

1 Inter-model variations in projected precipitation
2 change over the North Atlantic: Sea surface
3 temperature effect
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ABSTRACT

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16 Inter-model variations in future precipitation projection in the North Atlantic are
17 studied using 23 state-of-art models from Phase 5 of the Coupled Model Intercomparison
18 Project. Model uncertainty in annual-mean rainfall change is locally enhanced along the
19 Gulf Stream. The moisture budget analysis reveals that much of the model uncertainty in
20 rainfall change can be traced back to the discrepancies in surface evaporation change and
21 transient eddy effect among models. Results of the inter-model Singular Value
22 Decomposition (SVD) analysis show that inter-model variations in local sea surface
23 temperature (SST) pattern exert a strong control over the spread of rainfall projection
24 among models through the modulation of evaporation change. The first three SVD modes
25 explain more than 60% of the inter-model variance of rainfall projection and show
26 distinct SST patterns with mode-water-induced banded structures, reduced subpolar
27 warming due to ocean dynamical cooling and the Gulf Stream shift, respectively.

28 **1. Introduction**

29 Precipitation change under global warming is of great importance for society.
30 Achieving reliable projection of regional rainfall change remains a great challenge for
31 climate science since the sign and amplitude of precipitation change vary spatially [Ma
32 and Xie, 2013]. Uncertainty in future rainfall projection mainly derives from three
33 sources: radiative forcing, model uncertainty and internal variability. Among these three
34 sources, model uncertainty is dominant specifically for longer-term projections [Hawkins
35 and Sutton, 2011]. Model uncertainty in rainfall projection remains large in Phase 5 of
36 the Coupled Model Intercomparison Project (CMIP5) [Taylor et al., 2012], similar to that
37 in CMIP3 [Knutti and Sedlacek, 2013]. It is therefore essential to understand the physical
38 mechanism for the model uncertainty.

39 In the tropics, precipitation changes mainly follow the sea surface temperature (SST)
40 warming pattern [Xie et al., 2010], as a result of the offset between the wet-get-wetter
41 pattern and tropical circulation slowdown [Seager et al., 2010; Chadwick et al., 2013].
42 The SST warming pattern effect is apparent in El Niño-induced atmospheric anomalies
43 both in the tropics and extratropics [Zhou et al., 2014]. Furthermore, the inter-model
44 spread of SST warming pattern is important for both the inter-model divergence of
45 tropical precipitation change and circulation change [Ma and Xie, 2013].

46 Different from the tropical ocean where the mean-circulation-induced convergence
47 accounts for most of the precipitation distribution, rainfall in the midlatitudes is more
48 complicated, involving weather phenomena, strong influence of the SST front and large-
49 scale moisture advection. Transient eddies are important for precipitation, especially
50 along storm tracks [Hoskins and Valdes, 1990] in the boreal winter. A reduction in the

51 Meridional Overturning Circulation is associated with a substantial SST cooling over the
52 North Atlantic [Rahmstorf et al., 2015]. This SST pattern increases the meridional SST
53 gradient and baroclinic instability and hence strengthens the local storm track (Woollings
54 et al. 2012). The Gulf Stream transports a large amount of heat to the midlatitudes,
55 forming a long and narrow SST front that anchors a band of heavy rainfall and strong
56 evaporation [Yu, 2007]. The SST front effect is also apparent on synoptic eddies over the
57 North Atlantic [Kwon and Joyce, 2013]. The warm water transported by the Gulf Stream
58 from the tropics supplies much of the water vapor for precipitation via evaporation,
59 resulting in a close relationship between precipitation and evaporation in space. Large-
60 scale moisture advection peaks in the winter, dries the subtropical North Atlantic and
61 moistens the midlatitudes across the horizontal humidity gradient [Seager et al., 2010].
62 Furthermore, ocean heat transport associated with mode water dynamics [Xie et al., 2010;
63 Xie et al., 2011; Xu et al., 2012] is important for the formation of the midlatitudes SST
64 warming pattern over the North Pacific, forming banded structures in the subtropics [Xie
65 et al., 2010; Long et al., 2014]. Extratropical precipitation change is very similar between
66 Atmospheric General Circulation Model (AGCM) simulations forced with spatially
67 uniform and patterned SST warming [He et al., 2014]. The multi-model ensemble-mean
68 SST warming pattern they used under-estimates the spatial variations, especially over the
69 extratropical North Atlantic where the inter-model differences in SST climatology and
70 warming pattern are large (Fig. 1c). We show that the inter-model spread in SST pattern
71 explains much of the inter-model variations in precipitation change.

