other large-scale tropical phenomena that the
283 models simulate. Convectively coupled waves are relevant because they comprise the space-time structure of the MJO and can
284 influence its propagation by modulating the tropical circulation and the distribution of moisture that the MJO encounters
285 (Kiladis et al. 2009; Wang et al. 2019; Wei and Ren 2019; Berrington et al. 2022; Wang and Li 2022). Aspects of the waves,
286 such as their phase speed and what longitudinal domain they propagate in, vary depending on the seasonal-mean circulation
287 and convective activity (Roundy 2012; Yang et al. 2023). Hence, the amplification of the Walker Circulation by La Niña and
288 the weakening of it by El Niño (Fig. S1) provide a pathway for the perpetual ENSO forcings to modulate the waves and perhaps
289 the MJO. Applying similar methods to Wheeler and Kiladis (1999), we visualize the waves by computing the spectral power
290 of GPCP, ERA5, and model daily-averaged precipitation as a function of wavenumber and frequency. Figure 2 shows the
291 precipitation spectra for phenomena symmetric about the equator, taken over November-April and 15°S-15°N. Three
292 dispersion curves, as in Matsuno (1966), corresponding to equivalent depths of 10, 25, and 50 meters are also superimposed;
293 these curves are derived using the dispersion relations for equatorially trapped waves and they are co-located with modes of
294 organized convection, with larger equivalent depths corresponding to faster phase speeds.

295
296 The GPCP panels in row one draw on only three La Niña and three El Niño years, while the ERA5 panels in row two use just
297 six La Niña and eight El Niño years, resulting in signals that are noisier than the models' multi-decade averages. Power is

298 typically the highest amplitude in GPCP, then ERA5, and weakest in the models. Spectral signals associated with the eastward
299 propagating Kelvin wave move up and to the right on each panel, spanning sub-planetary low frequency ($\sim k = 3, 25$ days)
300 scales to synoptic ($k = 4+$) sub-weekly scales. Relative to GPCP and ERA5, the models underestimate the strength of the
301 Kelvin wave irrespective of the type of ENSO forcing, but have worse difficulties in the El Niño simulation. Power associated
302 with the westward propagating equatorial Rossby wave is evident on the left side of each panel between wavenumbers 1-10
303 and timescales of 10 days to five weeks. Overall, the models do a reasonable job of simulating the spectral amplitude of
304 equatorial Rossby waves, although it is too strong for some models (EMAC, MIROC-ESM), especially in their La Niña
305 simulations.

306



308 **Figure 2: Wavenumber-frequency spectrum of the symmetric component of 15°S-15°N November-April precipitation plotted as the**
309 **ratio between raw symmetric precipitation and a smoothed red noise background spectrum. The eastward (right) side of the**
310 **spectrum includes three Kelvin wave dispersion curves in black, of which the thickest curve corresponds to the equivalent depth of**
311 **12 meters, and the others to 25 and 50 meters, respectively. Similar dispersion curve plotting conventions are used on the westward**
312 **(left) side of the spectrum where the curves overlay the equatorial Rossby wave power. Column one corresponds to El Niño, column**
313 **two to La Niña, and column three to their difference, which is computed as (El Niño symmetric/El Niño background) minus (La**
314 **Niña symmetric/La Niña background). Computing the third column as (El Niño symmetric) minus (La Niña symmetric) yields**
315 **similar conclusions (not shown).**
316

317 **3.1.2 ENSO effects on convectively coupled waves**

318 The effect of ENSO phase on each wave is revealed by the rightmost column of Figure 2, which shows El Niño (column one)
319 minus La Niña (column two) differences, where red means stronger power during El Niño and blue means larger power during
320 La Niña. All models simulate stronger Kelvin waves in their El Niño simulation, particularly along the deeper equivalent depth
321 ($n = 25, 50$ meters) dispersion curves. This implies faster Kelvin wave phase speeds during El Niño. Examining the El Niño
322 column, the alignment of the Kelvin wave power along these particular curves is demonstrated by EC-EARTH3.3, GISS-E2-
323 2-G, LMDz6, MIROC-AGCM-LL, MIROC-ESM, MRI-ESM2.0 and (CESM1) WACCM5-110L, hereafter just “WACCM5-
324 110L.” The remaining models, ECHAM5sh and EMAC (both ECHAM-based models), differ in that their El Niño Kelvin wave
325 power is weighted towards higher zonal wavenumbers for frequencies below 0.2 cpd. GPCP and ERA5 shows large sporadic
326 increases in Kelvin wave power along deeper equivalent depth dispersion curves during El Niño compared to La Niña. It is
327 suggested that the Kelvin wave intensifies during observed El Niños because SST-driven increases in equatorial convection,
328 combined with a weakened upper-tropospheric westerly duct over the eastern Pacific, enhance the convective and dynamical
329 coupling that supports Kelvin wave growth. La Niña produces the reverse effect (Yang and Hoskins 2013; Yang et al. 2023).

330
331 In contrast, for the westward propagating equatorial Rossby wave, GPCP and ERA5 indicate stronger power during El Niño,
332 whereas all models simulate stronger Rossby wave power during La Niña. This discontinuity arises with models having weaker
333 El Niño Rossby wave spectral power than the GPCP and ERA5 composites (not shown), but stronger power given the La Niña
334 forcing. It should be noted that equatorial Rossbywaves exhibit strong regional variations in observations, varying with the
335 strength and position of seasonal-mean tropical westerlies and in response to extratropical Rossby waves entering low latitudes
336 (Yang and Hoskins 2016) factors we do not examine here. Such regional impacts may help explain why the GPCP and ERA5
337 composites differ from the model-simulated ENSO responses. Our observed composites show some consistency with Yang et
338 al. (2023) who found stronger equatorial Rossby waves during El Niño when the satellite-based OLR spectra were restricted
339 to the western hemisphere (their Fig. 13).

340
341 The ENSO-related differences in convectively coupled waves highlighted above motivate investigation into whether the MJO,
342 as represented in the models, also varies systematically by ENSO phase. Observational and modeling studies report that ENSO
343 modulates MJO propagation, amplitude, and regional impact, with notable differences in MJO structure between El Niño and

344 La Niña (Wei and Ren 2019; Fernandes and Grimm 2023). This raises the question of whether the model MJOs analyzed here
345 capture these ENSO-dependent features, a topic addressed in the next section.

346

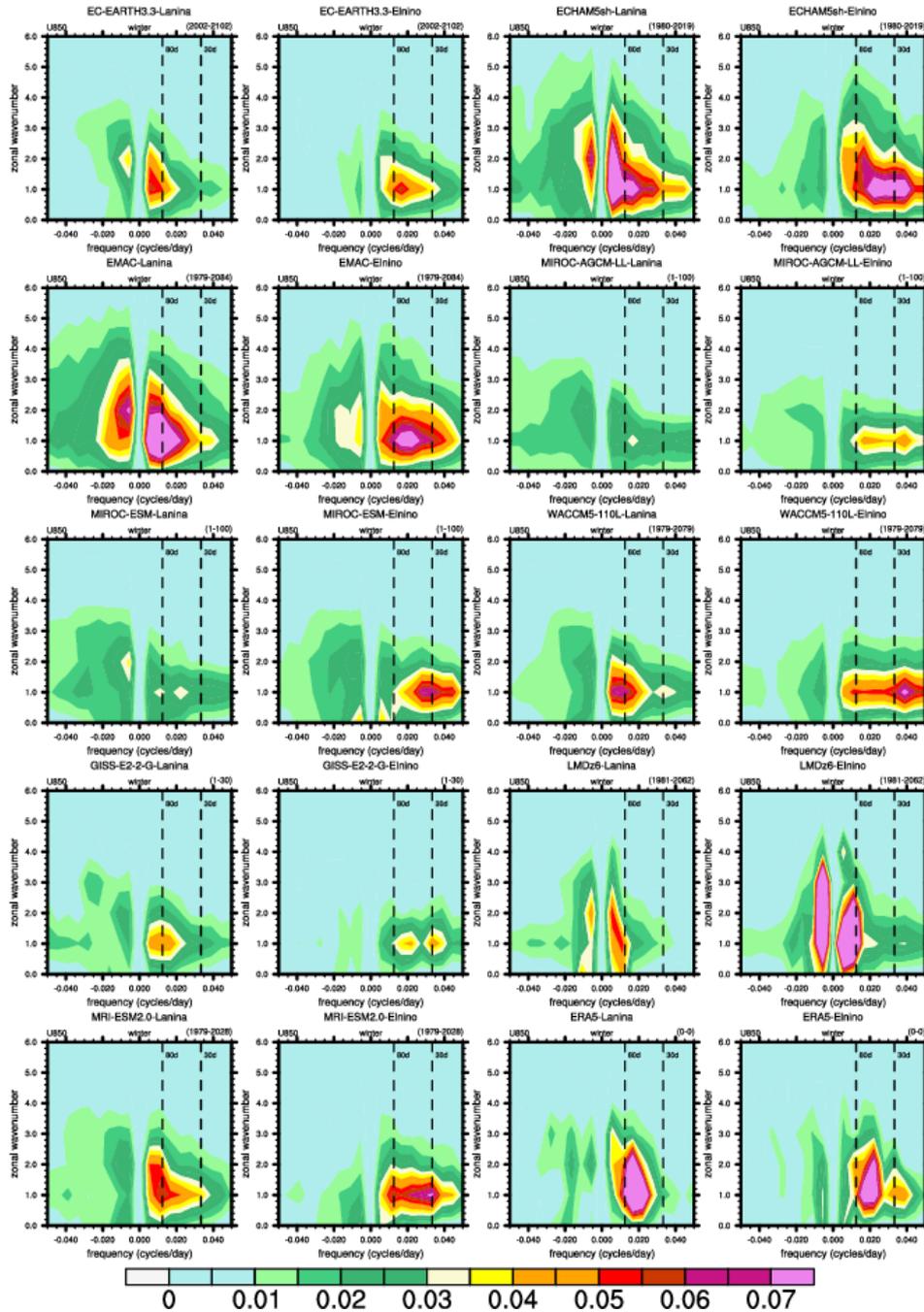
347 **3.1.3 MJO representation and ENSO dependence**

