

1 **The impact of the Madden-Julian Oscillation on hydrological extremes**

2 **Jian Peng^{1,2,3,4*}, Simon Dadson¹, Guoyong Leng^{5*}, Zheng Duan⁶,**

3 **Thomas Jagdhuber⁷, Weidong Guo^{2*}, Ralf Ludwig³**

4 ¹ School of Geography and the Environment, University of Oxford, OX1 3QY Oxford, UK;

5 ² Institute for Climate and Global Change Research, School of Atmospheric Sciences, Nanjing University, 210023
6 Nanjing, China;

7 ³ Department of Geography, University of Munich (LMU), 80333 Munich, Germany;

8 ⁴ Max Planck Institute for Meteorology, 20146 Hamburg, Germany;

9 ⁵ Environmental Change Institute, University of Oxford, OX1 3QY Oxford, UK;

10 ⁶ Chair of Hydrology and River Basin Management, Technical University of Munich, Munich 80333, Germany;

11 ⁷ Microwaves and Radar Institute, German Aerospace Center (DLR), 82234 Wessling, Germany;

12 * Authors to whom correspondence should be addressed: E-Mail: guowd@nju.edu.cn (Weidong Guo),
13 guoyong.leng@ouce.ox.ac.uk (Guoyong Leng), jian.peng@ouce.ox.ac.uk (Jian Peng).

14

15 **Abstract**

16 Extreme climate events such as severe droughts and floods have become more frequent and
17 widespread in the 21st Century. Recent studies have revealed the tele-connections between
18 Madden–Julian oscillation (MJO) and extreme precipitation over different regions such as South
19 America, India and China. This study investigates the influence of MJO on global extreme dry
20 and wet conditions, and how the strength of the relationship changes across the MJO phases over
21 the globe. The Evaporative Stress Index (ESI) calculated from global GLEAM
22 evapotranspiration dataset is used to represent extreme dry and wet conditions. Strong
23 correlations between MJO and extreme dry and wet conditions are found, particularly over
24 monsoon regions such as South Asia, South America and East Africa. The underlying
25 mechanism of the influence of MJO on extreme dry and wet conditions is associated with the
26 variation of precipitation, air temperature and soil moisture modulated by the MJO. The study
27 suggests that MJO impacts on extreme dry and wet conditions should be taken into account in
28 investigation of droughts/floods around the world particularly over monsoon areas.

29 **Keywords:** Madden–Julian oscillation, GLEAM, evaporative stress index, CCI soil moisture,
30 GPCP precipitation, ERA-Interim

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33 **1 Introduction**

34 Climate extremes such as floods and droughts have significant impacts on natural system,
35 human, society and economy in vulnerable regions (Easterling et al. 2000; Meehl et al. 2000;
36 Zhang et al. 2019). The Fifth Assessment Report of the Intergovernmental Panel on Climate
37 Change (IPCC) has indicated that the frequency, intensity, and duration of some climate extreme
38 events will increase by the end of this century (IPCC 2013). However, there are large

39 discrepancies on climate extreme events projections in Coupled Model Intercomparison Project
40 phase 5 (CMPI5) and the older CMIP3 collections (Knutti and Sedláček 2013; Sillmann et al.
41 2013). Therefore, understanding, modeling and predicting climate extremes has been identified
42 as one of the World Climate Research Programme (WCRP) Grand Challenges (Sillmann et al.
43 2017). Various processes determine the onset, duration and recovery of climate extremes
44 particularly floods and droughts at multiple temporal (seasonal, annual, and decadal) and spatial
45 scales (local, regional, continental) (Dai 2013; Frei et al. 2006; Sun et al. 2016a). In general,
46 climate extremes are strongly influenced by modes of climate variability such as El Niño–
47 Southern Oscillation (ENSO), the North Atlantic Oscillation (NAO) and the Pacific Decadal
48 Oscillation (PDO) (Kenyon and Hegerl 2010; Weisheimer et al. 2017; Zhang et al. 2010). In
49 recent years, the ENSO, NAO, and PDO have been widely studied and identified as the source of
50 climate predictability for global dry/wet conditions seasonal forecasting (Barlow et al. 2001;
51 Cayan et al. 1998). The current climate models are also able to simulate the gross characteristics
52 of these modes of climate variability (Sillmann et al. 2017). However, it is still challenging in
53 sub-seasonal climate extremes forecasting because of the poor ability of models to simulate
54 Madden–Julian Oscillation (MJO) (Inness and Slingo 2003; Robertson et al. 2018; Waliser et al.
55 2003). A recent report from Robertson et al. (2018) emphasized the importance of the MJO in
56 sub-seasonal forecasts over the tropics and remarkable progress of the representation of MJO in
57 the operational models. However, they also highlighted that the impacts of MJO on weather
58 through teleconnections are still not well captured by the models, which limits the sub-seasonal
59 to seasonal predictability of extreme weather and climate. Therefore, the projections of regional
60 climate extremes over South Asia, West Asia, Southeast Asia and Australia are highly uncertain
61 when associated with MJO.

62 The MJO is a tropical intra-seasonal oscillation that is characterized by a large scale
63 convection system propagating from the Indian Ocean to the Pacific Ocean with a speed of 5-10
64 ms^{-1} (Madden and Julian 1971; Zhang 2005). Typically, the MJO shows an intraseasonal
65 variability with time scale of 30-90 days, which is characterized by enhanced convection across
66 the western Indian Ocean and suppressed convection over the western Pacific Ocean. It has been
67 found that the MJO can modify the extratropical circulation by acting as diabatic heating source
68 (Matthews et al. 2004). It plays an important role in the global weather-climate system and
69 influences many weather and climate phenomena. Previous studies have found that the MJO has
70 effects on precipitation (Jones et al. 2004a), surface temperature (Zhou et al. 2016), snow cover
71 (Li et al. 2016), tropical cyclones (Frank and Roundy 2006), tornadoes (Thompson and Roundy
72 2013), fire (Reid et al. 2012), soil moisture (Peng et al. 2017a), El Niño Southern Oscillation
73 (Pohl and Matthews 2007), Wyrki jets (Han 2005) and several other weather and climate events
74 (Zhang 2013). Several studies have found a strong relationship between MJO and extreme
75 precipitation over specific regions such as South America, India and China (Joseph et al. 2009;
76 Lü et al. 2012; Shimizu et al. 2017). Although floods are caused by complex interactions
77 between atmosphere, ocean and land (Seager et al. 2013; Sheffield et al. 2009; Sun et al. 2016a;
78 Sun et al. 2016c), these studies revealed the impacts of ocean states on the occurrence of extreme
79 precipitation. In addition, several studies reported the impacts of MJO on droughts in China and
80 India (Joseph et al. 2009; Lü et al. 2012; Neena et al. 2011). All these previous studies have
81 concentrated on the impacts of MJO on either extreme precipitation or droughts over specific
82 geographic locations. However, the impacts of MJO on droughts and floods at global scale are
83 rarely explored. As MJO has been considered as a major source of intra-seasonal climate
84 predictability in many dynamical forecasting systems (Kang and Kim 2010; Liu et al. 2017b;

85 Sun et al. 2014; Tan et al. 2017; Waliser et al. 2003) and the planetary scale of MJO, a better
86 understanding of the influence of MJO on global extreme dry and wet conditions is essential for
87 improving the forecasting of sub-seasonal extreme precipitation and droughts. With the ability to
88 accurately identify, model and forecast the MJO, it will provide valuable information for
89 operational climate risk management.

