

1 **Improved tropical modes of variability in the NCEP Climate Forecast**

2 **System (version 2) via a stochastic multcloud model**

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## ABSTRACT

18 A new stochastic multi-cloud model (SMCM) convective parametrization,  
19 which mimics the interactions at sub-grid scales of multiple cloud types, is  
20 incorporated into the National Centers for Environmental Prediction (NCEP)  
21 Climate Forecast System version 2 (CFSv2) model (referred to as CFSsmcm  
22 hereafter) in lieu of the pre-existing simplified Arakawa-Schubert (SAS) cu-  
23 mulus scheme. A detailed analysis of the tropical intra-seasonal variability  
24 (TISV) and convectively coupled equatorial waves (CCEW), in comparison  
25 with the original (control) model and with observations, is presented here.  
26 The last 10-years of a 15-year long climate simulation are analyzed. Signifi-  
27 cant improvements are seen in the simulation of the Madden-Julian oscillation  
28 (MJO) and most of the CCEWs as well as the Indian summer monsoon (ISM)  
29 intra-seasonal oscillation (MISO). These improvements appear in the form  
30 of improved mechanisms and physical structure of these waves. This can be  
31 regarded as a validation of the central idea behind the SMCM according to  
32 which organized tropical convection is based on three cloud types namely, the  
33 congestus, deep and stratiform cloud decks that interacts with each other and  
34 form a building block for multiscale convective systems. An adequate account  
35 for the dynamical interactions of this cloud hierarchy thus constitutes an im-  
36 portant requirement for cumulus parameterizations to succeed in representing  
37 atmospheric tropical variability. SAS fails to fulfill this requirement evident in  
38 the unrealistic mechanisms and structures of the major intra-seasonal modes  
39 simulated by CFSv2 as documented here.

## 40 **1. Introduction**

41 The tropical atmosphere harbors a spectrum of dynamical modes that interact with each other  
42 and with the climate systems, on multiple spatial and temporal (Moncrieff and Klinker 1997; Ki-  
43 ladis et al. 2009; Lau and Waliser 2011). It is still debatable whether these different modes are part  
44 of a monster tropical convection belt or are they separate components (Toma and Webster 2010a,b;  
45 Serra et al. 2014). The Madden-Julian oscillation (MJO) (Zhang 2005, 2013) and monsoon intra-  
46 seasonal oscillations (MISO) (Goswami 2012) dominate the tropical variability on intra-seasonal  
47 time-scales and convectively coupled waves (CCEW) and tropical depressions of all sorts are seen  
48 on synoptic scales (Kiladis et al. 2009). While CCEWs are thought to be the moist analogs of  
49 equatorially trapped waves—linear modes of equatorial dynamics (Matsuno 1966; Takayabu 1994;  
50 Wheeler and Kiladis 1999), there is no dry dynamical equivalent mode for the MJO. The at-  
51 mospheric science community is still debating whether the MJO is a moisture-coupled planetary  
52 scale mode or some sort of a multi-scale convective envelope owing its existence to upscale en-  
53 ergy transfer from synoptic and mesoscale systems (Majda and Stechmann 2009a; Wang and Liu  
54 2011; Sobel and Maloney 2012; Thual and Majda 2015; Stachnik et al. 2015). Nonetheless, there  
55 is a consensus in the climate modeling community that a climate model’s ability to simulate the  
56 weather and climate realistically depends largely on its ability to simulate these intra-seasonal and  
57 synoptic scale modes (Lin et al. 2006; Hung et al. 2013; Jiang et al. 2015). This study aims to  
58 gauge, in this regard, the U. S. A. National Centers for Environmental Predictions Climate Fore-  
59 casting System, version 2 (CFSv2), in which the stochastic multicloud model (SMCM) convective  
60 parametrization of (Khouider et al. 2010, hereafter KBM10) is implemented (Deng et al. 2015),  
61 in comparison to the original CFSv2 model. In the sequel, the acronym CFSv2 is used to desig-

62 nate the original (control) model while the modified model, using the SMCM parameterization, is  
63 termed CFSsmcm.

64 Despite the significant progress of the last decade or so (Moncrieff et al. 2012, and references  
65 therein), present day global climate models (GCM) still show limited ability in simulating the  
66 MJO (Slingo et al. 1996; Lin et al. 2006; Kim et al. 2009; Hung et al. 2013; Jiang et al. 2015),  
67 MISO (Waliser et al. 2003; Lin et al. 2008b; Sabeerali et al. 2013; Sperber et al. 2013) and CCEWs  
68 (Lin et al. 2008a; Straub et al. 2010; Hung et al. 2013; Guo et al. 2015). The inefficiency of the  
69 present day climate models to simulate these tropical intraseasonal variability (TISV) modes stems  
70 from our limited understanding of tropical dynamics. Recent studies emphasize the importance  
71 of representing processes that are thought to be important for TSIV dynamics including moisture  
72 pre-conditioning, atmosphere ocean coupling, cloud radiative feedback, convective momentum  
73 transport, stratiform heating, and boundary layer dynamics (Lin et al. 2006; Straub et al. 2010;  
74 Jiang et al. 2015; Wang et al. 2016a). Nonetheless, there is a consensus in the climate modeling  
75 community that the fidelity in proper simulation of the MJO is a pinnacle metric to asses the fi-  
76 delity of a GCM (Waliser et al. 2009). Straub et al. (2010) found that 75% of the Coupled Model  
77 Intercomparison Project (CMIP) phase 3 models fail to realistically simulate the convectively cou-  
78 pled Kelvin waves. Although, the MJO and the CCEWs have a lot in common, improvement in  
79 one does not necessarily translate into improvement in the other despite the undeniable evidence  
80 that CCEWs, the MJO and mesoscale convective systems are embedded in and interact with each  
81 other across multiple temporal and spatial scales (Nakazawa 1988; Moncrieff and Klinker 1997;  
82 Gottschalck et al. 2013; Dias et al. 2013). Moreover, both the MJO and the MISO are believed to  
83 have an impact on the global weather and climate (Krishnamurthy and Kinter 2003; Zhang 2005,  
84 2013; Lau and Waliser 2011). From clustering synoptic systems (Goswami et al. 2003) to influ-  
85 encing ENSO development (Kirtman and Shukla 2000), TISV modes have profound effects on

86 the tropical variability, the impact being felt much beyond their own spatial and temporal scales.  
87 Therefore a model which simulates these TISV modes, viz., MJO, CCEW and MISO, realistically,  
88 is expected to simulate the mean state and many other aspects of the global climate with fidelity  
89 (Jiang et al. 2015; Kim et al. 2009; Waliser et al. 2009).

90 The inability of the present day climate models to accurately simulate th prominent TISV modes,  
91 is often attributed to their inability to simulate the mean climate state and vice versa (Slingo et al.  
92 1996; Sperber et al. 1997; Gadgil and Sajani 1998; Waliser et al. 2003; Sperber 2004; Lin et al.  
93 2006; Zhang et al. 2006). In this chicken-and-egg dilemma, the synoptic variability has got com-  
94 paratively less attention. Recently, using 36 coupled models (including 32 CMIP phase 5 models),  
95 Goswami and Goswami (2016) argued that the lack of simulated synoptic variability to be par-  
96 tially, at least, responsible for the precipitation dry bias in rain abundant regions of the tropics.  
97 The deterministic nature of the convective parameterization (CP) schemes, used in those models,  
98 are to be blamed, to some extent, for this as they fail to represent the stochastic nature of con-  
99 vection to trigger organization across multiple scales (Arakawa 2004; Frenkel et al. 2012; Peters  
100 et al. 2013). This limitation of the deterministic CP schemes got further exposed when stochastic  
101 (Buizza et al. 1999; Lin and Neelin 2000, 2002, 2003; Palmer 2001; Majda and Khouider 2002;  
102 Khouider et al. 2003; Plant and Craig 2008; Teixeira and Reynolds 2008; KBM10) and cloud re-  
103 solving (Grabowski and Smolarkiewicz 1999; Grabowski 2001; Khairoutdinov and Randall 2001;  
104 Randall et al. 2003; Satoh et al. 2008; Fudeyasu et al. 2008; Benedict and Randall 2009; Liu et al.  
105 2009) approaches have showed promise. While superparameterized and global cloud resolving  
106 models continue to evolve (Goswami et al. 2015; Yashiro et al. 2016; Fukutomi et al. 2016; Koop-  
107 erman et al. 2016), they remain computationally expensive and impractical. Stochastic approaches  
108 are getting more and more consideration (Deng et al. 2015, 2016; Ajayamohan et al. 2016; Davini  
109 et al. 2016; Goswami et al. 2016; Dorrestijn et al. 2016; Wang et al. 2016b; Gottwald et al. 2016;

110 Bengtsson and Kørnich 2016; Berner et al. 2016; Peters et al. 2017), as a computationally cheap al-  
111 ternative. In this paper, we use the first time the stochastic multcloud model (SMCM) of KBM10  
112 as a cumulus parameterization in a comprehensive GCM.

113 Previous studies involving the SMCM (KBM10; Khouider et al. 2011; Ajayamohan et al. 2013,  
114 2014, 2016; Deng et al. 2015, 2016) have shown considerable skill in simulating TISV. Using the  
115 deterministic multcloud model (DMCM) of Khouider and Majda (2006), hereafter KM06, (see  
116 also Khouider and Majda 2008b) as a cumulus parameterization in the National Center for Atmo-  
117 spheric Research (NCAR)'s High-Order Methods Modeling Environment (HOMME), at course  
118 GCM resolution, Khouider et al. (2011) demonstrated that the DMCM could simulate many ob-  
119 served features of TISV modes, such as the MJO and CCEWs. Ajayamohan et al. (2013) showed  
120 that when a warm pool like background is imposed, the same model exhibits realistic initiation  
121 and dynamics of the MJO via circumnavigating Kelvin waves. Deng et al. (2015, 2016) showed  
122 that when the SMCM is incorporated into HOMME, in an aqua-planet setup, it produces MJOs  
123 with dynamical features such as the front-to-rear vertical tilt and the quadruple vortex structure  
124 (Kiladis et al. 2005), and realistic intermittent variability. Ajayamohan et al. (2014, 2016) showed  
125 that the simulation of MISOs can be improved by incorporating the SMCM or its deterministic  
126 version in the NCAR-HOMME aqua-planet model. However, all of the above results are based on  
127 idealized-aquaplanet simulations. Therefore, implementing the SMCM in a fully coupled climate  
128 model is an obvious way forward. We took up the NCEP CFSv2 model, promoted by the Na-  
129 tional Monsoon Mission of the Ministry of Earth Sciences, India, and implemented SMCM in it.  
130 Namely, we have replaced the conventional convective parameterization used by CFSv2, which is  
131 the Simplified Arakawa-Schubert (SAS) (Pan and Wu 1995; Pattanaik et al. 2013), by the SMCM  
132 model. The details of this implementation including the parameter tuning can be found in Goswami  
133 et al. (2016), hereafter GKPM16, and Goswami et al. (2017), hereafter GKPM17.

134 The paper is organized as follows. A brief description of the SMCM model formulation is  
135 presented in Section 2. Section 3 describes the results, particularly emphasizing the dynamical and  
136 physical features of TISV modes, namely, the MJO, CCEW and MISO, as simulated by CFSsmcm  
137 in comparison to the control CFSv2 model and observations. Finally, some concluding remarks  
138 are given in Section 4.

## 139 **2. Model Equations, Data and Methodology**

140 At this experimental stage of the SMCM parameterization approach, the convective heating pro-  
141 file is based on three prescribed basis functions, which are designed to mimic the three dominating  
142 cloud types of tropical convection, namely, congestus, deep and stratiform (Johnson et al. 1999;  
143 Mapes et al. 2006). The SMCM divides each GCM grid box into a  $40 \times 40$  microscopic lattice.  
144 Each lattice site is either occupied by congestus, deep or stratiform cloud decks, or it is a clear  
145 sky site. Transitions from a lattice site with one type of cloud to another type occur according to  
146 a stochastic-Markov chain process whose transition probabilities depend on the large-scale state  
147 through a few convection predictors. New to the CFS implementation (GKPMM16; GKPMM17),  
148 the large scale predictors include the convective available potential energy (CAPE), convective  
149 inhibition (CIN), middle tropospheric (700 hPa) dryness/moistness (MTD) and vertical velocity  
150 at the top of the boundary layer ( $W$ ). Each microscopic lattice within a large-scale grid box sees  
151 the same large scale conditions. However, their evolution in time differ as the transition rules also  
152 depend on the previous state of a microscopic lattice, which provides time memory for the cu-  
153 mulus parameterization. The heating rates associated with the three cloud types are parameterized  
154 through closure formulas, depending on mid-level moisture and CAPE, that are proportional to the  
155 cloud area fractions obtained through the evolving stochastic lattice model. The three prescribed  
156 basis functions of the SMCM are amplified by the respective parameterized heating rates and the

157 amplified profiles add up to yield the total parameterized heating. The moisture and temperature  
158 tendencies are calculated from this parameterized total heating and then given back to the host  
159 model, which is CFSv2.

160 Specifically, the total convective heating is expressed as (Khouider et al. 2011):

$$Q_{tot}(z) = H_d\phi_d(z) + H_c\phi_c(z) + H_s\phi_s(z).$$

161 Here,  $H_c$ ,  $H_d$  and  $H_s$  are the parameterized heating rates associated with the three cloud types,  
162 congestus, deep, and stratiform, respectively, while  $\phi_d, \phi_c, \phi_s$  are the corresponding heating profile  
163 basis functions. Further we have

$$H_d = \sigma_d Q_d, H_c = \sigma_c Q_c, H_s = \sigma_s Q_s$$

164 with  $Q_d, Q_c, Q_s$  are the parameterized heating potentials depending deterministically on CAPE and  
165 midlevel moisture and  $\sigma_d, \sigma_c, \sigma_s$  are the stochastic area fractions (lattice coverage) occupied by the  
166 respective cloud types. These cloud area fractions, along with a fourth state, of sky condition with  
167 no clouds, describe a Markov jump stochastic process in the form of a multi-dimensional birth-  
168 death system whose transition probabilities depend explicitly on some key large scale predictors  
169 motivated by observations and physical intuition (KBM10; Frenkel et al. 2012; De La Chevrotière  
170 et al. 2015; Deng et al. 2016). The temperature and moisture convective tendencies are set accord-  
171 ing to  $Q_{tot}$ . While, as already mentioned, further details about the implementation of the SMCM  
172 convective parametrization in CFSv2 can be found in GKPM16 and GKPM17, we note here  
173 that except for replacing the SAS cumulus scheme with SMCM, the rest of CFSv2 configuration  
174 is unchanged. For instance, CFSsmcm still uses the same shallow cumulus scheme as CFSv2 but  
175 the parameterized cloud feedback is ignored. The latter may be included in future versions of CF-  
176 Ssmcm by taking advantage of the stochastic cloud area fractions. Details on the reference model  
177 CFSv2 are available in Saha et al. (2014).