348 Having found changes in the convectively coupled waves due to ENSO, perhaps the model MJOs also behave differently given
349 ENSO state. Broadly speaking, Figure 2 shows that the models include MJOs as indicated by the maxima in spectral power at
350 intraseasonal timescales ($\ll 0.1$ cpd) between eastward propagating wavenumbers 1-5. Holt et al. (2020) also found MJOs to
351 be simulated by these models in historical AMIP simulations (QBOi Exp 1). The highest MJO power in GPCP is concentrated
352 between wavenumbers 1 and 3 irrespective of ENSO phase and exceeds that of ERA5 and the models. The ENSO simulations
353 for EC-EARTH3.3, MIROC-AGCM-LL, MIROC-ESM, MRI-ESM2.0 correctly position MJO power between wavenumbers
354 1-3 whereas ECHAM5sh, EMAC (sharing convective parameterizations) and LMDz6 exhibit spectral power that is incorrectly
355 shifted towards higher wavenumbers. The amplitude of the MJO is comparatively small in GISS-E2-2-G, LMDz6, and
356 WACCM5-110L, which may adversely affect their stratospheric waves, since these three models adopted interactive GW
357 sources. The El Niño minus La Niña differences in the rightmost column of Figure 2 show that MJO spectral power is stronger
358 in the presence of the El Niño basic state. Differing from the other models, EC-EARTH3.3, ECHAM5sh, and EMAC have
359 fairly large El Niño minus La Niña MJO power differences at wavenumber 4 and 5.

360

361 Irrespective of ENSO phase, the amplitude of the model MJOs as shown in Fig. 2 is systematically weaker than in GPCP or
362 ERA5. This may have something to do with dividing each simulation's symmetric power by its respective background power,
363 the latter of which is contaminated, in a sense, by the perpetual ENSO conditions. Note that recomputing the third column of
364 Fig. 2 without dividing each El Niño and La Niña composite by their respective background does not change our conclusions
365 (not shown). To get around this potential issue with the background power and further inspect the model MJOs as opposed to
366 the convectively coupled waves, in Figures 3 we consider the westward and eastward wavenumber-frequency spectra of U850
367 taken over the intraseasonal timescale and over MJO-like zonal wavenumber scales. These analyses yield more holistic views
368 of the MJO than in Fig. 2 because they incorporate the MJO's signals in these fields that are both symmetric and antisymmetric
369 about the equator.

370

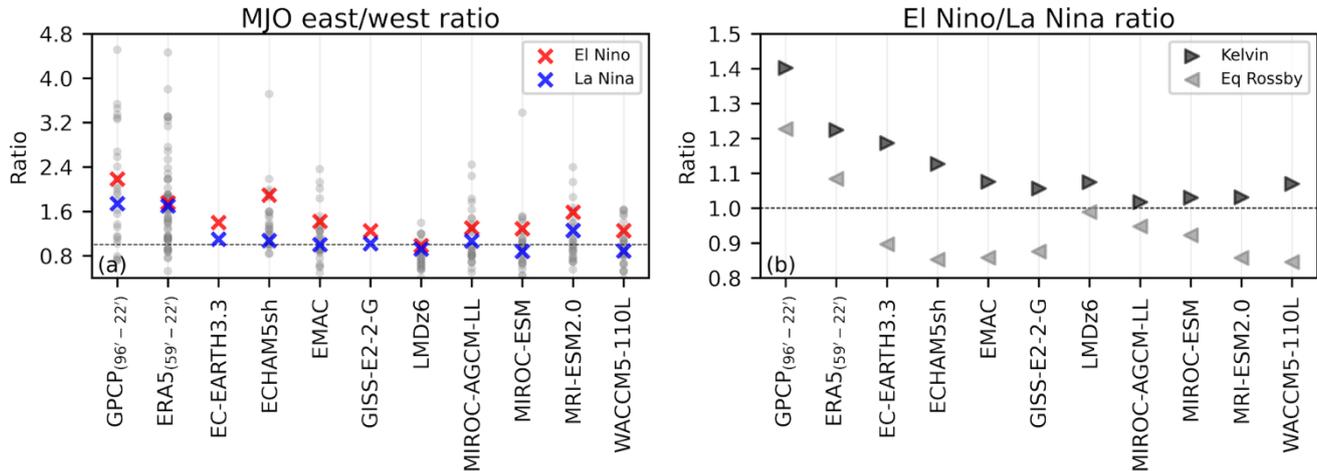


371
 372 **Figure 3: November-April wavenumber-frequency power spectra of 10°S-10°N averaged U850. Units of the U850 spectrum are**
 373 **m²/s² per frequency interval per wavenumber interval. We interpret the power at ~ 180-day frequencies as a by-product of using**
 374 **centered November-April segments.**

375
 376 As shown by ERA5, the MJO timescale variance of dynamical fields such as U850 are known to have a narrow spectral peak
 377 around zonal wavenumber-1 (Hendon and Salby 1994). The models are fairly good at reproducing this (Fig. 3), despite large

378 inter-model differences in terms of peak amplitudes. Independent of the ENSO phase, ERA5’s wavenumber-frequency shows
 379 the highest power between frequencies of 30-80 days (vertical dashes), a so-called “MJO band” (Ahn et al. 2017). Considering
 380 only the ERA5 MJO band, the MJO is stronger given lower zonal wavenumbers and longer periodicities during La Niña
 381 compared to El Niño. Similarly, the model spectra broadly indicate that the periodicity of the MJO decreases during El Niño
 382 and increases during La Niña. However, as was the case for the symmetric precipitation spectra (Fig. 2), there are notable
 383 differences in the amplitude of the spectral power between models and reanalysis. Of the models, MJO band amplitude in
 384 ECHAM5sh and EMAC is most like ERA5, whereas other models, particularly MIROC-AGCM-LL, MIROC-ESM, and
 385 GISS-E2-2-G significantly underestimate the strength of the MJO in this metric. Another issue is that models exhibit too large
 386 of an MJO signal for zonal wavenumbers three and up, which is unrealistic and common amongst GCMs (Ahn et al. 2017),
 387 with exaggerated high-wavenumber amplitudes in ECHAM5sh, EMAC, and LMDz6, as previously noted for precipitation.
 388 The clustering of models related to each other, such as MIROC and ECHAM-based models, affirm that the tropospheric wind
 389 response to imposed ENSO anomalies is strongly influenced by model configuration and convective schemes.