90 The aim of the present study is to investigate the spatial distribution of the land extreme
91 dry and wet conditions and its spatial and temporal variation connected to the MJO phases. To
92 our knowledge, this study is the first time to provide a global of view of the slow eastward
93 propagation of the MJO and the occurrence of extreme dry and wet conditions over the globe.
94 The study is mainly based on satellite products, and uses Evaporative Stress Index (ESI)
95 (Anderson et al. 2007) to represent global extreme dry and wet conditions. It should also be
96 noted that the definition of drought is much more complex than flooding. A wide range of
97 drought indexes has been developed to identify drought over the last decades (Zhang and He
98 2016; Zhang et al. 2015). The ESI represents the standardized anomaly of the evapotranspiration
99 (ET) fraction (actual ET/potential ET) and has been widely used for monitoring drought and
100 wetness conditions (Anderson et al. 2015; Choi et al. 2013; Otkin et al. 2014). The MJO has
101 been found to have significant influence on precipitation, air temperature and soil moisture
102 (Zhang 2013), which also to some extent represent the status of dry and wet conditions. Thus, the
103 statistical relationship between these climate variable and ESI are also quantified during MJO
104 events to reveal the mechanism of influence of the MJO on global extreme dry and wet
105 conditions. The following section introduces the details about data and methodology used for the
106 analysis. Section 3 presents and discusses the results. The conclusions drawn from this study are
107 summarized in Section 4.

108 **2 Data and Methodology**

109 2.1 Data

110 Both reanalysis and satellite datasets are used in this study. These datasets mainly include
111 Global Precipitation Climatology Project (GPCP) daily precipitation (Huffman et al. 2001),
112 Global Land Evaporation Amsterdam Model (GLEAM) daily evapotranspiration/potential
113 evapotranspiration (ET/PET) (Martens et al. 2017; Miralles et al. 2011), European Space
114 Agency's Climate Change Initiative (ESA CCI) soil moisture (Dorigo et al. 2015b; Liu et al.
115 2011) as well as wind speed and air temperature from ERA-Interim reanalysis datasets (Dee et
116 al. 2011). These datasets are briefly introduced below.

117 GPCP precipitation: the GPCP global precipitation dataset was produced based on rain gauge,
118 satellite and sounding data with the support of World Climate Research Program (WCRP) and
119 GEWEX activities (Huffman et al. 2001). The GPCP provides globally complete precipitation
120 dataset at 1° spatial resolution and daily time scale from October 1996 to present. The dataset has
121 been widely validated and applied in various studies (e.g., Hu et al. 2007; Sylla et al. 2013;
122 Trenberth et al. 2018).

123 GLEAM evapotranspiration: the GLEAM aims to estimate evapotranspiration from satellite
124 observations (Martens et al. 2017; Miralles et al. 2011). The actual evapotranspiration
125 components are based on Priestley and Taylor's (Priestley and Taylor 1972) potential
126 evapotranspiration equation constrained by a multiplicative stress factor. GLEAM provides both
127 actual evapotranspiration and potential evapotranspiration at daily temporal resolution and 0.25°
128 spatial resolution from 1980 to 2017. The GLEAM evapotranspiration datasets have been widely
129 validated against global FLUXNET measurements and applied for many hydro-meteorological
130 applications such as global land wetting and drying trend analysis, regional climate response to

131 greening of Earth, El Niño–La Niña cycle and recent trends in continental evaporation, and
132 drought monitoring (Forzieri et al. 2017; Greve et al. 2014; Lian et al. 2018; Martens et al. 2017;
133 Miralles et al. 2014; Richard et al. 2018; Vicente-Serrano et al. 2018).

134 ESA CCI soil moisture: the CCI soil moisture product was generated by fusion of existing active
135 and passive satellite-based soil moisture datasets within the framework of the ESA Climate
136 Change Initiative (Liu et al. 2011). It provides global soil moisture estimate at daily time scale
137 and at a spatial resolution of 0.25° from 1978 to 2016. Since the first release of the dataset in
138 2012, there are a wide range of validation and applications been conducted at either regional
139 scale or global scale (e.g., Dorigo et al. 2015a; Miralles et al. 2014; Peng and Loew 2017; Peng
140 et al. 2017d; Peng et al. 2016). Particularly, it has been applied widely for drought monitoring
141 and assessment (Liu et al. 2017a; Nicolai-Shaw et al. 2017; Yuan et al. 2015).

142 ERA-Interim reanalysis wind speed and air temperature: the ERA-Interim is a global
143 atmospheric reanalysis product that is generated with ECMWF 4-dimensional variational
144 analysis (4D-Var) data assimilation system. The ERA-Interim covers the period from 1979 to
145 present and provides 6-hourly data at a spatial sampling of about 0.7° . The ERA-Interim datasets
146 such as wind speed, air temperature and soil moisture have been widely evaluated and applied in
147 various studies (e.g., Betts et al. 2009; Peng et al. 2015; Szczypta et al. 2011).

148 The GPCP precipitation and ERA-Interim wind speed were applied to calculate the real-time
149 multivariate MJO (RMM) as described below. The GLEAM ET/PET were used to calculate
150 evaporative stress index for representing dry/wet conditions. The ESA CCI soil moisture and
151 ERA-Interim air temperature were applied to investigate the underlying mechanisms of MJO's
152 effects on extreme dry and wet conditions. To make all these datasets consistent, all of them

153 were reprocessed to $0.25^\circ \times 0.25^\circ$ grid size using bilinear interpolation method and to daily time
154 scale for a 17-year period (January 1997 to December 2013).

155 2.2 MJO index and composite analysis

156 The MJO is normally identified and tracked with an MJO index. One of the commonly
157 used MJO indices is the real-time multivariate MJO (RMM) index (Wheeler and Hendon 2004),
158 which is constructed based on empirical orthogonal functions (EOF) analysis of combined fields
159 of outgoing long-wave radiation (OLR) or precipitation and 850-hPa and 200-hPa zonal wind
160 anomalies (Crueger et al. 2013; Waliser et al. 2009). The first two normalized principal
161 components are usually referred as RMM1 and RMM2, which are in quadrature and represent a
162 propagating mode. The MJO eight phase life cycle is then defined depending on the sign and
163 amplitude of RMM1 and RMM2. The MJO event is identified when the RMM index amplitude
164 is larger than 1:

$$RMM = \sqrt{RMM1^2 + RMM2^2} \quad (1)$$

165 The MJO events are then separated into eight phases, corresponding to the location and strength
166 of the MJO convection from the Indian to the Central Pacific oceans (Wheeler and Hendon
167 2004). In this study, the GPCP precipitation and ERA-Interim zonal winds at 850-hPa and 200-
168 hPa were used to calculate RMM index and identify MJO events. In addition, the MJO has
169 strong seasonal variability with a primary convective signal appearing in the Northern
170 Hemisphere during boreal summer, and a convection anomaly centered in the Southern
171 Hemisphere during boreal winter (Kim et al. 2017; Zhang and Dong 2004; Zhou et al. 2012).
172 Therefore, the composite analysis in this study was examined based on two seasons: boreal
173 summer (May-October) and boreal winter (November-April).

174 2.3 Evaporative stress index and statistical analysis

175 The Evaporative Stress Index (ESI) (Anderson et al. 2007) was used in this study to
 176 exhibit global dry/wet conditions. The ESI represents the standardized anomaly of the
 177 evapotranspiration (ET) fraction (actual ET/potential ET) (Anderson et al. 2016). It is considered
 178 to be uniquely sensitive to changes in soil moisture and vegetation water content due to the
 179 association of ET with temperature, precipitation, radiation, and wind (Leng et al. 2017; Otkin et
 180 al. 2013). Therefore, the ESI has been widely used for monitoring drought and wetness
 181 conditions (Anderson et al. 2015; Choi et al. 2013; Otkin et al. 2014). The ESI is calculated as
 182 the standardized anomaly of ET fraction (Otkin et al. 2014):

$$ESI(t, x, y) = \frac{v(t, x, y) - \frac{1}{n} \sum_1^n v(t, x, y)}{\sigma(x, y)} \quad (2)$$

183 where v refers to ET fraction, $v(t, x, y)$ denotes the ET fraction at day t , n is the number of days,
 184 x, y is grid location, and $\sigma(x, y)$ is the standard deviation. Negative ESI values show the dry
 185 conditions, indicating vegetation that is stressed because of insufficient soil moisture. The
 186 recently released GLEAM v3.2 global ET and potential ET datasets provides a unique
 187 opportunity to calculate global spatial resolution ESI, which was used in the current study.