178 We have analyzed the last 10 years output from a 15 year CFSsmcm climate simulation in com-  
179 parison with a simulation of the same length and same initial conditions, done with the original  
180 model, CFSv2, using the SAS convection scheme, as a control run. As an observational bench-  
181 mark, we used outgoing long-wave radiation (OLR) from NOAA ( $2.5^\circ \times 2.5^\circ$ ; daily) (Liebmann  
182 and Smith 1996) and the thermo-dynamical and dynamical parameters from NCEP reanalysis  
183 ( $2.5^\circ \times 2.5^\circ$ ; daily) (Kalnay et al. 1996) to evaluate the model simulated climate, using either  
184 SMCM or SAS.

185 For both CFSsmcm and CFSv2 simulations, we used a horizontal resolution of T126, 64 vertical  
186 levels, and a time step of 10 minutes. We have extensively used the wavenumber-frequency filter-  
187 ing technique introduced and used by Kiladis et al. (2005, 2009) to isolate the different modes of  
188 tropical ISV and the CCEW's.

### 189 **3. Results**

190 In GKPM16, the CFSsmcm simulation is found to have a reasonably good mean state at least  
191 as good as the control CFSv2 model, if not better in some aspects, especially in places where  
192 CFSv2 is known to have significant biases. Given that CFSv2 is one of the better of the state-  
193 of-the-art climate models, this is a satisfactory result. In this section, the tropical intra-seasonal  
194 variability in the CFSsmcm simulation is documented. One standard metric to assess a model-  
195 simulated ISV is to plot the Takayabu-Wheeler-Kiladis (TWK for short) spectra (Takayabu 1994;  
196 Wheeler and Kiladis 1999). Figure 1 shows the TWK-spectra plotted for the model-simulated  
197 OLR for both CFSv2 and CFSsmcm and observations (NOAA OLR, Liebmann and Smith (1996)).

198 The observed modes have a wealth of literature available for their documentation (Takayabu  
199 1994; Wheeler and Kiladis 1999; Kiladis et al. 2009, and references therein). However, for the  
200 sake of completeness, it is worthwhile to list the prominent modes, corresponding to the most

201 significant peaks in Figure 1a. In the symmetric part of the spectrum, we have the eastward  
202 moving Madden-Julien oscillation (MJO) corresponding to the peak at wavenumbers 1 to 3 and  
203 time periods between 30 to 60 days, westward moving  $n = 1$  equatorial Rossby (ER) wave peak  
204 at wavenumbers  $-3$  to  $-4$  and time periods  $\sim 30$  days, Kelvin waves with an elongated peak  
205 spanning wavenumbers 2 to 7 and time periods 4 to 10 days, and  $n = 1$  westward inertia gravity  
206 waves (WIG) roughly around wavenumbers  $= -1$  to  $-15$  and a time period of 3 days. The an-  
207 tisymmetric part shows one dominant corresponding to westward mixed Rossby-gravity (MRG)  
208 waves, between wavenumbers 0 and  $-6$  and time periods of 3 to 6 days, and eastward inertia grav-  
209 ity waves (EIG) for wavenumbers 0 to 8 and time periods between 2 and 5 days. The remaining  
210 power blobs at negative wavenumbers sandwiched between the WIG and ER waves are believed to  
211 correspond to tropical depressions of all sorts including monsoon low pressure systems (Wheeler  
212 and Kiladis 1999).

213 The TISV modes are not prominent in the CFSv2 simulation as indicated by the lack of color  
214 contrast in the plots, Fig 1b and 1e. Except for the ER waves, CFSv2 underestimates the power for  
215 all the other prominent modes. Moreover, CFSv2 MJO peak has longer time-period than obser-  
216 vations (Figure 1b). Significant improvement is evident in the CFSsmcm simulated TWK-spectra  
217 (Figure 1e and 1f), including the MJO and especially the higher frequency CCEWs, mentioned  
218 above. The MJO period and strength has substantially improved. Also, the Kelvin wave  $n = 1$   
219 WIG power have clearly improved in the CFSsmcm. There is a discernible peak corresponding to  
220 MRG waves in the CFSsmcm run while it is inexistent in CFSv2. Nonetheless, CFSsmcm sim-  
221 ulates a weaker power for most of these modes, compared to observations, and thus there is still  
222 room for improvement.

223 While they obey a rough self similarity feature in vertical structure (Mapes et al. 2006; Kiladis  
224 et al. 2009; Khouider et al. 2011), the MJO and the other convectively coupled waves have different

225 propagation properties and different structural details and physical features. The MJO has been  
226 one of the most highly studied climate phenomena (Zhang 2005; Wang et al. 2016a, and references  
227 therein). A review of the CCEWs can be found in Kiladis et al. (2009). The state of the art models  
228 show limited ability in simulating these essential features of the tropical ISV (Lin et al. 2006,  
229 2008b,a; Straub et al. 2010; Hung et al. 2013; Guo et al. 2015; Wang et al. 2016b). Guo et al.  
230 (2015) argues that, there is a good chance that a model which simulates the CCEWs realistically  
231 would simulate a “good” MJO as well. Therefore, for an in-depth analysis of TISV in CFSsmcm  
232 simulations, we isolate the MJO and the different CCEWs applying space-time filtering (Kiladis  
233 et al. 2009) and examine the different features. We repeat the same exercise for observations and  
234 the control-CFSv2 simulations, for a proper assessment of the improvements.

### 235 *a. MJO*

236 We applied space-time filtering on different meteorological fields and we retained the aver-  
237 aged signal corresponding to wavenumber 1-9 and time-period 30-96 days following Kiladis et al.  
238 (2005). We isolated the MJO filtered anomalies for OLR and zonal and meridional wind fields  
239 for both the CFSsmcm and the control-CFSv2 simulations and observations (NOAA OLR and  
240 NCEP winds, Kalnay et al. (1996)). In Figure 2, the daily variance, for the full year, of the MJO  
241 filtered OLR anomalies are shown. In observations (Figure 2b), the maximum variance is seen  
242 over the warm waters of the Indian Ocean and western Pacific Ocean, with the peak located over  
243 the equatorial Bay of Bengal (BoB). In the western Pacific Ocean, the amplitude is asymmetric  
244 about the equator, tilting southward. This is possibly due to the interaction of the MJO with the  
245 warm waters of the Indonesian throughflow (Zhou and Murtugudde 2010; Zhang 2013). There is  
246 an isolated peak off the gulf of California. In the CFSsmcm simulation (Figure 2a), the variance  
247 over the Indian Ocean is reasonably well simulated with the peak slightly shifted south-westward.

248 Over the western Pacific, the pattern appears patchy with an underestimation towards the southern  
249 branch but it remains qualitatively similar to the observations. The peak off the gulf of Califor-  
250 nia is captured well, however one more isolated peak is visible over the south-east Atlantic. In  
251 the CFSv2 simulation (Figure 2c), the MJO variance splits into two streaks, distributed north and  
252 south of the equator (Figure 2b). Moreover the variance is marginally stronger over the west Pa-  
253 cific than over the Indian Ocean, unlike the observations and the CFSsmcm run. Similarity while a  
254 variance pattern is evident, a remnant of the double ITCZ problem is also seen. Overall, CFSsmcm  
255 simulated MJO daily variance has greatly improved qualitatively compared to the CFSv2-control  
256 simulation.

257 Figure 3 shows snapshots of an MJO phase composite in terms of the MJO filtered OLR, for  
258 different lead times. In order to construct the composite, the peak MJO dates are identified based  
259 on an MJO index, corresponding to the MJO filtered anomalies taken at a location of corresponding  
260 high variance. We have checked that, the results are resilient to changes in location of this index.  
261 The left column of Figure 3, shows the propagation of the MJO filtered OLR anomalies from  
262 observations. At 15 days lag, a blob of convection occurs over the west equatorial Indian Ocean  
263 (around 60°E). The blob makes a smooth migration eastward and reaches 180°E at a lead time of  
264 25 days. Thus, the convection describes 120 degrees of longitude in 40 days, which corresponds  
265 to a phase speed of approximately  $5 \text{ m s}^{-1}$ . During the decay process of the blob, it spreads out  
266 and separates onto two blobs south and north of the equator (Lead 25 days).

267 In the middle column of Figure 3, the phase composite of CFSsmcm simulated MJO filtered  
268 OLR anomalies are shown. The overall features of the propagation of convection are reasonably  
269 captured. However there are a few striking discrepancies. The first to catch the eye are a smaller  
270 spatial extent of the blob and a slower phase speed. Moreover, the active convection over the cen-  
271 tral Pacific in lag 5 days seems unrealistic but it is very weak. A closer look at this active convec-

272 tion over the central Pacific reveals a wavy pattern indicated by deepening and fading blue shading  
273 alternately. This possibly indicates a contamination of the MJO signal by some other modes of  
274 variability inherent to CFSsmcm. In comparison, for the CFSv2 MJO filtered OLR anomalies (the  
275 third column on the right of Figure 3), the detailed features are hardly resembling the observa-  
276 tions. The amplitude and organization of convection are very weak, making it extremely difficult  
277 to comment on the phase speed or any other physical property.

278 The propagation features of the MJO are arguably better characterized by the Hovmöller plots  
279 of the MJO composite averaged over the latitude band between  $10^{\circ}\text{S}$  to  $10^{\circ}\text{N}$ , shown in Figure  
280 4. The top three panels, Figure 4a, b and c, show the composites of the MJO filtered anoma-  
281 lies and the bottom three panels, Figure 4d, e and f, show the composite of the corresponding  
282 raw (unfiltered) data anomalies for CFSsmcm, NOAA OLR, and the control-CFSv2 simulation,  
283 respectively. Consistent with Figure 3 (left column), a smooth propagation is exhibited by the  
284 observed OLR data. A feature, which was not evident in Figure 3 and appears clearly in Figure  
285 4b, is the different phase speeds of the MJO, over the Indian Ocean and Pacific Ocean basins. The  
286 MJO phase speed is faster over the West Pacific compared to that over the Indian Ocean. In the  
287 CFSsmcm simulation (Figure 4a and 4d) the phase speed of the MJO appears slower than observed  
288 over the Indian Ocean and the organization is weaker over the West Pacific. Focusing on the MJO  
289 filtered anomaly composite Hovmöller plot (Figure 4a), the organization almost seems broken past  
290 the maritime continent and reappears in the central Pacific with a hint of eastward movement from  
291 there indicating a wavenumber 2 structure. However, when we observe closely the unfiltered com-  
292 posite in Figure 4d, the CFSsmcm simulation appears to capture the two different phase speeds on  
293 the two sides of the maritime continent, especially for the active phase of convection (blue shad-  
294 ing). In the CFSv2 simulations (Figure 4c and f), however, both the organization and amplitude  
295 are poorly simulated, consistent with Figure 3 (right column).

296 Figure 5 shows the circulation features at 850 hPa, top three panels (a, b and c), and at 200 hPa,  
297 bottom three panels (d, e and f), of the MJO filtered anomalies for the observations and the two  
298 model simulations, as indicated. The observed circulation pattern at 850 hPa (Figure 5b) shows  
299 a pair of Rossby gyres north (counter-clockwise) and south (clockwise) and slightly west of the  
300 convection peak location with a broad fetch of easterlies over the equatorial Pacific. Two other  
301 gyres of opposite signs are also visible east of the convection center but their centers are located  
302 far away from the equator—outside the displayed domain. This structure constitutes the famous  
303 quadruple structure of the MJO reported in many observational and theoretical studies of the MJO  
304 (Rui and Wang 1990; Hendon and Salby 1994; Majda and Biello 2003; Kiladis et al. 2005; Zhang  
305 2005; Majda and Stechmann 2009b). At 200 hPa (Figure 5e), subtropical quadruple rotational  
306 circulation enveloping the convective center and winds diverging out of the convection center are  
307 seen. The observed circulation patterns at 850hPa and 200hPa indicate a dominantly baroclinic  
308 (reversal of wind direction with height) vertical structure for the MJO filtered anomalous winds.

309 The CFSsmcm simulation (Figure 5a and d) indeed appears to capture this baroclinicity to a  
310 good extend. However, the circulation patterns themselves, both at 850hPa (Figure 5a) and at  
311 200hPa (Figure 5d), are not as well organized as in observations, which is consistent with the  
312 aforementioned wavenumber 2 type structure of the CFSsmcm simulated MJO. Nevertheless, CF-  
313 Ssmcm captures the major features considerably well: at 850hPa (200hPa), the Rossby gyres (anti-  
314 cyclonic circulations) meridionally placed at the two wings of the convection maxima at around  
315 70°E-80°E, underdeveloped in the southern (northern) hemisphere, with westerlies (easterlies)  
316 over the Indian Ocean basin and easterlies (westerlies) over the Pacific basin. For the CFSv2 sim-  
317 ulations (Figure 5c and f), the circulation patterns look too disorganized to make any conclusive  
318 remark.

319 Figure 6 shows the composite of OLR time series (top panels) and vertical structure of the  
320 MJO filtered anomalies, averaged over  $5^{\circ}\text{S}$ - $5^{\circ}\text{N}$ , for CFSsmcm (left column), observations (mid-  
321 dle column) and CFSv2 (right column). For a better visualization, each column is topped with its  
322 respective OLR variations to locate the convection maximum. It is not hard to see that CFSsmcm  
323 simulates the convection activity significantly more realistically than CFSv2. The observed fea-  
324 tures around the convection maximum (OLR minimum in Figure 6b), like the quadruple structure  
325 of horizontal wind in the zonal cross-section (Figure 6e), convergence (divergence) at the lower  
326 (upper) troposphere collocated with the OLR minimum (Figure 6h), leading (following) nega-  
327 tive (positive) humidity anomalies lead by a lower level moistening (Figure 6k), the collocated  
328 positive temperature anomalies (Figure 6n), the collocated updraft with surrounding subsidence  
329 (Figure 6q) and the collocated positive anomalous diabatic heating with an extension ahead of the  
330 convection maximum in the lower levels (Figure 6t), are reasonably well captured in the CFSsmcm  
331 simulation.

