

# Intermodel spread in Walker circulation responses linked to spread in moist stability and radiation responses

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## Abstract

The response of the Pacific Walker circulation (WC) to long-term warming remains uncertain. Here, we diagnose contributions to the WC response in comprehensive and idealized general circulation model (GCM) simulations. We find that the spread in WC response is substantial across both the Coupled Model Intercomparison Project (CMIP6) and the Atmospheric Model Intercomparison Project (AMIP) models, implicating differences in atmospheric models in the spread in projected WC strength. Using a moist static energy (MSE) budget, we evaluate the contributions to changes in the WC strength related to changes in gross moist stability (GMS), horizontal MSE advection, radiation, and surface fluxes. We find that the multimodel mean WC weakening is mostly related to changes in GMS and radiation. Furthermore, the *spread* in WC response is related to the spread in GMS and radiation responses. The GMS response is potentially sensitive to parameterized convective entrainment which can affect lapse rates and the depth of convection. We thus investigate the role of entrainment in setting the GMS response by varying the entrainment rate in an idealized GCM. The idealized GCM is run with a simplified Betts-Miller convection scheme, modified to represent entrainment. The weakening of the WC with warming in the idealized GCM is dampened when higher entrainment rates are used. However, the spread in GMS responses due to differing entrainment rates is much smaller than the spread in GMS responses across CMIP6 models. Therefore, further work is needed to understand the large spread in GMS responses across CMIP6 and AMIP models.

1 **Intermodel spread in Walker circulation responses**  
2 **linked to spread in moist stability and radiation**  
3 **responses**

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8 **Key Points:**

- 9 • The atmosphere plays an important role in setting the large spread in the Walker  
10 circulation (WC) response to warming in coupled models  
11 • Energetic analysis shows the WC response and its spread are strongly related to  
12 the responses of the gross moist stability and radiation  
13 • The responses of the WC and GMS exhibit some sensitivity to convective entrain-  
14 ment in an idealized general circulation model

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**Plain Language Summary**

The Walker circulation (WC), an east-west circulation over the tropical Pacific, has an uncertain response to climate warming. We diagnose contributions to the WC response in climate models. We find that the spread in WC responses is similar across atmosphere-only models as across models with both an atmosphere and ocean, implicating the atmosphere in the spread in WC response. We find that the WC response and its spread across models are mostly related to changes in gross moist stability (GMS) and radiation. The GMS measures the propensity of the atmospheric circulation to export energy, and is influenced by the vertical structure of temperature and winds. Changes in atmospheric radiation, especially those associated with clouds, amplify the effects of changes in GMS on the WC.

The GMS is affected by an uncertain parameter in climate models, the entrainment rate. The entrainment rate controls how much clouds mix with their environment. Using an idealized climate model, we learn that the weakening of the WC response is dampened with higher entrainment rates. However, the effect of different entrainment rates is too small to explain the large spread in GMS and WC responses across models; further work is needed to understand this large spread.

**1 Introduction**

The Pacific Walker circulation (WC) is an atmospheric zonal circulation over the equatorial Pacific Ocean. The WC transports energy from the West Pacific to the East Pacific (Trenberth & Stepaniak, 2003) in response to differing sea surface temperatures (SSTs) and net energy input to the atmosphere over the West and East Pacific. The WC can strongly influence precipitation over the tropical Pacific and also has nonlocal impacts. It is associated with a zonal surface pressure gradient over the Pacific Ocean, whose interannual variability comprises the Southern Oscillation. In addition to influencing the extratropical climate, it can respond to extratropical forcing (Kang et al., 2020). How the WC responds to a warming climate has been assessed using a combination of theory, observations, historical model trends, and model projections. Together, these lines of evidence give an unclear picture of the response of the WC to warming.

65 Observational and reanalysis products going back only a few decades indicate a strength-  
66 ening of the WC, while observations over a longer record indicate a weakening (Vecchi  
67 et al., 2006; Tokinaga et al., 2012; L’Heureux et al., 2013; Sohn et al., 2016; Wills et al.,  
68 2022). This discrepancy may be explained by the large role of internal variability which  
69 means that long time periods are needed to evaluate trends in the WC (Vecchi et al.,  
70 2006). Coupled climate model trends over the historical period of observed WC strength-  
71 ening are mixed, with some models indicating a weakening and others indicating a strength-  
72 ening, though no model strengthens to the same extent as observations (Sohn et al., 2016).  
73 Projections of a warm 21st century climate almost unanimously indicate a WC weak-  
74 ening, but with substantial spread in the degree of weakening (Vecchi & Soden, 2007).

75 There are a number of proposed mechanisms for the response of the WC to warm-  
76 ing, some of which suggest a weakening and some of which suggest a strengthening. Trop-  
77 ical convective mass fluxes are constrained to weaken overall with warming because pre-  
78 cipitation increases at a slower rate than specific humidity, which increases at a rate set  
79 by the Clausius-Clapeyron relationship (Held & Soden, 2006). However, it is not clear  
80 that local changes in the WC must follow overall changes in convective mass fluxes (Merlis  
81 & Schneider, 2011). Knutson and Manabe (1995) found a weakening of the WC in pro-  
82 jections despite an increase in precipitation in the ascent region. Increases in dry static  
83 stability, which are the result of changes in moist adiabatic lapse rate, are implicated in  
84 this weakening (Knutson & Manabe, 1995; Ma et al., 2012; Sohn et al., 2016). Further,  
85 differential increases in evaporative damping between the warm West Pacific and cool  
86 East Pacific weaken the SST gradient (Knutson & Manabe, 1995). Additionally, increased  
87 CO<sub>2</sub> directly weakens the tropical circulation through differences in masking of the CO<sub>2</sub>  
88 radiative forcing by deep clouds and water vapor between tropical ascent and descent  
89 regions (Merlis, 2015).

90 In contrast, an ocean dynamical thermostat mechanism, changes in anthropogenic  
91 aerosols, and southern ocean cooling may contribute a strengthening of the zonal SST  
92 gradient with warming (Clement et al., 1996; Heede & Fedorov, 2021; Hartmann, 2022).  
93 The ocean dynamical thermostat mechanism, which was proposed using a highly ideal-  
94 ized ocean model, describes a transient strengthening of the zonal SST gradient through  
95 (1) upwelling of relatively cool water in the equatorial East Pacific, thereby increasing  
96 the zonal SST gradient, and (2) increases in surface easterly winds which further increase  
97 this gradient (Clement et al., 1996). An analysis of coupled GCMs from CMIP3 found  
98 the upwelling portion of the mechanism to be operating but not the atmospheric por-  
99 tion of the mechanism because the surface easterly winds tend to weaken in the mod-  
100 els, and the net effect is a slight weakening of the zonal SST gradient (DiNezio et al., 2009).  
101 Further, analysis of changes in historical CMIP6 simulations from 1950 to 2014 suggests  
102 a relative cooling of the equatorial East Pacific due to changes in aerosols, contributing  
103 an initial strengthening tendency of the WC (Heede & Fedorov, 2021). Additionally, cool-  
104 ing of the southern ocean is linked with cooling of the tropical East Pacific, and may con-  
105 tribute to the observed strengthening of the zonal SST gradient (Hartmann, 2022).

106 Here we seek to understand the spread in WC response across GCM projections  
107 through an energetic approach. An MSE budget approach has previously been used to  
108 study tropical circulations (Neelin & Held, 1987; Chou & Neelin, 2004). We are partic-  
109 ularly motivated by the study of Wills et al. (2017) which used a moist static energy (MSE)  
110 budget to analyze the response of the WC to warming in simulations with an idealized  
111 GCM. Wills et al. (2017) found that the WC strength varies inversely with the gross moist  
112 stability (GMS) across a range of climates. GMS measures the efficiency of a circulation  
113 in exporting energy (Neelin & Held, 1987; Raymond et al., 2009). GMS has the advan-  
114 tage over the dry static stability, which has previously been used to explain changes in  
115 the WC (Knutson & Manabe, 1995; Sohn et al., 2016), that it can account for both dry  
116 adiabatic cooling and convective heating associated with ascent, and thus can be used  
117 in both the ascent and descent regions of the WC. For a given zonal gradient of net en-

118 ergetic input to the atmosphere, we expect an increase in GMS with warming to corre-  
 119 spond to a weaker WC (Wills et al., 2017). In general, we expect the GMS to increase  
 120 with warming owing predominantly to an increase in tropopause height (Chou et al., 2013).  
 121 In the observed atmosphere and in more realistic simulations, we expect a more com-  
 122 plicated relationship between GMS and WC responses than in the idealized simulations  
 123 of Wills et al. (2017). Nonetheless, we also find an inverse relationship between WC re-  
 124 sponse and changes in GMS in CMIP6 and AMIP models.

125 The close relationship we find between the responses of WC strength and GMS across  
 126 CMIP6 and AMIP simulations warrants further investigation into the response of GMS  
 127 to warming. We focus on the role of convective entrainment in setting the response of  
 128 the WC and GMS. In general, entrainment is the process by which a cloud or buoyant  
 129 plume mixes with the environment. Increasing entrainment affects GMS by (1) steepen-  
 130 ing the temperature lapse rate and (2) increasing the top-heaviness of vertical veloc-  
 131 ity profiles (Held et al., 2007; Singh & O’Gorman, 2013; Singh & Neogi, 2022). However,  
 132 it is difficult to represent entrainment in GCMs because it occurs on subgrid scales and  
 133 is difficult to measure directly (Romps, 2010). Following Wills et al. (2017), we use an  
 134 idealized GCM (Frierson et al., 2006; O’Gorman & Schneider, 2008) with a simplified  
 135 Betts-Miller (SBM) convection scheme (Frierson, 2007) to study the response of the WC  
 136 to warming. Here we modify the SBM scheme to represent entrainment so that we can  
 137 evaluate the role of entrainment in the WC and GMS changes across climates.

138 This paper has two aims: (1) diagnose the contributions to the mean and spread  
 139 of the WC response to warming in CMIP6 and AMIP simulations using an MSE bud-  
 140 get, and (2) evaluate the influence of entrainment on WC strength and its response to  
 141 warming in simulations with an idealized GCM. We address the first aim in Section 2  
 142 and the second aim in Section 3. We discuss and conclude in Section 4.

## 143 2 Response of WC to warming in CMIP6 and AMIP simulations

### 144 2.1 WC decomposition using GMS and the MSE budget

145 We diagnose the contributions to the response of the Walker circulation to warm-  
 146 ing across CMIP6 and AMIP models. We use monthly data of each variable and then  
 147 take the time and spatial average of calculated terms in a given climate before calculat-  
 148 ing the difference between warm and control climates. For the CMIP6 simulations, ‘con-  
 149 trol climate’ refers to the historical experiment for the years 1970-1999 and ‘warm cli-  
 150 mate’ refers to the SSP5-8.5 experiment for the years 2070-2099. For the AMIP simu-  
 151 lations, ‘control climate’ refers to the ‘amip’ experiment for the years 1979-2014 and ‘warm  
 152 climate’ refers to the ‘amip-future4K’ experiment for the years 1979-2014. The same en-  
 153 semble member is used for both control and warm experiments. The imposed SST field  
 154 of the ‘amip-future4K’ experiment is of a simulated warming, including a change in pat-  
 155 tern derived from coupled model experiments. The imposed SST field in ‘amip’ exper-  
 156 iments is the same across models. The imposed SST field in ‘amip-future4K’ experiments  
 157 is the same across models. We use one model from each modeling center, matching the  
 158 AMIP and CMIP6 models where possible. Some models were eventually excluded from  
 159 the analysis for missing data or excessive spectral ringing. The models used here are shown  
 160 in Table S1. Tropical-mean skin temperature warming from 20°S to 20°N is used to nor-  
 161 malize throughout (i.e., to calculate rates of change in % K<sup>-1</sup>).