390
 391



392
 393

Figure 4: (a) Ratio of eastward to westward MJO precipitation spectral power (from Figure 2) filtered for wavenumbers +/- 1 to 3 and 30 to 96 day frequencies. Red and blue X’s denote El Niño and La Niña, respectively, using the same notation for both the GPCP and ERA5 composite averages and the climatological values from the perpetual El Niño and La Niña simulations. Gray dots indicate interannual east/west ratios from GPCP, ERA5, and the QBOi Experiment 1 simulations (1979-2009 AMIP) by these same models; EC-EARTH3.3 and GISS-E2-2G did not run Experiment 1. (b) Ratio of Kelvin wave and equatorial Rossby wave power in El Niño to that in La Niña. Each wave is filtered using equivalent depths of 8 to 90 meters, with wavenumbers of 1 to 14 for the Kelvin wave and -1 to -10 for the equatorial Rossby wave.

400
 401
 402
 403
 404

To further explore the MJO timescale changes suggested by Figure 3, Figure 4a shows the ratio of eastward to westward MJO spectral power in symmetric precipitation from GPCP, ERA5, and the models. The E/W ratio measures how robust MJO eastward propagation is (Jiang et al. 2015; Ahn et al. 2017). El Niño and La Niña responses in GPCP and ERA5 (red and blue X’s) are derived from composites of previously defined strong ENSO seasons, while gray dots represent interannual E/W

405 ratios across all years (GPCP: 1996–2022; ERA5: 1959–2022), providing a baseline of internal variability. Both data sets show
406 modestly higher E/W ratios during El Niño than La Niña.

407
408 It is worth iterating that the models’ ratios are derived from the climatological means of the perpetual ENSO simulations, in
409 contrast with the event-based composites of GPCP and ERA5. Model gray dots are interannual ratios from QBOi Experiment
410 1 AMIP 1979-2009 simulations, which contextualizes internal variability. E/W ratios are higher in El Niño simulations,
411 shifting towards the upper tail of the AMIP-based sampling variability. Conversely, many of La Niña E/W ratios are around
412 one (dashed line), indicating subdued MJO propagation. The response to ENSO is relatively small in models with a weak
413 MJO, such as GISS, LMDz6 and MIROC-AGCM-LL.

414
415 Building on the observed ENSO-related changes in MJO propagation, Figure 4b further quantifies how convectively coupled
416 equatorial waves, Kelvin and equatorial Rossby, respond to ENSO forcing, given their known links to MJO variability. Figure
417 4b shows the ratio of El Niño to La Niña precipitation-based spectral power, filtered from Figure 2 for each wave type. GPCP
418 and ERA5 composites indicate that both waves are stronger during El Niño than La Niña. Models generally reproduce stronger
419 Kelvin waves in their El Niño simulations. However, equatorial Rossby waves are consistently stronger in La Niña simulations,
420 in contrast to GPCP and ERA5. This disagreement between observational products and models is difficult to attribute given
421 the differences in how spectra for each are computed and known biases in modeled precipitation spectra (Holt et al. 2020,
422 Experiment 1). This discrepancy arises with model equatorial Rossby waves being weaker in the El Niño simulation than in
423 GPCP and ERA5, but stronger than the observational benchmarks in the La Niña simulation (not shown).

424
425 The modulation of Kelvin and equatorial Rossby wave activity by ENSO, as evidenced in the historical records and models,
426 is expected to impact the MJO’s propagation speed and spatial structure (Wang et al. 2019). Since the balance between these
427 waves impacts the MJO’s eastward progression, it is helpful to assess how such dynamical changes are reflected in the
428 statistical properties of MJO events, specifically their lifetime and diversity, across different ENSO backgrounds.

429
430 *Table 2: The number of MJO events, their mean lifetimes and standard errors (reported in parentheses), and their mean amplitudes given*
431 *either perpetual El Niño or La Niña conditions in a model. An asterisk (*) next to a model name indicates that the RMMs were retrieved*
432 *using only 250 and 850 hPa zonal wind. Different from the models, for ERA5, MJO event statistics are calculated using 8 El Niño and 6 La*
433 *Niña winters subsampled from the entire 1959-2022 RMM record.*

434

	EN events (#/dec)	LN events (#/dec)	EN lifetime (days)	LN lifetime (days)	EN amplitude	LN amplitude
ERA5	25	22	38.25 (2.27)	47.07 (3.97)	1.36	1.41

EC-EARTH3.3	18.07	13.80	37.96 (1.25)	46.73 (2.01)	1.47	1.40
ECHAM5sh	26.20	15.87	28.59 (1.18)	42.94 (3.26)	1.43	1.38
EMAC	23.83	14.75	29.77 (0.88)	45.57 (1.93)	1.37	1.45
GISS-E2-2-G*	20.56	11.46	32.23 (1.87)	37.06 (3.50)	1.42	1.31
LMDz6*	11.30	9.81	37.01 (2.20)	32.47 (1.69)	1.55	1.41
MIROC-AGCM-LL	23.48	19.26	30.31 (0.78)	30.90 (1.11)	1.46	1.49
MIROC-ESM	27.77	18.15	27.96 (0.57)	29.90 (1.15)	1.38	1.34
MRI-ESM2.0	21.53	16.70	35.02 (1.39)	39.35 (1.88)	1.43	1.40
WACCM5-110L	26.00	16.87	27.13 (0.77)	36.99 (1.31)	1.40	1.49

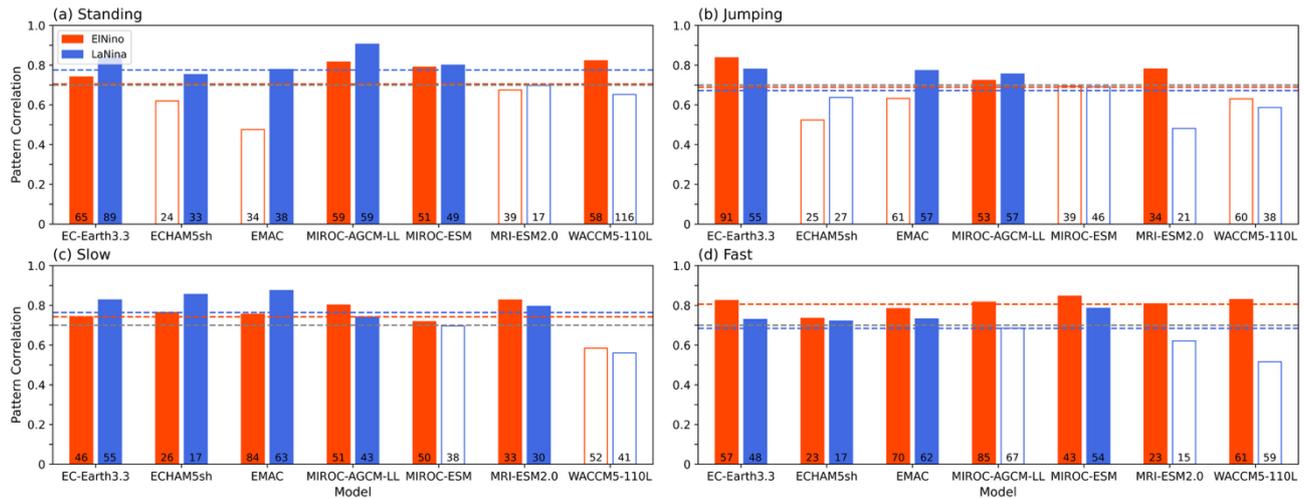
435

436 **3.1.5 MJO lifetime and event diversity**

437 Based on ERA5, La Niña events are roughly nine days longer than El Niño events on average. Wei and Ren (2019) found La
438 Niña to support both high-frequency (lifetime ~40 days) and low frequency (lifetime ~80 days) MJOs, which conceivably
439 explains the much larger ERA5 lifetime standard errors during La Niña compared with El Niño. Strikingly, the difference in
440 lifetime and its standard error between ENSO phases is nearly ubiquitous across the models. With the exception of LMDz6,
441 La Niña lifetimes are between 0.59 (MIROC-AGCM-LL) and 15.8 (EMAC) days longer than El Niño lifetimes. Models in
442 similar families, for instance ECHAM5sh and EMAC as well as MIROC-AGCM-LL and MIROC-ESM, typically have similar
443 magnitude differences in their Pohl and Matthews (2007) statistics between ENSO phases. All models simulate more MJO
444 events during El Niño, which is consistent with ERA5, however the difference in the number of events between ENSO phases
445 is generally larger in the models than in ERA5. MJO amplitude is only marginally larger during La Niña based on ERA5
446 whereas six of the nine models have larger amplitudes during El Niño.

