

1 **Sensitivity of Tropical Extreme Precipitation to Surface**  
2 **Warming in Aquaplanet Experiments Using a Global**  
3 **Nonhydrostatic Model**

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8 **Key Points:**

- 9
- 10 • The sensitivity of tropical extreme precipitation to warming is larger than the Clausius-  
11 Clapeyron rate
  - 12 • Results are more sensitive to horizontal resolution when convection is parametrized  
13 than explicitly resolved
  - 14 • Super Clausius-Clapeyron rates are mainly due to dynamical changes, but appear  
unrelated to changes in indicators of convective organization

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## 15 Abstract

16 Increases of atmospheric water vapor holding capacity with temperature ( $7-8\%K^{-1}$ ,  
 17 CC-rate) can lead to increasing Extreme Precipitation (EP). Observations show that trop-  
 18 ical EP has increased during the last five decades with a rate higher than in the extra-  
 19 tropics. Global climate models (GCM's) diverge in the magnitude of increase in the trop-  
 20 ics, and cloud-resolving models (CRM's) indicate correlations between changes in trop-  
 21 ical EP and organization of deep convection. We conducted global-scale aquaplanet ex-  
 22 periments at a wide range of resolutions with explicit and parameterized convection to  
 23 bridge the gap between GCM's and CRM's. We found increases of tropical EP beyond  
 24 the CC rate, with similar magnitudes when using explicit convection and parametrized  
 25 convection at the resolution it is tuned for. Those super-CC rates are produced due to  
 26 strengthening updrafts where extreme precipitation occurs, and they do not exhibit re-  
 27 lations with changes in convective organization.

## 28 Plain Language Summary

29 Theory and observations indicate tropical extreme precipitation might increase with  
 30 global warming. Projections from climate models agree on increases in the extratrop-  
 31 ics, but not in the tropics. More idealized simulations indicate links between increases  
 32 of tropical extreme precipitation and changes in the spatial organization of the meteo-  
 33 rological systems producing those extremes. Using a novel model approach, we found  
 34 that tropical extreme precipitation increases with warming more than expected due to  
 35 increases in the dynamics of the extreme precipitation systems, whereas changes of the  
 36 spatial organization have a small role.

## 37 1 Introduction

38 The Clausius-Clapeyron relation provides a theoretical starting point for under-  
 39 standing the response of extreme precipitation to a warming climate. At lower tropo-  
 40 spheric temperatures this relation predicts a saturation specific humidity change of ap-  
 41 proximately  $7-8\%K^{-1}$  (CC rate) (Trenberth et al., 2003). With such an increase of  
 42 saturation water vapor in the atmosphere it is likely that the amount of precipitation  
 43 from events where most of the water vapor precipitates out will increase with warming,  
 44 and thus this value represents a basic scaling for the sensitivity of extreme precipitation  
 45 to warming (Berg et al., 2013). This behavior of extreme precipitation events stands in  
 46 contrast with global mean changes in precipitation with warming, which is instead linked  
 47 to enhanced longwave cooling of the troposphere. This cooling is mostly balanced by en-  
 48 hanced convective heating through release of latent heat, leading to increasing mean global  
 49 precipitation at a rate lower than, and not physically related to, the CC rate (Newell et  
 50 al. (1975); Mitchell et al. (1987); Boer (1993)).

51 Observations provide evidence partly in favor of this scaling of extreme precipita-  
 52 tion. Since the 1950s there have been statistically significant increases in the number of  
 53 extreme precipitation events in more regions than there have been statistically signif-  
 54 icant decreases (Hartmann et al., 2013). However, observations indicate that sensitiv-  
 55 ities depend also on the type of precipitation with higher values than CC rate (super-  
 56 CC) for convective precipitation (Berg et al., 2013). This discrepancy arises naturally  
 57 between midlatitudes, where extreme precipitation is usually associated with frontal ac-  
 58 tivity and midlatitude storms (Kodama et al., 2019), and the tropics where convection  
 59 is the main driver. This was shown by O’Gorman (2015), who found that daily extremes  
 60 in the tropics are more sensitive to climate warming than those in the extratropics and  
 61 suggested one possible cause is from dynamical origin - changes in vertical motion.