162 We develop a framework for diagnosing contributions to changes in WC strength  
 163 using the MSE budget. The WC strength is measured by  $-\bar{\omega}_{w-e} = -p_s^{-1} \int \omega_{w-e} dp$  where  
 164  $p_s$  is surface pressure,  $\omega$  is vertical velocity in pressure coordinates, the overbar indicates  
 165 a vertical average in pressure over the depth of the atmosphere, and  $w-e$  denotes a hor-  
 166 izontal average over a western Pacific box minus a horizontal average over an eastern Pa-  
 167 cific box. We use the same boxes as Vecchi et al. (2006) when evaluating the CMIP6 and

168 AMIP models. That is, both boxes extend from 5°S to 5°N. The western Pacific box ex-  
 169 tends from 80°E to 160°E and the eastern Pacific box extends from 160°W to 80°W. The  
 170 western Pacific box includes a small portion of the Indian ocean. WC strength is cal-  
 171 culated by taking spatial and time averages of monthly  $\omega$  to create two profiles: one for  
 172 the western box and one for the eastern box. These profiles are then vertically integrated  
 173 and differences between west and east are taken. For figures and results including the  
 174 idealized GCM, we will refer to ‘ascent region’ and ‘descent region’ instead of ‘western  
 175 box’ and ‘eastern box’, but these should be interpreted equivalently.

176 We difference the MSE budget in the time average between the western and east-  
 177 ern boxes to give

$$\left\langle \omega \frac{\partial h}{\partial p} \right\rangle_{w-e} \approx - \langle \mathbf{u} \cdot \nabla h \rangle_{w-e} + R_{w-e} + S_{w-e}, \quad (1)$$

178 where  $\langle \cdot \rangle$  indicates a mass-weighted vertical integral, the subscript  $w-e$  indicates the  
 179 difference between western and eastern boxes,  $\mathbf{u}$  are horizontal winds,  $R$  is the sum of  
 180 net longwave and shortwave radiative fluxes into the atmosphere (including at both the  
 181 surface and top of atmosphere),  $S$  is the sum of upward surface fluxes of latent and sen-  
 182 sible heat, and  $h = c_p T + gz + Lq$  is MSE where  $c_p$  is the heat capacity of dry air,  $T$  is  
 183 temperature,  $g$  is acceleration due to gravity,  $z$  is height,  $L$  is latent heat of vaporiza-  
 184 tion, and  $q$  is specific humidity. All four terms in Equation 1 are implicitly taken to be  
 185 time averages in a given climate assuming a statistical steady state, and we are neglect-  
 186 ing sub-monthly eddy terms, whose differences between climates are small (not shown).

187 There are numerous definitions of GMS in the literature. Similar to Wills et al. (2017),  
 188 a definition of GMS appropriate for the WC is used here, denoted  $\text{GMS}_{\text{WC}}$ .  $\text{GMS}_{\text{WC}}$  is  
 189 the ratio of vertical advection of MSE, differenced between the western and eastern boxes,  
 190 to the WC strength and is given by

$$\text{GMS}_{\text{WC}} \equiv -g \frac{\left\langle \omega \frac{\partial h}{\partial p} \right\rangle_{w-e}}{\bar{\omega}_{w-e}}. \quad (2)$$

191 We further introduce  $\hat{\omega} = \frac{\omega}{\bar{\omega}_{w-e}}$  as the shape of the vertical-velocity profile to give the  
 192 simple form

$$\text{GMS}_{\text{WC}} = -g \left\langle \hat{\omega} \frac{\partial h}{\partial p} \right\rangle_{w-e}, \quad (3)$$

193 so that  $\text{GMS}_{\text{WC}}$  can be thought of as depending on the shape of the vertical velocity pro-  
 194 file and the MSE stratification, rather than directly on the WC strength. Our definition  
 195 of  $\text{GMS}_{\text{WC}}$  is similar to what Wills et al. (2017) calls GMS or  $\mathcal{M}$  with two differences.  
 196 First, instead of taking a zonal anomaly, we take the difference between the western and  
 197 eastern Pacific boxes. Second, we use a different definition of WC strength. Wills et al.  
 198 (2017) defined the WC strength by the zonally-anomalous vertical velocity at the level  
 199 of its maximum,  $\omega_{max}^*$ . Instead, we use vertically averaged  $\omega$  and the difference between  
 200 the western and eastern Pacific boxes, as described above.

201 In order to derive a diagnostic expression for WC strength from the MSE budget,  
 202 we combine Equations 1 and 2 to give

$$-\bar{\omega}_{w-e} \approx g \frac{- \langle \mathbf{u} \cdot \nabla h \rangle_{w-e} + R_{w-e} + S_{w-e}}{\text{GMS}_{\text{WC}}}. \quad (4)$$

203 Considering a perturbation due to climate change gives an expression for the fractional  
 204 change in WC strength as a function of changes in  $\text{GMS}_{\text{WC}}$ , horizontal MSE advection,  
 205 surface heat fluxes, and radiation:

$$\delta \bar{\omega}_{w-e} \approx -\delta \text{GMS}_{\text{WC}} - \frac{\Delta \langle \mathbf{u} \cdot \nabla h \rangle_{w-e}}{\left\langle \omega \frac{\partial h}{\partial p} \right\rangle_{w-e}} + \frac{\Delta R_{w-e}}{\left\langle \omega \frac{\partial h}{\partial p} \right\rangle_{w-e}} + \frac{\Delta S_{w-e}}{\left\langle \omega \frac{\partial h}{\partial p} \right\rangle_{w-e}}. \quad (5)$$

206 Here and throughout the paper,  $\Delta$  indicates a response to warming,  $\delta$  is the fractional  
 207 response to warming given by  $\delta X = \frac{\Delta X}{X}$ . We evaluate  $X$  in the denominator as the av-  
 208 erage between the control and warm climates and  $\left\langle \omega \frac{\partial h}{\partial p} \right\rangle_{w-e}$  in Equation 5 is also eval-  
 209 uated as the average between the control and warm climates to avoid cross terms. There-  
 210 fore no additional approximations are introduced between Equations 4 and 5. Equation  
 211 5 is evaluated by first calculating the the energy budget terms of Equation 1, then cli-  
 212 matologies for each month of the year taken for each term, then differences between cli-  
 213 mates are taken where applicable, and then spatial and annual means are taken for the  
 214 western and eastern Pacific boxes. Lastly, the terms in Equation 5 are evaluated. The  
 215 terms on the RHS of Equation 5 are the contributions to the WC response from changes  
 216 in  $\text{GMS}_{\text{WC}}$ , horizontal advection, radiation, and surface heat fluxes, respectively. Equa-  
 217 tions 1, 4, and 5 are approximations to the extent that there are errors due to, for ex-  
 218 ample, finite differencing in calculating advection terms and neglect of sub-monthly eddy  
 219 terms. The neglect of sub-monthly eddy terms introduces a substantial residual in a given  
 220 climate (Equation 4) but only a small residual for the differences between climates (Equa-  
 221 tion 5).

222 We further decompose the radiation contribution into a contribution from changes  
 223 in WC strength and a contribution not related to changes in WC strength using a lin-  
 224 ear regression of radiation as a function of WC strength. The regression is taken across  
 225 the 12 climatological monthly means for each model and climate and is given by

$$R_{w-e} \approx r_1 \bar{\omega}_{w-e} + R_0, \quad (6)$$

226 where  $r_1$  and  $R_0$  are regression coefficients. Having fit  $r_1$  and  $R_0$  using the seasonal cy-  
 227 cle, we now return to the average over all months in each climate and take the difference  
 228 between climates to give

$$\Delta R_{w-e} \approx r_1 \Delta \bar{\omega}_{w-e} + \Delta r_1 \bar{\omega}_{w-e} + \Delta R_0. \quad (7)$$

229 We continue to use averages between control and warm climates for terms that are not  
 230 differences between climates so that no cross terms are introduced between Equations  
 231 6 and 7. The first term on the RHS is interpreted as the contribution to  $\Delta R_{w-e}$  which  
 232 is linked with changes in WC strength, and the sum of the last two terms on the RHS  
 233 is interpreted as the contribution to  $\Delta R_{w-e}$  which is not linked with changes in WC strength.

## 234 2.2 WC response and decomposition in CMIP6

235 In order to diagnose contributions to changes in WC strength in coupled GCMs,  
 236 we apply the decomposition given by Equation 5 to each CMIP6 model. Figure 1a shows  
 237 the decomposition in the multimodel mean and the spread across models, and Figure  
 238 S1 shows the decomposition in individual CMIP6 models. We find that the WC weak-  
 239 ens in all models, with a weakening ranging from a 6%  $\text{K}^{-1}$  to 20%  $\text{K}^{-1}$ . The multimodel  
 240 mean weakening of 12%  $\text{K}^{-1}$  is greater than the 5 to 10%  $\text{K}^{-1}$  estimated by Vecchi and  
 241 Soden (2007) using changes in  $\omega$  at 500 hPa and this is partly because we normalize by  
 242 changes in tropical mean SST warming rather than global-mean surface warming as in  
 243 Vecchi and Soden (2007).

244 Looking at Figures 1a and S1, we notice that the relative roles of each mechanism  
 245 in setting the WC response can vary substantially across models, but a few important  
 246 commonalities emerge. The response of  $\text{GMS}_{\text{WC}}$  contributes a weakening of the WC in  
 247 all models. That is,  $\text{GMS}_{\text{WC}}$  increases with warming in all models, consistent with Chou  
 248 et al. (2013). The contribution from changes in  $\text{GMS}_{\text{WC}}$  ranges from a weakening of 4  
 249 to 18%  $\text{K}^{-1}$ . The total radiation contribution also contributes a weakening in all mod-  
 250 els, ranging from a weakening of 1 to 18%  $\text{K}^{-1}$ . The total radiation contribution is well  
 251 approximated by the sum of the WC-linked and not WC-linked portions, with the WC-

252 linked portion dominating in the multimodel mean (Figure 1c). Thus, the weakening con-  
 253 tribution from radiation in the multimodel mean is largely due to an amplifying feed-  
 254 back of radiation on WC response (cf. Peters and Bretherton (2005)). Looking at Fig-  
 255 ure S1, EC-Earth3 is an outlier model for the radiation contribution but it is not an out-  
 256 lier for WC response because this model has a small contribution from GMS changes.  
 257 If the EC-Earth3 model is neglected, the radiation contribution has a spread of 1 to 11%  
 258  $\text{K}^{-1}$ .