447

448 Analyzing MJO diversity offers further insight into its propagation characteristics. k-means clustering of empirical OLR
449 Hovmöller diagrams reveals four dominant archetypes: standing, jumping, slow, and fast propagating MJOs (Wang et al.
450 2019). Associated SST composites show standing MJOs tend to coincide with La Niña, fast MJOs with El Niño, while jumping
451 and slow events show no clear ENSO linkage (Back et al. 2024). The experimental design allows us to test whether certain
452 archetypes become more robust under persistent ENSO forcing. Fast and slow events can occur during either ENSO phase
453 (Yadav and Straus 2017), suggesting some sensitivity of the established archetype-ENSO associations to internal variability.
454 Figure 5 lists the number of each archetype's events simulated by each model (at the base of each bar). Four of seven models
455 produce more standing events during La Niña and four models produce more jumping events during El Niño, results that do
456 not indicate a consistent ENSO dependence. In contrast, slow and fast MJOs are modestly more frequent in El Niño simulations
457 for six of seven models.



460
461 **Figure 5: Pattern correlations between the ERA5 and simulated time-longitude (Hovmöller) 10°S-10°N convective OLR anomaly**
462 **composites (OLR < -5 W/m²) corresponding to each of the four k-means clusters MJO archetypes defined in Wang et al. (2019).**
463 **Multi-model average El Niño and La Niña correlations are shown by red and blue dashed lines, respectively. The gray dashed line**
464 **marked at a correlation of 0.7 is a heuristic threshold (Back et al. 2024) to decide when a particular model’s MJO archetype is well**
465 **captured by a model and bars are shaded in red or blue when correlation exceeds this threshold. Number of events is printed at the**
466 **bottom of each bar.**

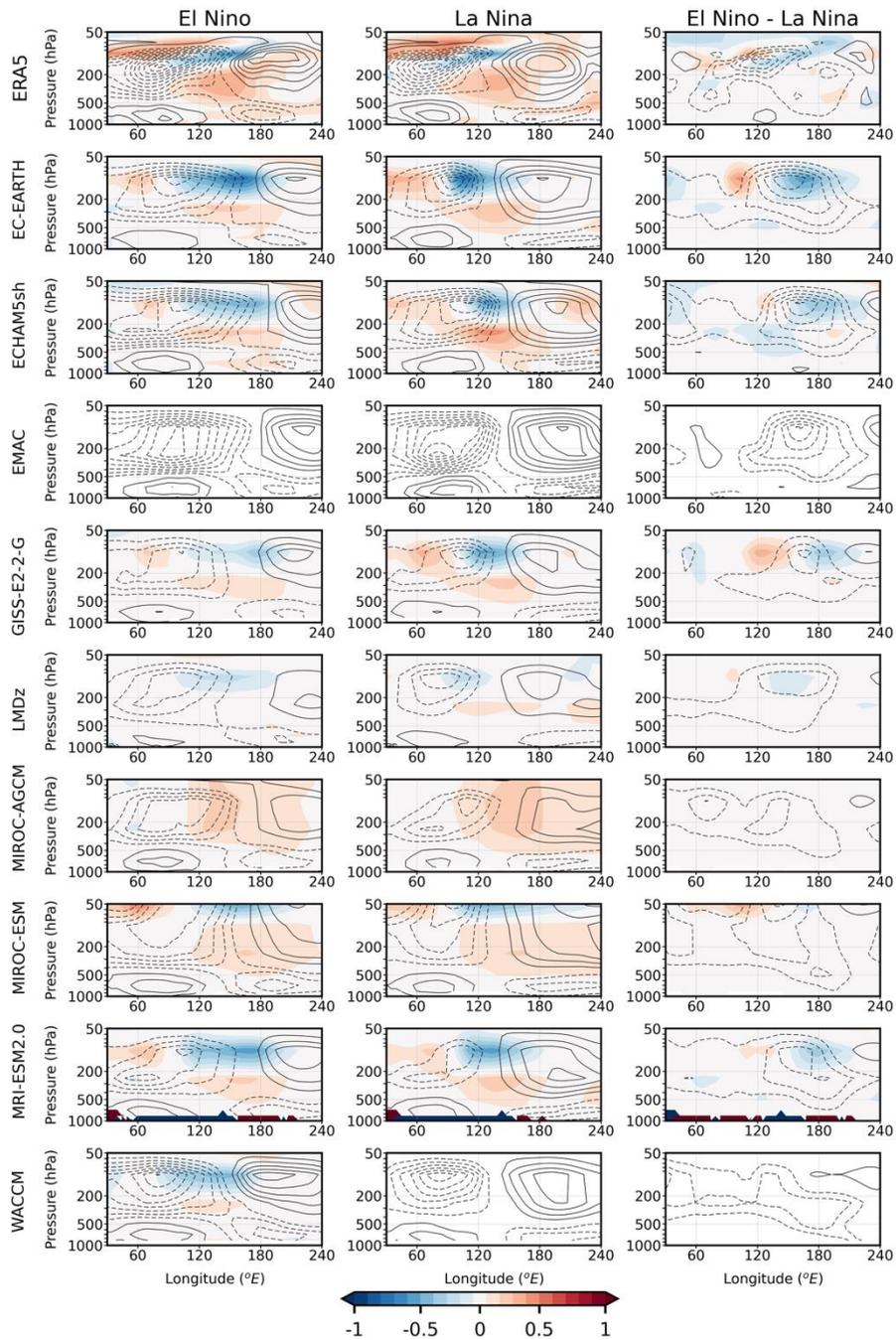
467
468 To assess how closely model archetypes resemble those in reanalysis, we calculate pattern correlations between ERA5 and
469 model time-longitude tropical OLR anomaly composites for each cluster. This evaluates how well models capture observed
470 MJO archetypes and whether representation improves under ENSO forcing. Prior multi-model studies (Back et al. 2024)
471 suggest a pattern correlation of 0.7 as a threshold for distinguishing well-simulated OLR Hovmöllers. For the standing MJO
472 (Fig. 5a), five of 14 models fall below this threshold and are shaded white. Although roughly a third of simulations poorly
473 represent the standing cluster, model skill improves during La Niña, with the multi-model mean correlation exceeding the El
474 Niño value by 0.09. For jumping MJOs (Fig. 5b), correlations fall below 0.7 in over half the models, indicating widespread
475 difficulty in simulating this archetype. As jumping MJOs are not thought to be ENSO-sensitive, they are not analyzed further.

476
477 Studies suggest slowly propagating MJOs often coincide with La Niña, though the associated SST pattern is weak and
478 statistically insignificant (Wang et al. 2019; Back et al. 2024). Similarly, model skill is slightly higher in La Niña simulations,
479 but likely indistinguishable from El Niño (Fig. 5c). In contrast, fast MJOs are better represented in every El Niño simulation,
480 with the multi-model mean correlation exceeding the La Niña mean by 0.11 (Fig. 5d). In summary, model skill in representing
481 fast and standing MJOs tends to be higher during El Niño and La Niña simulations, respectively, and models simulate a
482 modestly higher number of fast events in their El Niño simulation.

484 **3.1.6 MJO vertical structure and ENSO dependence**

485 The vertical structure of the MJO differs between slow and fast propagating events (Wang et al. 2019). To consider this
486 further, we regress the latitudinally averaged 10°S-10°N zonal wind and temperature from ERA5 and the models onto their
487 phase 3/4 RMM indices as in Hendon and Abhik (2018) and form pressure-longitude cross-sections (Figure 6). Phases 3/4,
488 when the MJO convection is over the western Maritime Continent, are of interest because ENSO modulates the low-
489 frequency circulation here through its effect on the Walker Circulation, giving it a pathway to influence MJO propagation
490 (Sun et al. 2019; Suematsu and Miura 2022). Irrespective of the ENSO phase, the MJO in ERA5 exhibits a quadrupole
491 structure in zonal wind, all of which is centered around a tropospheric warming at 140°E that peaks in amplitude near 300
492 hPa (cf. Jiang et al. 2015).

493



494

495 **Figure 6: Pressure-longitude cross-sections of the 10°S-10°N zonal wind and temperature regressed onto the Phase 3/4 RMMs as in**
 496 **Hendon and Abhik (2018). Black contours show zonal wind (intervals of +/- 0.5, 1.5, 2.5 m/s...) and temperature is shaded between**
 497 **-1 and 1 °C. EMAC is missing temperature and we omit WACCM's La Niña temperature due to a conspicuous artifact.**

498

499 Although the ERA5 El Niño and La Niña composites are similar overall, subtracting the two reveals that they differ due to the
 500 El Niño composite including a stronger Kelvin wave, evidenced by the stronger tropospheric easterlies to the east of MJO

501 convection (cf. Fig. 2). In addition, the composites reveals a dry Kelvin wave signature, its characteristic features being a cold
502 cap temperature anomaly in the UTLS, which is slightly out of phase with easterly zonal wind anomalies, all of which tilt
503 eastward with increasing height above ~200 hPa and westward with increasing height below (Straub and Kiladis 2002; Kim
504 et al. 2013; Yuni et al. 2019; Nakamura and Takayabu 2022). Judging by the longitude of the ERA5 100 hPa cold cap maximas,
505 the dry Kelvin wave embedded in the composite El Niño MJO is shifted further east compared to its La Niña equivalent.