72 The present study examines the sources and mechanism of inter-model spread in
73 precipitation projection in the North Atlantic, based on 23 CMIP5 model projections

74 (Table S1 in the supporting information). We show that model uncertainty in annual-
75 mean rainfall change is locally enhanced along the Gulf Stream. Our moisture budget
76 analysis reveals that the uncertainty mainly originates from the inter-model discrepancies
77 in evaporation change and transient eddy effect. This is different from the tropical ocean
78 case where the changes in mean convergence dominate the spread of rainfall change
79 among models. The effect of local SST warming pattern on model uncertainty in rainfall
80 projection is examined with the inter-model Singular Value Decomposition (SVD)
81 analyses.

82 The rest of the paper is organized as follows. Section 2 describes the data and
83 methods. Section 3 discusses the sources of model uncertainty in annual-mean
84 precipitation projection. Section 4 investigates the role of local SST change in the
85 discrepancy of annual-mean rainfall change among models and extends the analysis to
86 the boreal winter and boreal summer. Section 5 is a summary.

87 **2. Data and Methods**

88 The monthly outputs of preindustrial control (piControl) runs, historical simulations
89 (1850-2005) and Representative Concentration Pathway 4.5 (RCP4.5, 2006-2100) runs in
90 23 CMIP5 models are analyzed. Future climate change (denote as δ/Δ) is calculated by
91 subtracting the 50-year mean of 1950-1999 (present climatology) in historical simulation
92 from the 2050-2099 mean (RCP4.5 climatology) in the RCP4.5 run and then normalized
93 by the domain mean ($80^{\circ}\text{W}-0^{\circ}$, $20^{\circ}\text{N}-60^{\circ}\text{N}$) SST warming in each model to highlight the
94 uncertainty in spatial pattern. Internal variability causes uncertainty in projections of
95 regional climate in the midlatitudes [Deser et al., 2012] and contributes to the total model
96 uncertainty. To evaluate the contribution from internal variability, we first calculate 100-

97 year rainfall trends for every 50 years based on the 50-year running mean time series of
 98 piControl run in each model. Then one trend is randomly selected per model and used to
 99 calculate the inter-model standard deviation at each grid point. To obtain robust results,
 100 we repeat the random selection and standard deviation calculation 100 times and average
 101 all the resultant inter-model standard deviations as the model uncertainty induced by
 102 internal variability. All model outputs are interpolated onto a common grid of 2.5°
 103 latitude × 2.5° longitude. Only one member run (r1i1p1) per model is analyzed to ensure
 104 equal weight for each model. Note that the near-surface specific humidity in 2 models
 105 and wind speed in 4 models are not available (see Table S1).

106 The moisture budget derived from the water vapor conservation equation for
 107 monthly time average is [Trenberth and Guillemot, 1995; Seager et al., 2010]:

$$108 \quad \rho_w g(P - E) = - \int_0^{p_s} (\mathbf{u} \cdot \nabla q) dp - \int_0^{p_s} (q \nabla \cdot \mathbf{u}) dp + residual. \quad (1)$$

109 Here P is precipitation, E is evaporation, ρ_w is the density of water, q is specific
 110 humidity, \mathbf{u} is the horizontal vector wind, p is pressure and the subscript s denotes the
 111 surface value. The first term on the right-hand is moisture advection by the monthly mean
 112 circulation and the second term is the wind convergence term and the residual is largely
 113 due to transient eddy effect.

114 For climate change, we neglect the small nonlinear terms. Equation (1) can be
 115 approximated as:

$$116 \quad \delta P = \delta E - \frac{1}{\rho_w g} \int_0^{p_s} (\mathbf{u} \cdot \nabla \delta q) dp - \frac{1}{\rho_w g} \int_0^{p_s} (\delta \mathbf{u} \cdot \nabla q) dp - \frac{1}{\rho_w g} \int_0^{p_s} (\delta q \nabla \cdot \mathbf{u}) dp - \frac{1}{\rho_w g} \int_0^{p_s} (q \nabla \cdot \delta \mathbf{u}) dp + residual. \quad (2)$$

117 Terms involving δq are referred to as thermodynamical contribution, and terms involving
 118 $\delta \mathbf{u}$ as dynamical contribution [Seager et al., 2010]. Thus the thermodynamical and

119 dynamical components each have two subcomponents due to moisture advection and
 120 wind convergence. The moisture budget analysis is an effective way to diagnose causes
 121 of precipitation change, and will be applied to the analysis of inter-model variations in
 122 this study.