### 259 **2.3 WC response and decomposition in AMIP**

260 In order to isolate the atmospheric contribution to the spread in WC response, we  
 261 analyze the response of the WC in AMIP simulations using the ‘amip’ and ‘amip-future4K’  
 262 experiments. Recall that all of the ‘amip’ experiments have the same imposed SST dis-  
 263 tribution as one another and all of the ‘amip-future4K’ experiments have the same im-  
 264 posed SST distribution as one another, so these experiments isolate the role of the at-  
 265 mosphere in causing intermodel differences independent from differences in SST.

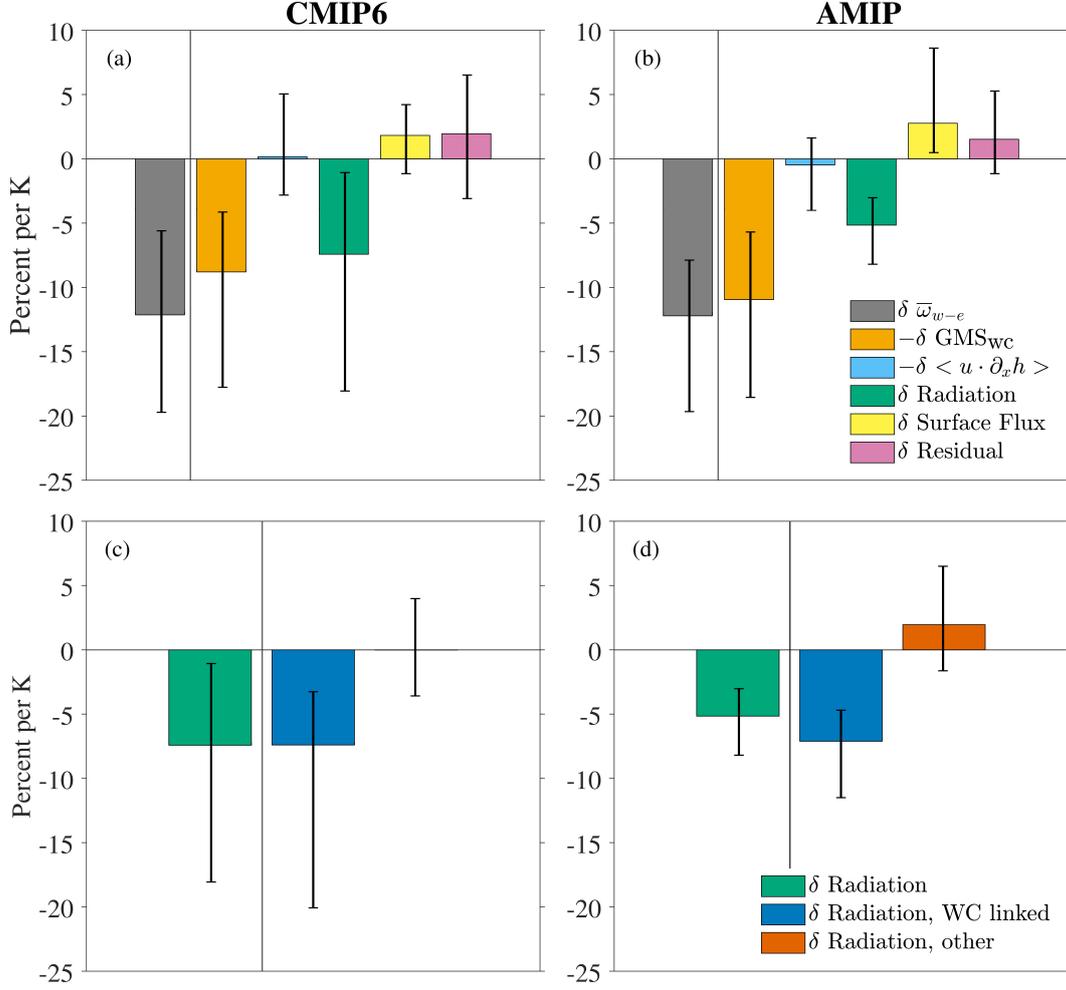
266 As we did with the CMIP6 models, we apply the decomposition given by Equa-  
 267 tion 5 to each AMIP model. Figure 1b shows the decomposition in the multimodel mean  
 268 and the spread across models, and Figure S2 shows the decomposition in individual AMIP  
 269 models. Even with the same SST response across models, there is spread in the weak-  
 270 ening response of the WC from 8 to 20%  $\text{K}^{-1}$  which is similar to the range for the CMIP6  
 271 simulations which are coupled with interactive oceans. Similar to the CMIP6 simulations,  
 272 the WC response is dominated by changes in  $\text{GMS}_{\text{WC}}$  and radiation; both contribute a  
 273 weakening in all AMIP simulations. The contribution from changes in  $\text{GMS}_{\text{WC}}$  range from  
 274 a weakening of 6 to 19%  $\text{K}^{-1}$ , while the contribution from changes in radiation range  
 275 from a weakening of 3 to 8%  $\text{K}^{-1}$ . The range of radiation contributions is not much smaller  
 276 than that of the CMIP6 models when the outlier EC-Earth3 model, which does not ap-  
 277 pear in AMIP, is removed from CMIP6. Further, the radiation contribution is dominated  
 278 by changes in the WC-linked portion in the multimodel mean (Figure 1d). The spread  
 279 due to changes in surface heat fluxes is larger in AMIP than in CMIP6, which may be  
 280 the result of artificially imposing SSTs. While the substantial spread in WC response  
 281 across AMIP models does not rule out some role for the ocean in setting the spread in  
 282 CMIP6, it does suggest an important role of the atmosphere in setting the spread in CMIP6  
 283 response.

### 284 **2.4 Models with CMIP and AMIP equivalents**

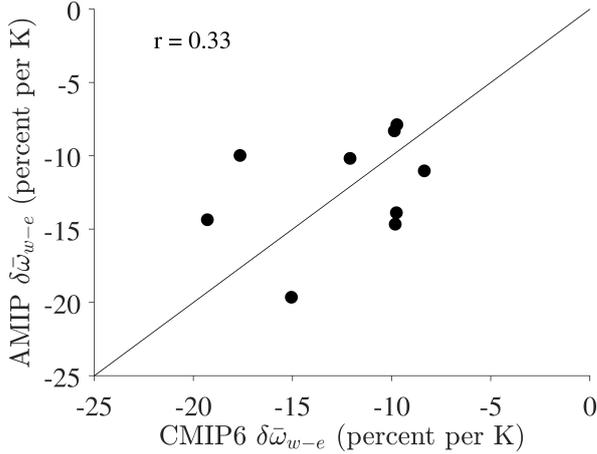
285 Our results so far indicate that the spread in WC responses across CMIP6 mod-  
 286 els is comparable to the spread across AMIP models. However, the two ensembles con-  
 287 sist of different sets of models. There are nine models with both AMIP and CMIP6 coun-  
 288 terparts. Figure 2 compares the WC responses for these nine models. The WC responses  
 289 are positively correlated between AMIP and CMIP6 with a correlation coefficient of 0.33.  
 290 The positive correlation suggests atmospheric processes active in AMIP are contribut-  
 291 ing to some of the spread in CMIP6 models. Further, the models are evenly distributed  
 292 above and below the one-to-one line, which suggests that there is not a single mecha-  
 293 nism associated with ocean-atmosphere coupling, such as the Bjerknes feedback, caus-  
 294 ing differences of a consistent sign between CMIP6 models and their AMIP counterparts.

### 295 **2.5 Contributions of western and eastern boxes**

296 We also decompose each term in Equation 5 into contributions from changes over  
 297 the West and East Pacific. Figure S3 shows this decomposition for CMIP6 models, and  
 298 Figure S4 shows this decomposition for AMIP models. In both CMIP6 and AMIP mod-  
 299 els, changes in  $\delta\bar{\omega}_{w-e}$  have weakening contributions from changes over both the West  
 300 and East Pacific, with a larger contribution from the East Pacific in the multimodel mean.



**Figure 1.** Contributions to multimodel mean response of WC to warming in (a) CMIP6 and (b) AMIP simulations. WC response (gray) is the sum of the contributions from each term on the RHS of Equation 5. The radiation contribution in (c) CMIP6 and (d) AMIP is decomposed into the portion that is linked to WC strength (dark blue) and the portion that is not linked with WC strength (dark orange). The radiation decomposition is performed using the seasonal cycle and Equation 7. The whiskers cover the entire spread across models for each term.



**Figure 2.** Scatterplot of WC responses in CMIP6 and AMIP simulations for the nine models that are present in both ensembles. The black line is a reference line with a slope of 1.

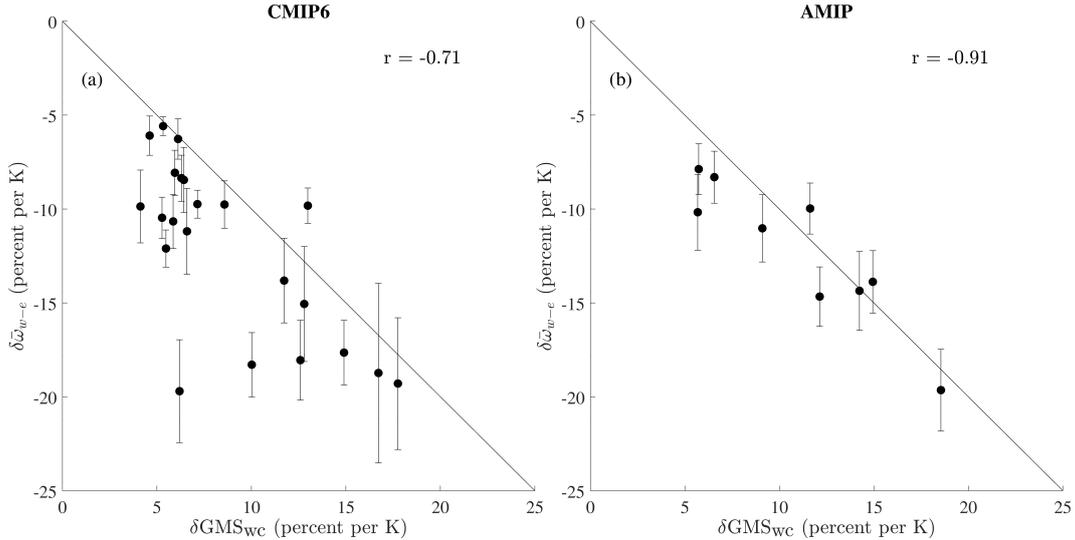
301 In both ensembles, the radiation response over the West Pacific contributes a weaken-  
 302 ing in all models and the GMS response over the East Pacific contributes a weakening  
 303 in all models. GMS also contributes a weakening over the West Pacific in the multimodel  
 304 mean and in most models. The radiation contribution over the East Pacific is uncertain.

## 305 2.6 Relationship between WC and GMS responses

306 Given their large contributions, we investigate the roles of changes in  $GMS_{WC}$  and  
 307 radiation on WC strength in the remainder of Section 2.