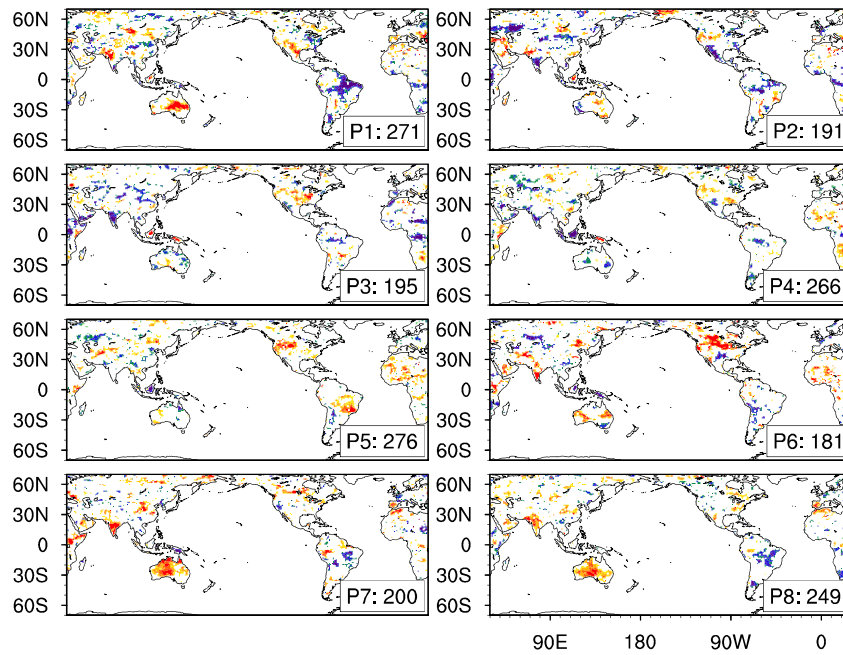
188 The spatial pattern correlation analysis was used to investigate the relation between
 189 dry/wet conditions and precipitation, soil moisture and air temperature. The statistical
 190 significance of the correlation analysis and composite analysis was tested using the Student's t -
 191 test with confidence level of 95%.

192 **3 Results and discussion**

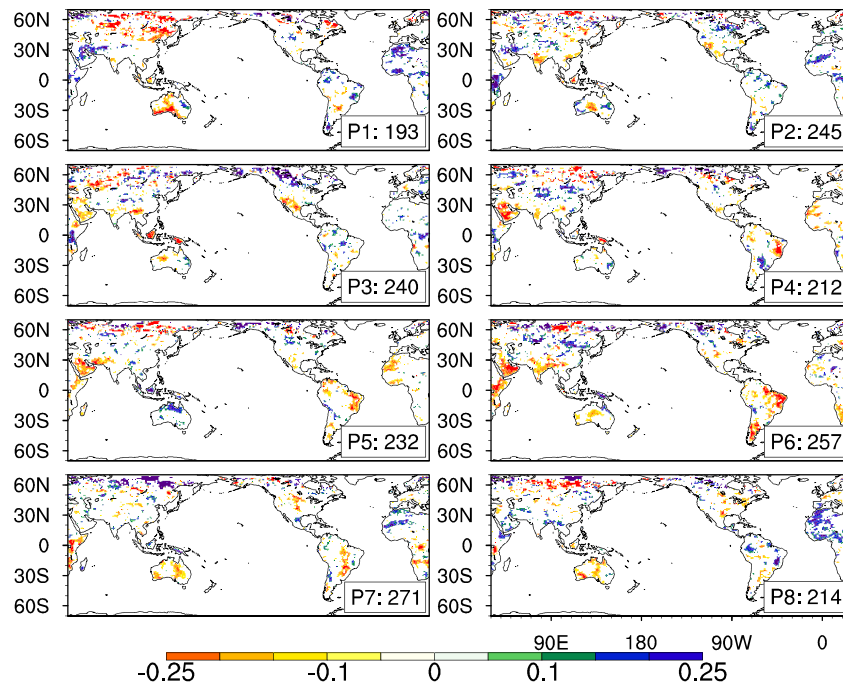
193 3.1 Analysis of global dry/wet conditions over different phases of MJO

194 The composite analysis of the impacts of MJO on extreme dry and wet conditions was explored
195 during MJO days. Figure 1 shows the composite maps of the ESI over eight MJO phases for
196 summer and winter respectively. It should be noted that only areas with P value less than 0.05
197 are shown, indicating the results are statistically significant at 95% level. In general, the figure
198 shows that the MJO has impacts on the variation of dry and wet patterns in many parts of the
199 world particularly over monsoon regions such as South Asia, southern Brazil, North American,
200 Australia, as well as East Africa. The presence of dry and wet conditions and their variations
201 across MJO phases over these areas agree well with the variations of monsoon systems (Zhang
202 2013). For example in boreal summer, the dryness was observed in South Asia in MJO phase 1,
203 while wetness was shown in MJO phase 2, 3 and 4. From phase 6 to 8, the dryness was observed
204 again. The intraseasonal variations of dryness and wetness correspond well with the variations of
205 summer monsoon over India caused by MJO (Joseph et al. 2009; Pai et al. 2011). In addition, the
206 dryness and wetness of non-monsoon regions such as East Africa is also affected by MJO. The
207 wetness is shown in phase 3 and 4 while dryness is found in phase 6 and 7 during boreal winter,
208 which is likely related to the corresponding heavy rain and light rain. What is the mechanism
209 through which MJO can influence dry and wet conditions? Previous studies have found that the
210 MJO plays an import role in modulating precipitation (e.g., Jones et al. 2004b), surface air
211 temperature (e.g., Vecchi and Bond 2004) and soil moisture (e.g., Peng et al. 2017a). The
212 impacts of MJO on these variables are related to modifying the meridional overturning
213 circulations (He et al. 2011), and exciting Rossby wave trains caused by heating anomaly (Zheng
214 et al. 2018), and moisture transport (Jia et al. 2011), and forcing zonally-propagating equatorial

215 Rossby and Kelvin waves (Janicot et al. 2009). Through these teleconnections, the MJO
216 modulate the precipitation, surface air temperature and soil moisture over the globe (Donald et
217 al. 2006; Matsueda and Takaya 2015; Peng et al. 2017a). For example, the MJO can influence
218 the extreme precipitation in USA through a phenomenon known as “atmospheric river”
219 transporting moisture from the tropical central Pacific to the west coast of the USA (Dettinger
220 2011). The impacts of MJO on dry and wet conditions found in the current study might be
221 related to the abovementioned teleconnection mechanism. One explanation for dry condition
222 induced by MJO relates to the deficiency of rainfall. In addition, the air temperature above
223 normal further increases the soil moisture evaporation. For MJO effects on wet conditions can be
224 explained by tropical-extratropical teleconnections, through which the MJO influences extreme
225 rainfall and air temperature and further lead to extreme wet conditions. For example, the ESI
226 signal shown in Figure 1 is consistent with precipitation patterns found by previous studies
227 (Donald et al. 2006; Zhang 2013). In particular, the onset of the South Asian monsoon is found
228 to occur more likely (50%-80%) in MJO phases 2 and 3, which will induce a spike in rainfall
229 over South Asia during boreal summer (Zhang 2013). Another example is east Africa, where
230 Pohl and Camberlin (2006) found 72% of extreme rainfall occurs near coastal regions when the
231 MJO center is over the Indian Ocean (phase 2) during boreal winter. These MJO induced
232 extreme precipitation events can also be observed in figure 1, where extreme wet conditions
233 expressed by ESI anomaly occur over South Asia in MJO phase 2 and 3 during boreal summer,
234 and east Africa in phase 2 during boreal winter. To test the above assumption that the dry and
235 wet condition is due to the combined impacts of MJO on precipitation, air temperature and soil
236 moisture, a correlation analysis between dryness/wetness and precipitation, soil moisture and air
237 temperature is conducted in the next section.