123 Change in evaporation involves either change in sea-air humidity gradient (denote as
 124 dq), or wind speed, or both [Yu, 2007]. Sea-air humidity gradient is defined as the
 125 difference between the saturation specific humidity at the sea surface temperature (q_s)
 126 and the near-surface (at the 2m height in the models) atmospheric specific humidity (q_a):

$$127 \quad dq = q_s - q_a.$$

128 **3. Sources of model uncertainty in precipitation change**

129 Figure 1 displays model uncertainty, estimated as the inter-model standard
 130 deviation, of precipitation change, contribution from internal variability, SST change and
 131 the six components of rainfall change in Eq. (2) in the North Atlantic. The tropics are
 132 included for comparison. There are two distinct regions with maximum uncertainty in
 133 rainfall projection: an extratropical band extending from the subtropics to high latitudes,
 134 and the tropical Atlantic (Figure 1a). Model uncertainty is generally larger than the
 135 ensemble-mean change, especially over regions where the agreement on the sign of
 136 rainfall change among model is low (Fig. S1). In deed, the domain mean (80°W - 0° ,
 137 20°N - 60°N , ocean only) of the signal-to-noise ratio, defined as the absolute value of the
 138 ensemble-mean divided by the inter-model standard deviation ($\frac{|\Delta P|}{\sigma(\Delta P')}$), of annual-mean
 139 rainfall change is only 0.63. Here Δ denotes climate change, the prime the deviation from
 140 the ensemble-mean change, and σ the standard deviation. For the model uncertainty in

141 projected rainfall change over 100 years in RCP4.5 run, the contribution from internal
142 variability is small (Figure 1b).

143 In the extratropical North Atlantic, the discrepancy in evaporation change among
144 models associated with large inter-model difference in SST warming is important
145 (Figures 1c, d). The SST warming pattern can efficiently affect the sea-air humidity
146 gradient and wind speed change, especially along the Gulf Stream where evaporation is
147 large (Yu, 2007). The second major source of uncertainty is the inter-model spread in
148 transient eddy effect (Figure 1i), which is important for midlatitudes rainfall and related
149 to SST gradient [Woollings et al., 2012]. The inter-model variations in the
150 thermodynamical and dynamical contributions in Eq. (2) are relatively small and mainly
151 origin from the differences in simulating the large horizontal humidity gradient and the
152 Gulf Stream-induced wind convergence (Figures 1e-h). In the tropical Atlantic, by
153 contrast, model uncertainty in rainfall projection is dominated by the dynamical
154 contribution due to wind convergence (Figure 1g). Thus mechanisms for inter-model
155 spread in precipitation projection are totally different between the tropical and
156 extratropical North Atlantic. Here we focus on the extratropical North Atlantic and will
157 discuss the model uncertainty in the tropics elsewhere.

158 **4. Effect of local SST effect on precipitation change**

159 We examine the dominant pattern of inter-model co-variability by the SVD
160 method. Figure 2 shows the first three inter-model SVD modes between $\Delta P'$ and $\Delta SST'$
161 and the regressions of $\Delta E'$, sea-air humidity gradient change and scalar surface wind
162 speed change onto the PCs of $\Delta SST'$.

163 The first SVD mode (SVD1) of $\Delta P'$ displays banded structures that tilt in the
164 northeast-southwest direction, associated with a banded SST pattern that resembles the
165 SST warming pattern due to mode water change [Xie et al., 2010]. The spatial correlation
166 between $\Delta P'$ and $\Delta SST'$ patterns is 0.79 for the SVD1, indicating physical significance of
167 the covariance. Indeed, the regressed evaporation pattern closely resembles the $\Delta P'$
168 pattern with a spatial correlation of 0.89, suggestive of a robust relationship between the
169 SST-induced evaporation change and the rainfall projection.

170 To verify the role of mode water in the formation of the banded SST pattern in the
171 SVD1 mode, we select a specific model (ACCESS1-0), in which the banded structures of
172 $\Delta P'$, $\Delta SST'$ and $\Delta E'$ are pronounced (Figure 3). The spatial correlations of ΔP with ΔSST
173 and ΔE are high in this model (Figures 3a,b) at 0.63 and 0.74, respectively. Changes in
174 sea-air humidity gradient and surface wind speed display similar patterns to the
175 evaporation change, confirming their effect on evaporation. Furthermore, the upper ocean
176 current displays banded structures, with warm (cold) advection from lower (higher)
177 latitudes causing enhanced (reduced) SST warming (Figure 3a).