308 We expect  $GMS_{WC}$  to vary inversely with WC strength because a larger increase  
 309 in GMS indicates a larger weakening of the atmospheric circulation for a given energetic  
 310 forcing. Figure 3 shows that the relationship between responses of WC strength and  $GMS_{WC}$   
 311 in CMIP6 and AMIP models are consistent with this expectation: the WC weakens and  
 312  $GMS_{WC}$  increases in all models, with a tendency for greater weakening of the WC with  
 313 a greater increase in  $GMS_{WC}$ . The correlation coefficient is -0.71 across the CMIP6 mod-  
 314 els and -0.91 across the AMIP models. Most models fall below the line through the ori-  
 315 gin with a slope of  $-1$  because changes in radiation also contribute to a weakening of  
 316 the Walker circulation. There is a greater spread in the radiation contribution across CMIP6  
 317 models than AMIP models (Figure 1), so the correlation between  $GMS_{WC}$  response and  
 318 WC response is lower across CMIP6 models than across AMIP models. The outlier CMIP6  
 319 model located near (6,-20) is EC-Earth3, which has the largest radiation contribution  
 320 of any CMIP6 model (Figure S1).

321 Figure 3 also shows a measure of the standard error of the WC response for each  
 322 model. The WC response in Figure 3 is shown as the fractional change in WC strength  
 323 normalized by surface temperature response, given by  $100 \left( \frac{\Delta\bar{\omega}_{w-e}}{\bar{\omega}_{w-e}} \right) / \Delta T_s$ . We calculate  
 324 the standard error of the change in WC strength,  $\Delta\bar{\omega}_{w-e}$ , as  $\frac{\sqrt{\text{std}(\bar{\omega}_{w-e}^{warm})^2 + \text{std}(\bar{\omega}_{w-e}^{ctrl})^2}}{\sqrt{n}}$ ,  
 325 where  $n$  is the number of simulation years in each climate and  $\text{std}()$  indicates a standard  
 326 deviation across model years. This standard error calculation assumes WC strength is  
 327 independent between different model years and climates. We then normalize by multi-  
 328 plying by  $100/(\bar{\omega}_{w-e}/\Delta T_s)$  so that the standard error has the same units as the plot-  
 329 ted value. The standard errors are sufficiently small that we can be sure that the inter-  
 330 model spread in WC response is not just due to unforced variability.



**Figure 3.** Relationship between responses of  $\text{GMS}_{\text{WC}}$  and WC strength for (a) CMIP6 and (b) AMIP simulations. The error bars indicate a measure of the standard error of the WC response calculated as described in Section 2.6. The black lines are reference lines with slopes of -1.

331 Wills et al. (2017) showed a similar inverse relationship between WC strength and  
 332 GMS in idealized GCM simulations. The strong anticorrelation between responses of WC  
 333 strength and  $\text{GMS}_{\text{WC}}$  indicates that the WC-GMS relationship holds in more complex  
 334 simulations and warrants further investigation into the response of  $\text{GMS}_{\text{WC}}$  to warming.

### 335 2.7 $\text{GMS}_{\text{WC}}$ decomposition

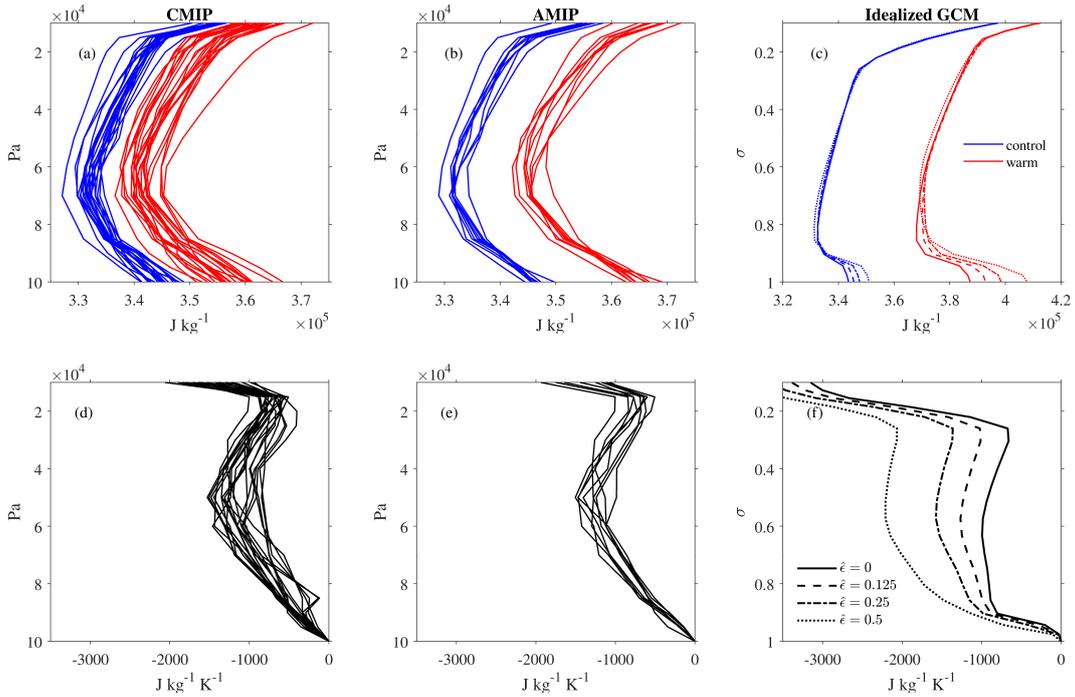
336 In order to better understand the response of  $\text{GMS}_{\text{WC}}$  to warming in CMIP6 and  
 337 AMIP models, we decompose the  $\text{GMS}_{\text{WC}}$  response into contributions due to changes in  
 338 vertical velocity and MSE profiles. Looking at Equation 3, the fractional change in  $\text{GMS}_{\text{WC}}$   
 339 with warming has contributions from changes in the shape of the vertical velocity pro-  
 340 file  $\hat{\omega}$  and changes in the MSE profile through  $\partial h / \partial p$  as follows:

$$\delta \text{GMS}_{\text{WC}} \approx \frac{\left\langle \Delta \hat{\omega} \frac{\partial h}{\partial p} \right\rangle_{w-e}}{\left\langle \hat{\omega} \frac{\partial h}{\partial p} \right\rangle_{w-e}} + \frac{\left\langle \hat{\omega} \Delta \frac{\partial h}{\partial p} \right\rangle_{w-e}}{\left\langle \hat{\omega} \frac{\partial h}{\partial p} \right\rangle_{w-e}}. \quad (8)$$

341 There is a small residual because monthly climatologies of  $\hat{\omega}$  and  $\partial h / \partial p$  are used in cal-  
 342 culating the numerator.

343 Figures 4a,b,d,e compare ascent-region MSE profiles and their response to warm-  
 344 ing in CMIP6 and AMIP models. The response of surface MSE is subtracted from each  
 345 response profile since it is the vertical gradient of MSE which affects GMS. For the CMIP6  
 346 and AMIP models, profiles are averaged over the area of the western Pacific box. Fig-  
 347 ures 5a,b,d,e compare  $\hat{\omega}$  profiles and their response to warming. All response profiles are  
 348 normalized by tropical-mean SST warming. Figure 4 reveals that MSE increases with  
 349 warming and Figure 5 reveals that  $\hat{\omega}$  profiles have a tendency to shift upward with warm-  
 350 ing consistent with the increase in tropopause height and the upward shift of the gen-  
 351 eral circulation with warming (Singh & O’Gorman, 2012).

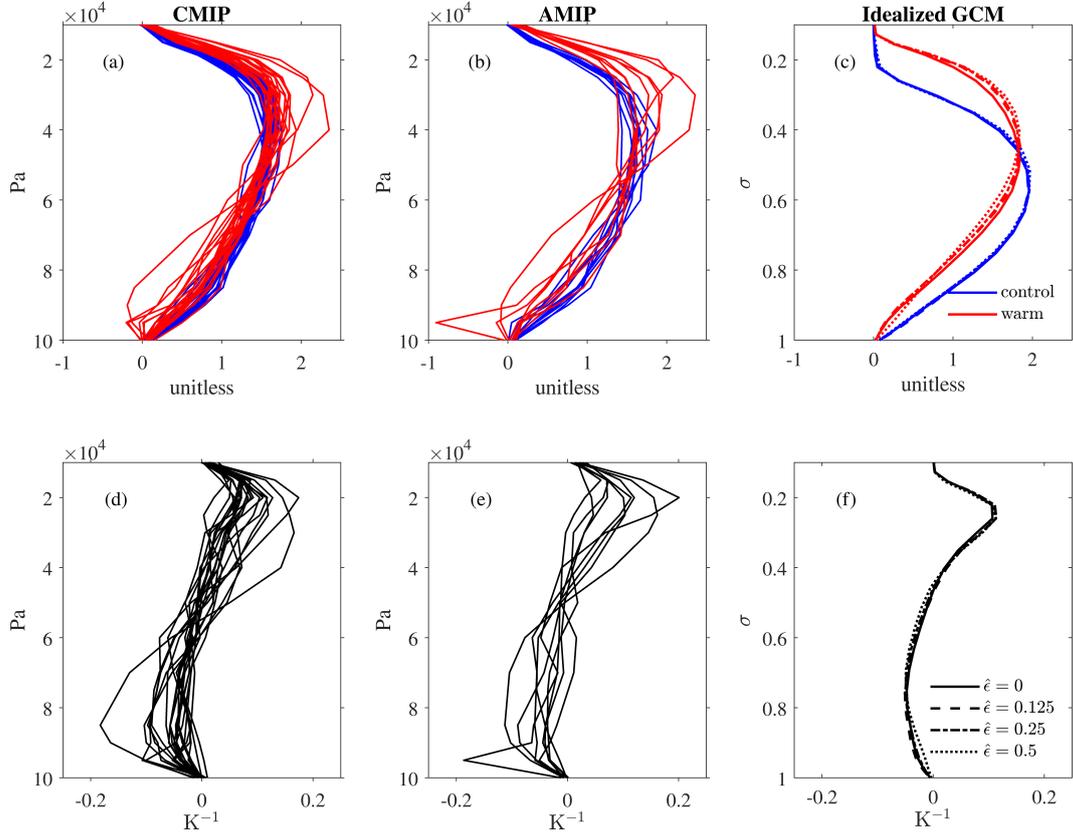
352 The results of the decomposition of  $\delta \text{GMS}_{\text{WC}}$  from Equation 8 are shown for the  
 353 multimodel means in Figure 6, for each CMIP6 model in Figure S5, and for each AMIP



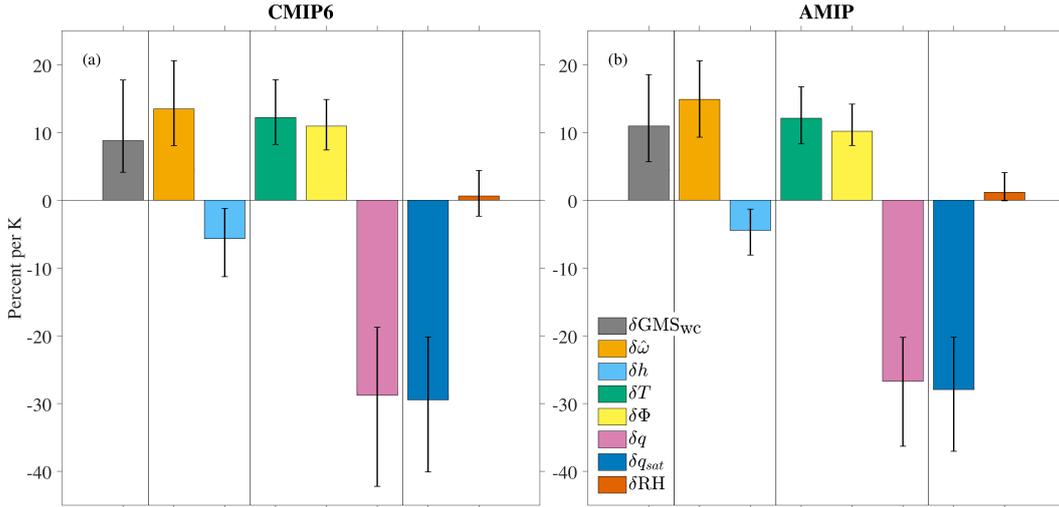
**Figure 4.** Ascent-region MSE profiles (a-c) and their response to warming (d-f) in CMIP6 (a,d), AMIP (b,e), and the idealized GCM (c,f). In panels (d-f), surface MSE responses for each profile is subtracted so that all profiles go through zero at the surface. The response profiles (d-f) are normalized by the tropical-mean SST response. CMIP6 and AMIP profiles are in pressure coordinates and idealized profiles are in sigma coordinates. CMIP6 and AMIP profiles are averaged over the ascent region of the WC defined here as the western box, and idealized GCM profiles are averaged over the boundary of the ascent region to be consistent with the boundary  $GMS_{wc}$  introduced in Section 3.5 (see text for details).