62 General Circulation Model (GCM) simulations also produce a general increase of  
 63 extremes, the strength of which depends on latitude (O’Gorman & Schneider, 2009b; O’Gorman,  
 64 2012). In models from the Coupled Model Intercomparison Project (CMIP) phase 3 for

example, extratropical sensitivities consistently predict that precipitation extremes increase more slowly with surface air temperature than atmospheric water vapor content; however, tropical changes are not consistent among models, with sensitivities ranging from  $1.3\%K^{-1}$  to  $30\%K^{-1}$  (O’Gorman & Schneider, 2009a). These studies suggest that the discrepancy in the tropics may arise from inaccurate simulation of upward velocity during convection. Bhattacharya et al. (2017) suggested that to improve modeled tropical precipitation extremes, it is essential to better represent the upward velocity associated with those extremes. Increasing horizontal resolution may be a way to improve the simulation of convection in GCM’s where precipitation is not resolved by the coarse grid and has to be parametrized. However, those convective parametrization schemes are sensitive to horizontal model resolution and time-step length (Li et al., 2011a, 2011b; Yang et al., 2014; Lu et al., 2014; Benedict et al., 2017; Williamson, 2013) and thus the sensitivity of extreme precipitation to warming varies not just among individual models but also across horizontal resolutions with a single model.

Given the long-standing structural uncertainties among CGM’s, Cloud Resolving Models in idealized setups of Radiative-Convective Equilibrium (RCE) (Manabe and Strickler (1964); Nakajima and Matsuno (1988); Tompkins and Craig (1998)) have been used to study tropical convection and sensitivities of extreme precipitation to warming. On such setups and under certain conditions RCE can become unstable (Nilsson & Emanuel, 1999) and lead to spontaneous spatial organization of convection. In RCE simulations, it has been shown that extreme precipitation increases close to the CC rate if self-aggregation is absent (Romps, 2011; Muller et al., 2011) or if the degree of organization does not change (Bao et al., 2017); whereas super-CC behaviour has been found when self-aggregation increases with warming (Singleton & Toumi, 2013; Pendergrass et al., 2016; Bao et al., 2017; Bao & Sherwood, 2019).

Here we use a less idealized set of aquaplanet simulations to study the uncertainties of changes in tropical extreme precipitation by using a nonhydrostatic atmospheric GCM in rotating configuration with a meridional gradient of temperature (Neale & Hoskins, 2000; Medeiros et al., 2015). Model description and methods are presented in Section 2. In Section 3, we compare the sensitivity of tropical extreme precipitation to warming between simulations with parametrized and explicit convection and its resolution dependency, study contributors to those sensitivities, and look for relationships between the change of convective organization and the change of precipitation extremes. Finally, we present the conclusions in Section 4.

## 2 Model setup and methods

Simulations were performed using the ICOSahedral Nonhydrostatic Atmospheric general circulation model (ICON-A). ICON-A is built using the Max Planck Institute physical parametrization package, which originates from the ECHAM6.3 general circulation model (Mauritsen et al., 2019) and with adaptations to account for the change in the dynamical core and a new turbulence parametrization. A full description is given in Giorgetta et al. (2018).

The experiments were conducted using the aquaplanet configuration, which uses the Qobs zonally symmetric SST as surface boundary conditions (Neale & Hoskins, 2000). Owing to its simplicity (e.g. no topography, land-sea contrasts, surface heterogeneities), this configuration helps to understand the physical atmospheric processes driving the changes of extremes in response to global warming (Li et al., 2011a). Moreover, because diurnal insolation and the radiatively active species are held at equinoctial and hemispherically symmetric geometry, the model statistics are zonally and hemispherically symmetric, which helps to identify significant signals using relatively short integrations.

A range of simulations at different horizontal resolutions were performed with parametrization of convection on and off (explicit convection) for the control and uniformly increased SST of 4K (Table 1). We used explicit convection at resolutions lower than 10 km since even at coarse resolutions without parametrization of convection, models are able to pro-

**Table 1.** Experiment resolutions, time step lengths and cases with parameterized and explicit convection. The resolution is the approximate side length of squares with the same area as the average triangle in the ICON grid.

Grid name	Resolution [km]	Time step [min]	Parameterized	Explicit
R2B4	160	7.5	yes	yes
R2B5	80	3.75	yes	yes
R2B6	40	0.83	yes	yes
R2B8	10	0.25	no	yes

118 duce large scale features related to convection (Webb et al., 2015; Retsch et al., 2019).  
 119 The simulations were initialized analytically and used time-invariant boundary condi-  
 120 tions for SST, spectral solar irradiation, well mixed greenhouse gases CO<sub>2</sub>, CH<sub>4</sub>, N<sub>2</sub>O,  
 121 CFC's and O<sub>3</sub> concentration. All experiments were run for four years and we treated  
 122 the first year as spin-up, except for *R2B8* where the simulation length was limited to six  
 123 months with one month of spin-up.