(a)

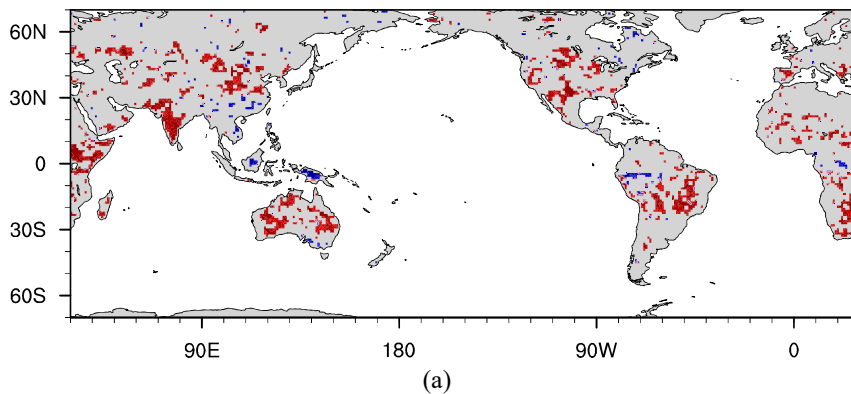


(b)

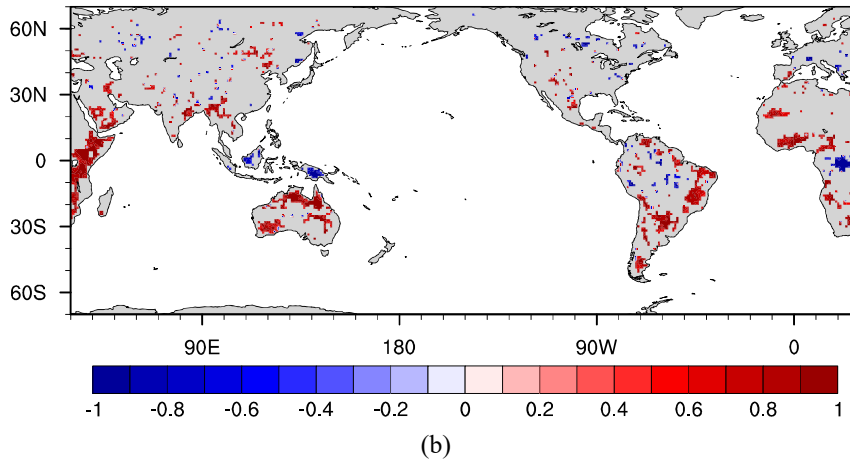
Figure 1: Composites of Evaporative Stress Index anomalies over eight MJO phases for boreal summer (a) and boreal winter (b). The P together with a number represents the MJO phase and the corresponding number of days. It should be noted that only the areas with statistically significant results ($p < 0.05$) are shown. The ESI values larger than 0 refers to wet conditions while the values less than 0 implies dry conditions.

247 3.2 Correlation analysis between dry/wet conditions and rainfall, soil moisture and air
 248 temperature

249 The correlation coefficient between dry/wet patterns represented by ESI and precipitation, soil
 250 moisture and air temperature over the globe and across MJO phases are shown in Figure 2, 3 and
 251 4, respectively. The grey areas refer to masked areas with $p > 0.05$. In general, the dryness and
 252 wetness over the globe during MJO events are related to precipitation, soil moisture and air
 253 temperature. The positive correlation is found between dryness/wetness and precipitation, as well
 254 as soil moisture, while air temperature has negative correlation with dry/wet patterns. It is likely
 255 due to that dry conditions favor more solar radiation and less evaporative cooling, which is
 256 consistent with published studies such as (Trenberth and Shea 2005). The relative high
 257 correlation was observed in monsoon areas such as South Asia, Australia, and South America
 258 where MJO has strong effects. In addition, the correlation also varies with seasons and variables.
 259 For example, the high correlation only appears in boreal summer for South Asia, which
 260 corresponds well to the monsoon season in South Asia (Xavier et al. 2014). And significant
 261 correlation with precipitation and soil moisture but not with air temperature was found in
 262 Australia, which suggests that precipitation, soil moisture and air temperature contribute
 263 unequally to the dryness/wetness caused by MJO.

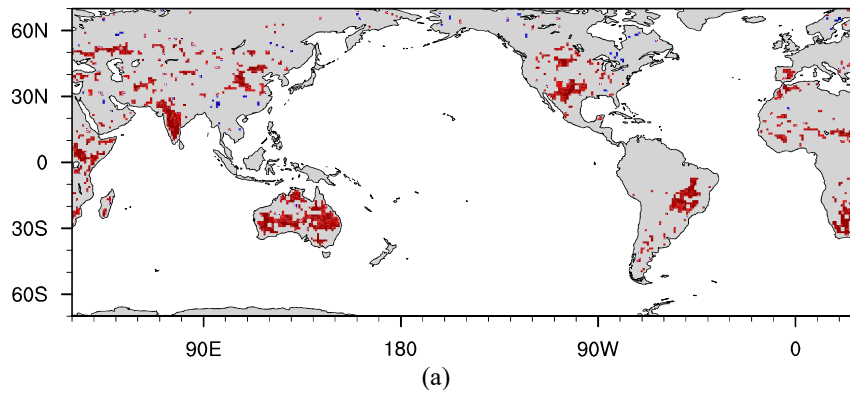


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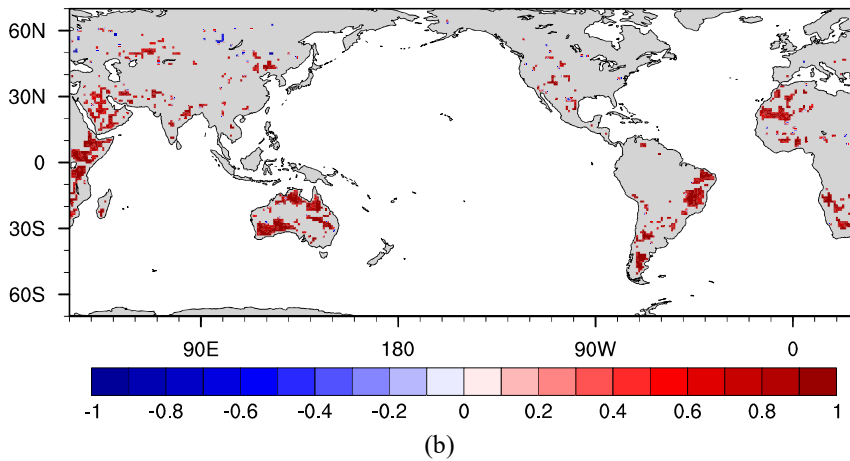


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Figure 2: The correlation coefficient (R) between Evaporative Stress Index and GPCP precipitation anomaly composites across the MJO phases at each grid point: (a) boreal summer and (b) boreal winter. Only areas with statistically significant results ($p < 0.05$) are shown.