354 model in Figure S6. The  $\hat{\omega}$  contribution is positive and considerably larger in magnitude  
 355 than the MSE profile contribution for both CMIP6 and AMIP. The positive contribu-  
 356 tion from changes in  $\hat{\omega}$  is consistent with the increase in GMS from increasing tropopause  
 357 height and the associated upward shift of  $\hat{\omega}$  (Chou et al., 2013; Wills et al., 2017). We  
 358 also see a partially-compensating negative contribution from changes in MSE profile. Changes  
 359 in MSE profile are also influenced by the upward shift. Not taking into account the up-  
 360 ward shift in all variables simultaneously is a limitation of the decomposition used here.

361 Using the definition of  $h = c_p T + gz + Lq$ , the  $h$  profile contribution can be lin-  
 362 earlyly decomposed into contributions from changes in temperature ( $T$ ), geopotential height  
 363 ( $z$ ) and specific humidity ( $q$ ). Further, the changes in specific humidity can be decom-  
 364 posed into its contributions from changes in saturation specific humidity ( $q_{sat}$ ) and rel-  
 365 ative humidity (RH), according to  $\Delta q \approx \Delta RH q_{sat} + RH \Delta q_{sat}$ , where again there is a  
 366 small residual since climatologies of each term are used. Figures 6, S2, and S4 show that  
 367 changes in  $h$  profile tend to have small net contributions to changes in  $GMS_{wc}$ , but this  
 368 is the result of compensation between strong positive contributions from changes in  $T$   
 369 and  $\Phi$  and a strong negative contribution from changes in specific humidity. The con-  
 370 tribution from changes in specific humidity, which acts to decrease the  $GMS_{wc}$ , is mostly  
 371 the result of changes in saturation specific humidity. Note that our contributions from  
 372 changes in  $T$ ,  $z$ , and  $q$  assume constant  $\hat{\omega}$ , and thus our contributions differ from the con-  
 373 tributions found in Wills et al. (2017) in which the increase in tropopause height was in-



**Figure 5.** Profiles of  $\hat{\omega}$  (a-c) and their response to warming (d-f) in CMIP6 (a,d), AMIP (b,e), and the idealized GCM (c,f). The response profiles (d-f) are normalized by the tropical-mean SST response. CMIP6 and AMIP profiles are in pressure coordinates and idealized GCM profiles are in sigma coordinates. All profiles represent the average over the ascent region minus the average over the descent region (see text for details). For the CMIP6 and AMIP simulations, the ascent and descent regions are the western and eastern boxes, respectively.



**Figure 6.** Contributions to the multimodel mean response of  $GMS_{wc}$  to warming in (a) CMIP6 and (b) AMIP simulations. The response of  $GMS_{wc}$  to warming (gray) is decomposed into contributions from changes in shape of vertical velocity profile (light orange) and changes in MSE (light blue) as in Equation 8. The MSE contribution is further decomposed into contributions from changes in temperature (green), geopotential height (yellow), and humidity (pink). The humidity contribution is further decomposed into contributions from changes in saturation specific humidity (dark blue) and relative humidity (dark orange). The whiskers cover the entire spread across models for each term.

374 cluded. Figure 6 also shows that intermodel spread in changes in both the MSE profile  
 375 and the shape of the vertical velocity profile contribute to the intermodel spread in the  
 376  $GMS$  response.

### 377 2.8 Relationship between WC and radiation responses

378 Radiation responses contributes a weakening of the WC in all AMIP and CMIP  
 379 models. Figures 1c,d show that the WC-linked portion of the radiation response domi-  
 380 nates over the portion not linked with the WC. We further decompose the radiation con-  
 381 tribution into top of atmosphere (TOA) and surface contributions, SW and LW contri-  
 382 butions, and clear-sky and cloud-radiative effects (CRE), for a total of eight terms (Fig-  
 383 ures S7 and S8). We further decompose these eight terms into their WC-linked and other  
 384 contributions by adapting the regression used for Figure 1 (Figures S9 and S10).

385 Across CMIP6 and AMIP models, CRE dominates over clear-sky contributions in  
 386 both magnitude and spread across models. In particular, the CRE of TOA LW, TOA  
 387 SW, and surface SW are dominant with the largest intermodel spread. For each of these  
 388 three contributions, the WC-linked portion dominates across CMIP6 and AMIP mod-  
 389 els. These results suggest that changes in clouds associated with the WC dominate the  
 390 spread in the radiation contribution across models and, in general, amplify the weaken-  
 391 ing of the WC with warming in models. That the radiation and circulation influence one  
 392 another is consistent with the findings of Peters and Bretherton (2005) and Silvers and  
 393 Robinson (2021).

394 Keeping in mind that cloud-radiative effects amplify the WC response, we next fur-  
 395 ther investigate changes in  $GMS_{wc}$ , which is the other primary contributor to the WC  
 396 response. In particular, we examine the effect of convective entrainment in the  $GMS_{wc}$

response given that convective entrainment can affect both the MSE profile and the shape of the vertical velocity profile.

### 3 The role of entrainment in setting GMS and WC strength in idealized GCM simulations

#### 3.1 Why consider entrainment?

In order to further evaluate the spread in WC strength response, we study the role of entrainment in setting the WC strength and its response to warming in an idealized GCM. Entrainment is a parameterized process which is difficult to quantify in observations. However, entrainment can have a substantial effect on the climate, especially in the tropics (Singh & O’Gorman, 2013; Miyawaki et al., 2020). Entrainment affects the temperature lapse rate: a higher entrainment rate tends to steepen the temperature lapse rate in the lower and mid troposphere in GCM simulations (Held et al., 2007; Keil et al., 2021). Variations in temperature lapse rate with entrainment will also affect specific humidity, and both the temperature and humidity profiles influence the MSE profile, a key portion of the GMS. Further, entrainment can increase the top-heaviness of vertical velocity profiles (Singh & Neogi, 2022) which again strongly influences the GMS (Inoue et al., 2021). Therefore, we test the effect of entrainment on the GMS and WC using idealized simulations with different values of an entrainment parameter. These idealized simulations allow us to establish a causal relationship between imposed changes in stratification (from changes in entrainment) and the effect on GMS and WC strength, and are thus complementary to the CMIP6 and AMIP results which are diagnostic. Other processes such as radiation also contribute to differences in the WC response and should be studied in future work.

#### 3.2 Idealized GCM simulations

Idealized simulations of the Walker circulation are run using an idealized moist atmospheric GCM based on the GFDL spectral dynamical core following Frierson et al. (2006) with details as in O’Gorman and Schneider (2008). The idealized GCM lacks land, a seasonal cycle, and cloud and water-vapor radiative feedbacks. The lower boundary is a thermodynamic mixed-layer ocean with a depth of 1 m. The horizontal convergence of the ocean energy flux is specified through a  $Q$  flux. There is a zonal-mean component of the  $Q$  flux with a maximum magnitude of  $30 \text{ W m}^{-2}$  and a latitudinal width parameter of  $16^\circ$  following Equation 1 of Merlis and Schneider (2011). Through missing a cosine latitude factor, this zonal-mean  $Q$  flux formulation induces a small global-mean sink of energy (Merlis et al., 2013) which is not expected to strongly affect the results presented here.

Following Wills et al. (2017), the WC is driven by a zonally anomalous component of the  $Q$  flux with an elliptic convergent region in the ‘western’ hemisphere (leading to atmospheric ascent) and an equal and opposite divergent region (leading to atmospheric descent) in the ‘eastern’ hemisphere, both centered on the equator. The zonally anomalous  $Q$  flux,  $Q^*$ , has the form

$$Q^* = Q_1 \exp \left[ -\frac{(\lambda - \lambda_W)^2}{2\sigma_\lambda^2} - \frac{\phi^2}{2\sigma_\phi^2} \right] - Q_1 \exp \left[ -\frac{(\lambda - \lambda_E)^2}{2\sigma_\lambda^2} - \frac{\phi^2}{2\sigma_\phi^2} \right], \quad (9)$$

where  $\lambda$  is longitude,  $\phi$  is latitude,  $Q_1 = 50 \text{ W m}^{-2}$  is the amplitude of the zonally anomalous  $Q$  flux,  $\lambda_E = 270^\circ$  is the longitude of the center of the descent region,  $\lambda_W = 90^\circ$  is the longitude of the center of the ascent region,  $\sigma_\lambda = 12.5^\circ$  is proportional to the zonal extent of the anomaly, and  $\sigma_\phi = 8^\circ$  is proportional to the meridional extent of the anomaly. The sign of the zonally anomalous  $Q$  flux is modified from Wills et al. (2017) such that positive indicates a flux from ocean to atmosphere at steady state. The imposed zon-

443 ally anomalous  $Q$  flux is plotted in Figure S11. We define the ascent region as the el-  
 444 liptic area within the  $10 \text{ W m}^{-2}$   $Q$ -flux contour and the descent region as the elliptic area  
 445 within the  $-10 \text{ W m}^{-2}$   $Q$ -flux contour. We refer to these as ‘west’ and ‘east’ and con-  
 446 tinue to use the  $w-e$  subscript because the ascent region is meant to represent the West  
 447 Pacific and the descent region is meant to represent the East Pacific.

448 The idealized simulations are spun up for four years, and the analysis is performed  
 449 on the following eight years of simulation output. The convection scheme is a modifi-  
 450 cation of the simplified Betts-Miller (SBM) convection scheme of Frierson (2007), which  
 451 relaxes temperature profiles to a moist adiabat and relative humidity to 70% in convec-  
 452 ting regions. Here, we modify the SBM scheme by introducing a non-dimensional entrain-  
 453 ment parameter  $\hat{\epsilon}$  such that the convection scheme relaxes to the temperature profile of  
 454 an *entraining* plume when  $\hat{\epsilon} > 0$ . Our entraining SBM scheme reduces to the SBM con-  
 455 vection scheme when  $\hat{\epsilon} = 0$ . Details about the modification to represent entrainment  
 456 are given in Appendix A.