506
507 Consistent with recent studies, the surface easterlies positioned east of the MJO convection are indeed stronger during El Niño
508 in reanalysis (Wang et al. 2019; Wei and Ren 2019). We attribute the amplification of these easterlies to the Kelvin wave’s
509 signature in wind, which better bridges the MJO lower tropospheric easterlies over the Pacific with the upper tropospheric
510 easterly outflow over the Indian Ocean; compare the 500 hPa zonal winds at 150°E between El Niño and La Niña. This
511 enhanced continuity of the MJO easterlies during El Niño is a robust feature amongst the models and is particularly clear in
512 the El Niño minus La Niña composites of EC-EARTH3.3, ECHAM5sh, MIROC-AGCM-LL, MIROC-ESM and MRI-
513 ESM2.0.

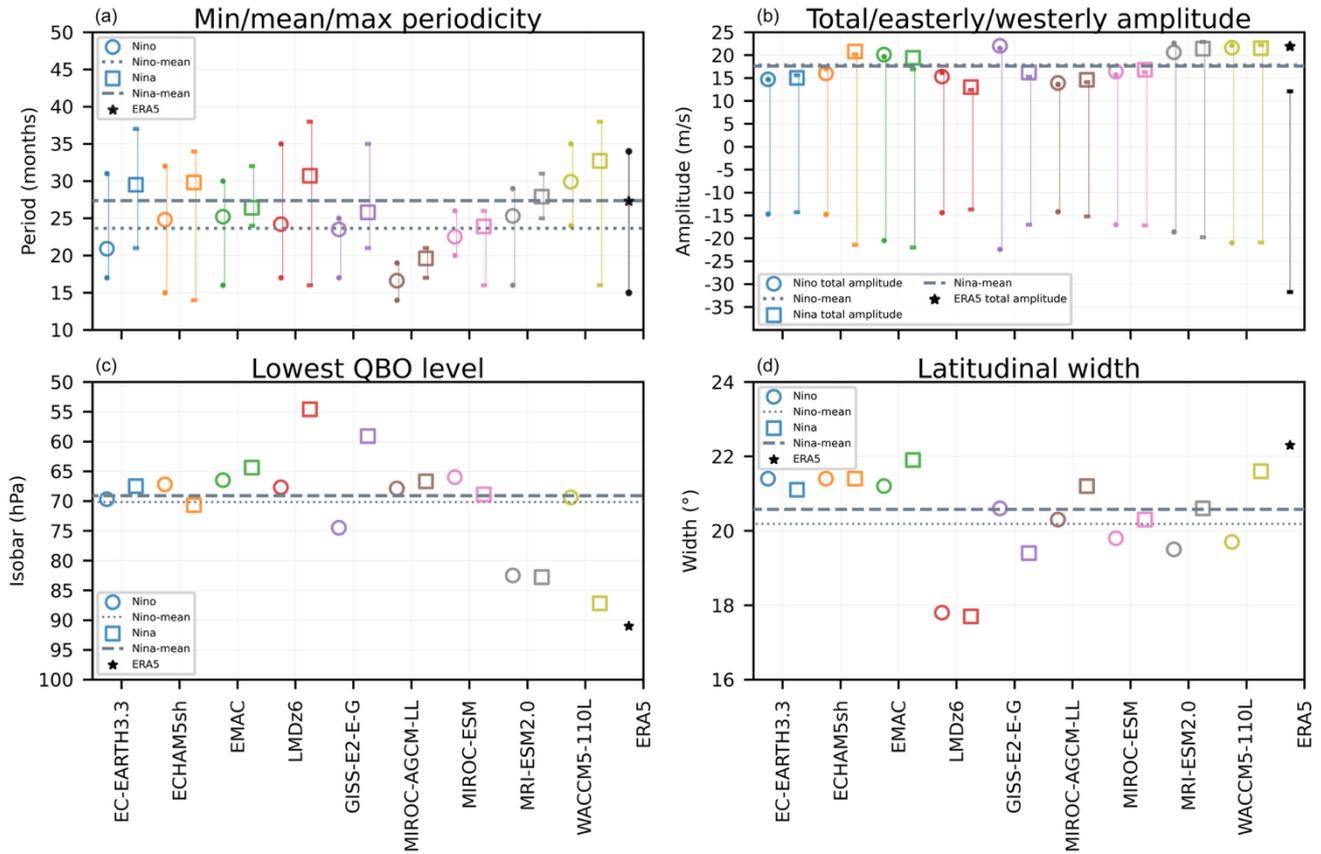
514
515 It is also possible that the amplification of the equatorial Rossby wave during La Niña (cf. Figure 2) projects onto the MJO’s
516 vertical structure. These waves (when located in the eastern hemisphere) have a first baroclinic structure in zonal wind that
517 consists of low-level westerlies and upper-level easterlies (Kiladis et al. 2009; Yuni et al. 2019; Nakamura and Takayabu
518 2022). Following from the robust amplification of the equatorial Rossby wave across the models during La Niña, it was
519 hypothesized that the low-level westerlies west of the MJO convection would be stronger during La Niña than El Niño like in
520 Wei and Ren (2019). This does not appear to be the case though and no first baroclinic zonal wind structure stands out in the
521 El Niño minus La Niña composites. The signal of the equatorial Rossby wave does, however, appear to be visible in the
522 temperature field. These waves are associated with a mid to upper tropospheric warming that is centered around 300 hPa
523 (Kiladis et al. 2009, Fig. 18c). This region of the upper troposphere is warmer in all of the La Niña simulations, with the
524 exception of MIROC-ESM in which the warming is marginally stronger in the El Niño composite.

525

526 **3.2 The lack of QBO-MJO interaction**

527 The results in the previous section indicate that ENSO modulates the MJO’s propagation, promoting faster MJOs during El
528 Niño and the opposite during La Niña. However, it is possible that aliased signals from the spontaneously generated QBOs are
529 embedded in the aforementioned results. Therefore, in this section we look for evidence of QBO-MJO interaction. As a first
530 step, the representation of the QBO is documented using previously defined metrics, with a specific interest in quantifying the
531 “lowest level” that the QBO descends to in the lower stratosphere. Insufficient descent is a known bias, which may hinder the
532 QBO from modulating other potentially important variables near the tropopause such as temperature (Richter et al. 2020; Kim
533 et al. 2020). Similar to Schenzinger et al. (2017), the lowest level that QBO reaches is found by averaging the QBO Fourier

534 amplitude (see Methods) over 5°S-5°N, identifying the maximum amplitude (fixed at 20 hPa here), and then finding the isobar
 535 in the lower stratosphere where the amplitude equals 10% of the maximum.
 536



537
 538 **Figure 7: Periodicity, amplitude, lowest level of descent, and latitudinal width of the QBO in El Niño (circles) and La Niña (squares)**
 539 **simulations. (a) mean periodicity, where whiskers extend to the minimum and maximum period, El Niño mean is dotted, La Niña**
 540 **mean is dashed, and ERA5 is shown in black. (b) circles and squares show total amplitudes (“TT amplitude” from Richter et al.**
 541 **2020), tops of whiskers show westerly amplitudes, and bottoms of whiskers show easterly amplitudes. (c) lowest isobar the QBO**
 542 **descends to, the level at which the QBO Fourier amplitude falls to 10% of its maximum. (d) latitudinal width of QBO, full width at**
 543 **half amplitude maximum of a Gaussian fit to the QBO Fourier amplitude**

544
 545 The QBO descends to 92.2 hPa in ERA5 (Fig. 7c). Of the simulated QBOs, the majority do not reach beneath 70 hPa, indicating
 546 that they are likely too high in altitude to influence the tropical atmosphere beneath 100 hPa as observed (Tegtmeier et al.
 547 2020). One outlier is the WACCM5-110L’s La Niña simulation whose lowest isobar of 87 hPa is fairly similar to the ERA5
 548 benchmark. Nonetheless, sensitivity tests with this simulation in which MJO amplitude is computed as a function of lower
 549 stratospheric QBO phase reveals its MJO to be insensitive to the QBO in the observed way (not shown), which may also be