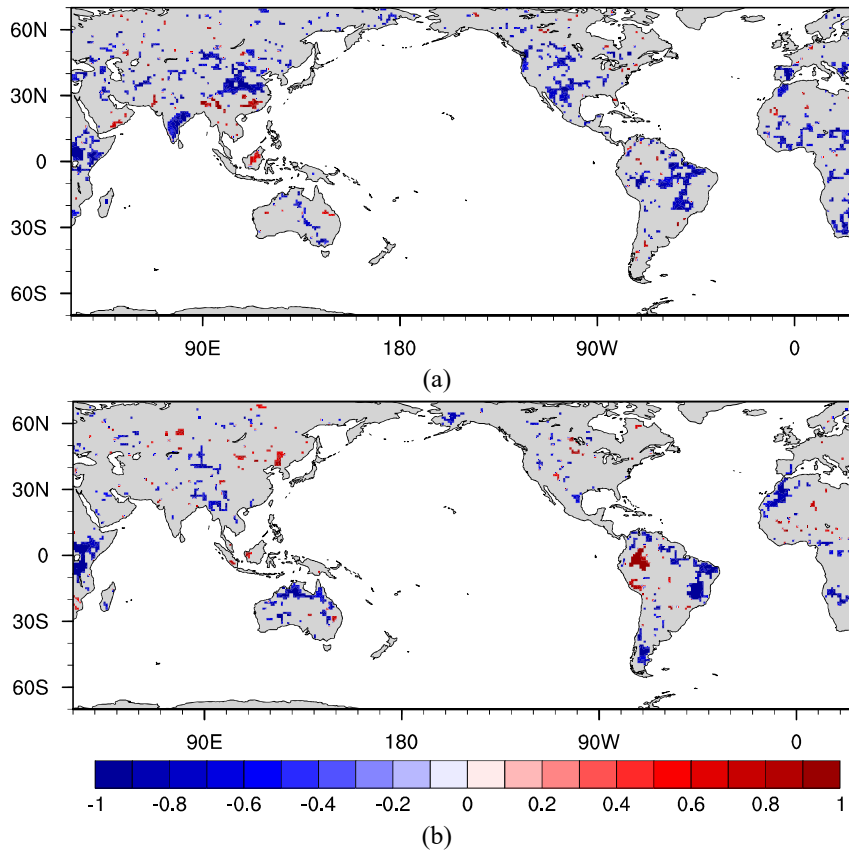


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282 Figure 3: The correlation coefficient (R) between Evaporative Stress Index and ESA CCI soil moisture anomaly
 283 composites across the MJO phases at each grid point: (a) boreal summer and (b) boreal winter. Only areas with
 284 statistically significant results ($p < 0.05$) are shown.
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289 Figure 4: The correlation coefficient (R) between Evaporative Stress Index and ERA Interim air temperature
 290 anomaly composites across the MJO phases at each grid point: (a) boreal summer and (b) boreal winter. Only areas with
 291 statistically significant results ($p < 0.05$) are shown.
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299 Furthermore, the average correlation coefficient values between dry/wet conditions and
 300 precipitation, soil moisture and air temperature over the whole globe and regions with strong
 301 MJO impacts are shown in Table 1. These selected regions are South Asia (5°N-27°N, 70°E-
 302 90°E), east part of South America (5°S-40°S, 35°W-60°W), Australia (11°S-45°S, 110°E-
 303 155°E), and East Africa (10°N-30°S, 32°E-56°E). Compared to the R values over the entire
 304 globe, the R values are increased over these sub-regions, further indicating that the strong MJO

305 impacts on these areas. The R values between dryness/wetness and soil moisture are higher than
306 0.699 for all these regions and slightly better than that for precipitation and air temperature,
307 which agrees well with the characteristics of ESI. Overall, the results presented here suggest that
308 the MJO has impacts on intraseasonal variations of both soil moisture and precipitation. The
309 variation of soil moisture is partly caused by its relation to precipitation, because the R value
310 between soil moisture and precipitation was found to vary over MJO phases, and the soil
311 moisture is also related to other climate variables such as temperature and evaporation. Overall,
312 the results mentioned above clearly show that the MJO has impacts on the extreme dry and wet
313 conditions of the globe particularly over areas with strong MJO signals. The variation of this
314 dryness/wetness across the MJO phases are directly related to precipitation, soil moisture and air
315 temperature. The results suggest that the MJO impacts should be taken into account in the
316 investigation of dryness/wetness around the world particularly over monsoon areas. It should
317 also be noted that there are uncertainties associated with satellite-based products such as GPCP
318 precipitation, ESA CCI soil moisture and GLEAM ET/PET datasets. For studies that aiming to
319 quantify the impacts of MJO on dry and wet conditions over regional scale, the uncertainties of
320 satellite-based products should be taken into account.

321

322 Table 1. Averaged correlation between dryness/wetness and precipitation, soil moisture, and air
 323 temperature across MJO phases for Globe and selected regions.

		Globe	South Asia	South America	East Africa	Australia
Precipitation	R (Summer)	0.376	0.733	0.557	0.791	0.608
	R (Winter)	0.279	0.547	0.454	0.776	0.677
Soil moisture	R (Summer)	0.473	0.791	0.699	0.788	0.78
	R (Winter)	0.466	0.711	0.725	0.808	0.719
Air temperature	R (Summer)	-0.415	-0.622	-0.502	-0.61	-0.257
	R (Winter)	-0.246	-0.088	-0.508	-0.576	-0.6

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326 4 Conclusions

327 To the best of our knowledge, this study is the first to reveal the impacts of MJO on the
 328 global extreme dry and wet conditions based on observations. All previous works have focused
 329 on either extreme precipitation or droughts over specific geographic locations. Given the
 330 planetary scale of the MJO, the current study first developed an observational analysis to obtain a
 331 global view of the relation between the MJO and global dryness and wetness that were
 332 represented by ESI drought index. It is found that MJO has nearly global influences on the
 333 extreme dry and wet conditions particularly over monsoon regions (e.g. South Asia, South
 334 America and East Africa) via the effects of MJO on precipitation, soil moisture and air
 335 temperature, which are induced by modifying the meridional overturning circulations, and
 336 exciting Rossby wave trains caused by heating anomalies, and moisture transport, and forcing
 337 zonally-propagating equatorial Rossby and Kelvin waves. The results suggest that the impacts of
 338 MJO should be accounted for in studies of droughts and floods in areas where MJO has strong

339 impacts such as South Asia. In addition, the findings of the present study provide the basis for
340 drought analysis and MJO-based monitoring and prediction systems, which would support
341 decision-making in climate sensitive sectors such as drought monitoring and agricultural
342 management. However, it should be noted that the current study only provides an observational
343 analysis of the impacts of MJO on global extreme dry and wet conditions. The applied satellite-
344 based products such as GLEAM ET/PET, ESA CCI soil moisture still have uncertainties, which
345 should be accounted for in future studies to quantify their impacts on related dryness and
346 wetness.

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349 Range Weather Forecasts (ECMWF) and Integrated Climate Data Centre (ICDC) for providing
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354 **References**

- 355 Anderson, M.C., Norman, J.M., Mecikalski, J.R., Otkin, J.A., & Kustas, W.P. (2007). A climatological study of
356 evapotranspiration and moisture stress across the continental United States based on thermal remote sensing: 2.
357 Surface moisture climatology. *Journal of Geophysical Research: Atmospheres*, 112, n/a-n/a
- 358 Anderson, M.C., Zolin, C.A., Hain, C.R., Semmens, K., Yilmaz, M.T., & Gao, F. (2015). Comparison of satellite-
359 derived LAI and precipitation anomalies over Brazil with a thermal infrared-based Evaporative Stress Index for
360 2003–2013. *Journal of Hydrology*, 526, 287-302
- 361 Anderson, M.C., Zolin, C.A., Sentelhas, P.C., Hain, C.R., Semmens, K., Tugrul Yilmaz, M., Gao, F., Otkin, J.A., &
362 Tetrault, R. (2016). The Evaporative Stress Index as an indicator of agricultural drought in Brazil: An assessment
363 based on crop yield impacts. *Remote Sensing of Environment*, 174, 82-99
- 364 Barlow, M., Nigam, S., & Berbery, E.H. (2001). ENSO, Pacific decadal variability, and US summertime
365 precipitation, drought, and stream flow. *Journal of Climate*, 14, 2105-2128
- 366 Betts, A.K., Köhler, M., & Zhang, Y. (2009). Comparison of river basin hydrometeorology in ERA-Interim and
367 ERA-40 reanalyses with observations. *Journal of Geophysical Research: Atmospheres*, 114
- 368 Cayan, D.R., Dettinger, M.D., Diaz, H.F., & Graham, N.E. (1998). Decadal variability of precipitation over western
369 North America. *Journal of Climate*, 11, 3148-3166