332 The westward tilt (Zhang 2005; Kiladis et al. 2005) prominent in the zonal wind (Figure 6d),  
333 convergence (Figure 6g), relative humidity (Figure 6j) and temperature anomalies (Figure 6m),  
334 is also captured to a good extent by CFSsmcm. However, the CFSsmcm fields are somewhat  
335 noisy and have weaker amplitudes. CFSv2 shows limited skill in capturing these features lack-  
336 ing severely in simulating the adequate organization and amplitude. Noteworthy, in the CFSv2  
337 simulation, the anomalies corresponding to the suppressed phase of convection (OLR maximum)  
338 looks more prominent than the active phase. This is clearly visible on the top panels where CFSv2  
339 exhibits a strong OLR positive peak ahead of the convection center and not much of a minimum  
340 OLR peak, contrary to CFSsmcm, which is consistent with the observations, with a caveat that  
341 the OLR maximum is ahead of the minimum in the observation but the former lags the latter in  
342 CFSsmcm. Nonetheless, this difference is perhaps simply an artifact of the compositing technique.

343 *b. CCEW*

344 In order to verify whether the structure and propagation features of the simulated CCEW modes  
345 are well simulated, here we isolate the different individual modes by applying a space-time filtering  
346 and examine the different features as done for the MJO. The space-time filters used here are the  
347 same as the ones used in Kiladis et al. (2009), except for the  $n = 0$  EIG waves for which we  
348 used a narrower region, limited by wavenumbers 1 to 3, frequencies 0.166 to 0.55, and equivalent  
349 depth curves  $H = 12$  m and  $H = 50$  m. A broader filter as in Kiladis et al. (2009) makes the EIG  
350 signal contaminated with Kelvin waves, as the latter appear stronger in the CFCsmcm simulation.  
351 An alternative would be to separate the solution into symmetric and antisymmetric parts but we  
352 refrained from doing that here because it is not standard practice.

353 Figure 7 shows the daily variances corresponding to the different modes of the CCEW spec-  
354 trum for CFSsmcm simulation (left column), observations (middle column) and CFSv2 simula-  
355 tion (right column). The maximum variance for all the displayed modes are observed to be over  
356 the west to central Pacific region. In the CFSsmcm simulations (left column panels) the over-  
357 all pattern and amplitude of the different variances are well captured however the peak variance  
358 is slightly shifted westward for almost all the modes. Variance for the Kelvin, EIG and MRG  
359 modes are slightly under-estimated whereas, that of the ER and WIG are slightly over-estimated.  
360 CFSv2 (right column panels) severely underestimates the daily variance for all the modes, except  
361 ER waves, which are overestimated, on the contrary. CFSsmcm also simulates an overestimated  
362 ER daily variance, but the CFSv2 overestimation is larger. The black lines are drawn to highlight  
363 the maximum variance region, over which the composite anomalies are averaged to explore the  
364 propagation features of the different CCEWs in Figure 8.

365 The propagation features, shown in Figure 8, are captured reasonably well by CFSsmcm: east-  
366 ward propagating Kelvin and EIG waves and westward propagating ER, WIG and MRG waves.  
367 The phase speeds are simulated marginally slowly (more so for the EIG waves), except for the  
368 MRG waves. The westward shift of the maximum variance, observed in CFSsmcm simulations  
369 in Figure 7, are now prominently visible. In fact, the slower phase speeds are maybe connected  
370 to this westward shift. Overestimation of ER waves is also more evident now. Except for the  
371 ER waves, CFSv2 simulated CCEWs have weak amplitudes. The most striking improvements are  
372 seen in the simulation of the Kelvin, inertia gravity and MRG waves by CFSsmcm, compared to  
373 CFSv2.

374 The 850 hPa and 200hPa composite circulation patterns corresponding to the peak phase of  
375 the different CCEWs are shown in Figures 9 and 10, respectively, superimposed on the corre-  
376 sponding OLR anomalies. The observed circulation features are reasonably well simulated by the  
377 CFSsmcm simulation, including equatorial low level westerlies and low level easterlies conver-  
378 gent to the active convection center for the Kelvin wave (Figure 9b), a the train of cyclonic and  
379 anticyclonic circulation patterns flanking both sides of the equator and collocated with the active  
380 and suppressed centers of convection for the ER wave (Figure 9e), and a train of cyclonic and anti-  
381 cyclonic circulations over the equator binding the convective centers located on the four quadrants  
382 of the circulation pattern for the MRG wave (Figure 9n). The simultaneous meridional and zonal  
383 convergent streamlines in the Kelvin wave composites in the CFSsmcm and to some extent in the  
384 observation are consistent with the structure of Kelvin waves evolving in a meridional jet shear  
385 background (Roundy 2008; Ferguson et al. 2010; Han and Khouider 2010). For the ER (Figure  
386 9d) and MRG (Figure 9m) waves, the convection is underdeveloped south of the equator. Also  
387 the location of the simulated convective centers corresponding to the ER wave (Figure 9d), in the  
388 northern hemisphere, are shifted considerably south of the observed locations (Figure 9e).

389 In the CFSv2 simulated climate, the pattern of the OLR anomalies are realistically captured,  
390 however, they are underestimated, except for the ER waves, which is consistent with Figures 7  
391 and 8. Also except for the ER (Figure 9f) and MRG (Figure 9o), the model misses the major  
392 circulation features. The WIG (Figure 9l) waves particularly look very poorly simulated.

393 At 200hPa, the observed winds (Figure 10, middle column) are reversed relative to 850hPa winds  
394 due to baroclinicity, and are relatively stronger than at 850hPa. The improvements in the winds at  
395 200hPa, are consistent with the improvements seen in the 850hPa level, in CFSsmcm simulations.  
396 To avoid, redundancy we are keeping away from a detailed description of the features observed  
397 in the 200hPa level. Realistic circulations at lower (850hPa) and upper (200hPa) levels indicate  
398 better heating profiles associated with these modes. For CFSv2 as well, the circulation patterns at  
399 200hPa level appears consistent with the 850hPa pattern, in terms of baroclinicity. However, like  
400 in the 850hPa level, the simulated winds are weaker and the circulation patterns lack organization.  
401 Overall, the convection and circulation patterns associated with the different CCEW modes (in  
402 their peak phase) are simulated significantly better in CFSsmcm than in CFSv2 climate.

### 403 *c. MISO*

404 The Indian summer monsoon (ISM) intra-seasonal oscillations (ISO) or MISOs constitute a  
405 major component of the tropical climate variability. Like the MJO, the state-of-the-art climate  
406 models find it difficult to simulate MISOs as well (Lin et al. 2008b; Sabeerali et al. 2013). There is  
407 still a debate on whether the boreal summer MISO mode is distinct from the eastward propagating  
408 MJO mode apart from the fact that the former is prominent in the boreal summer while the latter  
409 is dominant in winter (Lau and Chan 1986; Kikuchi et al. 2012). Nonetheless, the MJO has an  
410 equatorially trapped spatial structure, whereas the MISO shows an off-equatorial structure with  
411 strong convective activity over the south Asian region. In fact, the challenges of simulating MJO

412 and MISO are similar. It is believed that they are both conditioned by a proper representation of  
413 organized convection as in essence they are both a byproduct of the latter. In Section 3a, we have  
414 seen that CFSsmcm has significantly improved the simulation of the MJO compared to the host  
415 model. Moreover, GKPM16 has shown that the distribution of rainfall has also improved in the  
416 CFSsmcm simulation, especially over India.

417 In order to investigate the ISM intra-seasonal variability, we plot the north-south version of the  
418 TWK-spectra and the conventional east-west TWK-spectra (Figure 11), but for the boreal summer  
419 season only. Noteworthy, for the north-south TWK-spectra (Figure 11a, c and e), wavenumber 1  
420 corresponds to 50 degrees of latitude (from 20°S to 30°N). In observations (Figure 11a), we notice  
421 a northward propagating mode with time-period of about 45 days and centered at wavenumber 1.  
422 For comparable time period and wavenumber, a southward component is also noted, but with less  
423 power.

424 Comparing the model simulations (CFSsmcm in Figure 11c and CFSv2 in Figure 11e) with ob-  
425 servations, we can see that both models capture the northward and southward propagating compo-  
426 nents but with a longer time period of about 60 days, though the CFSsmcm signal seems to extend  
427 to higher frequencies. Also, the power in the MISO modes is slightly underestimated, more so  
428 in CFSsmcm simulations. In the east-west TWK spectra (Figure 11b), the dominant power, seen  
429 around wavenumbers 1 to 2 and time period 45-days, correspond to the eastward moving ISOs  
430 or MJOs. The power in the 45-day time period in both the north-south and east-west spectra is  
431 consistent with the fact that MISOs predominantly propagate northeastward (Lau and Chan 1986;  
432 Goswami 2012). Eastward moving Kelvin waves and westward ER waves are also seen in the  
433 spectra (Figure 11b). Power in the 10-20 day range propagating westward indicate 10-20 day  
434 high-frequency ISOs (Goswami 2012).

435 In the two model-simulations, CFSsmcm simulated spectra (Figure 11f) looks more realistic than  
436 that of CFSv2 (Figure 11d). CFSv2 simulates unrealistic eastward power at higher wavenumbers.  
437 Also it simulates spurious power all along the positive wavenumber axis. The westward ER wave  
438 power is overestimated by CFSv2 and it peaks at a much longer period  $\sim 60$  days. Also it simulates  
439 a weak power for the Kelvin waves, which is consistent with Figure 1b and the 10-20 day westward  
440 ISO power is underestimated.

441 The eastward power at 60-days in Figure 11f is consistent with the power at the same time-period  
442 in Figure 11e. This indicates a possibility of realistically simulating the northeastward movement  
443 of the monsoon trough by CFSsmcm. Noteworthy, the lack of eastward power about wavenumber  
444 1 (Figure 11d), in the backdrop of CFSv2 simulating reasonable power corresponding to associ-  
445 ated northward propagation, is unrealistic and raises suspicion about the simulated propagation  
446 mechanism. CFSv2 fails to simulate the desired power for the westward propagating 10-20 day  
447 high frequency monsoon ISOs and it is slightly overestimated in the CFSsmcm simulations. In the  
448 remainder of this section, we analyze the 45-day MISO or simply the MISO dynamical structure  
449 and physical properties as simulated by the two models. In order to isolate the MISO anomalies,  
450 we apply space-time filtering as we have done in Sections 3a and 3b. However, we apply the filter  
451 only for the boreal summer data. Based in the spectra shown in Figure 11 we use the filter with the  
452 time-period range 20-100 days and wavenumber range 0 to 4. The anomalies isolated for MISO  
453 are plotted in Figures 12, 13 and 14.

454 The boreal summer MISO daily OLR variance is plotted in Figure 12a, b and c for CFSsmcm,  
455 observation (NOAA OLR data), and CFSv2, respectively. In observations (Figure 12b), the maxi-  
456 mum variance is located in the northern Indian Ocean and West Pacific, with the peak at the head  
457 of the Bay of Bengal and the high variance contours displaying a northwest-southeast orienta-  
458 tion. Noteworthy, all the high variance zones are over the oceanic regions in both models and

459 observations. The titled orientation of the variance pattern is missing in both model-simulations  
460 (CFSsmcm in Figure 12a and CFSv2 in Figure 12c). CFSsmcm fails to capture the peak in the  
461 head of the Bay of Bengal region. It is shifted over the Arabian Sea, instead. In fact, the whole  
462 CFSsmcm MISO variance, over the west Pacific, is shifted eastward and it is slightly overesti-  
463 mated. This eastward variance shifting is symptomatic and it is utterly consistent with that of the  
464 MJO and CCEWs in Figures 2 and 7. It won't be surprising if the ocean model is the culprit and a  
465 thorough investigation of this matter is warranted.

466 CFSv2 simulates an overestimated peak at the head of the Bay of Bengal and also highly over-  
467 estimates the variance over the west Pacific. This is consistent with the work of Goswami et al.  
468 (2014, 2015) who showed that CFSv2 tends to overestimate the low frequency ISV. As per the  
469 variance plots, both models have difficulties in simulating the daily variance of the MISO though  
470 the CFSsmcm simulation has noticeable improvements. The most significant improvements in-  
471 clude an extended power over continental India as in the observations and a significant reduction  
472 of the faulty power over the Western Pacific and the Bay of Bengal.

473 The northward propagation feature of MISO is examined by plotting meridional Hovmöller  
474 diagrams, averaged over 65°E-95°E, of a composite of the MISO filtered OLR anomalies. The  
475 composite is constructed based on an index located over the Bay of Bengal. In observations  
476 (Figure 12e), the MISOs start migrating approximately from 10°S and go up to 25°N with a phase  
477 speed of about 1.5°Lat per day. CFSsmcm (Figure 12d) captures this phase speed realistically and  
478 the convection starts migrating from about 10°S as in the observations. The amplitude, however,  
479 is weaker. CFSv2 simulated MISOs (Figure 12f) also appear to be weaker than the observations  
480 but they also have a slower northward propagation and the migration starts right at the equator,  
481 unlike in the observations and in CFSsmcm. The convection pattern south of 10°S looks spurious  
482 in the CFSv2 simulation.

483 The composite circulation patterns for the peak MISO phase at 850 hPa superposed on the  
484 corresponding OLR anomalies are shown Figure 12g, h and i for CFSsmcm, observations (OLR  
485 from NOAA and winds from NCEP) and CFSv2, respectively. In the observations (Figure 12h),  
486 a monsoon trough like organization is evident in the OLR anomalies. This is accompanied by a  
487 Rossby gyre-type pair of cyclonic circulations at 850 hPa with a fetch of easterlies emanating from  
488 the northern Pacific Ocean blowing over India and strong westerlies over the Indian Ocean. These  
489 features are similar to what we had observed in the circulation patterns for the MJO peak phase  
490 shown in Figure 5b, at least for the Rossby gyres and Equatorial westerlies somewhat lagging the  
491 convection core. A third gyre can be seen in north-eastern India.

492 The CFSsmcm (Figure 12g) simulated OLR anomalies rather appear to have a “blob” like struc-  
493 ture instead of a monsoon trough-like orientation in the sense that it is not extended in the north-  
494 westward direction. However it captures the cyclonic circulation slightly north of the convection  
495 maximum reasonably well but it underestimates the one to the south; it is somewhat shifted to the  
496 west allowing the north-westerly winds to penetrate into the Arabian sea. This is perhaps con-  
497 nected with the lack of elongation of the OLR signal. The easterlies over the Pacific are captured  
498 reasonably well, but they look wobbly. The third gyre is shifted North East. Again, these biases  
499 are somewhat similar to the issues discussed while describing CFSsmcm simulated circulation  
500 pattern for the peak MJO phase (Figure 5a).