178 This upper ocean current change is tightly coupled with the mode water change
179 [Xie et al., 2011; Xu et al., 2012]. The mode water is a thick layer of water with vertically
180 uniform properties, whose change affects the upper ocean pycnocline and circulation
181 [Kobashi and Kubokawa, 2012]. In the North Atlantic, the subtropical mode water
182 mainly forms in the deep winter mixed layer south of the Gulf Stream [McCartney and
183 Talley, 1982; Hanawa and Talley, 2001]. Figures 3c,d show the vertical sections along
184 $42^\circ W$ of seawater temperature and zonal current. The mode water of vertical uniform
185 temperature appears in 25° - $40^\circ N$ at depths of 200-400m in the present climatology

186 (Figure 3c, black contours). It forces the upper thermocline (e.g. the 20°C isotherm) to
187 shoal and generates eastward (westward) zonal current band at its north (south) side
188 (Figure 3d) via thermal wind relation. Note that the strong zonal velocity in 40°N-45°N is
189 part of the large-scale gyre unrelated to the mode-water. In the RCP4.5 climatology
190 (Figure 3c, white contours), the mode water shifts northward, as indicated by the bulge of
191 the 18°C isotherm. The northward shift of mode water causes the upper thermocline to
192 deepen (shoal) to the south (north). This results in a cooling around 35°N in the upper
193 ocean, which is quite unusual against the background of thermodynamic warming. The
194 subsurface causes an anomalous eastward (westward) current to the south (north)
195 (Figures 3a,d). Note that the zonal velocity change is negligible below 600m, suggesting
196 that the changes in large-scale gyre circulation are secondary.

197 The SVD2 mode shows negative rainfall change corresponding to the
198 substantially reduced SST warming over the subpolar region and short banded structures
199 south of 45°N (Figure 2b). This negative subpolar SST indicates the importance of the
200 ocean dynamical cooling effect associated with the deep-water formation [Manabe et al.,
201 1990; Long et al., 2014]. The SVD3 mode represents a Gulf Stream shift pattern in the
202 inter-model variations of SST (Figure 2c), as revealed by the two neighboring elongated
203 bands extending from the west to the east with opposite signs.

204 The spatial correlations of $\Delta P'$ pattern with $\Delta SST'$ pattern and the regressed $\Delta E'$
205 pattern are prominent in all the first three SVD modes (see Table S2). This happens
206 because variables important to the evaporation, like sea-air humidity difference [Cayan,
207 1992; Zhang and Mcphaden, 1995; Yu, 2007] and surface wind speed [Chelton and Xie,
208 2010], are all influenced by the SST. The effect of the SST pattern on changes in sea-air

209 humidity gradient and surface wind speed are clear in the North Atlantic (Figures 2d-f),
210 positively correlated with the SST pattern. This positive correlation between SST and
211 surface wind speed patterns indicates the ocean warming drives the wind response
212 [Chelton and Xie, 2010]. Besides, the effects of sea-air humidity gradient and surface
213 wind speed reinforce each other on evaporation. Consequently, inter-model variations in
214 SST warming pattern exert a strong control on the inter-model divergence of precipitation
215 change over the extratropical North Atlantic.

216 The first three modes account for 61% and 71% of the inter-model variance of
217 $\Delta P'$ and $\Delta SST'$, respectively. We also examined the next 7 SVD modes of $\Delta P'$ and $\Delta SST'$
218 and found that positive relationship between them is robust in almost all modes (Table
219 S2). This further confirms the role of the inter-model spread of local SST in explaining
220 the model uncertainty in precipitation change.

221 Precipitation in the North Atlantic has a robust seasonal cycle with the peak in the
222 boreal winter (DJF, December-January-February), associated with similar seasonal
223 variability in evaporation [Yu, 2007]. The inter-model standard deviations of $\Delta P'$, $\Delta E'$
224 and $\Delta SST'$ are much larger in DJF than JJA (June-July-August) (Figure 4). This indicates
225 that much of the inter-model discrepancy of precipitation change develops in the boreal
226 winter. The spatial distributions of inter-model standard deviations in these three
227 variables are very similar in DJF but substantially different in JJA. In JJA, for example,
228 the inter-model standard deviation of $\Delta P'$ is largest off the U.S. east coast but the
229 maximum of the inter-model standard deviations of $\Delta E'$ and $\Delta SST'$ are found far apart to
230 the northeast in the subpolar region (Figures 4d-f). Furthermore, spatial correlation
231 between patterns of $\Delta P'$ and $\Delta SST'$ in the inter-model SVD analysis is high in DJF but

232 low in JJA (Table S2), illustrating that the local SST effect on rainfall change reaches the
233 maximum in the boreal winter. All the above analyses highlight the importance of
234 improving simulations of SST warming pattern, especially in the boreal winter, for
235 reliable precipitation projection.