- 370 Choi, M., Jacobs, J.M., Anderson, M.C., & Bosch, D.D. (2013). Evaluation of drought indices via remotely sensed
371 data with hydrological variables. *Journal of Hydrology*, 476, 265-273
- 372 Crueger, T., Stevens, B., & Brokopf, R. (2013). The Madden–Julian Oscillation in ECHAM6 and the Introduction of
373 an Objective MJO Metric. *Journal of Climate*, 26, 3241-3257
- 374 Dai, A. (2013). Increasing drought under global warming in observations and models. *Nature Climate Change*, 3, 52
- 375 Dee, D.P., Uppala, S.M., Simmons, A.J., Berrisford, P., Poli, P., Kobayashi, S., Andrae, U., Balmaseda, M.A.,
376 Balsamo, G., Bauer, P., Bechtold, P., Beljaars, A.C.M., van de Berg, L., Bidlot, J., Bormann, N., Delsol, C.,
377 Dragani, R., Fuentes, M., Geer, A.J., Haimberger, L., Healy, S.B., Hersbach, H., Hólm, E.V., Isaksen, L., Kållberg,
378 P., Köhler, M., Matricardi, M., McNally, A.P., Monge-Sanz, B.M., Morcrette, J.J., Park, B.K., Peubey, C., de
379 Rosnay, P., Tavolato, C., Thépaut, J.N., & Vitart, F. (2011). The ERA-Interim reanalysis: configuration and
380 performance of the data assimilation system. *Quarterly Journal of the Royal Meteorological Society*, 137, 553-597
- 381 Dettinger, M. (2011). Climate change, atmospheric rivers, and floods in California—a multimodel analysis of storm
382 frequency and magnitude changes 1. *JAWRA Journal of the American Water Resources Association*, 47, 514-523
- 383 Donald, A., Meinke, H., Power, B., Maia, A.d.H., Wheeler, M.C., White, N., Stone, R.C., & Ribbe, J. (2006).
384 Near-global impact of the Madden-Julian Oscillation on rainfall. *Geophysical Research Letters*, 33
- 385 Dorigo, W., Gruber, A., De Jeu, R., Wagner, W., Stacke, T., Loew, A., Albergel, C., Brocca, L., Chung, D., &
386 Parinussa, R. (2015a). Evaluation of the ESA CCI soil moisture product using ground-based observations. *Remote
387 Sensing of Environment*, 162, 380-395
- 388 Dorigo, W.A., Gruber, A., De Jeu, R.A.M., Wagner, W., Stacke, T., Loew, A., Albergel, C., Brocca, L., Chung, D.,
389 Parinussa, R.M., & Kidd, R. (2015b). Evaluation of the ESA CCI soil moisture product using ground-based
390 observations. *Remote Sensing of Environment*, 162, 380-395
- 391 Easterling, D.R., Meehl, G.A., Parmesan, C., Changnon, S.A., Karl, T.R., & Mearns, L.O. (2000). Climate extremes:
392 observations, modeling, and impacts. *Science*, 289, 2068-2074
- 393 Forzieri, G., Alkama, R., Miralles, D.G., & Cescatti, A. (2017). Satellites reveal contrasting responses of regional
394 climate to the widespread greening of Earth. *Science*, 356, 1180-1184
- 395 Frank, W.M., & Roundy, P.E. (2006). The Role of Tropical Waves in Tropical Cyclogenesis. *Monthly Weather
396 Review*, 134, 2397-2417
- 397 Frei, C., Schöll, R., Fukutome, S., Schmidli, J., & Vidale, P.L. (2006). Future change of precipitation extremes in
398 Europe: Intercomparison of scenarios from regional climate models. *Journal of Geophysical Research:
399 Atmospheres*, 111
- 400 Greve, P., Orłowsky, B., Mueller, B., Sheffield, J., Reichstein, M., & Seneviratne, S.I. (2014). Global assessment of
401 trends in wetting and drying over land. *Nature geoscience*, 7, 716
- 402 Han, W. (2005). Origins and Dynamics of the 90-Day and 30–60-Day Variations in the Equatorial Indian Ocean.
403 *Journal of Physical Oceanography*, 35, 708-728
- 404 He, J., Lin, H., & Wu, Z. (2011). Another look at influences of the Madden-Julian Oscillation on the wintertime
405 East Asian weather. *Journal of Geophysical Research: Atmospheres*, 116
- 406 Hu, Y., Li, D., & Liu, J. (2007). Abrupt seasonal variation of the ITCZ and the Hadley circulation. *Geophysical
407 Research Letters*, 34

- 408 Huffman, G.J., Adler, R.F., Morrissey, M.M., Bolvin, D.T., Curtis, S., Joyce, R., McGavock, B., & Susskind, J.
409 (2001). Global Precipitation at One-Degree Daily Resolution from Multisatellite Observations. *Journal of*
410 *Hydrometeorology*, 2, 36-50
- 411 Inness, P.M., & Slingo, J.M. (2003). Simulation of the Madden–Julian oscillation in a coupled general circulation
412 model. Part I: Comparison with observations and an atmosphere-only GCM. *Journal of Climate*, 16, 345-364
- 413 IPCC (2013). *Climate change 2013: the physical science basis: Working Group I contribution to the Fifth*
414 *assessment report of the Intergovernmental Panel on Climate Change*. Cambridge University Press
- 415 Janicot, S., Mounier, F., Hall, N.M., Leroux, S., Sultan, B., & Kiladis, G.N. (2009). Dynamics of the West African
416 monsoon. Part IV: Analysis of 25–90-day variability of convection and the role of the Indian monsoon. *Journal of*
417 *Climate*, 22, 1541-1565
- 418 Jia, X., Chen, L., Ren, F., & Li, C. (2011). Impacts of the MJO on winter rainfall and circulation in China. *Advances*
419 *in Atmospheric Sciences*, 28, 521-533
- 420 Jones, C., Waliser, D.E., Lau, K., & Stern, W. (2004a). Global occurrences of extreme precipitation and the
421 Madden-Julian Oscillation: Observations and predictability. *Journal of Climate*, 17, 4575-4589
- 422 Jones, C., Waliser, D.E., Lau, K., & Stern, W. (2004b). Global occurrences of extreme precipitation and the
423 Madden–Julian oscillation: Observations and predictability. *Journal of Climate*, 17, 4575-4589
- 424 Joseph, S., Sahai, A.K., & Goswami, B.N. (2009). Eastward propagating MJO during boreal summer and Indian
425 monsoon droughts. *Climate Dynamics*, 32, 1139-1153
- 426 Kang, I.-S., & Kim, H.-M. (2010). Assessment of MJO Predictability for Boreal Winter with Various Statistical and
427 Dynamical Models. *Journal of Climate*, 23, 2368-2378
- 428 Kenyon, J., & Hegerl, G.C. (2010). Influence of modes of climate variability on global precipitation extremes.
429 *Journal of Climate*, 23, 6248-6262
- 430 Kim, D., Kim, H., & Lee, M.-I. (2017). Why does the MJO detour the Maritime Continent during austral summer?
431 *Geophysical Research Letters*, 44, 2579-2587
- 432 Knutti, R., & Sedláček, J. (2013). Robustness and uncertainties in the new CMIP5 climate model projections. *Nature*
433 *Climate Change*, 3, 369
- 434 Leng, P., Li, Z.L., Duan, S.B., Tang, R., & Gao, M.F. (2017). A method for deriving all-sky evapotranspiration from
435 the synergistic use of remotely sensed images and meteorological data. *Journal of Geophysical Research:*
436 *Atmospheres*
- 437 Li, W., Guo, W., Hsu, P.-c., & Xue, Y. (2016). Influence of the Madden–Julian oscillation on Tibetan Plateau snow
438 cover at the intraseasonal time-scale. *Scientific Reports*, 6, 30456
- 439 Lian, X., Piao, S., Huntingford, C., Li, Y., Zeng, Z., Wang, X., Ciais, P., McVicar, T.R., Peng, S., Ottlé, C., Yang,
440 H., Yang, Y., Zhang, Y., & Wang, T. (2018). Partitioning global land evapotranspiration using CMIP5 models
441 constrained by observations. *Nature Climate Change*, 8, 640-646
- 442 Liu, M., Xu, X., Xu, C., Sun, A.Y., Wang, K., Scanlon, B.R., & Zhang, L. (2017a). A new drought index that
443 considers the joint effects of climate and land surface change. *Water Resources Research*, 53, 3262-3278