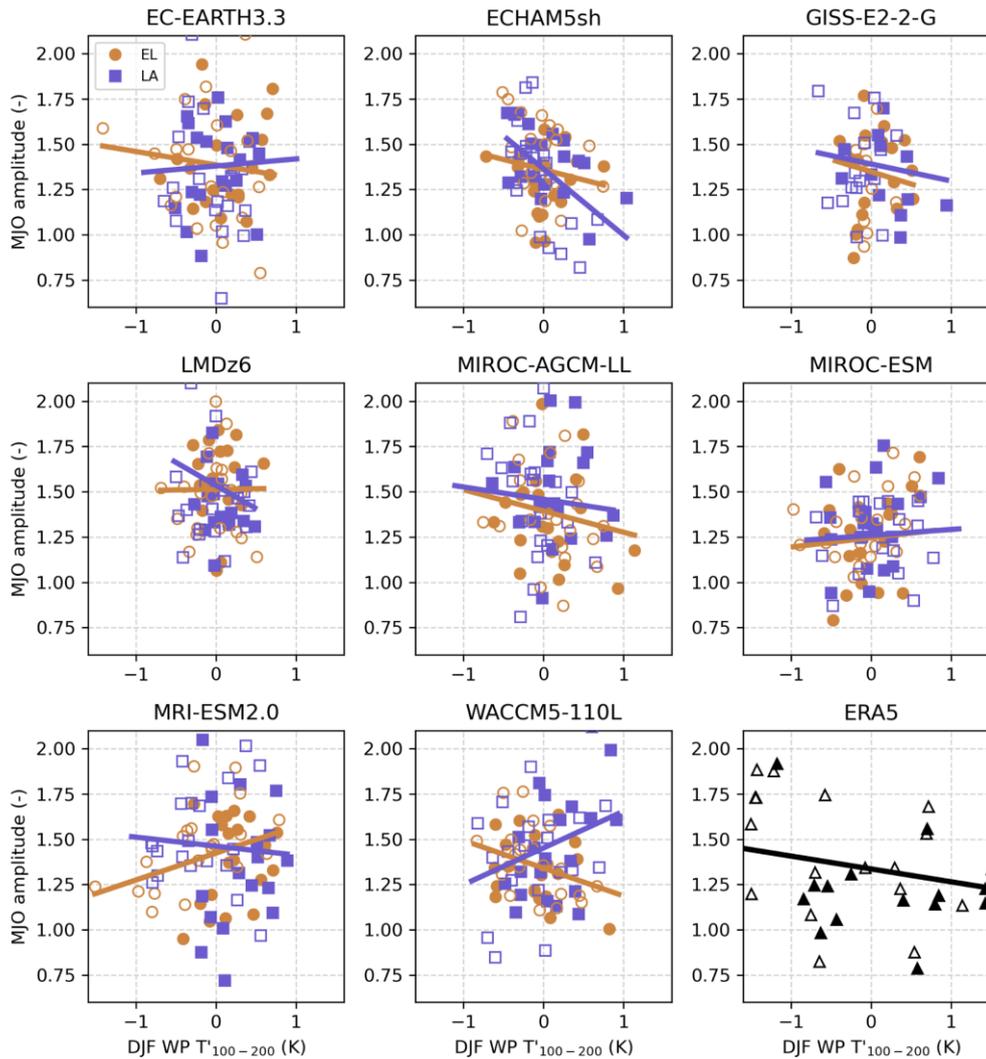
550 attributable to the weak MJO signal in the model. ENSO phase does not have consistent effects on what lowest isobar a given
551 model's QBO reaches. For example, GISS-E2-2-G and LMDz6 favor much stronger descent of the QBO into the lower
552 stratosphere during El Niño whereas WACCM5-110L's ENSO simulations reflect a strong opposite signed response.

553

554 QBO periodicity varies systematically with ENSO phase in these simulations, being shorter during La Niña and longer during
555 El Niño (Fig. 7a); see Kawatani et al. (2025) for a detailed analysis of this result. However, none of the QBO spatial metrics
556 exhibit consistent ENSO-phase dependence across models. For example, total QBO amplitude is marginally stronger during
557 La Niña in five of the nine models (Fig. 7b). El Niño and La Niña total QBO amplitudes differ by less than 1 m/s for all models
558 except ECHAM5sh, GISS-E2-2-G, and LMDz6. ECHAM5sh favors a stronger QBO amplitude during La Niña, owing to
559 intensified QBO easterlies and westerlies during this ENSO phase. Conversely, GISS-E2-2-G, and LMDz6 favor stronger
560 easterly, westerly, and total QBO amplitudes during El Niño. Of the 12 simulations corresponding to the six other models, EC-
561 EARTH3.3, EMAC, MIROC-AGCM-LL, MIROC-ESM, MRI-ESM2.0, and WACCM5-110L, the magnitude of their easterly
562 and westerly QBO amplitudes is stronger during La Niña in eight of the 12 simulations. Although the models ubiquitously
563 underestimate the latitudinal extent of the QBOs relative to ERA5, six of nine models have wider QBO latitudinal extents
564 during La Niña (Fig. 7d). This may be noteworthy because the boreal winter polar stratospheric wind response to the QBO is
565 stronger when the QBO is wider (Hansen et al. 2013) and there is a preference for this teleconnection to happen during La
566 Niña over the observed record (Kumar et al. 2022).

567

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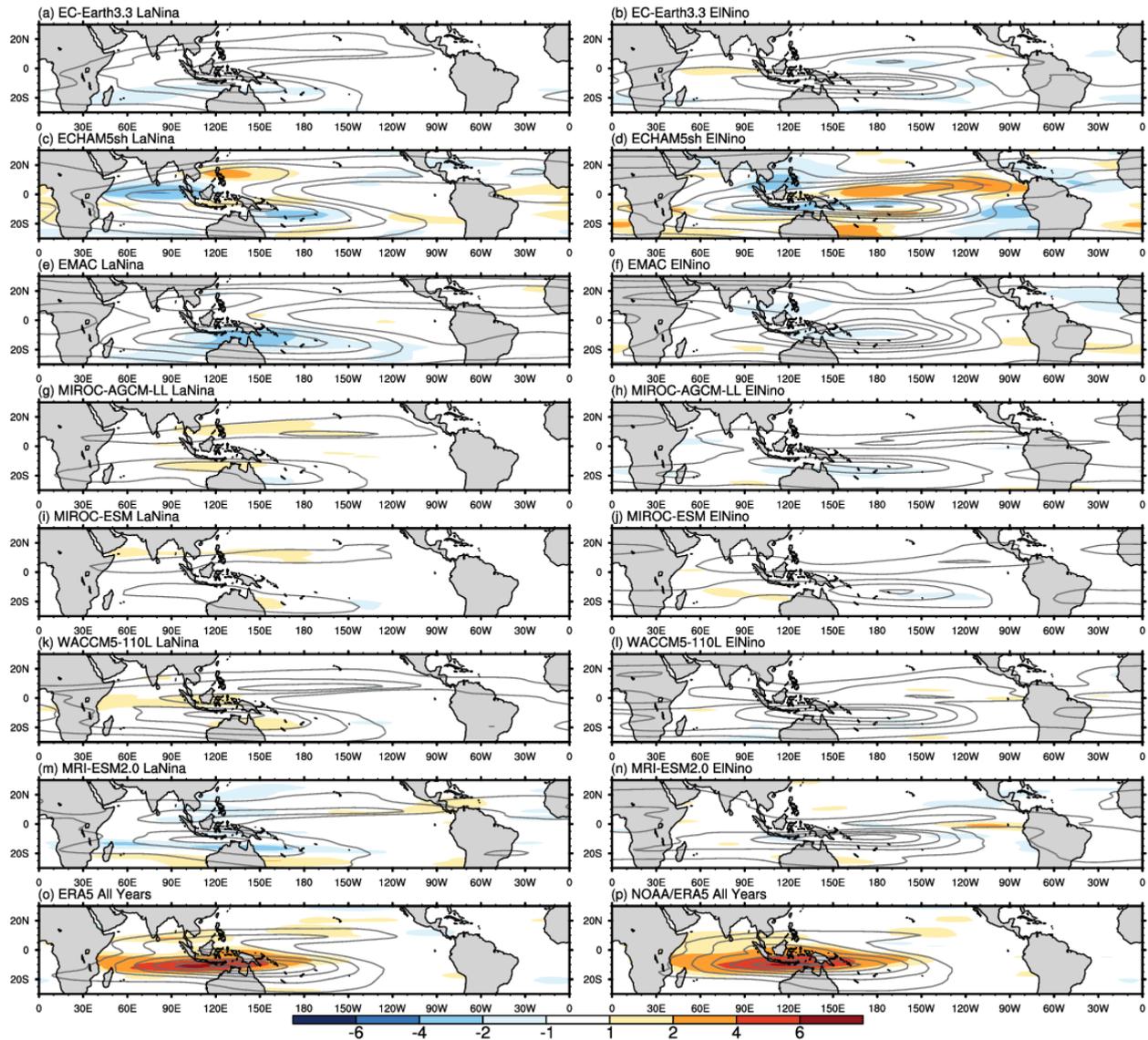