501 In the CFSv2 simulation (Figure 12i), the peak of the OLR anomalies are heavily shifted east-  
502 ward compared to the observations. Moreover the simulated monsoon trough-like OLR anomaly  
503 pattern has an overestimated meridionally oriented component extending to 30°S. Nevertheless,  
504 the CFSv2 simulated circulation pattern, looks reasonably simulated and somewhat better than  
505 CFSsmcm (comparing Figure 12g and i) except for the fact that the Southern tail of the whole  
506 pattern is shifted to the west.

507 The winds over the west Pacific are observed to be dominantly westerlies, at 200hPa. Comparing  
508 the observed winds at 850hPa and 200hPa we note, the circulations are neither dominantly baro-  
509 clinic nor barotropic. This is a feature of the MISO that is different from MJO, which is dominantly  
510 baroclinic (Figures 5b and e). This in fact makes the MISOs a dynamically complex component of  
511 the tropical climate and a difficult feature for the climate models to simulate. CFSsmcm simulates  
512 the 200hPa circulation patters for the peak MISO phase with considerable fidelity. However, the  
513 cyclonic circulation north-west of the convection maximum looks unrealistic. Nevertheless, com-  
514 paring the 200hPa circulation relative to the 850hPa winds in the CFSsmcm simulations, the model  
515 seems to capture the baroclinic-barotropic nature of the MISO circulation reasonably well. CFSv2  
516 simulated winds at 200hPa (Figure 12l) shows limited ability in simulating the major observed  
517 features. The fact that, CFSv2 simulates a “too much” meridional orientation of the convective  
518 band by simulating the 850hPa circulation with considerable success while missing the major cir-  
519 culation features at the 200hPa level suggests the possibility of an unrealistic dynamics in the  
520 model.

521 In Figure 13, the vertical structure of the MISO mode is examined. This figure is similar to Fig-  
522 ure 6, but for MISO. The panels in Figure 13 show the height-latitude cross sections of different  
523 fields averaged over 70°E-90°E. The top panels show the meridional variation of the correspond-  
524 ing OLR with the minimum indicating the peak convection. In observations, the convection peak  
525 is seen at around 7°N (Figure 13h). The impression of the cross-equatorial south-westerly ISM  
526 low level jet is seen in Figure 13i, where the zero (meridional) shear line is slightly north (10°N) of  
527 the convection maximum. Around the same location, “convergence below and divergence aloft”  
528 feature is seen in Figure 13j. Positive moisture anomalies with a significant southward tilt domi-  
529 nate the atmosphere south of 17°N and negative anomalies northward beyond 17°N (Figure 13k).  
530 At about 10°N, negative temperature anomalies are seen at the lower troposphere (below 600hPa)

531 and positive anomalies aloft (Figure 13l). Updrafts throughout the atmospheric column are col-  
532 located with the convection maximum (Figure 13m). The updraft maximum is led by downdraft  
533 northward and followed (from the south) by a region of mild updraft in the middle troposphere  
534 and downdraft in the lower and upper troposphere. The diabatic heating shows positive anoma-  
535 lies collocated with the convection peak, the maximum heating being at 400-500hPa (Figure 13n).  
536 The positive anomalies are led by negative anomalies northward and followed by mild positive  
537 anomalies in the middle troposphere. The observational features noted above are consistent with  
538 the shear-vorticity driven northward propagation mechanism of MISOs (Jiang et al. 2004).

539 CFSsmcm reasonably captures the major features, as noted in the OLR meridional profile, zonal  
540 wind, convergence and temperature anomalies (Figures 13a, b, c and e). The only major concerns  
541 of the CFSsmcm simulation are the dry moisture bias immediately south of the equator at about  
542 5°S (Figure 13d) and the very narrow and highly overestimated values of updrafts (Figure 13f) and  
543 diabatic heating rates (Figure 13g). In the CFSv2 simulation (Figure 13 o-u) all the fields are found  
544 to have major biases. The biases in the zonal wind, convergence and moisture fields are particularly  
545 grave in the backdrop of the fact that CFSv2 simulates reasonable northward propagating MISOs.  
546 The lack of barotropic shear vorticity line and the northward tilt of the moisture anomalies are  
547 particularly disturbing. It seems like CFSv2 captures the northward propagation of MISO for the  
548 wrong reasons.

549 There is significant evidence that it is the low level moisture convergence north of the convection  
550 maximum that drives the convection northward. Jiang et al. (2004) argued that, a heat source in  
551 the presence of an easterly mean flow leads to a cyclonic barotropic vorticity centered slightly to  
552 the north, which in turn drives the frictional convergence in the boundary layer, consistent with  
553 the finding of De La Chevrotière and Khouider (2017) who coupled the SMCM to an idealized  
554 three layer zonally symmetric model monsoon-like simulation. In a recent study, Hazra and Kr-

555 ishnamurthy (2015) argued that moisture anomalies may provide the necessary preconditioning  
556 to promote the northward propagation of MISOs, a mechanism analogous to the preconditioning  
557 mechanism in the case of MJOs (Jiang et al. 2011; Khouider et al. 2011). Abhik et al. (2013) also  
558 argues in favor of preconditioning mechanism for northward propagation of the MISOs. Note-  
559 worthy, the SMCMM framework is in fact designed to mimic the congestus preconditioning in the  
560 tropics (KM06; KBM10). Motivated by these arguments, we have explored the preconditioning  
561 mechanism for the northward propagation of the MISOs.

562 Figure 14 shows the composite phase-wise latitude-height cross-sections (averaged over 70°E-  
563 90°E) of diabatic heating superimposed on the moisture anomalies. The observations are shown on  
564 the left column. The red contours show the heating associated with the MISO convection, which  
565 starts over the equator and propagates poleward till 20°N. The heating shows a top heavy vertical  
566 structure with the peak heating observed around 400-500hPa level attaining a maximum value of  
567  $2 \text{ K day}^{-1}$  in phases Lag 0 and Lag +5. The associated specific humidity field, shown in shading,  
568 indicates a bottom heavy profile with positive moisture anomalies leading the heating maximum in  
569 the lower troposphere synonymous of moisture preconditioning ahead of the convection (indicated  
570 by the heating contours) driving the convection northward. Both the heating and moisture fields  
571 exhibit a north-south vertical tilt, leaning backward at the upper troposphere.

572 The CFSsmcm simulation (in the middle column) captures this tilted structure reasonably well,  
573 in both the heating and moisture fields despite a few discrepancies, such as an overestimation of the  
574 heating maximum, an earlier peak, and a limited poleward extension. Nevertheless, the moisture  
575 preconditioning ahead of the convection maximum is captured well. This preconditioning feature  
576 is missing in the CFSv2 simulation (extreme right column). The positive and negative heating  
577 contours are in phase with the positive and negative shadings of specific humidity, respectively.

578 Arguably, the titled vertical structure is backward (and more prominent in the negative heating  
579 contours) compared to the observations.

580 The heating is overestimated in the CFSv2 simulation as in the CFSsmcm simulation, however  
581 in CFSv2 the poleward propagation is observed to reach 20°N like the observations. However,  
582 the moisture and heating maxima are in phase and CFSv2 seems to lack the main moisture pre-  
583 conditioning mechanism, which raises the same questions as the the MJO. In the absence of the  
584 preconditioning mechanism, what is the mechanism responsible for the northward propagation of  
585 the MISOs in the CFSv2 simulated climate? Noteworthy, CFSsmcm simulated MISO northward  
586 propagation mechanism appears to be consistent with the hypothesis of Jiang et al. (2004) and it  
587 is realistic. The realistic simulation of the moisture preconditioning in CFSsmcm climate is un-  
588 doubtedly related to the prescribed cloud-type trilogy in the SMCM formulation and its ability to  
589 simulate the other TISV modes.

#### 590 **4. Discussion and Conclusion**

591 A 15 year simulation with NCEP's coupled climate model CFSv2, in which a new stochastic  
592 multcloud model (SMCM) cumulus scheme was implemented (GKPMM16; GKPMM17), CF-  
593 Ssmcm for short, was analyzed here against a control simulation of the same length and same  
594 initial conditions, in terms of its ability to capture the main modes of tropical variability on syn-  
595 optic and intra-seasonal scales, including the MJO, CCEWs, and MISO. NOAA OLR (Liebmann  
596 and Smith 1996) and NCEP reanalysis fields (Kalnay et al. 1996) are utilized as an observational  
597 benchmark. SMCM aims to capture the statistics of the subgrid variability of the three cloud  
598 types, cumulus congestus, deep and stratiform (KBM10, Frenkel et al. 2012; Peters et al. 2013;  
599 De La Chevrotière et al. 2015), that are observed to characterize multi scale tropical convective  
600 systems, including the MJO and CCEWs (e.g., Johnson et al. 1999; Mapes et al. 2006; KM06). As

601 such CFSsmcm captures most of the spectrum of tropical intra-seasonal variability with great fi-  
602 delity including many of their physical and dynamical features while the control model performed  
603 very poorly overall. Though there is still room for further improvements, the performance of CF-  
604 Ssmcm is somehow expected based on the previous successes of the SMCM in the context of an  
605 aquaplanet atmospheric GCM (Khouider et al. 2011; Ajayamohan et al. 2013, 2014, 2016; Deng  
606 et al. 2015, 2016) and the fact that the SMCM is rooted from the thoroughly documented theo-  
607 retical framework of the multcloud model for convectively coupled waves (KM06; Khouider and  
608 Majda 2007, 2008a,b; Han and Khouider 2010).

609 The first striking improvement is seen in terms of the Takayabu-Wheeler-Kiladis diagram  
610 (Takayabu 1994; Wheeler and Kiladis 1999) in Figure 1. While the control run, CFSv2, has a lim-  
611 ited skill in this regard, CFSsmcm shows significant improvements essentially by adding power to  
612 the Kelvin, MRG, WIG, and EIG waves, most of which are weaker or inexistent in the control run.  
613 The MJO frequency is also improved. Nonetheless, the superiority of the CFSsmcm simulation  
614 is more appreciated when digging deeper and looking at the physical and dynamical features of  
615 these waves.

616 The physics and dynamics of the MJO are presented in Section 3a. In terms of the geographi-  
617 cal distribution of MJO variance, while both simulations exhibit a fair amount of power over the  
618 bulk area of the tropical warm pool, they both show some limitations when compared to NOAA  
619 OLR. While CFSsmcm suffers from a severe westward shift of the variance maximum, the control  
620 run exhibits an unrealistic double peak, each of which are located on either side of the equator  
621 somewhat reminiscent of the double ITCZ problem. One of the most visible striking outperfor-  
622 mance of CFSsmcm comes in terms of the propagation of MJO filtered OLR composites in Fig 3.  
623 While CFSsmcm shows a clear propagating blob of low OLR, with the right phase speed and ge-  
624 ographical location and amplitude as in the observation, CFSv2 fails miserably in this regard. The

625 same consistent behavior is seen in the Hovmöller plots in Figure 4. Also the famous quadruple  
626 vortex structure and associated baroclinicity of the MJO (Rui and Wang 1990; Hendon and Salby  
627 1994; Majda and Biello 2003; Kiladis et al. 2005; Zhang 2005; Majda and Stechmann 2009b)  
628 are reasonably captured by CFSsmcm while the horizontal flow structure of CFSv2 is completely  
629 disorganized.

630 The vertical structure in Figure 6 raises the question whether the MJO-power spectrum peak  
631 exhibited by the CFSv2 simulation in Figure 1 has anything in common with the MJO as a physi-  
632 cal mode of tropical variability. CFSv2 lacks the most fundamental dynamical and morphological  
633 features such as the absence of a pronounced OLR minimum or any of the fundamental character-  
634 istics of the dynamical fields while CFSsmcm compares relatively well to the observations in all  
635 aspects, including the backward tilt in moisture, horizontal wind, and temperature (Kiladis et al.  
636 2005, for e.g.). In particular, the persistence of low-level moistening and congestus (low-level)  
637 heating, during the suppressed phase of the MJO for about two to three weeks (60-80 deg divided  
638 by  $5 \text{ m s}^{-1}$ ) prior to the MJO active convection, which is observed in both the CFSsmcm simu-  
639 lation and the reanalysis MJO plots but absent in the control CFSv2 MJO, as illustrated in Figure  
640 6, is consistent with the idea that congestus heating serves to moisten the environment prior to  
641 deep convection as demonstrated by in situ observation from the CINDY/DYNAMO observation  
642 campaign (Johnson and Ciesielski 2013; Bellenger et al. 2015; Chen et al. 2015) and detailed nu-  
643 merical and theoretical studies (Derbyshire et al. 2004; Waite and Khouider 2010; Hohenegger  
644 and Stevens 2013; Hirons et al. 2013).

645 The faithful representation of the MJO's main physical and dynamical features in the CFSsmcm  
646 simulation stems from the design principles of the stochastic multcloud model based on the self-  
647 similar morphology and dynamics of multiscale tropical convective systems (KM06; KBM10)  
648 and is consistent with previous studies using the deterministic and stochastic multcloud model

649 (Khouider et al. 2011; Ajayamohan et al. 2013, 2014, 2016; Deng et al. 2015, 2016). The su-  
650 periority of the stochastic simulation, as opposed to global simulations using the deterministic  
651 MCM, comes from the fact that the stochastic model is able to simultaneously put variability at a  
652 wide range of scales, ranging from meso- to planetary scales in an intermittent fashion (Frenkel  
653 et al. 2012, 2013; Deng et al. 2016) although, the main linear instabilities, for the considered  
654 parameter regimes, exhibited by the (deterministic) multi-cloud model, occur at synoptic scales  
655 (KM06; Khouider and Majda 2008a,b; Han and Khouider 2010; Khouider et al. 2012).

656 The simulation of Kelvin waves has always been found to be good in the CFSsmcm simulation  
657 even during the tuning of the model (Goswami et al. 2017). We presume that these improvements  
658 in the CCEWs come by virtue of a better simulation of the convective heating profiles, which take  
659 into account the proper dynamical interactions of the three cloud types with the large-scale mois-  
660 ture and other thermodynamical fields. Although currently we do not have a solid evidence to sup-  
661 port this claim, the improvements in the associated circulation patterns, shown in Figures 9 and 10,  
662 backs this well, consistently with the design principle of the multcloud model (KM06; Khouider  
663 and Majda 2008a,b; Han and Khouider 2010; Khouider et al. 2011, 2012).