457 The longwave optical depth distribution is specified as a function of latitude and  
 458 pressure and then scaled by a factor  $\alpha$  (O’Gorman & Schneider, 2008). Two climates are  
 459 simulated: a control climate with a default longwave optical depth ( $\alpha = 1$ ) and a warm  
 460 climate with doubled longwave optical depth ( $\alpha = 2$ ). From the control to the warm  
 461 climate there is a large warming with a global-mean SST increase of 11.2K and a tropical-  
 462 mean ( $20^\circ\text{S}$  to  $20^\circ\text{N}$ ) SST increase of 9.1K in the simulations without entrainment. We  
 463 also considered additional  $\alpha$  values and, consistent with Wills et al. (2017), we found that  
 464 WC strength scales nearly linearly with temperature. Therefore, it is reasonable to com-  
 465 pare our results (when normalized per K) to the CMIP6 and AMIP models with less warm-  
 466 ing. The ocean  $Q$  flux is held constant as the climate warms. We run the idealized model  
 467 for simulations of a control climate and a warm climate with four values of the entrain-  
 468 ment parameter  $\hat{\epsilon}$ , for a total of eight simulations. The four values of  $\hat{\epsilon}$  are 0 (no entrain-  
 469 ment), 0.125, 0.25, and 0.5.

### 470 3.3 Spread in MSE and $\hat{\omega}$ profiles

471 Before evaluating responses of WC strength and GMS to warming across entrain-  
 472 ment rates in the idealized GCM, it is useful to examine the  $\hat{\omega}$  and MSE profiles and their  
 473 responses to warming (Figures 4 and 5) since these affect the GMS response. Figure 4  
 474 compares the ascent-region MSE profiles and their responses to warming in CMIP6 mod-  
 475 els, AMIP models, and across entrainment rates in the idealized GCM. Recall that the  
 476 response of surface MSE is subtracted from each profile since it is the vertical gradient  
 477 of MSE which affects GMS. For the idealized GCM, MSE profiles are averaged over the  
 478 *boundary* of the elliptic ascent region, consistent with the upcoming GMS analysis. Note  
 479 that the gray radiation scheme used in the idealized GMS leads to biases in vertical tem-  
 480 perature structure as compared to more complex radiation schemes (Tan et al., 2019).  
 481 These biases likely influence the MSE profiles in Figures 4c,f. Figure 5 compares  $\hat{\omega}$  pro-  
 482 files and their responses to warming in CMIP6 models, AMIP models, and across entrain-  
 483 ment rates in the idealized GCM. Recall that response profiles are normalized by tropical-  
 484 mean SST warming.

485 Focusing on the sensitivity to entrainment in the idealized GCM, Figures 4c and  
 486 5c show that entrainment has a bigger effect in the warmer climate than in the control  
 487 climate. The greater sensitivity to entrainment in a warmer climate is because entrain-  
 488 ment in the convection scheme acts on the difference between the MSE of the environ-  
 489 ment and that of saturated rising air, and this difference is larger in the warm climate.  
 490 Figure 4c reveals that increases in entrainment have a tendency to steepen the MSE lapse  
 491 rate, especially in the lower troposphere, and that this steepening is greater in the warmer  
 492 climate. Figure 5c reveals that  $\hat{\omega}$  profiles have a tendency to shift upward with warm-  
 493 ing, and this upward shift is enhanced by convective entrainment. The enhancement in

494 the upward shift of  $\hat{\omega}$  with higher entrainment rates is broadly consistent with Singh and  
 495 Neogi (2022), who found that entrainment tends to make vertical velocity profiles more  
 496 top heavy.

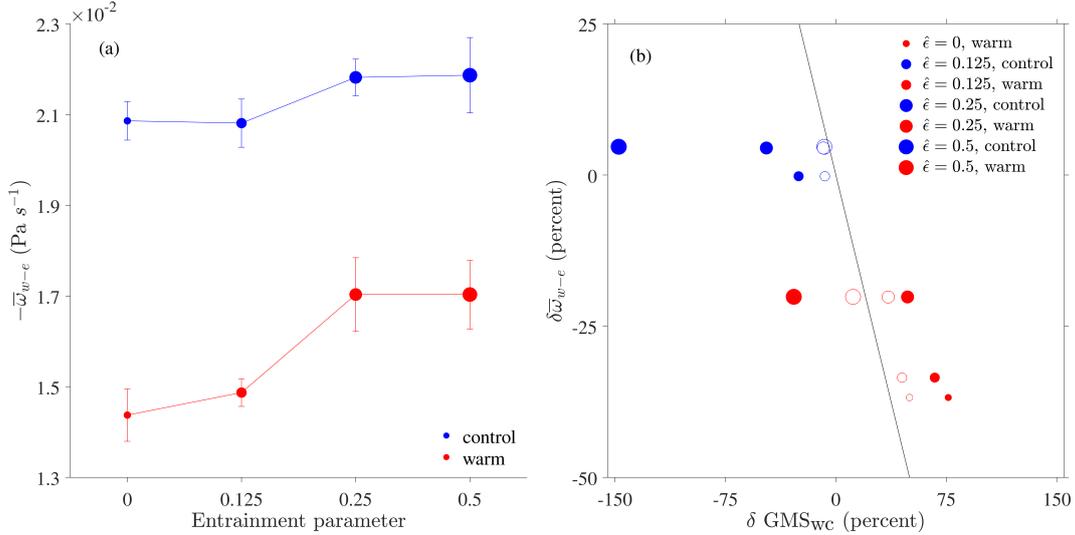
497 Comparing the idealized GCM to CMIP6 and AMIP, we find some important sim-  
 498 ilarities in the response to warming including an increase in MSE, a steeping of the lapse  
 499 rate of MSE in the lower troposphere, and an upward shift of the  $\hat{\omega}$  profile. We also find  
 500 that the spread in MSE profile response across entrainment rates in the idealized GCM  
 501 is substantial and somewhat larger than the spread in MSE profile response across CMIP6  
 502 and AMIP models. We hypothesize that the sensitivity to entrainment in the upper tro-  
 503 posphere may be exaggerated because the convection scheme used in the idealized sim-  
 504 ulation is based on a single plume with one fixed entrainment profile, whereas with a spec-  
 505 trum of plumes the air that reaches the upper-troposphere is only weakly affected by en-  
 506 trainment. Interestingly, the spread across control-climate MSE profiles in CMIP6 and  
 507 AMIP models is larger than the spread across entrainment rates in the idealized GCM,  
 508 but the opposite is true for the response of MSE profiles to warming. In contrast to the  
 509 MSE profiles, we find that the spread in  $\hat{\omega}$  profile response across entrainment rates in  
 510 the idealized simulations is very small as compared to the spread in CMIP6 and AMIP  
 511 simulations. Thus we expect  $\hat{\omega}$  changes to play a much bigger role for the spread in GMS  
 512 and WC response in CMIP6 and AMIP compared to the variation across entrainment  
 513 rates in the idealized GCM simulations.

### 514 **3.4 Sensitivity of WC strength to warming and entrainment in ideal-** 515 **ized simulations**

516 The WC strength is defined as the negative of the average value of  $\omega$  in the ascent  
 517 region minus the average value of  $\omega$  over the descent region. Further, we estimate the  
 518 uncertainty in WC strength by using the WC strength in each of the eight simulated years  
 519 to calculate the standard error for the eight-year average. The WC strength and its stan-  
 520 dard error are plotted in Figure 7a for each of the idealized GCM simulations. In gen-  
 521 eral, the WC is weaker in the warm climate than in the control climate, consistent with  
 522 the CMIP6 and AMIP simulations. WC strength increases with increasing entrainment  
 523 in both climates, but the sensitivity to entrainment is greater in the warm climate. As  
 524 a result, the WC weakens with warming more at lower entrainment rates than it does  
 525 at higher entrainment rates. While entrainment does affect the response of the WC to  
 526 warming, the spread due to variations in entrainment of  $1.6\% \text{ K}^{-1}$  (Figure 8) is not as  
 527 large as the spread due to differences across models in CMIP6 ( $14\% \text{ K}^{-1}$ ) or AMIP ( $12\%$   
 528  $\text{K}^{-1}$ ). Figures 4 and 5 suggest that this is because variations in entrainment only cap-  
 529 ture the size of the spread in MSE profile response, but not the size of the spread in  $\hat{\omega}$   
 530 response. Further, this may be partly because radiative feedbacks are not as fully rep-  
 531 resented in the idealized model as they are in the CMIP6 and AMIP models, and our  
 532 analysis in Sections 2.2 and 2.3 suggests that they have an amplifying effect on the WC  
 533 response.

### 534 **3.5 GMS in idealized simulations**

535 From Wills et al. (2017), the Walker circulation strength varies inversely with a GMS  
 536 measure similar to  $\text{GMS}_{\text{WC}}$  in this idealized GCM when entrainment is set to zero. Here  
 537 we determine whether this relationship between WC strength and GMS responses holds  
 538 with variations in entrainment. Looking at Equation 1, we notice that in the idealized  
 539 simulations the sum of changes in  $R_{w-e}$  and  $S_{w-e}$  is negligible because the  $Q$  flux at the  
 540 surface is fixed and changes in radiation are very nearly zonally uniform because the sim-  
 541 ulations do not have cloud-radiative effects or water vapor-radiative feedback. There-  
 542 fore, the radiative and surface flux terms vanish from Equations 1 and 5 when applied  
 543 to the idealized simulations. Consequently, in the idealized simulations, Equation 5 re-



**Figure 7.** (a) WC strength versus entrainment for a control climate with default longwave optical depth (blue) and a warm climate with double longwave optical depth (red) in the idealized GCM simulations. Error bars show the standard error. (b) Relationship between GMS<sub>WC</sub> response and WC response to warming and changes in entrainment in the idealized GCM simulations. Delta indicates the fractional change from the reference case of the control climate ( $\alpha = 1$ ) with zero entrainment ( $\hat{\epsilon} = 0$ ). Filled circles indicate the response of GMS<sub>WC</sub> and open circles indicate the response of boundary GMS<sub>WC</sub>, where boundary GMS<sub>WC</sub> is defined by Equation 11. Black line is a reference line with slope of -1. Blue symbols indicate that the perturbed climate is a control climate and red symbols indicate that the perturbed climate is a warm climate.

544 duces to

$$545 \quad \delta \overline{\omega}_{w-e} \simeq -\delta \text{GMS}_{\text{WC}} - \frac{\Delta \langle \mathbf{u} \cdot \nabla h \rangle_{w-e}}{\langle \omega \frac{\partial h}{\partial p} \rangle_{w-e}}, \quad (10)$$

546 where  $\delta$  is a fractional response and  $\Delta$  is a difference between simulations in response  
 547 to warming or changes in entrainment parameter. Equation 10 is an excellent approx-  
 548 imation, and thus there is an inverse relationship between WC strength and GMS<sub>WC</sub> if  
 changes in the horizontal MSE advection term are small.