- 444 Liu, X., Wu, T., Yang, S., Li, T., Jie, W., Zhang, L., Wang, Z., Liang, X., Li, Q., Cheng, Y., Ren, H., Fang, Y., &
 445 Nie, S. (2017b). MJO prediction using the sub-seasonal to seasonal forecast model of Beijing Climate Center.
 446 *Climate Dynamics*, *48*, 3283-3307
- 447 Liu, Y.Y., Parinussa, R.M., Dorigo, W.A., De Jeu, R.A.M., Wagner, W., van Dijk, A.I.J.M., McCabe, M.F., &
 448 Evans, J.P. (2011). Developing an improved soil moisture dataset by blending passive and active microwave
 449 satellite-based retrievals. *Hydrology and Earth System Sciences*, *15*, 425-436
- 450 Lü, J., Ju, J., Ren, J., & Gan, W. (2012). The influence of the Madden-Julian Oscillation activity anomalies on
 451 Yunnan's extreme drought of 2009–2010. *Science China Earth Sciences*, *55*, 98-112
- 452 Madden, R.A., & Julian, P.R. (1971). Detection of a 40–50 Day Oscillation in the Zonal Wind in the Tropical
 453 Pacific. *Journal of the Atmospheric Sciences*, *28*, 702-708
- 454 Martens, B., Miralles, D.G., Lievens, H., van der Schalie, R., de Jeu, R.A.M., Fernández-Prieto, D., Beck, H.E.,
 455 Dorigo, W.A., & Verhoest, N.E.C. (2017). GLEAM v3: satellite-based land evaporation and root-zone soil moisture.
 456 *Geosci. Model Dev.*, *10*, 1903-1925
- 457 Matsueda, S., & Takaya, Y. (2015). The global influence of the Madden–Julian oscillation on extreme temperature
 458 events. *Journal of Climate*, *28*, 4141-4151
- 459 Matthews, A.J., Hoskins, B.J., & Masutani, M. (2004). The global response to tropical heating in the Madden–Julian
 460 oscillation during the northern winter. *Quarterly Journal of the Royal Meteorological Society*, *130*, 1991-2011
- 461 Meehl, G.A., Zwiers, F., Evans, J., Knutson, T., Mearns, L., & Whetton, P. (2000). Trends in extreme weather and
 462 climate events: issues related to modeling extremes in projections of future climate change. *Bulletin of the American
 463 Meteorological Society*, *81*, 427-436
- 464 Miralles, D.G., Holmes, T.R.H., De Jeu, R.A.M., Gash, J.H., Meesters, A.G.C.A., & Dolman, A.J. (2011). Global
 465 land-surface evaporation estimated from satellite-based observations. *Hydrol. Earth Syst. Sci.*, *15*, 453-469
- 466 Miralles, D.G., Van Den Berg, M.J., Gash, J.H., Parinussa, R.M., De Jeu, R.A., Beck, H.E., Holmes, T.R., Jiménez,
 467 C., Verhoest, N.E., & Dorigo, W.A. (2014). El Niño–La Niña cycle and recent trends in continental evaporation.
 468 *Nature Climate Change*, *4*, 122
- 469 Neena, J.M., Suhas, E., & Goswami, B.N. (2011). Leading role of internal dynamics in the 2009 Indian summer
 470 monsoon drought. *Journal of Geophysical Research: Atmospheres*, *116*
- 471 Nicolai-Shaw, N., Zscheischler, J., Hirschi, M., Gudmundsson, L., & Seneviratne, S.I. (2017). A drought event
 472 composite analysis using satellite remote-sensing based soil moisture. *Remote Sensing of Environment*, *203*, 216-
 473 225
- 474 Otkin, J.A., Anderson, M.C., Hain, C., & Svoboda, M. (2013). Examining the Relationship between Drought
 475 Development and Rapid Changes in the Evaporative Stress Index. *Journal of Hydrometeorology*, *15*, 938-956
- 476 Otkin, J.A., Anderson, M.C., Hain, C., & Svoboda, M. (2014). Examining the relationship between drought
 477 development and rapid changes in the evaporative stress index. *Journal of Hydrometeorology*, *15*, 938-956
- 478 Pai, D.S., Bhate, J., Sreejith, O.P., & Hatwar, H.R. (2011). Impact of MJO on the intraseasonal variation of summer
 479 monsoon rainfall over India. *Climate Dynamics*, *36*, 41-55
- 480 Peng, J., & Loew, A. (2017). Recent Advances in Soil Moisture Estimation from Remote Sensing. *Water*, *9*, 530

- 481 Peng, J., Loew, A., & Crueger, T. (2017a). The relationship between the Madden-Julian oscillation and the land
482 surface soil moisture. *Remote Sensing of Environment*, *203*, 226-239
- 483 Peng, J., Loew, A., Merlin, O., & Verhoest, N.E. (2017d). A review of spatial downscaling of satellite remotely
484 sensed soil moisture. *Reviews of Geophysics*, *55*, 341-366
- 485 Peng, J., Loew, A., Zhang, S., Wang, J., & Niesel, J. (2016). Spatial Downscaling of Satellite Soil Moisture Data
486 Using a Vegetation Temperature Condition Index. *IEEE Transactions on Geoscience and Remote Sensing*, *54*, 558-
487 566
- 488 Peng, J., Niesel, J., Loew, A., Zhang, S., & Wang, J. (2015). Evaluation of Satellite and Reanalysis Soil Moisture
489 Products over Southwest China Using Ground-Based Measurements. *Remote Sensing*, *7*, 15729
- 490 Pohl, B., & Camberlin, P. (2006). Influence of the Madden–Julian oscillation on East African rainfall. I:
491 Intraseasonal variability and regional dependency. *Quarterly Journal of the Royal Meteorological Society*, *132*,
492 2521-2539
- 493 Pohl, B., & Matthews, A.J. (2007). Observed Changes in the Lifetime and Amplitude of the Madden–Julian
494 Oscillation Associated with Interannual ENSO Sea Surface Temperature Anomalies. *Journal of Climate*, *20*, 2659-
495 2674
- 496 Priestley, C.H.B., & Taylor, R.J. (1972). On the Assessment of Surface Heat Flux and Evaporation Using Large-
497 Scale Parameters. *Monthly Weather Review*, *100*, 81-92
- 498 Reid, J.S., Xian, P., Hyer, E.J., Flatau, M.K., Ramirez, E.M., Turk, F.J., Sampson, C.R., Zhang, C., Fukada, E.M., &
499 Maloney, E.D. (2012). Multi-scale meteorological conceptual analysis of observed active fire hotspot activity and
500 smoke optical depth in the Maritime Continent. *Atmos. Chem. Phys.*, *12*, 2117-2147
- 501 Richard, W., Sonia, I.S., Martin, H., Jinfeng, C., Philippe, C., Delphine, D., Joshua, E., Christian, F., Simon, N.G.,
502 Lukas, G., Alexandra-Jane, H., Thomas, H., Akihiko, I., Nikolay, K., Hyungjun, K., Guoyong, L., Junguo, L.,
503 Xingcai, L., Yoshimitsu, M., Catherine, M., Christoph, M., Hannes Müller, S., Kazuya, N., Rene, O., Yadu, P.,
504 Thomas, A.M.P., Yusuke, S., Sibyll, S., Erwin, S., Justin, S., Tobias, S., Joerg, S., Qihong, T., Wim, T., Yoshihide,
505 W., Xuhui, W., Graham, P.W., Hong, Y., & Tian, Z. (2018). Evapotranspiration simulations in ISIMIP2a—
506 Evaluation of spatio-temporal characteristics with a comprehensive ensemble of independent datasets.
507 *Environmental Research Letters*, *13*, 075001
- 508 Robertson, A.W., Camargo, S.J., Sobel, A., Vitart, F., & Wang, S. (2018). Summary of workshop on sub-seasonal to
509 seasonal predictability of extreme weather and climate. *npj Climate and Atmospheric Science*, *1*, 20178
- 510 Seager, R., Ting, M., Li, C., Naik, N., Cook, B., Nakamura, J., & Liu, H. (2013). Projections of declining surface-
511 water availability for the southwestern United States. *Nature Clim. Change*, *3*, 482-486
- 512 Sheffield, J., Andreadis, K.M., Wood, E.F., & Lettenmaier, D.P. (2009). Global and Continental Drought in the
513 Second Half of the Twentieth Century: Severity–Area–Duration Analysis and Temporal Variability of Large-Scale
514 Events. *Journal of Climate*, *22*, 1962-1981
- 515 Shimizu, M.H., Ambrizzi, T., & Liebmann, B. (2017). Extreme precipitation events and their relationship with
516 ENSO and MJO phases over northern South America. *International Journal of Climatology*, *37*, 2977-2989
- 517 Sillmann, J., Kharin, V., Zwiers, F., Zhang, X., & Bronaugh, D. (2013). Climate extremes indices in the CMIP5
518 multimodel ensemble: Part 2. Future climate projections. *Journal of Geophysical Research: Atmospheres*, *118*,
519 2473-2493