664 Except for equatorial Rossby waves, CFSv2 shows very little to no power in terms of the dis-  
665 tribution of OLR variance of CCEWs as shown in Figure 7 while CFSsmcm performs relatively  
666 well in this regard. However, there are some visible discrepancies when comparing CFSsmcm to  
667 the observations, including a westward shift of the maximum variance, particularly for the Kelvin  
668 and EIG waves, and differences in amplitude. The westward variance shift is consistent with that  
669 of the MJO and it won't be surprising if they have the same common origin. Curiously, these are  
670 all eastward moving signals. The propagation characteristics and horizontal structures of these  
671 waves are equally well captured by the CFSsmcm simulation according to Figures 8 through 10.  
672 It has to be noted at this point that these are T126 simulations and some features of these waves

673 (such as their convective cores) are represented by less than 5 grid points in one horizontal direc-  
674 tion. CFSv2 does a good job in representing the structure and propagation of the equatorial Rossby  
675 waves consistent with the spectral power plot in Figure 1.

676 Last but not the least, the capability of CFSsmcm to capture the physical and dynamical features  
677 of the Indian MISO is assessed in Section 3c and Figures 11 through 14. First, from Figure 12, the  
678 distribution of the MISO-filtered OLR variance is captured relatively well compared to the control  
679 run that puts too much power over the western Pacific and the Bay of Bengal. Also the northward  
680 propagation over the Indian Ocean and continental India, which appears to be too slow and has a  
681 too weak amplitude and starts migrating right at the equator instead of  $10^{\circ}\text{S}$ , in the CFSv2 control  
682 run, is considerably corrected in the CFSsmcm run. Moreover, while the vertical structure of this  
683 mode is well captured by CFSsmcm, compared to observation as shown in Figure 13, hinting to the  
684 shear vorticity–moisture preconditioning mechanism (Jiang et al. 2004; Abhik et al. 2013; Hazra  
685 and Krishnamurthy 2015) being at work, the CFSv2 MISO signal has too little in common with  
686 this mechanism. Arguably, the northward propagating ISO in CFSv2 obeys completely different  
687 physics than what actually occurs in nature and the same can be said about its MJO. Indeed,  
688 the fact that CFSsmcm captures the physical and dynamical features of the main tropical modes  
689 of variability is not a matter of serendipity but can be rooted to the theoretical foundation and  
690 empirical evidence of the SMCM (KBM10; Peters et al. 2013; De La Chevrotière et al. 2015)  
691 and its parent deterministic multcloud model (KM06) which is build based on intuition guided by  
692 observations (Lin and Johnson 1996; Johnson et al. 1999). The results shown here are yet another  
693 demonstration that tropical convective variability is both multiscale and self-similar in nature and  
694 most of it can be explained by the complex interactions of the three key cloud types, congestus,  
695 deep, and stratiform, with the dynamical and moisture fields, by shaping up the vertical structure  
696 of the diabatic heating, on multiple time and spatial scales.

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1024 j.dynatmoce.2009.12.003.