549 To evaluate the role of horizontal MSE advection, we compare changes in WC strength  
 550 and GMS<sub>WC</sub>. Figure 7b shows that GMS<sub>WC</sub> response does not have the expected inverse  
 551 relationship with WC response (although this does hold approximately for the zero en-  
 552 trainment case that was also considered by Wills et al. (2017)), indicating that changes  
 553 in horizontal advection terms are important in Equation 10. This is problematic because  
 554 although we have some understanding of how entrainment affects the vertical MSE ad-  
 555 vection term through MSE and vertical velocity profiles, we do not have a similar un-  
 556 derstanding for horizontal MSE advection. In order to reduce the role of horizontal ad-  
 557 vection in our analysis, we define a version of GMS<sub>WC</sub> appropriate for the WC in our ide-  
 558 alized simulations called the “boundary GMS<sub>WC</sub>.”

559 The boundary GMS<sub>WC</sub> is defined using MSE averaged over the *boundaries* of the  
 560 WC ascent and descent regions which are defined in our idealized simulations based on  
 561 contours of the zonally anomalous  $Q$  flux ( $Q^*$ ). Between the surface and top of atmo-  
 562 sphere, the  $Q^*$  contours create an elliptic cylinder for each region. We define  $h_b$  as the

563 average value of  $h$  around the elliptic contour at each level and each time, so that  $h_b$  does  
 564 not vary in latitude or longitude. The boundary  $\text{GMS}_{\text{WC}}$ , or  $\text{GMS}_{\text{WC}}^b$  is then defined as

$$\text{GMS}_{\text{WC}}^b = -g \frac{\left\langle \omega \frac{\partial h_b}{\partial p} \right\rangle_{w-e}}{\bar{\omega}_{w-e}}. \quad (11)$$

565 Only MSE is averaged over the boundary to give  $h_b$ . Terms with the subscript  $w - e$   
 566 are averaged over the areas of the elliptic ascent and descent regions. Intuitively, bound-  
 567 ary  $\text{GMS}_{\text{WC}}$  is helpful because it removes the effect of horizontal variations within the  
 568 ascent and descent regions and focuses on the MSE variations on the boundaries of the  
 569 ascent and descent regions that matter for export and import of energy out of and in to  
 570 these regions.

571 To further see why the boundary  $\text{GMS}_{\text{WC}}$  is helpful, we decompose  $h$  at a given verti-  
 572 cal level as the sum of  $h_b$  and a residual,  $h'$  such that  $h = h_b + h'$ . Considering the  
 573 ascent region, the advection terms can now be written

$$\left\langle \omega \frac{\partial h}{\partial p} \right\rangle_w + \langle \mathbf{u} \cdot \nabla h \rangle_w = \left\langle \omega \frac{\partial h_b}{\partial p} \right\rangle_w + \left\langle \omega \frac{\partial h'}{\partial p} \right\rangle_w + \langle \mathbf{u} \cdot \nabla h' \rangle_w, \quad (12)$$

574 where we have used that  $h_b$  does not vary horizontally. A similar result holds for the de-  
 575 scent region. In order for  $\left\langle \omega \frac{\partial h_b}{\partial p} \right\rangle_w$  to dominate the right-hand side, we need  $h'$  advect-  
 576 ion,  $\left\langle \omega \frac{\partial h'}{\partial p} \right\rangle_w + \langle \mathbf{u} \cdot \nabla h' \rangle_w = \langle \nabla_{3d} \cdot (\mathbf{u}_{3d} h') \rangle_w$ , to be negligible. By the divergence the-  
 577 orem, this will be the case if  $\mathbf{u}_{3d} h'$  is close to zero on the boundary of the elliptic cylin-  
 578 der, which will be the case if  $h'$  is close to zero on this boundary, meaning that the  $h$  con-  
 579 tours at each vertical level align with the -10 and 10  $\text{W m}^{-2}$  surface  $Q^*$  contours used  
 580 to define the boundary. At latitudes near the equator, we expect the  $h$  contours to roughly  
 581 align with the  $Q^*$  contours because  $Q^*$  is forcing anomalous warming and moistening  
 582 in the ascent region and anomalous cooling and drying in the descent region. If this is  
 583 approximately the case, then Equation 12 and the equivalent for the descent region gives  
 584 that

$$\left\langle \omega \frac{\partial h}{\partial p} \right\rangle_{w-e} + \langle \mathbf{u} \cdot \nabla h \rangle_{w-e} \simeq \left\langle \omega \frac{\partial h_b}{\partial p} \right\rangle_{w-e}. \quad (13)$$

585 Continuing to assume that  $h'$  is close to zero on the boundary of the elliptic cylinder and  
 586 repeating the derivation of Equation 10 but using  $\text{GMS}_{\text{WC}}^b$  gives that

$$\delta \bar{\omega}_{w-e} \simeq -\delta \text{GMS}_{\text{WC}}^b. \quad (14)$$

587 We evaluate the extent to which Equations 13 and 14 hold by looking at Figure  
 588 7b. We can see that the relationship between WC response and boundary  $\text{GMS}_{\text{WC}}$  re-  
 589 sponse is much closer to the slope  $-1$  line than the relationship between WC response  
 590 and  $\text{GMS}_{\text{WC}}$ . The extent to which the WC and boundary  $\text{GMS}_{\text{WC}}$  responses depart from  
 591 the slope  $-1$  line is due almost entirely to the neglect of  $h'$  advection because Equation  
 592 10 is nearly exact in the idealized simulations.

593 Our results show that the boundary  $\text{GMS}_{\text{WC}}$  is a better metric than  $\text{GMS}_{\text{WC}}$  for  
 594 understanding the WC response across entrainment rates and climates in the idealized  
 595 GCM. By contrast, it was sufficient to use the  $\text{GMS}_{\text{WC}}$  in the analysis of the CMIP6 and  
 596 AMIP simulations. Horizontal MSE advection does provide a contribution in the CMIP6  
 597 and AMIP simulations, but the multimodel mean of this contribution is close to zero and  
 598 the model spread is not as big as the spread in the  $\text{GMS}_{\text{WC}}$  contribution (Figures 1a,b).  
 599 The lesser role for the horizontal advection term in the CMIP6 and AMIP simulations  
 600 may be because of differences in the structure of the WC. For example, the lesser role  
 601 may be because of differences in the pattern of heat fluxes for the warm pool as com-  
 602 pared to the elliptical anomaly in the idealized simulations or because the range of en-  
 603 trainment parameters is not as wide across CMIP6 and AMIP models as across the ide-  
 604 alized GCM simulations.

### 3.6 Boundary $\text{GMS}_{\text{WC}}$ response to warming and decomposition

Finally, we evaluate the response of boundary  $\text{GMS}_{\text{WC}}$  to warming and compare it to the response of the WC. Looking at Figure 8, we find that the responses of boundary  $\text{GMS}_{\text{WC}}$  and WC strength are of opposite sign, consistent with the inverse relationship found in Wills et al. (2017) and in the CMIP6 and AMIP models in Sections 2.2 and 2.3 (although those results used  $\text{GMS}_{\text{WC}}$  rather than boundary  $\text{GMS}_{\text{WC}}$ ). Further, both the weakening of the WC and the increase in boundary  $\text{GMS}_{\text{WC}}$  with warming dampen with increasing entrainment rate. However, the decreases in WC strength are mostly smaller than the increases in boundary  $\text{GMS}_{\text{WC}}$ , and this reflects that the boundary  $\text{GMS}_{\text{WC}}$  does not fully account for contributions from changes in the horizontal MSE advection.

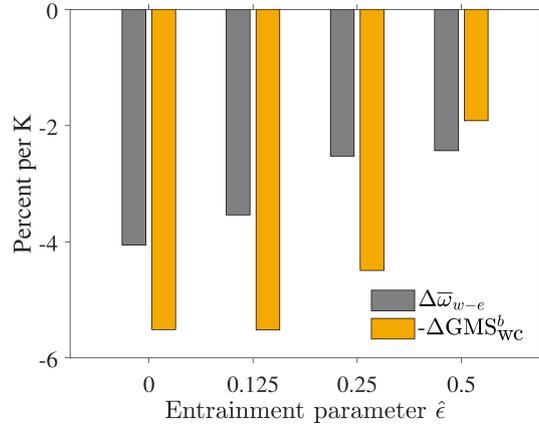
We decompose the response of boundary  $\text{GMS}_{\text{WC}}$  to warming in the idealized simulation as was done in Section 2.7 but here we replace  $\text{GMS}_{\text{WC}}$  with boundary  $\text{GMS}_{\text{WC}}$  in Equation 8. Similar to the CMIP6 and AMIP results, the  $\Delta\hat{\omega}$  contribution is positive and larger in magnitude than the negative  $\Delta h$  contribution (Figure 9). The  $\Delta h$  contribution is again the result of compensation between positive contributions due to temperature and geopotential height changes and a negative contribution from humidity changes. Again, the contribution from changes in humidity is dominated by changes in saturation specific humidity.

As the entrainment rate is increased, the increase in boundary  $\text{GMS}_{\text{WC}}$  with warming becomes weaker. This is mostly related to the  $\Delta h$  contribution becoming more negative, but it is partially compensated for by the  $\Delta\hat{\omega}$  contribution becoming more positive. The more negative changes in  $\Delta h$  are as expected given that entrainment makes the atmosphere less stable and has a greater effect in the warmer climate than the control climate (Singh & O’Gorman, 2013). Looking at Figures 5c,f, since entrainment has more of an effect on  $\hat{\omega}$  in the warmer climate, increasing the entrainment rate will also make the  $\Delta\hat{\omega}$  contribution more positive. Thus increasing entrainment does dampen the increase in boundary  $\text{GMS}_{\text{WC}}$  with warming as was expected initially, but there is less of an effect than would occur if only changes in MSE were considered.

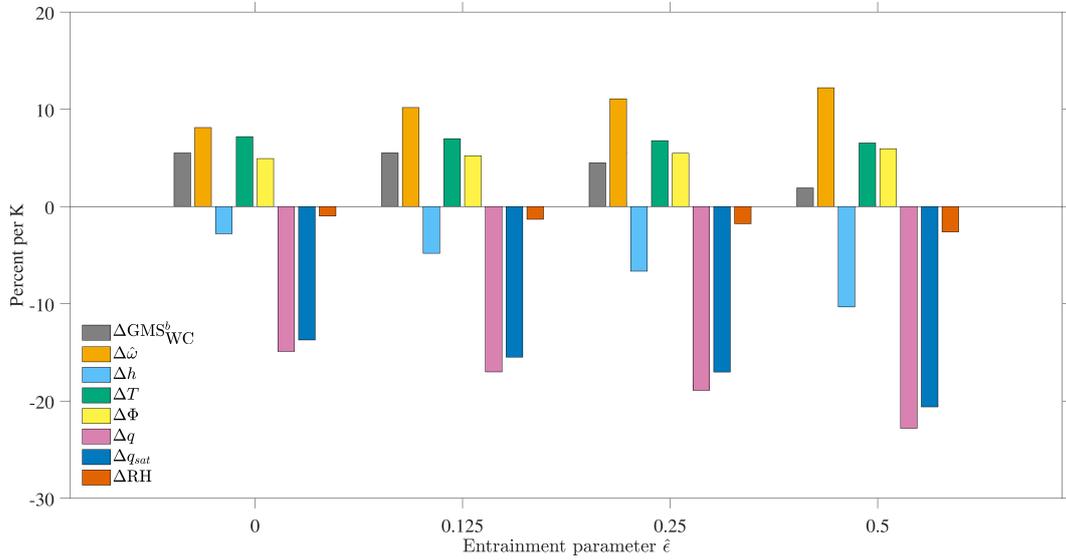
Figure 9 shows that changes in specific humidity are the main reason that the  $\Delta h$  contribution becomes more negative as the entrainment rate increase, while the contribution from changes in temperature does not vary noticeably across entrainment rates. Using the Clausius-Clapeyron relationship, the greater contribution from changes in specific humidity with increasing entrainment is consistent with temperature lapse rates steepening with increasing entrainment, and more so in a warmer climate (Held et al., 2007; Singh & O’Gorman, 2013). But why do changes in lapse rates with increasing entrainment not affect the temperature contribution? It appears to be because entrainment also affects the control-climate boundary  $\text{GMS}_{\text{WC}}$  and Figure 9 shows the fractional response to warming. If instead absolute changes in boundary  $\text{GMS}_{\text{WC}}$  with warming are considered (Figure S12), the temperature contribution does become less positive as the entrainment rate is increased as expected.