124 Daily zonal mean precipitation for explicit and parameterized experiments are shown  
 125 in supporting information Figure S1. As is to be expected the global mean precipitation  
 126 increases with warming, whereas the zonal distribution of precipitation follows a double  
 127 ITCZ structure for explicit convection simulations. When resolution is increased, pre-  
 128 cipitation tends to be more zonally distributed with a displacement of the ITCZ away  
 129 from the equator and an increase of midlatitude precipitation at the expense of tropi-  
 130 cal precipitation, particularly for *R2B6* and *R2B8*. In general, however, the zonal dis-  
 131 tribution of precipitation in the parameterized convection experiments is somewhat er-  
 132 ratic across resolutions, and the ITCZ behaviours in our experiments differ from those  
 133 of Retsch et al. (2019) who used an earlier version of ICON-A and found a single ITCZ  
 134 structure for explicit convection simulations at resolutions *R2B4*, *R2B5* and *R2B6*, while  
 135 a double ITCZ prevailed for parameterized convection. Since extremes in the tropics are  
 136 more likely to occur within the ITCZ, simulated shifts in the large-scale tropical circu-  
 137 lation might obscure our results and so in the following we shall discuss precipitation ex-  
 138 tremes in the entire tropics from 30°S and 30°N.

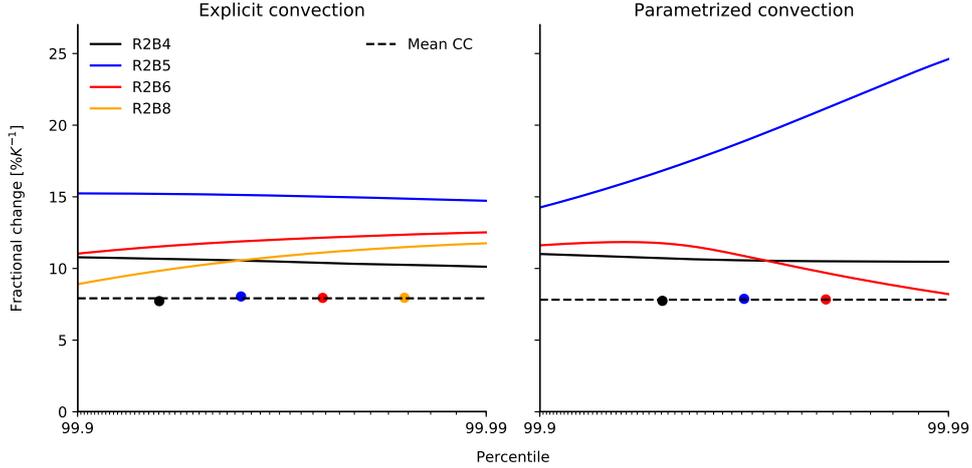
139 To study changes in tropical Extreme Precipitation (EP) with warming, we define  
 140 EP as the cases of grid points between 30°S to 30°N over the entire period exceeding  
 141 the *i*th percentile of daily precipitation:

$$142 \quad EP(i) = \frac{1}{n} \sum_0^n Pr_n \geq Pr_i, \quad (1)$$

143 where *i* varied from 99.9 to 99.99, *Pr* is daily precipitation and *Pr<sub>i</sub>* is the *i*th per-  
 144 centile of daily precipitation. The selection of this metric allows us to capture the be-  
 145 haviour of the precipitation distribution tail instead of focusing on one particular per-  
 146 centile. With it we calculate the Sensitivity of tropical Extreme Precipitation to warm-  
 147 ing (SEP) as the fractional change in EP ( $\delta EP(i)/EP(i)$ ) normalized by change in tem-  
 148 perature ( $\delta T$ ):

$$149 \quad SEP(i) = \frac{\delta EP(i)}{\delta T \cdot EP(i)} = \frac{EP(i)_{4K} - EP(i)_{CTL}}{4K \cdot EP(i)_{CTL}}, \quad (2)$$

150 where the subscripts *CTL* and *4K* denote control and 4K experiments, respectively.



**Figure 1.** Solid lines show sensitivities of tropical extreme precipitation to warming. Fractional changes of lower tropospheric saturation specific humidity for each simulation are displayed as dots and the dashed line shows the mean across resolutions.