236 **5. Summary**

237 We have investigated the model uncertainty in precipitation projection under global
238 warming and the local SST effect in the North Atlantic in CMIP5 models. For both
239 annual- and seasonal-mean precipitation changes, inter-model spread is generally larger
240 than the ensemble-mean change (Fig. S1), lowering the confidence in both the sign and
241 magnitude of the future projections. Model uncertainty in rainfall projection is large
242 along the Gulf Stream. Similar enhanced inter-model variability in precipitation change is
243 also found in other west boundary current regions, such as the Kuroshio and its extension
244 and the Agulhas Current (not shown), where local evaporation supplies much of the water
245 vapor for precipitation and latent heating for transient eddy activity. This occurs because
246 local SST effects of sea-air humidity gradient and surface wind speed reinforce each
247 other on evaporation changes in west boundary current regions [Yu, 2007; Chelton and
248 Xie, 2010]. As a result, inter-model variations in local SST change account for much of
249 the inter-model difference of precipitation change. The inter-model SVD analysis
250 between changes in precipitation and SST confirms this result. The local SST change
251 effect on the inter-model diversity of precipitation change spreads in a large number of
252 inter-model SVD modes, indicating the difficulty for extracting the local SST influence
253 with a few leading modes.

254 We performed moisture budget analysis for model uncertainty in rainfall projection.
255 The inter-model discrepancies in evaporation change and transient eddy effect are two
256 dominant sources in the extratropical North Atlantic. Further analyses show that model
257 uncertainty in precipitation and evaporation changes reach the maximum in DJF when the
258 effect of the inter-model variations in SST change is the strongest. The effect of the mean
259 atmospheric circulation change is dominant for the model uncertainty in rainfall change
260 in the tropical ocean, but is secondary in the extratropical North Atlantic.

261 Our results imply that reducing the inter-model spread in SST change, especially in
262 the boreal winter, can greatly improve the consistency of precipitation projection among
263 models. Ocean dynamics is essential in the formation of the SST warming pattern in the
264 midlatitudes, including mode-water-induced banded structures and the reduced subpolar
265 warming over the deep-water formation region. Work is needed to improve the
266 understanding of key physical processes towards greater inter-model consistency in SST
267 warming pattern.