## 4 Conclusions

We have evaluated the response of the Walker circulation to warming in comprehensive and idealized GCM simulations using an energetic perspective, with an emphasis on the spread in the response across GCM projections. A surprising result of our study is that the spread across AMIP models, which all have the same imposed SST, is similar to the spread across CMIP6 models, which are coupled to a dynamic ocean. The spread of WC response in the AMIP models is  $12\% \text{ K}^{-1}$  and the spread in CMIP6 models is  $14\% \text{ K}^{-1}$ . Still, the strong role of the atmosphere does not preclude a role of the ocean since the spread from each component separately need not sum to the total spread of the coupled system. In addition, the ascent and descent regions of the WC are not in exactly



**Figure 8.** Response of WC strength (gray) to warming compared with minus the response of boundary  $\text{GMS}_{wc}^b$  (orange) in idealized GCM simulations with varying entrainment rates.



**Figure 9.** Same as Figure 6 but for idealized GCM simulations with varying entrainment rates and using the boundary  $\text{GMS}_{wc}^b$  instead of  $\text{GMS}_{wc}$ .

655 the same location in each GCM which may complicate the comparison of CMIP6 sim-  
 656 ulations with the AMIP simulations in which the SST response is imposed the same way  
 657 in all models. A potential candidate for the spread across AMIP models not considered  
 658 here is the role of differences in resolution across models, given that WC strength is sen-  
 659 sitive to resolution in an idealized atmospheric GCM (Silvers & Robinson, 2021).

660 In an MSE budget analysis of WC strength in CMIP6 and AMIP simulations, a  
 661 weakening of the WC is related primarily to increases in  $GMS_{WC}$ , and this weakening  
 662 is amplified by changes in radiation. The gross moist stability thus emerges as a key fac-  
 663 tor, consistent with the heuristic idea that for a given energy transport, a higher GMS  
 664 is associated with a weaker circulation. Changes in horizontal MSE advection and sur-  
 665 face latent and sensible fluxes play a smaller role. We find a large spread in WC response  
 666 to warming across CMIP6 and AMIP models, with  $GMS_{WC}$  response anticorrelated with  
 667 WC response. The spread in  $GMS_{WC}$  response in AMIP models is  $13\% K^{-1}$ , and its spread  
 668 in CMIP6 models is similar at  $14\% K^{-1}$ .

669 The role of radiation is substantial in both CMIP6 and AMIP models. In the CMIP6  
 670 models, there is a multimodel mean weakening of the WC of  $12\% K^{-1}$  with a multimodel  
 671 mean contribution of  $7\% K^{-1}$  from radiation. In the AMIP models, there is a multimodel  
 672 mean weakening of the WC of  $12\% K^{-1}$  with a multimodel mean contribution of  $5\% K^{-1}$   
 673 from radiation. The radiation contribution is always the same sign as the  $GMS_{WC}$  con-  
 674 tribution; that is, contributing a weakening. Further, the decomposition of the radiation  
 675 contribution (Figures 1c,d) indicates a strong role of WC-linked changes in radiation across  
 676 CMIP6 and AMIP models. We find that cloud radiative feedbacks are amplifying the  
 677 WC responses in CMIP6 and AMIP models, and such feedbacks have been previously  
 678 found to affect the WC strength (e.g., Peters and Bretherton (2005)).

679 The  $GMS_{WC}$  response to warming involves changes in the vertical profiles of MSE  
 680 and vertical velocity. Both the vertical profile of MSE and the shape of the vertical ve-  
 681 locity profile contribute to the spread across CMIP6 and AMIP simulations in  $GMS_{WC}$   
 682 response. They are both sensitive to convective entrainment which is an uncertain and  
 683 parameterized process in GCMs. Therefore, we evaluate the role of entrainment in set-  
 684 ting  $GMS_{WC}$  and WC strength in an idealized GCM. To do so, we modify the simpli-  
 685 fied Betts Miller convection scheme of Frierson (2007) to include a simple representation  
 686 of entrainment. We find that horizontal MSE advection plays an important role in the  
 687 WC in some simulations, which is complicating because we do not have a theory for the  
 688 relationship between entrainment and horizontal advection. To address this, we define  
 689 a boundary  $GMS_{WC}$  which approximately includes the role of horizontal MSE advection  
 690 while not involving horizontal velocities and horizontal MSE gradients. Rather, the bound-  
 691 ary  $GMS_{WC}$  involves vertical advection of MSE profiles averaged over the boundary of  
 692 each of the ascent and descent regions. We find that the WC weakens with warming, but  
 693 less so at higher entrainment rates. This is consistent with increases in boundary  $GMS_{WC}$   
 694 that get weaker with increasing entrainment. The effect of increased entrainment on bound-  
 695 ary  $GMS_{WC}$  response can be understood through the fact that entrainment tends to make  
 696 the atmosphere less stable in terms of the vertical profile of MSE, and it does so to a greater  
 697 extent in the warmer climate. However, entrainment also affects the shape of the vertical-  
 698 velocity profile, and this tends to weaken the effect of entrainment on boundary  $GMS_{WC}$ .  
 699 The results from the idealized GCM provide a demonstration of a causal linkage between  
 700 an imposed change in thermal stratification and resulting changes in WC strength in a  
 701 way that is consistent with what would be expected from the energetic analysis.

702 We conclude that the atmosphere plays a key role in setting the spread in WC re-  
 703 sponse to warming, especially through changes in  $GMS_{WC}$  and cloud-radiative feedbacks.  
 704 Convective entrainment influences boundary  $GMS_{WC}$  response and thus the WC response  
 705 in the idealized GCM. However, the spread in  $GMS_{WC}$  response across CMIP6 and AMIP  
 706 models is primarily from intermodel differences in vertical velocity profiles and these in-  
 707 termodel differences are much bigger than the spread in vertical velocity profiles that re-

708 sults from changing entrainment in the idealized GCM. Thus it seems unlikely that dif-  
709 ferences in representation of entrainment are the dominant source of spread across CMIP6  
710 and AMIP models. Rather, other influences on vertical velocity profiles are likely a ma-  
711 jor cause of the substantial spread in WC response in GCMs. The projected response  
712 of vertical velocity profiles to climate warming over the tropical oceans has been linked  
713 to changes in the horizontal pattern of boundary-layer temperature, including through  
714 their Laplacian (Lindzen & Nigam, 1987; Back & Bretherton, 2009; Duffy et al., 2020).  
715 There is no spread in the SST change in the AMIP simulations, but the Laplacian of boundary-  
716 layer temperature change is not fully determined by the SST change (Duffy et al., 2020).  
717 What determines the changes in the shape of the vertical velocity profiles in the East  
718 and West Pacific in particular should be investigated in future work.

## Appendix A The entraining simplified Betts-Miller convection scheme

The SBM convection scheme of Frierson (2007) relaxes temperature profiles to a moist adiabat. Here, the scheme is modified such that temperature profiles are relaxed to a that of an *entraining* plume. The target humidity profile is calculated as in the original scheme using the target temperature profile (based on the entraining plume) and a reference relative humidity of 70%. The entrainment rate,  $\epsilon$ , varies inversely with height and is given by  $\epsilon = \frac{\hat{\epsilon}}{z}$ , where  $\hat{\epsilon}$  is a non-dimensional entrainment parameter and  $z$  is height. The convection scheme represents an ensemble of clouds, each of which detrains at a different level, which is crudely represented by the inverse relationship with  $z$ . The temperature lapse rate is assumed to be dry-adiabatic below the lifted condensation level (LCL). Above the LCL,

$$\frac{\partial h_s}{\partial z} = -\epsilon(h_s - h_e), \quad (\text{A1})$$

where  $h_s = c_p T + gz + Lr_s$  is the saturation MSE,  $r_s$  is the saturation mixing ratio, and  $h_e$  is the environmental MSE. Here we use the GCM's gridbox MSE to represent the environmental MSE. Using the definition of  $h_s$  gives

$$c_p \frac{dT}{dz} + g + L \frac{dr_s}{dz} = -\epsilon(h_s - h_e). \quad (\text{A2})$$

Using  $r_s = r_s(T, p)$  and applying the hydrostatic equation gives

$$c_p \frac{dT}{dz} + g + L \frac{\partial r_s}{\partial T} \frac{dT}{dz} - L\rho g \frac{\partial r_s}{\partial p} = -\epsilon(h_s - h_e). \quad (\text{A3})$$

Next, group like terms to give

$$(c_p + L \frac{\partial r_s}{\partial T}) \frac{\partial T}{\partial z} + g - L\rho g \frac{\partial r_s}{\partial p} = -\epsilon(h_s - h_e). \quad (\text{A4})$$

Rearranging to solve for  $\frac{\partial T}{\partial z}$  gives

$$\frac{\partial T}{\partial z} = \frac{-\epsilon(h_s - h_e) - g + L\rho g \frac{\partial r_s}{\partial p}}{c_p + L \frac{\partial r_s}{\partial T}}. \quad (\text{A5})$$

Following the original scheme, we approximate the partial derivatives of  $r_s$  with respect to pressure and temperature as  $\partial r_s / \partial p = -r_s / p$  and  $\partial r_s / \partial T = Lr_s / (R_v T^2)$ , respectively, where  $R_v$  is the gas constant for water vapor. Substituting these two expressions into Equation A5 and applying the ideal gas law gives

$$\frac{\partial T}{\partial z} = \frac{-\epsilon(h_s - h_e) - g(1 + \frac{Lr_s}{RT})}{c_p + \frac{L^2 r_s}{R_v T^2}}. \quad (\text{A6})$$

Using the hydrostatic equation and the ideal gas law gives the lapse rate of the entraining plume above the LCL

$$\frac{\partial T}{\partial \ln p} = \frac{\frac{RT}{gc_p} \epsilon (h_s - h_e) + \frac{RT}{c_p} + \frac{Lr_s}{c_p}}{1 + \frac{L^2 r_s}{c_p R_v T^2}}. \quad (\text{A7})$$

Notice that the temperature profile for the entraining plume reduces to a moist adiabat when  $\epsilon = 0$ .

## Open Research

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749 multiple funding agencies who support CMIP6 and ESGF. Each model and the variant  
 750 ID of the simulations used is listed in Table S1. The modified version of the GFDL ide-  
 751 alized moist spectral atmospheric model and the analysis scripts used for this work are  
 752 available on Zenodo at <https://doi.org/10.5281/zenodo.6620842>.

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