569 **Figure 8:** Scatter plot of warm-pool averaged (10°S - 10°N , 45°E - 180°E) tropopause stability anomalies (100 hPa minus 200 hPa
570 temperature) versus December-February MJO amplitude. Lines represent the slope of the regression line during El Niño (orange)
571 or La Niña (purple). Easterly and westerly QBO phases, which are delineated by the sign of the December-February 5°S - 5°N 50 hPa
572 zonal mean zonal-wind, are denoted by open and filled markers, respectively.
573
574

575 While the aforementioned QBO metrics help to broadly characterize the form of each model's QBO, they are non-time-varying
576 quantities and have less value for better understanding seasonal phenomena such as the predominantly boreal winter QBO-
577 MJO interaction. To incorporate the effects of seasonality, scatterplots of December-February warm-pool averaged tropopause
578 stability versus December-February MJO amplitude are made as a function of QBO phase for each of the simulations
579 (Klotzbach et al. 2019). MJO amplitude is expected to increase as tropopause stability decreases (Son et al. 2017; Klotzbach
580 et al. 2019), as is apparent based on ERA5 (Fig. 8). This metric is relevant for considering QBO-MJO coupling because the
581 QBO's effect on lower stratospheric stability (e.g., Densmore et al. 2019) is one of the suspected physical mechanisms coupling

582 the QBO and MJO. In general, the models do not reproduce the observed inverse relation between MJO amplitude and 100-
583 200 hPa stability and stratifying the results either by ENSO phase or QBO phase does not change this. The linear response of
584 the MJO amplitude to the tropopause stability, which is negative in observations, does not change consistently with the ENSO
585 phase. It is worth noting that the mean DJF stability in the models is generally close to that of ERA5 (-28.7 K) except for the
586 GISS model, which showcases a smaller gradient (-23.3 K), and all models appear to underestimate its interannual variability,
587 as evident by the smaller range compared to ERA5 (recall however that imposed SSTs do not vary interannually). Furthermore,
588 the stratification by QBO phase (with EQBO associated with lower stability values, and vice versa for the WQBO) is also
589 small or absent in models, possibly due to a limited influence of the QBO at tropopause heights (Serva et al., 2022). Further
590 sensitivity tests were done to see if the model QBOs simulate a QBO-MJO amplitude relationship when the MJO is in a
591 particular phase (e.g., Lim and Son 2020; Lawrence et al. 2023); no systematic effect was detected (not shown).

MJO Activity (OLR σ)



592
 593 **Figure 9:** Gray contours show the MJO activity defined as the standard deviation of MJO-filtered OLR for each model and ENSO
 594 phase, as well as for ERA5 OLR (9a) and NOAA OLR (9p) . The color-filled contours show the MJO-QBO relationship as the
 595 difference in MJO activity for the eastward minus westward 50 hPa QBO phases for each model and using the ERA5 wind in 9a
 596 and 9p.

597
 598 To further evaluate the representation of QBO-MJO coupling in the models, Figure 9 presents the effect of QBO phase on
 599 MJO activity (see Methods). As shown in previous studies (e.g., Kim et al. 2020), the models do not capture the observed

600 QBO-MJO relationship, which has maximum signal over the maritime continent region (Fig. 9m) illustrating the enhancement
601 of MJO activity during easterly QBO phase. With La Niña forcing, MIROC-AGCM-LL and WACCM5-110L show a weak
602 positive signal over the Maritime Continent (Fig. 9g, k), however EC-EARTH3.3, ECHAM5sh, and EMAC exhibit rather
603 different responses. In MIROC models, the MJO activity is further shifted off-equator, possibly hampering any QBO control
604 on MJO convection. No change in the MJO activity by QBO phase is evident in the El Niño simulations either, except for
605 some negative signal in the central and eastern Pacific in ECHAM5sh, which may be due to an irregular QBO in this experiment
606 (Kawatani et al., 2025). There is, however, a clear eastward shift of the MJO activity towards the Pacific during El Niño,
607 corroborating the observational work of Kessler (2001) and the climate model based study of Tam and Lau (2005).

608 **4 Discussion and conclusions**

609 The observed interannual variability of the MJO is influenced by multiple parts of the climate system. Due to their impact on
610 the tropical troposphere and prominent fluctuations at interannual timescales, ENSO and the QBO are known drivers of the
611 MJO's year to year variability, however it is difficult to definitively isolate their influence on the MJO because of how short
612 and noisy the observational record is (Randall et al. 2023). Building on previous work, our aim here is to evaluate the extent
613 to which model behavior is consistent with previously reported ENSO-MJO relationships by analyzing the representation of
614 the MJO in nine climate models forced by prescribed perpetual El Niño and La Niña conditions, each with a spontaneously
615 generated QBO. While the imposed SST anomalies represent an upper bound for the observed record, they are relevant given
616 the projected intensity increase of ENSO extremes.

617
618 Although the models exhibit difficulties simulating the MJO, several previously reported effects of the ENSO phase on the
619 MJO are corroborated by this coordinated set of experiments. These effects include faster propagation of the MJO during El
620 Niño versus slower propagation during La Niña, manifesting as shorter and longer lifetimes, respectively, stronger amplitude
621 of the MJO during El Niño, and east-west shifting of the MJO timescale variance towards the east Pacific during El Niño and
622 towards the west Pacific and Indian Ocean during La Niña.

623
624 It is likely that the high-amplitude SST forcings used here, particularly the amplified El Niño forcing, contribute to the
625 magnitude of these ENSO associated MJO changes, consistent with prior work linking exceptionally warm Pacific SSTs to
626 enhanced or farther-east MJO propagation (e.g., Marshall et al. 2016). To assess this sensitivity to ENSO amplitude, in one
627 analysis we compare a broad measure of MJO quality and propagation, the MJO E/W ratio, between the perpetual ENSO
628 simulations and each model's 1979–2009 AMIP simulations, the latter of which are more representative of typical ENSO
629 amplitudes. Across models, the amplified El Niño forcing increases the E/W ratio relative to La Niña, shifting it toward the
630 upper tail of the AMIP-based internal variability distribution and indicating more robust eastward MJO propagation,. All
631 perpetual ENSO E/W still ratios remain within the spectrum of their AMIP-based values, underscoring that the modeled

632 responses, while produced by extreme forcings, are consistent with how the MJO behaves under normal SSTs. For GPCP and
633 ERA5, however, their composite El Niño and La Niña MJO E/W ratios sit centrally within the distribution of sampling
634 variability (Fig. 4a), which supports Slingo et al. (1999) and Hendon et al. (1999)’s conclusion that ENSO does not strongly
635 modulate seasonal-mean MJO activity. Thus, the ENSO-MJO sensitivity expressed in the models is less clearly present in
636 observations, implying either reduced sensitivity in nature or an exaggerated response to the imposed SST anomalies.