- 520 Sillmann, J., Thorarinsdottir, T., Keenlyside, N., Schaller, N., Alexander, L.V., Hegerl, G., Seneviratne, S.I.,
 521 Vautard, R., Zhang, X., & Zwiers, F.W. (2017). Understanding, modeling and predicting weather and climate
 522 extremes: Challenges and opportunities. *Weather and Climate Extremes*, 18, 65-74
- 523 Sun, Q., Miao, C., AghaKouchak, A., & Duan, Q. (2016a). Century-scale causal relationships between global
 524 dry/wet conditions and the state of the Pacific and Atlantic Oceans. *Geophysical Research Letters*, 43, 6528-6537
- 525 Sun, S., Chen, H., Ju, W., Yu, M., Hua, W., & Yin, Y. (2014). On the attribution of the changing hydrological cycle
 526 in Poyang Lake Basin, China. *Journal of Hydrology*, 514, 214-225
- 527 Sun, S., Chen, H., Wang, G., Li, J., Mu, M., Yan, G., Xu, B., Huang, J., Wang, J., & Zhang, F. (2016c). Shift in
 528 potential evapotranspiration and its implications for dryness/wetness over Southwest China. *Journal of Geophysical
 529 Research: Atmospheres*, 121, 9342-9355
- 530 Sylla, M., Giorgi, F., Coppola, E., & Mariotti, L. (2013). Uncertainties in daily rainfall over Africa: assessment of
 531 gridded observation products and evaluation of a regional climate model simulation. *International Journal of
 532 Climatology*, 33, 1805-1817
- 533 Szczypta, C., Calvet, J.C., Albergel, C., Balsamo, G., Boussetta, S., Carrer, D., Lafont, S., & Meurey, C. (2011).
 534 Verification of the new ECMWF ERA-Interim reanalysis over France. *Hydrology and Earth System Sciences*, 15,
 535 647-666
- 536 Tan, M.L., Ibrahim, A.L., Cracknell, A.P., & Yusop, Z. (2017). Changes in precipitation extremes over the Kelantan
 537 River Basin, Malaysia. *International Journal of Climatology*, 37, 3780-3797
- 538 Thompson, D.B., & Roundy, P.E. (2013). The Relationship between the Madden–Julian Oscillation and U.S.
 539 Violent Tornado Outbreaks in the Spring. *Monthly Weather Review*, 141, 2087-2095
- 540 Trenberth, K.E., Cheng, L., Jacobs, P., Zhang, Y., & Fasullo, J. (2018). Hurricane Harvey Links to Ocean Heat
 541 Content and Climate Change Adaptation. *Earth's Future*, 6, 730-744
- 542 Trenberth, K.E., & Shea, D.J. (2005). Relationships between precipitation and surface temperature. *Geophysical
 543 Research Letters*, 32
- 544 Vecchi, G.A., & Bond, N.A. (2004). The Madden-Julian Oscillation (MJO) and northern high latitude wintertime
 545 surface air temperatures. *Geophysical Research Letters*, 31
- 546 Vicente-Serrano, S.M., Miralles, D.G., Domínguez-Castro, F., Azorin-Molina, C., Kenawy, A.E., McVicar, T.R.,
 547 Tomás-Burguera, M., Beguería, S., Maneta, M., & Peña-Gallardo, M. (2018). Global Assessment of the
 548 Standardized Evapotranspiration Deficit Index (SEDI) for Drought Analysis and Monitoring. *Journal of Climate*,
 549 31, 5371-5393
- 550 Waliser, D., Lau, K., Stern, W., & Jones, C. (2003). Potential predictability of the Madden–Julian oscillation.
 551 *Bulletin of the American Meteorological Society*, 84, 33-50
- 552 Waliser, D., Sperber, K., Hendon, H., Kim, D., Maloney, E., Wheeler, M., Weickmann, K., Zhang, C., Donner, L.,
 553 Gottschalck, J., Higgins, W., Kang, I.-S., Legler, D., Moncrieff, M., Schubert, S., Stern, W., Vitart, F., Wang, B.,
 554 Wang, W., & Woolnough, S. (2009). MJO Simulation Diagnostics. *Journal of Climate*, 22, 3006-3030
- 555 Weisheimer, A., Schaller, N., O'Reilly, C., MacLeod, D.A., & Palmer, T. (2017). Atmospheric seasonal forecasts of
 556 the twentieth century: multi-decadal variability in predictive skill of the winter North Atlantic Oscillation (NAO)
 557 and their potential value for extreme event attribution. *Quarterly Journal of the Royal Meteorological Society*, 143,
 558 917-926

- 559 Wheeler, M.C., & Hendon, H.H. (2004). An All-Season Real-Time Multivariate MJO Index: Development of an
560 Index for Monitoring and Prediction. *Monthly Weather Review*, 132, 1917-1932
- 561 Xavier, P., Rahmat, R., Cheong, W.K., & Wallace, E. (2014). Influence of Madden-Julian Oscillation on Southeast
562 Asia rainfall extremes: Observations and predictability. *Geophysical Research Letters*, 41, 4406-4412
- 563 Yuan, X., Ma, Z., Pan, M., & Shi, C. (2015). Microwave remote sensing of short-term droughts during crop growing
564 seasons. *Geophysical Research Letters*, 42, 4394-4401
- 565 Zhang, B., AghaKouchak, A., Yang, Y., Wei, J., & Wang, G. (2019). A water-energy balance approach for multi-
566 category drought assessment across globally diverse hydrological basins. *Agricultural and forest meteorology*, 264,
567 247-265
- 568 Zhang, B., & He, C. (2016). A modified water demand estimation method for drought identification over arid and
569 semiarid regions. *Agricultural and forest meteorology*, 230-231, 58-66
- 570 Zhang, B., Zhao, X., Jin, J., & Wu, P. (2015). Development and evaluation of a physically based multiscale drought
571 index: The Standardized Moisture Anomaly Index. *Journal of Geophysical Research: Atmospheres*, 120, 11,575-
572 511,588
- 573 Zhang, C. (2005). Madden-Julian Oscillation. *Reviews of Geophysics*, 43
- 574 Zhang, C. (2013). Madden–Julian Oscillation: Bridging Weather and Climate. *Bulletin of the American
575 Meteorological Society*, 94, 1849-1870
- 576 Zhang, C., & Dong, M. (2004). Seasonality in the Madden–Julian Oscillation. *Journal of Climate*, 17, 3169-3180
- 577 Zhang, X., Wang, J., Zwiers, F.W., & Groisman, P.Y. (2010). The influence of large-scale climate variability on
578 winter maximum daily precipitation over North America. *Journal of Climate*, 23, 2902-2915
- 579 Zheng, C., Kar-Man Chang, E., Kim, H.-M., Zhang, M., & Wang, W. (2018). Impacts of the Madden–Julian
580 Oscillation on Storm-Track Activity, Surface Air Temperature, and Precipitation over North America. *Journal of
581 Climate*, 31, 6113-6134
- 582 Zhou, S., L’Heureux, M., Weaver, S., & Kumar, A. (2012). A composite study of the MJO influence on the surface
583 air temperature and precipitation over the continental United States. *Climate Dynamics*, 38, 1459-1471
- 584 Zhou, Y., Lu, Y., Yang, B., Jiang, J., Huang, A., Zhao, Y., La, M., & Yang, Q. (2016). On the relationship between
585 the Madden-Julian Oscillation and 2 m air temperature over central Asia in boreal winter. *Journal of Geophysical
586 Research: Atmospheres*, 121, 13,250-213,272
- 587