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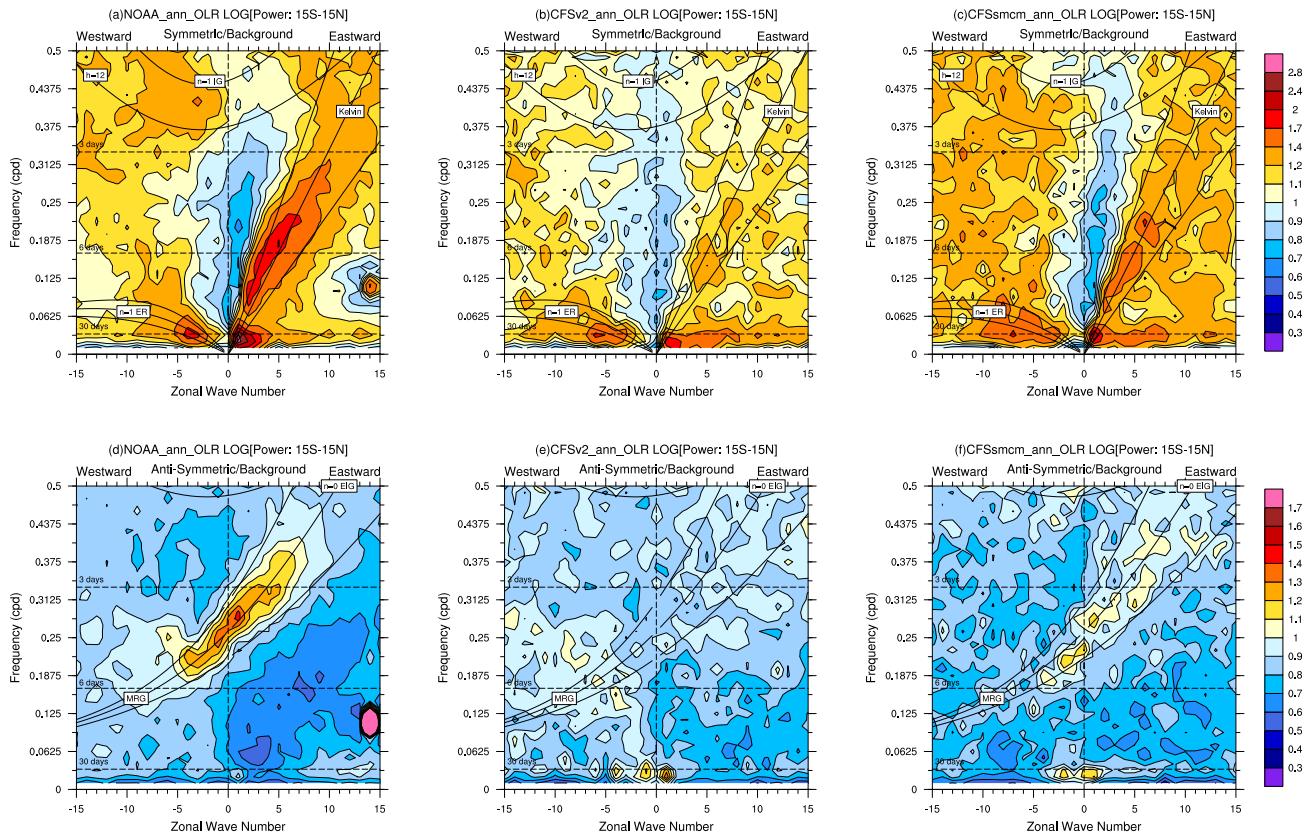
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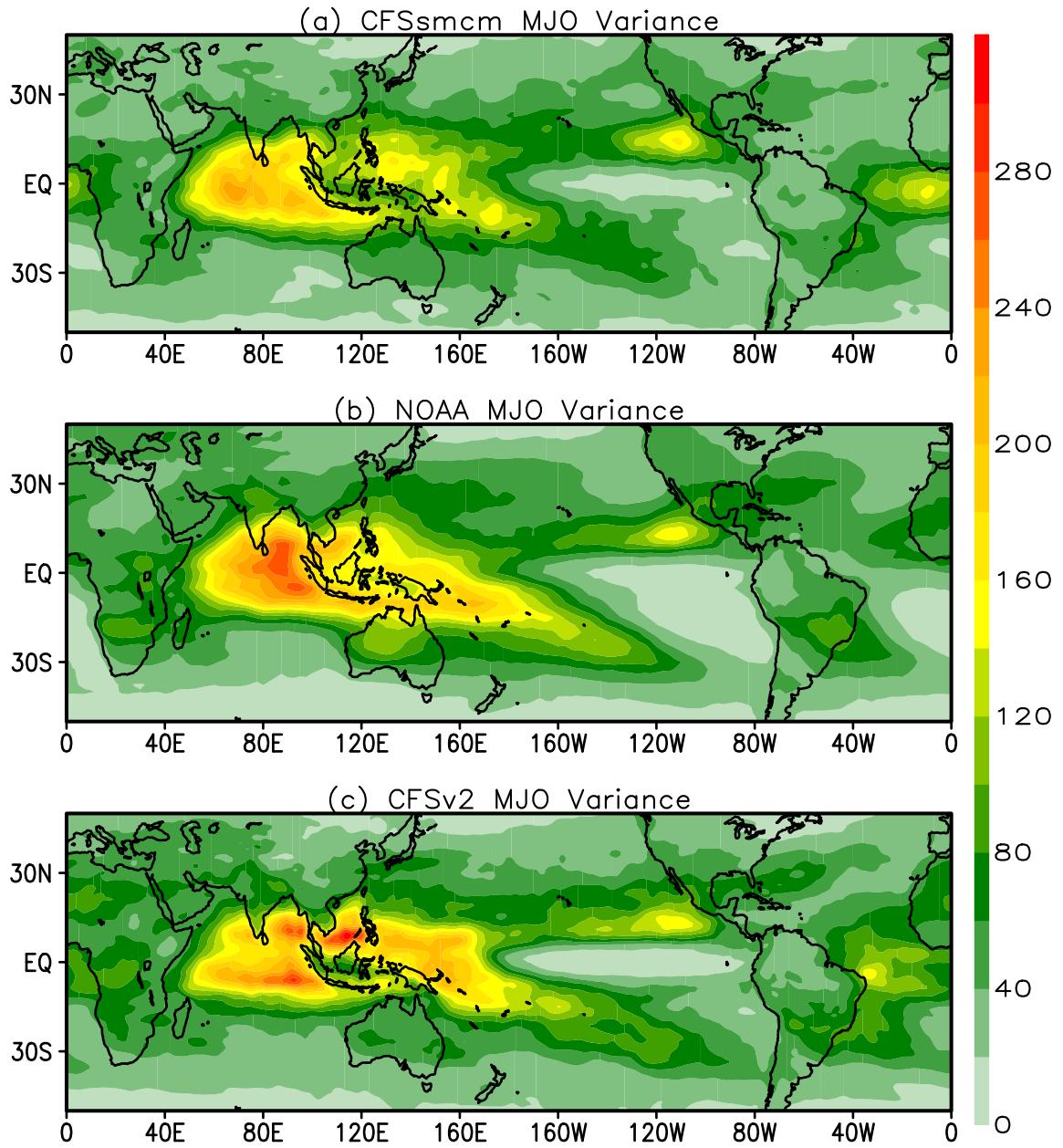
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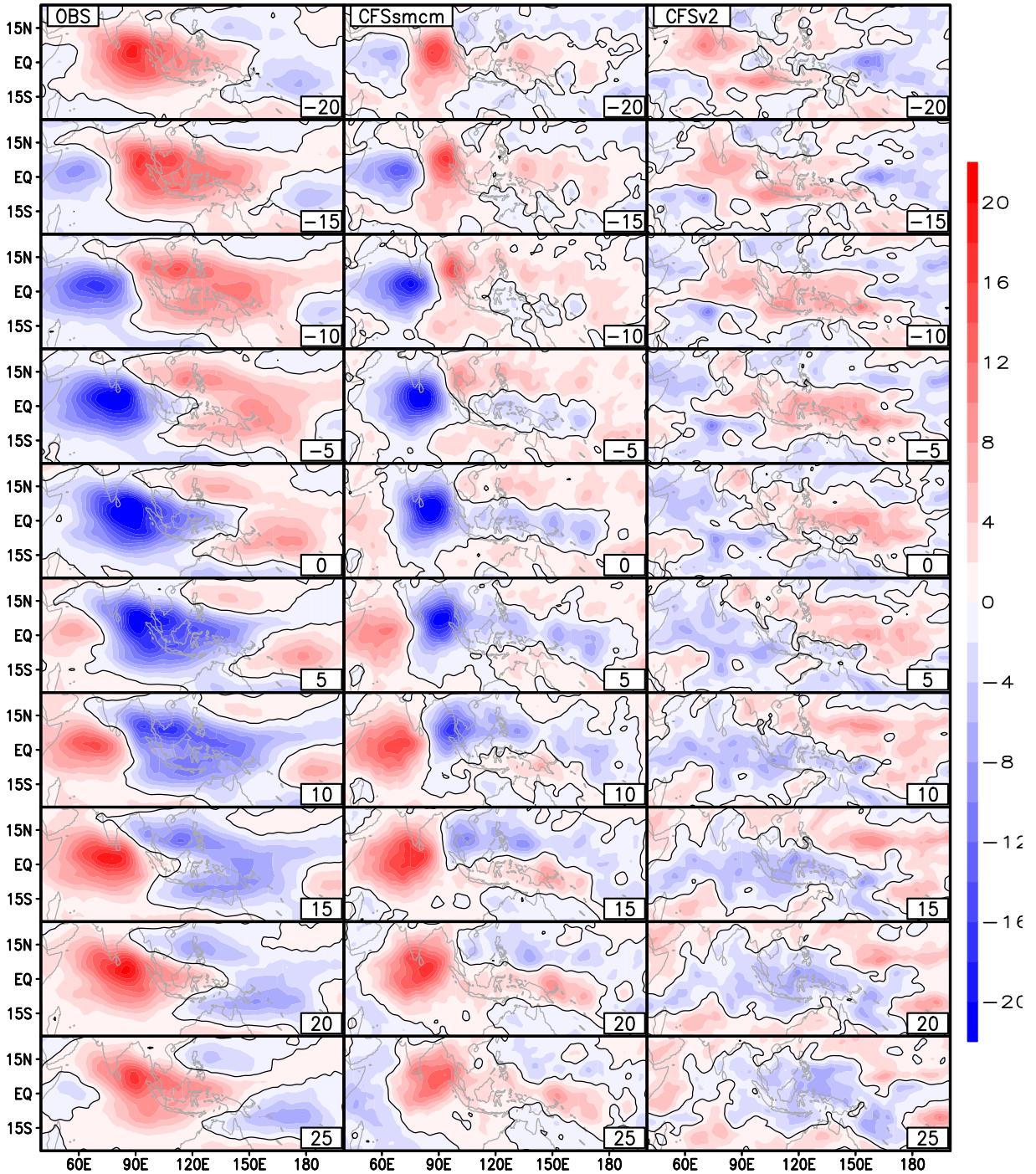
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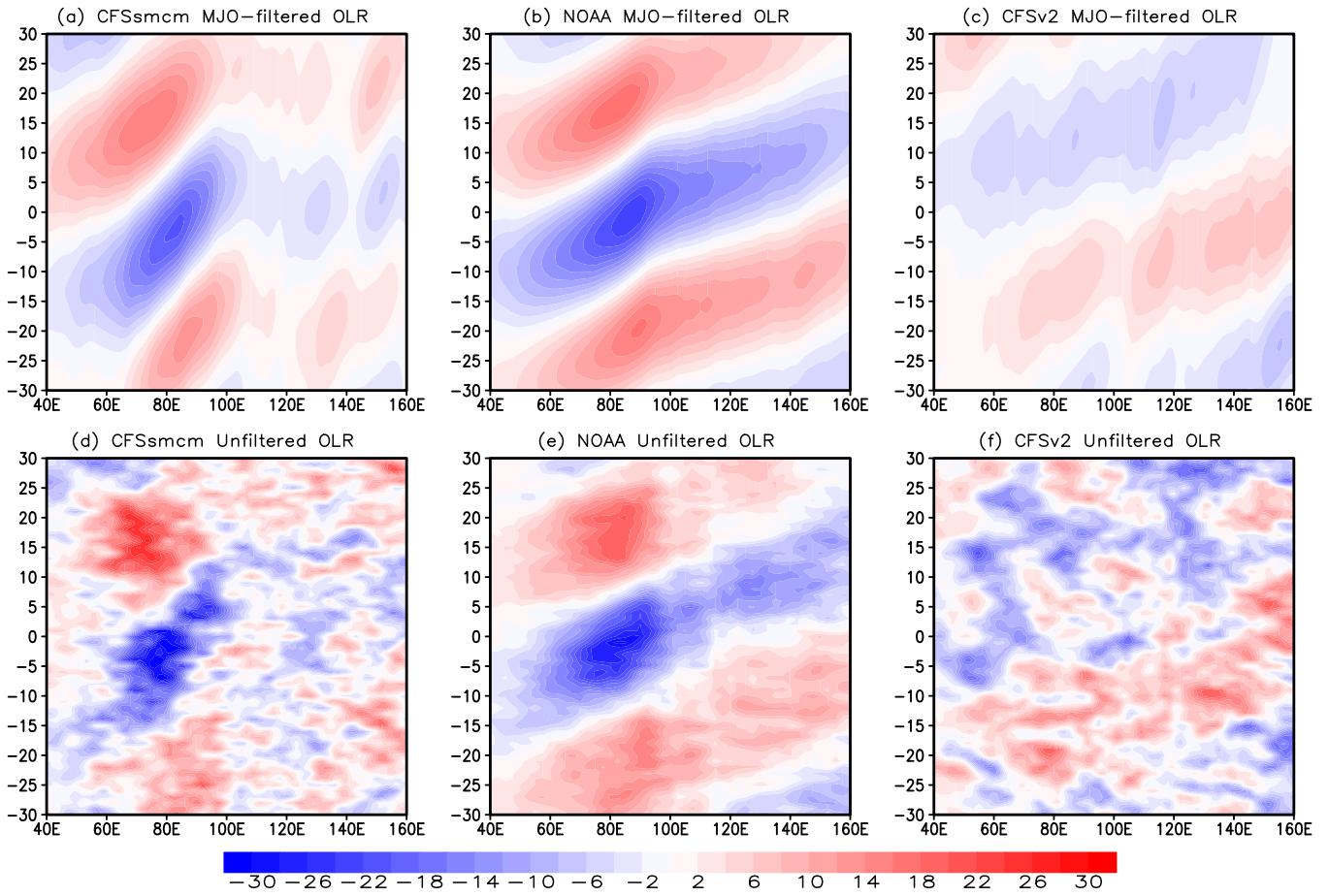
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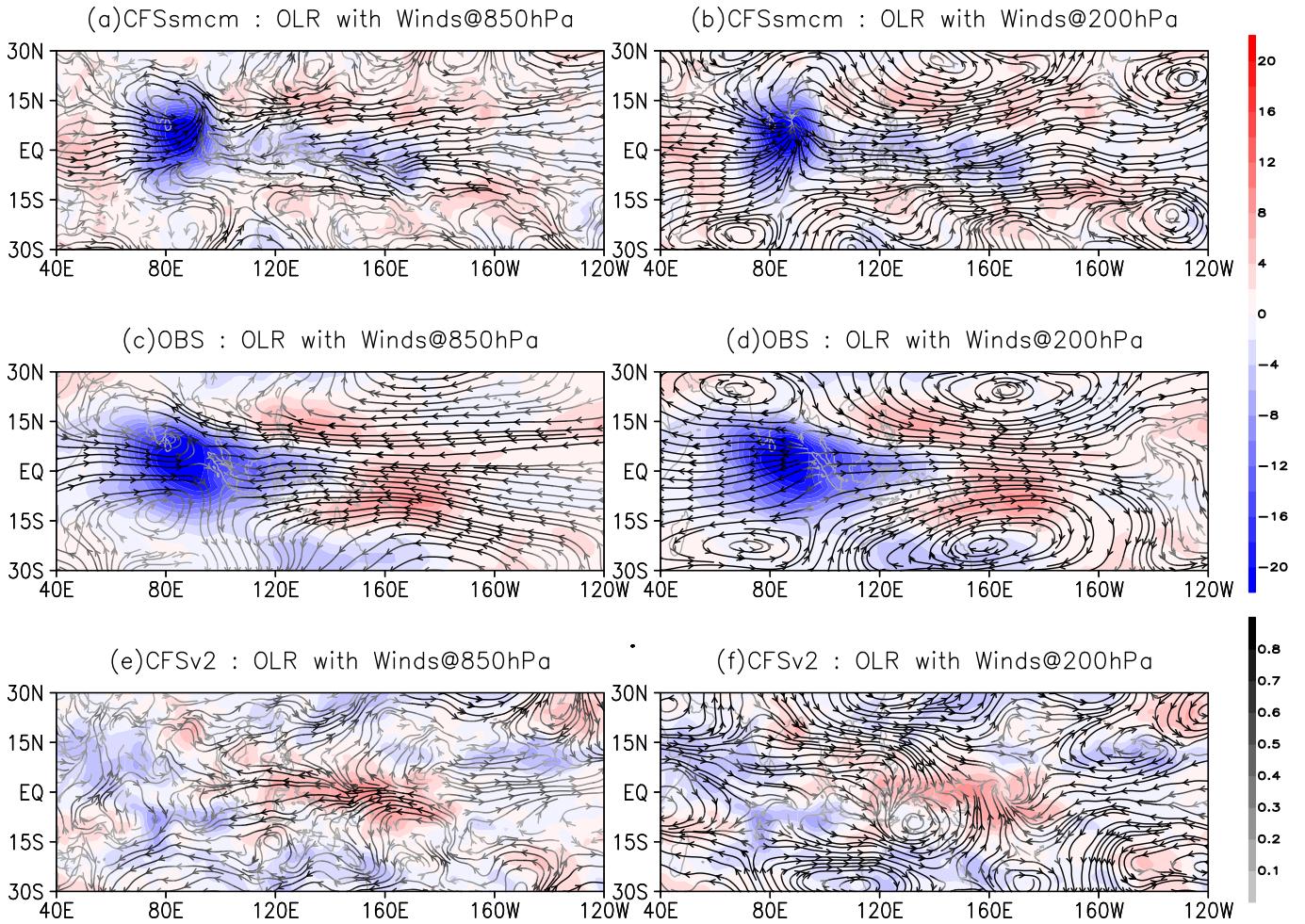
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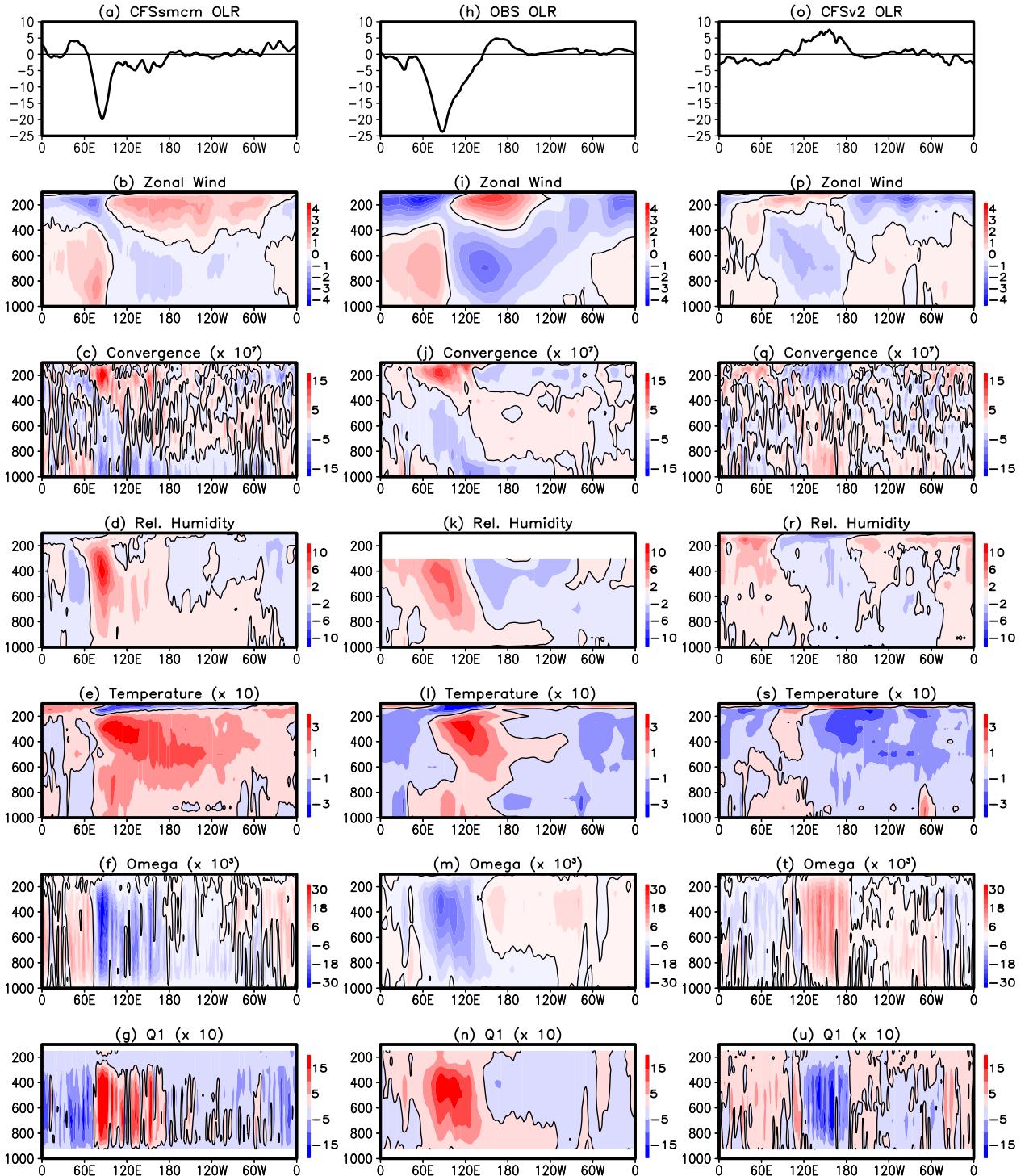
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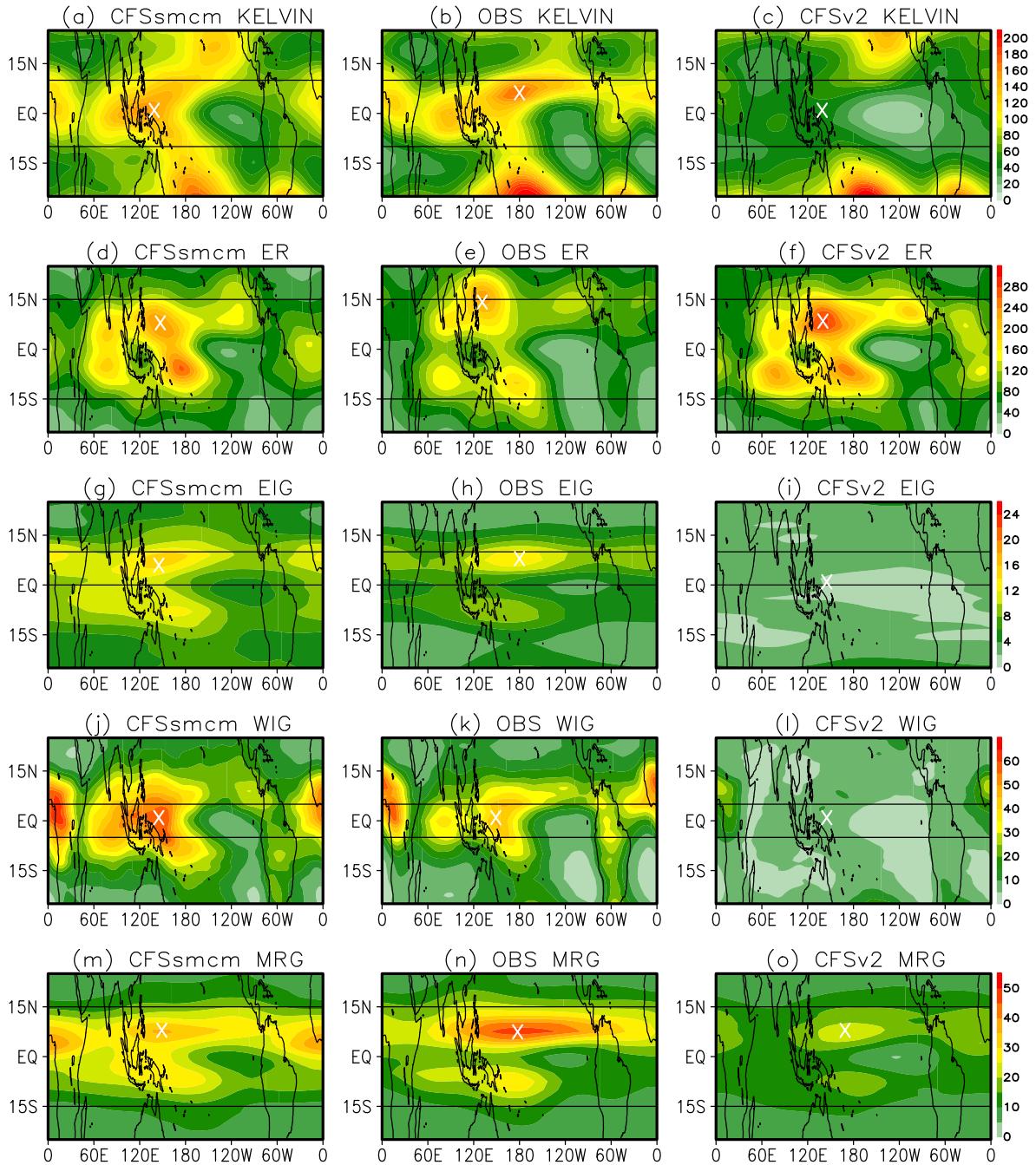
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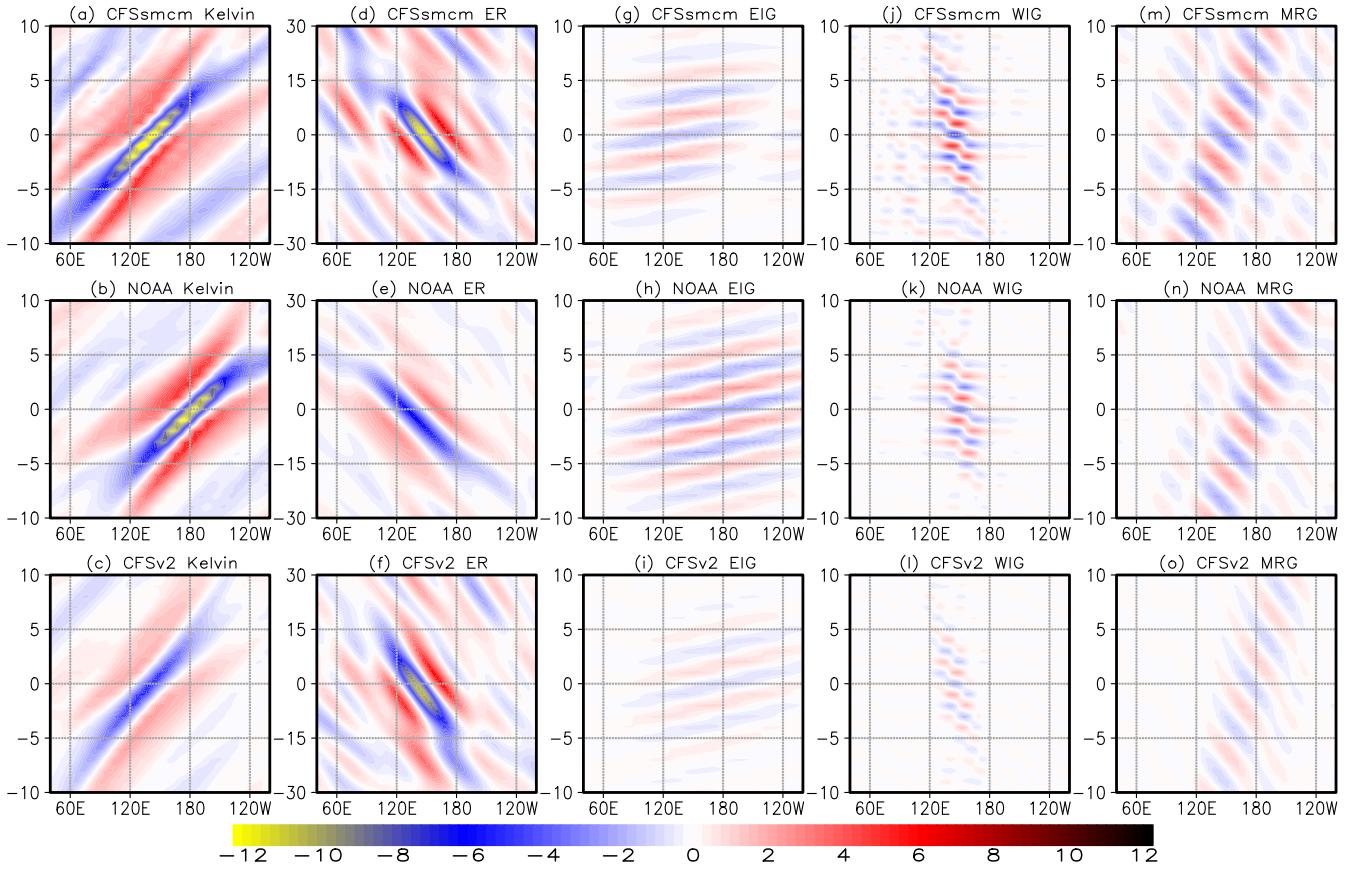
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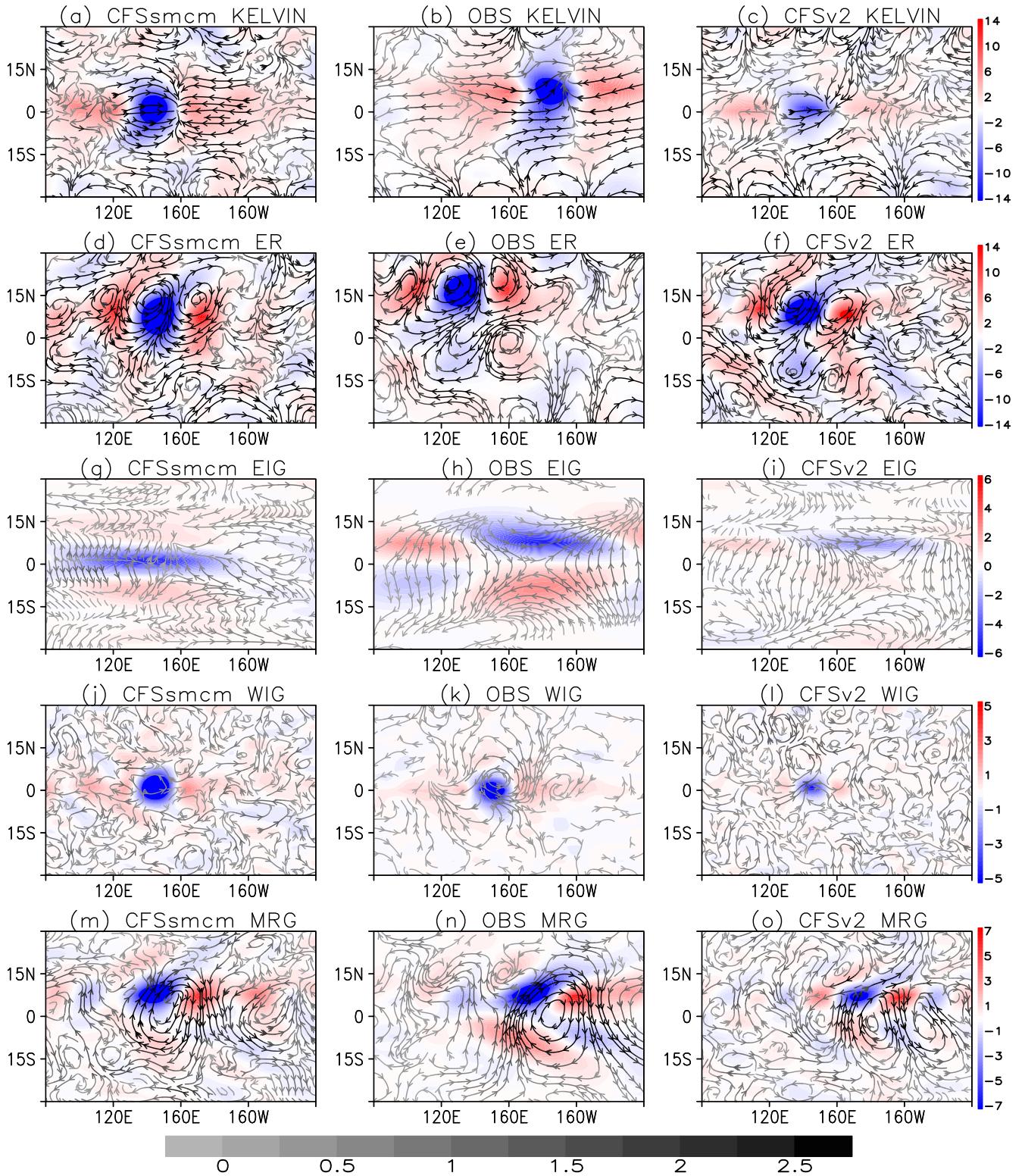
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1105 FIG. 7. Daily variance of different equatorial waves, for OLR ( $(W\ m^{-2})^2$ ) anomalies filtered respectively for  
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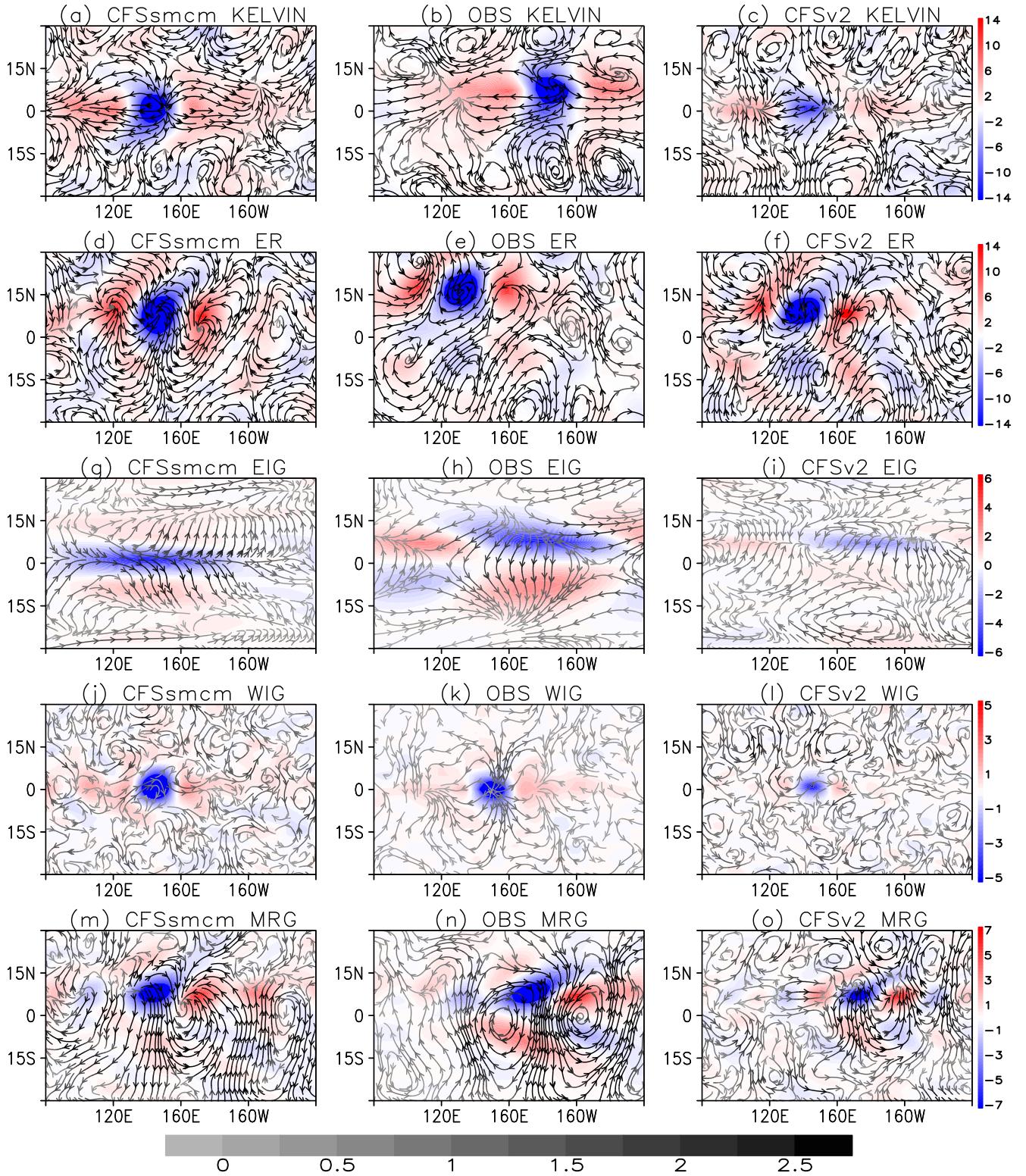
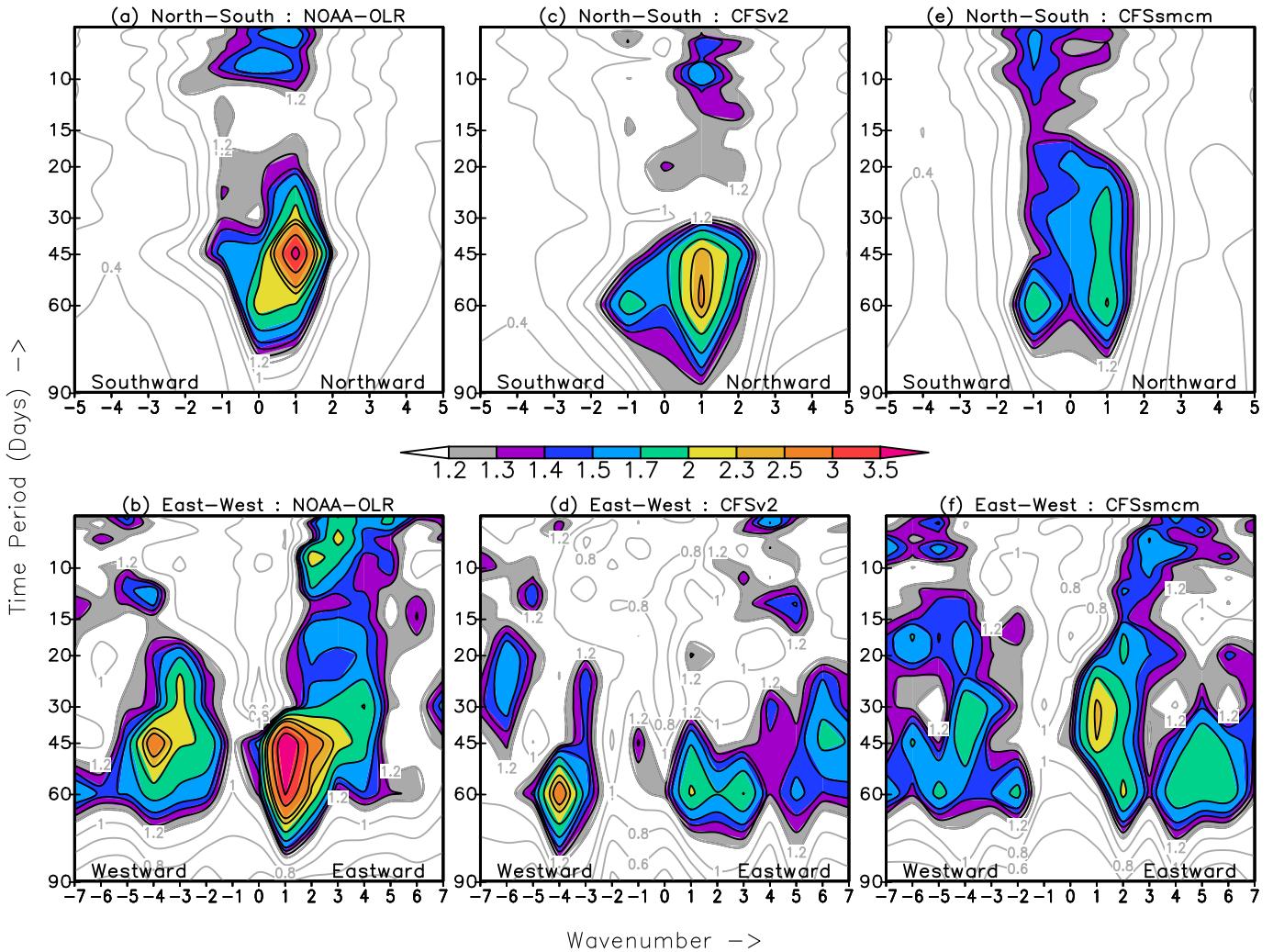
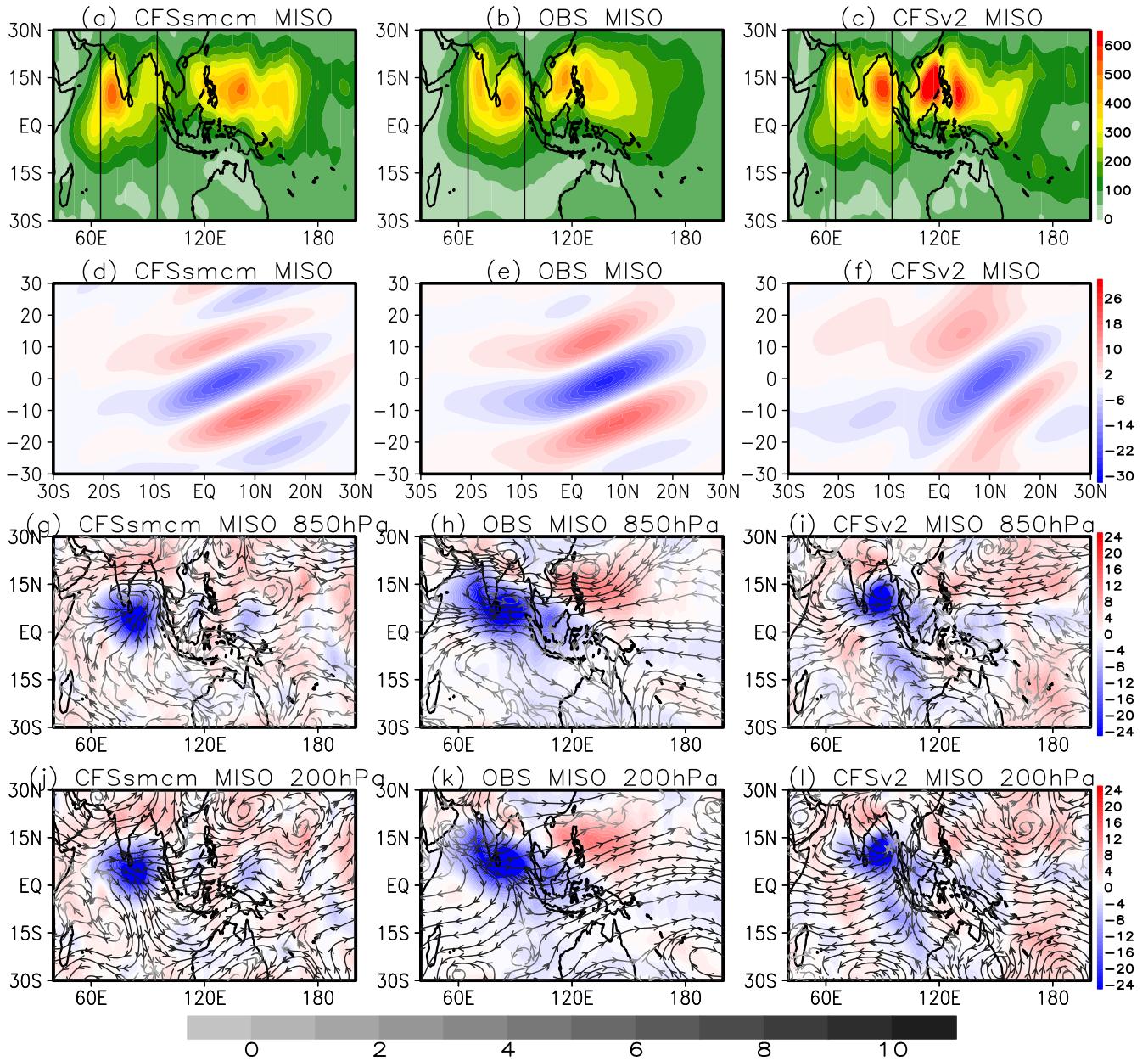