637
638 Beyond the E/W ratio, other indicators of MJO activity demonstrate strong responses to ENSO forcing in these simulations,
639 differing from these studies. Hendon et al. (1999) found that increased MJO activity coincides with an increased number of
640 MJO events and enhanced intraseasonal convective activity around the Maritime Continent. While these attributes of the MJO
641 were largely insensitive to SSTs in their study, the models here unambiguously simulate more events in their perpetual El Niño
642 simulation (Table 2), which are of stronger MJO activity than their La Niña equivalents (Figure 9). We suspect that the
643 distinction between our results and previous research is related to the timescale over which the oceanic component of ENSO
644 modulates the atmosphere. Recent studies show that while the likelihood of MJO occurrence and its propagation speed are
645 only weakly correlated with tropical intraseasonally filtered SSTs, they are strongly correlated with low-frequency (e.g., > 90
646 days) SSTs (Suematsu and Miura 2018; 2022). In light of this and our use of a simplified climate system in which smoothed
647 monthly SSTs are prescribed, intraseasonal and interannual SST fluctuations are explicitly ignored, and the downward impact
648 of the MJO on intraseasonal SSTs (Zhang and Gottschalk 2002; Hendon et al. 2007; Newman et al. 2009) is not simulated, we
649 deduce that the different basic state circulations set up by the indefinite ENSO forcings enables distinct MJOs. This
650 interpretation of the low-frequency SSTs as an important modulator of the MJO aligns with studies which have attributed
651 variability in the MJO’s propagation to ENSO (Wei and Ren 2019; Wang et al. 2019; Dasgupta et al. 2021; Back et al. 2024)
652 as well as studies employing climate models in which MJO propagation can be modulated by changing the horizontal gradients
653 of the background SST field (Kang et al. 2013; Jiang et al. 2020).

654
655 We note that, besides SST amplitude, the representation of MJOs in models appears to be particularly sensitive to convective
656 schemes, as discussed by Rind et al. (2020) for GISS, by Park et al. (2019) for WACCM, by Miura et al. (2012) for MIROC
657 and Holt et al. (2020) for other Phase 1 QBOi models. The impact of the convective scheme, for example, appears oversized
658 compared to horizontal resolution, which is finely resolved in EC-EARTH3.3, MIROC-AGCM-LL, MRI-ESM2.0, and
659 WACCM, none of which are exceptional at reproducing the MJO’s wavenumber frequency spectra or E/W ratios. Moreover,
660 models within the same family (MIROC-based, ECHAM-based) exhibit similar MJO spectra (Fig. 2, Fig. 4), despite having
661 intra-family differences in horizontal and vertical resolution (Table 2; Holt et al. 2020, Fig. 18). As shown by Orbe et al. (2020)
662 for the WACCM and GISS models, refinement of key parameterization has potential to improve simulated MJOs.

663
664 As in the observational record, it is possible that aliasing from the QBO is superimposed on what are thought to be MJO
665 changes due to ENSO. However, we find that the climate models considered here show little evidence of QBO-MJO coupling.

666 This may be due to various factors, such as biases in the descent of the QBO, and the location and characteristics of tropical
667 convection, which appears to be less organized in some of the models compared to GPCP. While this limits our ability to
668 diagnose the mechanisms linking the QBO and MJO, it suggests that, in these experiments, the MJO changes we detect are
669 primarily driven by the ENSO forcings. Experiments with specified rather than internally generated QBO aspects can help
670 understanding the processes at play, which remain elusive (Martin et al., 2023; Huang et al. 2023). One speculative reason for
671 the missing QBO-MJO signal is the representation of upper tropospheric Kelvin waves. Observations show that during EQBO,
672 the Kelvin wave associated with the MJO penetrates higher into the lower stratosphere, coinciding with an eastward-tilting
673 cold anomaly in the UTLS (Hendon and Abhik 2018). We find a similar dry Kelvin wave signature in our ERA5 El Niño
674 minus La Niña MJO vertical structure composite, but in models, the UTLS wave structure is much more coarsely resolved
675 (Fig. 6). Such deficiencies, perhaps linked to vertical resolution (Holt et al. 2020, Fig. 18a), lower-stratospheric wind biases
676 that develop in the absence of sufficient QBO descent (Fig. 7c), or perhaps just weak MJOs, may weaken this dry Kelvin wave
677 component and, in turn, suppress QBO-MJO interaction.

678
679

680 Compared with the La Niña simulations, all El Niño simulations include amplified Kelvin waves whereas equatorial Rossby
681 waves intensify in the presence of perpetual La Niña conditions. Consistent with the reported relationship between these
682 waves and the MJO (Wei and Ren 2019), all models simulate faster MJO propagation in their El Niño simulation. This is
683 further supported by the MJO diversity analysis, which reveals that models reproduce the observed fast and standing MJO
684 archetype OLR Hovmöllers well in the presence of perpetual El Niño and La Niña conditions, respectively. In addition, the
685 MJO's phase 3/4 vertical structures highlight that lower tropospheric easterlies do intensify to the east of the MJO's major
686 convection during El Niño across most models, which we interpret to result from the intensification of the Kelvin wave.

687

688 The strengthening of model equatorial Rossby and Kelvin waves under La Niña and El Niño forcing, respectively, is broadly
689 consistent with observational studies (Wei and Ren 2019; Wang et al. 2019), yet the striking uniformity of these responses
690 across models and underlying mechanisms are not fully understood. Both waves exhibit substantial longitudinal variation in
691 observations, and their behavior depends on multiple aspects not considered in detail here, including seasonal-mean
692 convection, ambient zonal wind, and the extent to which extratropical planetary waves can intrude into the tropics (Yang et al.
693 2023). These factors, in turn, are further shaped by the model-dependent tropospheric responses to the internally generated
694 QBOs (Naoe et al. 2025). Because these degrees of freedom are not isolated in our study, we cannot yet diagnose with
695 confidence why the models produce such consistent equatorial Rossby and Kelvin wave responses to the ENSO forcings.
696 Nevertheless, the fact that this coherence emerges despite these unconstrained influences underscores the robustness of the
697 simulated wave responses and motivates a closer examination of the waves in subsequent work.

698 While the modeled relationship between Kelvin waves, equatorial Rossby waves, and ENSO is well supported by previous
699 studies employing empirical data and reanalysis, this is the first time, to the best of our knowledge, that this relationship has

700 been ubiquitously affirmed by a coordinated set of climate model experiments with prescribed strong ENSO forcings. The
701 robustness of this result across models suggests that it is worthwhile considering how these wave responses to ENSO influence
702 other parts of the climate system. For instance, Kelvin waves are a source of resolved wave forcing for the QBO (Baldwin et
703 al. 2001; Taguchi 2010; Pahlavan et al. 2021) and more rapid descent of the QBO's westerly shear zones during El Niño in
704 observations has been attributed to their intensification (Das and Pan 2016). The periodicity of the QBO in the El Niño
705 simulations is in fact shorter than in the La Niña simulations across all models considered here (Kawatani et al., 2025),
706 however, the extent to which Kelvin waves are responsible for this as opposed to other waves (e.g., Kawatani et al. 2019), is
707 yet to be quantified across all of the models. The convectively coupled wave responses presented here may also be relevant
708 for better understanding of the observed ENSO diversity. El Niño events vary in type and intensity due to the influence of
709 westerly wind bursts, which introduce asymmetry and irregularity into ENSO's phase changes (Chen et al. 2015). Westerly
710 wind bursts are more frequent during the convective phases of equatorial Rossby waves and the MJO, especially strong MJOs
711 (Puy et al. 2016). Hence, the atmospheric responses to ENSO, such as the amplifications of the MJO during El Niño (Figs. 2-
712 4) and of the convectively coupled Rossby wave during La Niña, have a pathway to influence ENSO's oceanic component.

714 **Code availability**

715
716 The code used for the wavenumber-frequency analysis is publicly available through the National Center for Atmospheric
717 Research (NCAR) Command Language (NCL): <https://www.ncl.ucar.edu/Applications/mjoclivar.shtml>. A reproduction of
718 NCL's Wheeler-Kiladis (1999) routine in Python is available here: https://github.com/brianpm/wavenumber_frequency.
719 Python code, which assesses QBO morphology is available here: [https://github.com/NOAA-GFDL/MDTF-
720 diagnostics/blob/main/diagnostics/stc_qbo_ens/stc_qbo_ens.py](https://github.com/NOAA-GFDL/MDTF-diagnostics/blob/main/diagnostics/stc_qbo_ens/stc_qbo_ens.py).

722 **Data availability**

723 Storage for the QBOi multi-model data set is provided by the Centre for Environmental Data Analysis (CEDA) whose data
724 and processing service is called JASMIN. Interested users must obtain a JASMIN login account and take the necessary steps
725 to access the QBOi group workspace within JASMIN, which contains the perpetual ENSO simulations. Certain derived model
726 products (e.g., the MJO RMMs) may be made available upon request.

728 **Author contributions**

729 DE, FS, JC, SYB, CO, and JR contributed to the conceptualization of this study. DE, FS, JC, and SYB performed the data
730 analyses and produced the figures and tables. All authors contributed to the review and editing of this manuscript.

732 **Competing interests**

733 The authors declare that they have no conflict of interest.

734

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755

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