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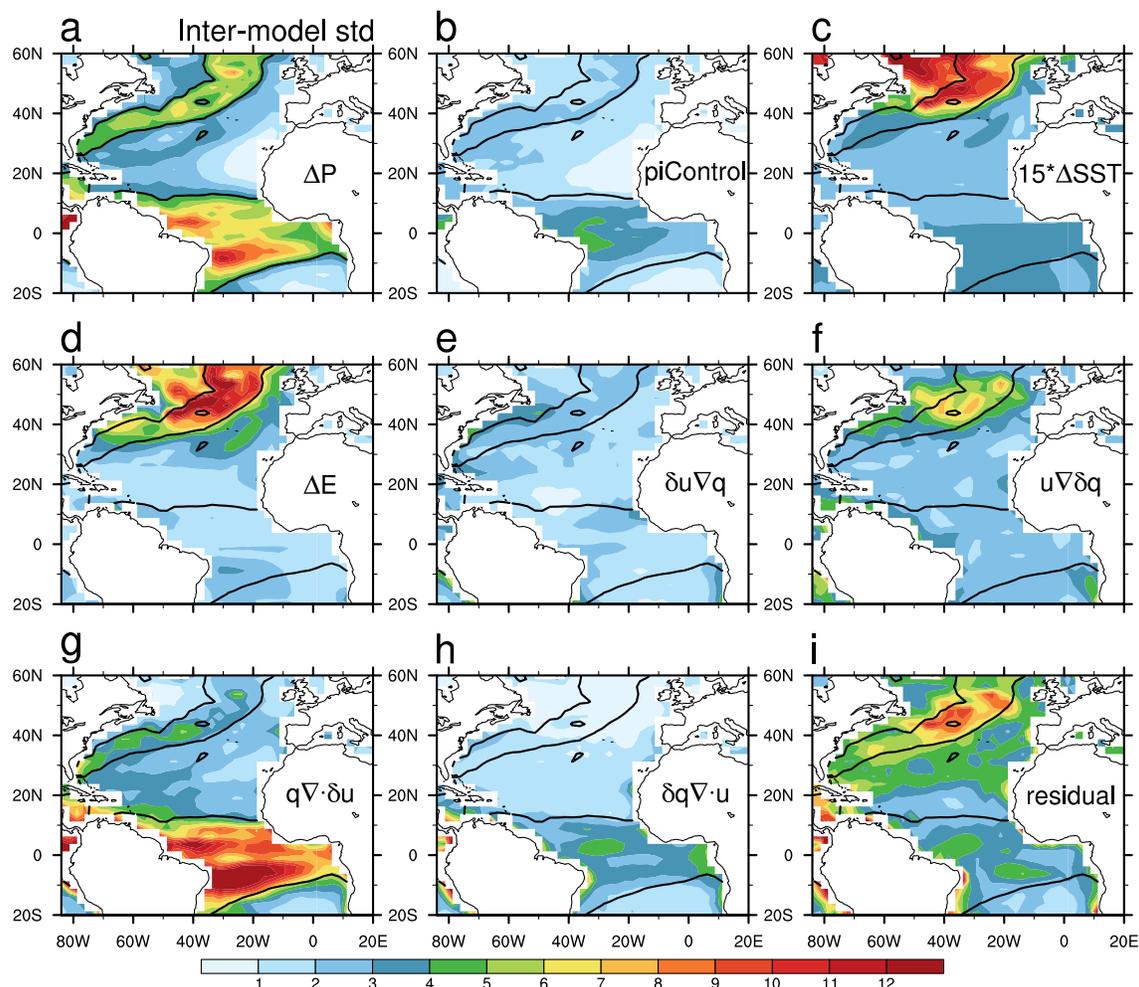
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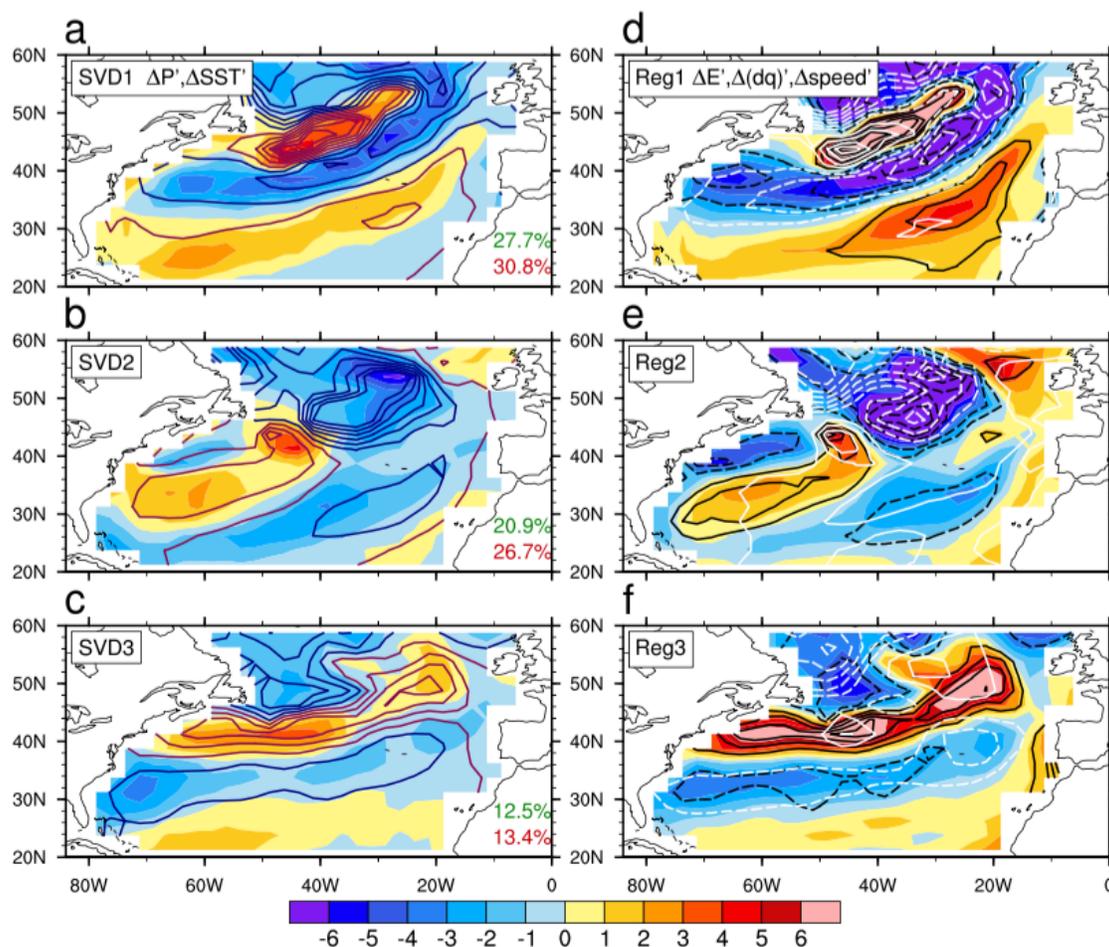
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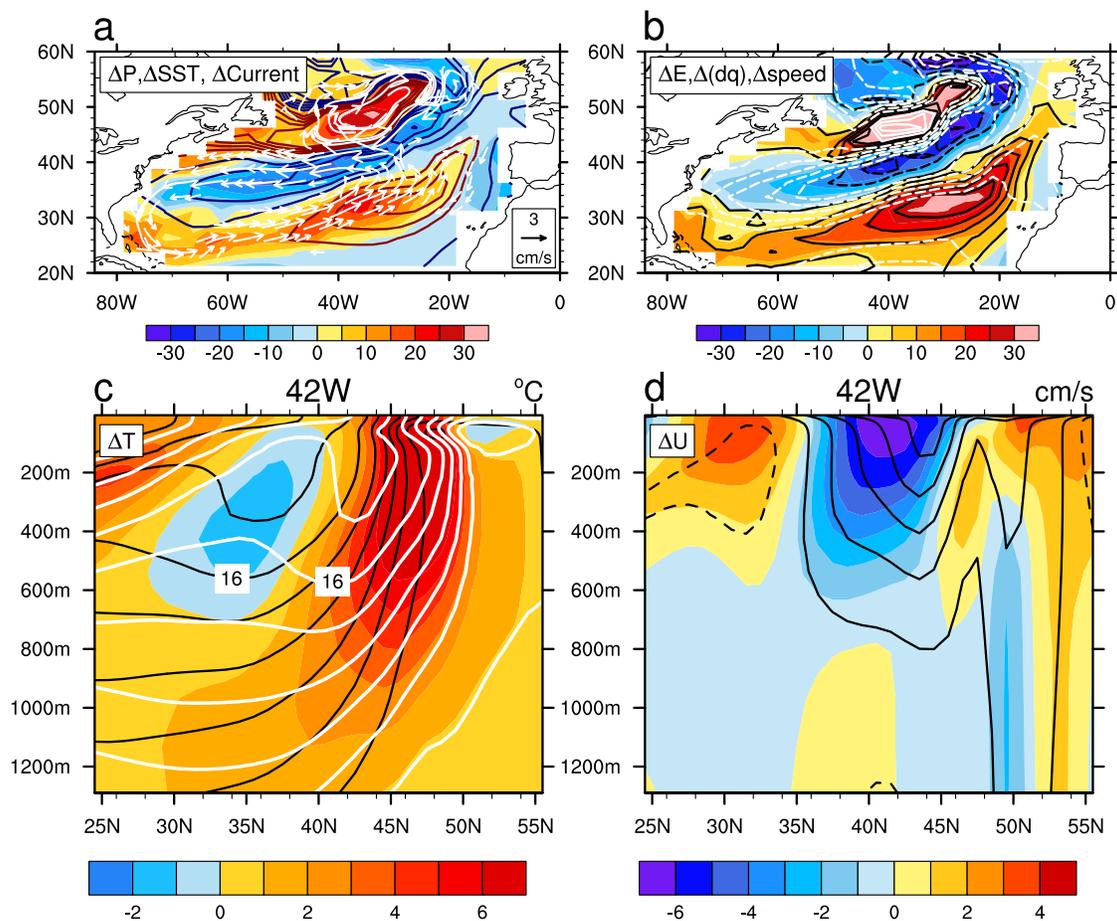
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339 **Fig. 1.** Inter-model standard deviation of (a) precipitation change (ΔP , mm/month), (b)
 340 contribution from internal variability, (c) SST change (ΔSST , $^{\circ}C$), (d) evaporation change
 341 (ΔE), (e) dynamical contribution (with change in mean circulation) due to moisture
 342 advection, (f) thermodynamical contribution (with change in specific humidity) due to
 343 moisture advection, (g) dynamical contribution due to wind convergence, (h)
 344 thermodynamical contribution due to wind convergence, and (i) residual. Black contours
 345 indicate value at 4 mm/month in panel (a). All results are normalized by the domain mean
 346 ($80^{\circ}W-0^{\circ}$, $20^{\circ}N-60^{\circ}N$) SST warming. Note that ΔSST is multiplied by a factor of 15 for
 347 display.