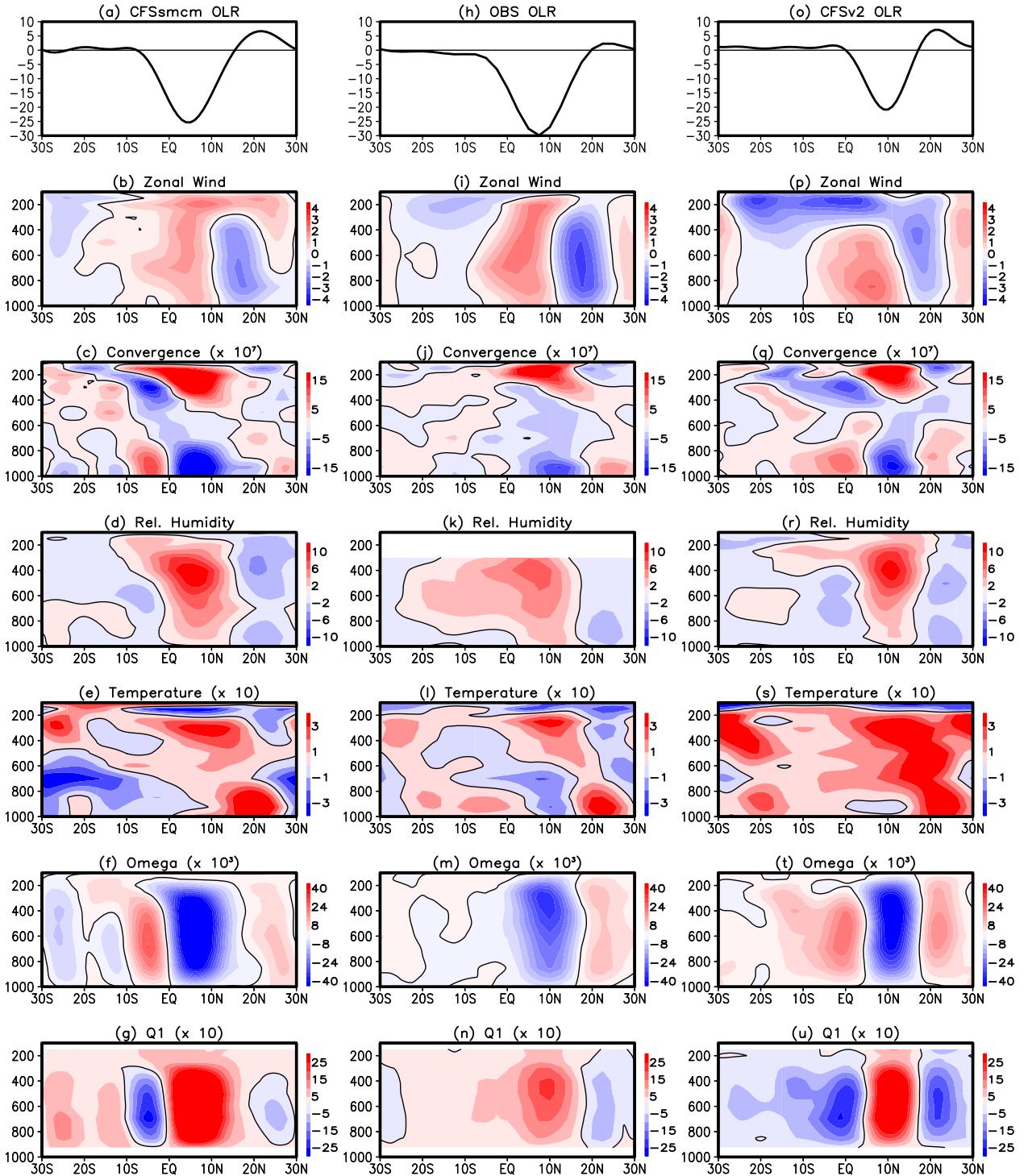
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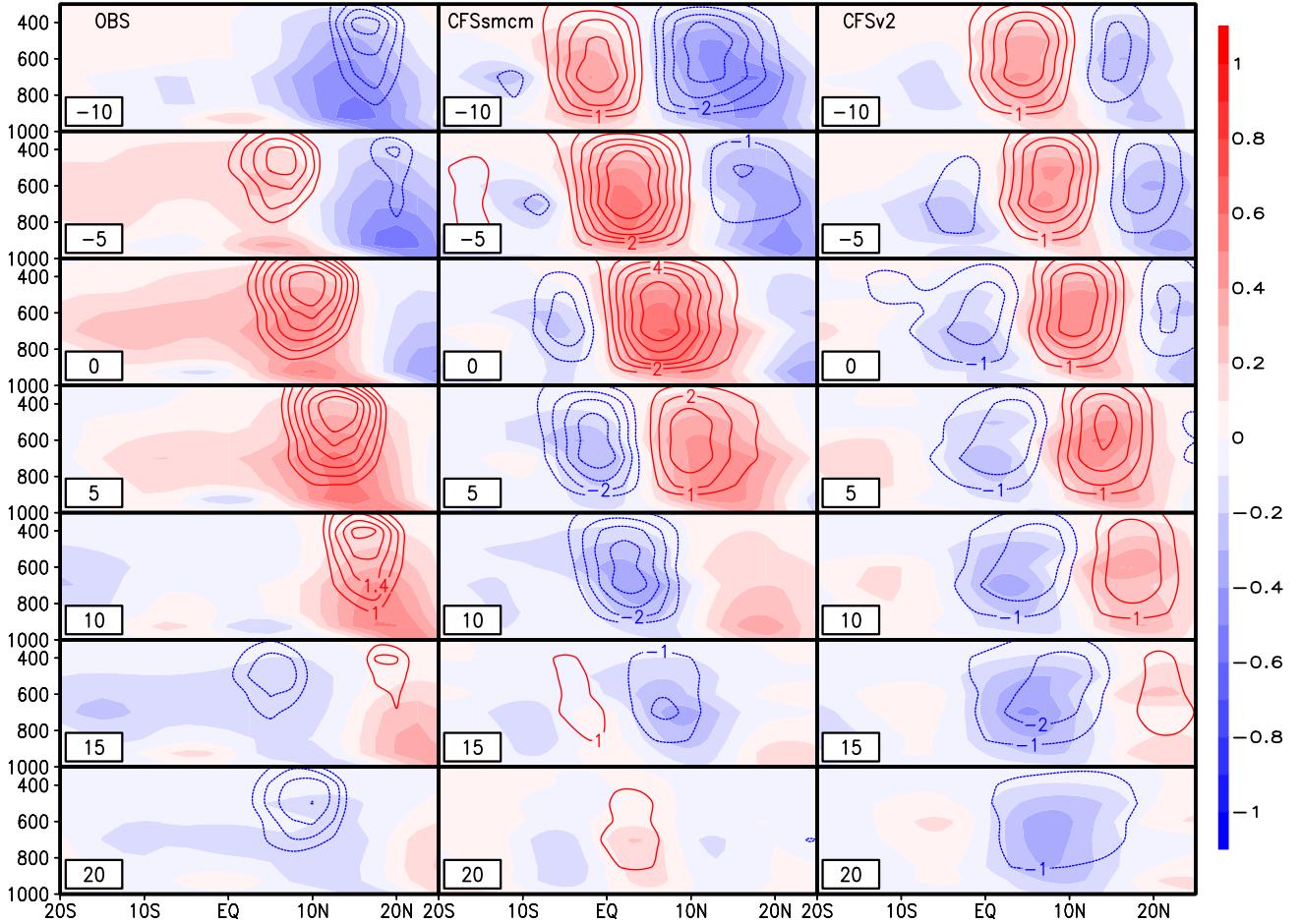