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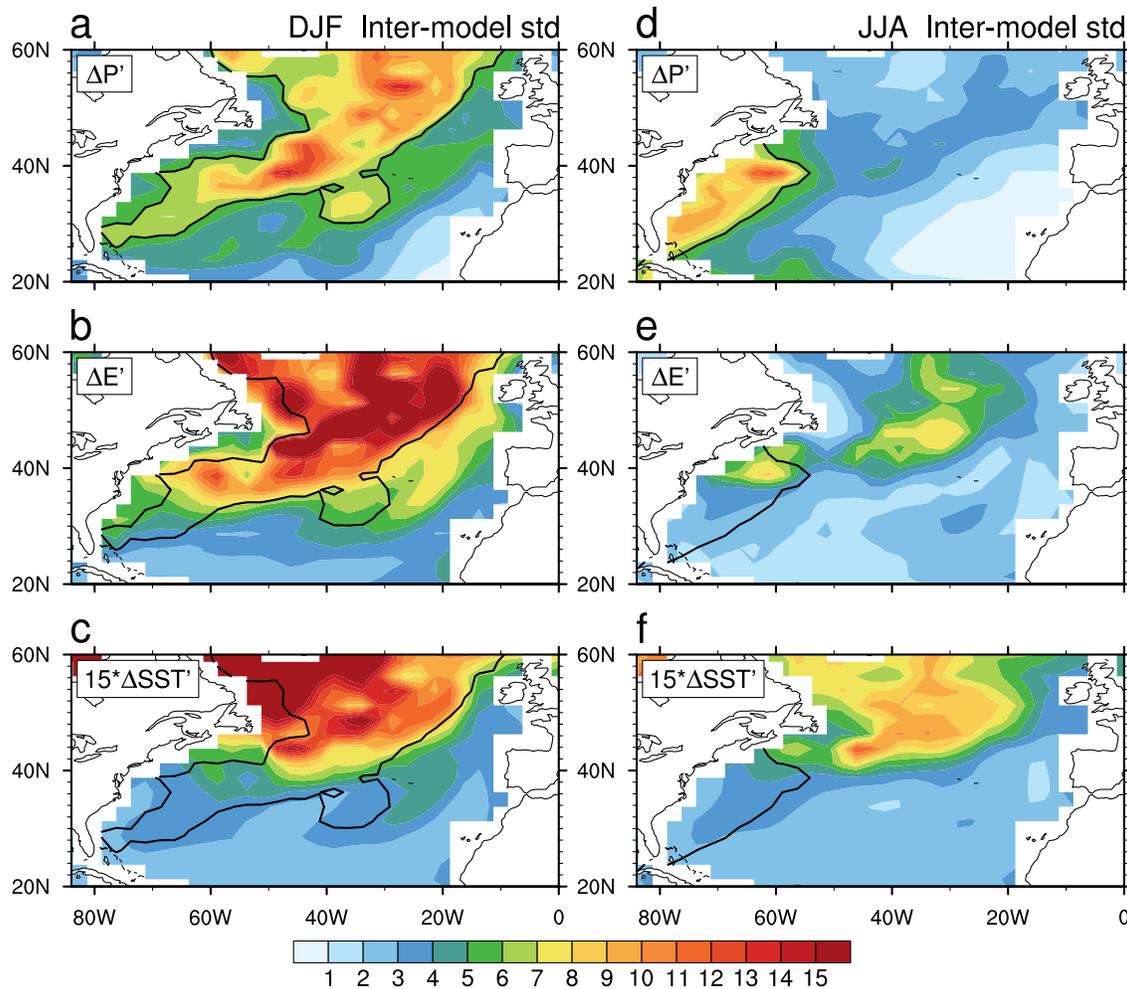
349 **Fig. 2.** (Left panels) First three modes of inter-model SVD between $\Delta P'$ (color shaded)
 350 and $\Delta SST'$ (contours, CI = 0.05°C) in RCP4.5 run. The explained variances for $\Delta P'$
 351 (green letters) and $\Delta SST'$ (red) are marked at the bottom right of each panel. (Right
 352 panels) Corresponding regressions of $\Delta E'$ (color shaded), sea-air humidity gradient
 353 change $[\Delta(dq)']$, black contours, CI = 0.03g/kg] and surface wind speed change ($\Delta speed'$,
 354 white contours, CI = 0.02m/s). Zero contours omitted for clarity. The prime indicates
 355 deviation from the ensemble-mean change.



356

357 **Fig. 3.** (a) ΔP (color shaded, mm/month), ΔSST (contours, $^{\circ}C$) and upper 50m ocean
 358 current change (vectors, cm/s); (b) ΔE (color shaded, mm/month), sea-air humidity
 359 gradient change [black contours, CI = 0.2g/kg] and surface wind speed change (white
 360 contours, CI = 0.1m/s)) in ACCESS1-0 RCP4.5 run. Vectors smaller than 1.5cm/s are
 361 omitted for clarity. Vertical transection along 42 $^{\circ}$ W of present (black contours) and
 362 future (white contours) climatologies, and future-present difference (color shaded): (c)
 363 seawater temperature ($^{\circ}C$) and (d) zonal velocity (cm/s). CI = 2 $^{\circ}C$ for temperature and
 364 2cm/s for zonal velocity.

365



366

367 **Fig. 4.** Inter-model standard deviations of future projections in DJF and JJA, colors
 368 shaded are $\Delta P'$ (mm/month) in (a and d), $\Delta E'$ (mm/month) in (b and e) and $\Delta SST'$ ($^{\circ}\text{C}$)
 369 in (c and f). Black contours indicate value at 6 mm/month in panel (a).