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 1128 and (l) CFSv2



1129 FIG. 13. Latitude-height cross section (averaged for 70°E - 90°E) of MISO composite of the MISO filtered  
 1130 OLR anomalies and the corresponding anomalous zonal wind ( $\text{m s}^{-1}$ ), convergence ( $\text{s}^{-1}$ ), relative humidity (%),  
 1131 temperature (K), vertical velocity ( $\text{Pa s}^{-1}$ ), and diabatic heating rate ( $\text{K day}^{-1}$ ). CFSsmcm simulation is shown in  
 1132 the left hand side column, observation (NOAA OLR) in the middle and CFSv2 in the right hand side.

### Heating countoured over SpHum



1133 FIG. 14. MISO phase composite of latitude-height cross section (averaged for 70°E - 90°E) of the MISO  
 1134 filtered diabatic heating rate( $\text{K day}^{-1}$ ) (contours) superimposed on the corresponding Specific Humidity ( $\text{g kg}^{-1}$ )  
 1135 (in shading). Observations are shown in the left hand side column, CFSsmcm in the middle and CFSv2 in the  
 1136 right hand side. Phase-lag stamps are seen in the right hand bottom corner. Note: diabatic heating rate contour  
 1137 intervals for the observations is  $0.2 \text{ K day}^{-1}$ , while the same for the CFSsmcm and CFSv2 simulations is  $2 \text{ K}$   
 1138  $\text{day}^{-1}$ .