### 3 Results

#### 3.1 Sensitivity of tropical extreme precipitation to warming

Sensitivities of tropical extreme precipitation to warming are displayed as solid lines in Figure 1. Extremes from both explicit and parametrized convection experiments increase with warming. For explicit convection simulations, sensitivities vary from  $9\%K^{-1}$  to  $15\%K^{-1}$  with a tendency to converge for the strongest extremes. This tendency is not clear for the experiments with parametrized convection, where at low resolution (R2B4) we observe a nearly constant sensitivity value, then at R2B5, it ranges from  $14\%K^{-1}$  to  $24\%K^{-1}$ ; and finally for the highest resolution (R2B6) it drops for the strongest extremes.

Fractional changes of lower tropospheric saturation specific humidity with warming in the tropics fall close to the CC rate (dots in Figure 1). In all simulations, extreme increases are higher than those fractional changes (super-CC, compare solid lines and dots). This suggests that not just increases in the capacity of the atmosphere to hold water vapor have an impact on extremes, but other processes contribute too. However, we note that for parametrized convection the amplitude of this difference varies with resolution more than it does for explicit convection. Since the convective parametrization implemented in ICON-A has been tuned for a grid resolution of R2B4, spurious behaviors at resolutions R2B5 and R2B6 resolutions might occur. This is mainly because the convective scheme is tuned to remove convective instability on a certain timescale. When resolution is increased from R2B4, the model dynamics may produce finer scale instabilities more rapidly than the parametrization can remove these, resulting in explicitly resolved updrafts or so-called grid-point storms (Williamson, 2013). We speculate that the erratic behavior of the higher resolution simulations is caused by the inadvertent competition between parameterized convection and partially resolved convective clouds. Given this, we restrain our analysis to explicit convection simulations and R2B4 with parametrized convection. Increases of extreme precipitation from parametrized convection at the resolution it is tuned for (R2B4) are similar to the increases with explicit convection, particularly to that at R2B4. This indicates a low model sensitivity of extreme precipita-

180 tion changes with warming to the activation of convective parametrization for this res-  
 181 olution.

182 **3.2 Contributors to increases of tropical extreme precipitation**

183 Next we will use a scaling derived by Muller et al. (2011), which assumes that changes  
 184 in precipitation efficiency with warming are negligible and separates changes in extremes  
 185 in terms of changes in dynamics, through vertical motion (here pressure velocity,  $\omega$ ), and  
 186 in thermodynamics, through the vertical gradient of saturation specific humidity, respec-  
 187 tively. Derivation and testing of the scaling are presented in supporting information S1.

$$\begin{aligned}
 188 \quad SEP \approx & \underbrace{\frac{\left[ \delta(\omega) \frac{\partial q_s}{\partial p} \right]}{\delta T \cdot \left[ \omega \frac{\partial q_s}{\partial p} \right]}}_{\text{Dynamic}} + \underbrace{\frac{\left[ \omega \delta \frac{\partial q_s}{\partial p} \right]}{\delta T \cdot \left[ \omega \frac{\partial q_s}{\partial p} \right]}}_{\text{Thermodynamic}}. \quad (3)
 \end{aligned}$$

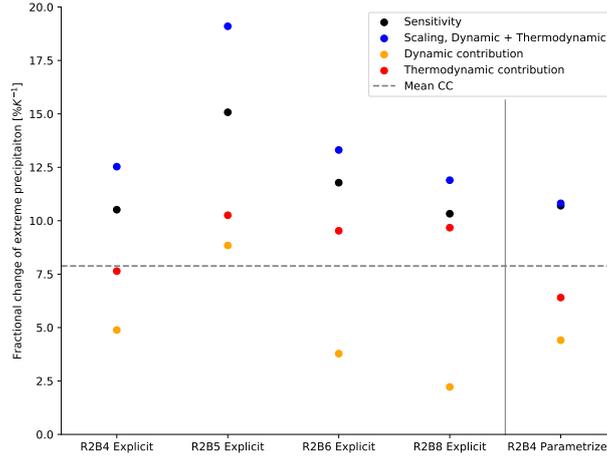
189 From the components of the scaling, we can identify that increases of tropical extreme  
 190 precipitation derive from convective circulation strength through increases in  $\omega$  and/or  
 191 increases of vertical gradient of saturation specific humidity with warming, which are ex-  
 192 pected to follow the CC rate as long as the strongest vertical gradients of saturation spe-  
 193 cific humidity are located in the lower troposphere.

194 Results of the scaling (Equation 3) are displayed in Figure 2. We note a similar be-  
 195 haviour across resolutions with explicit convection and R2B4 with parametrized convec-  
 196 tion, in that increases of extreme precipitation result from both positive dynamics and  
 197 thermodynamics contributions. In all simulations the dynamic contribution is positive,  
 198 therefore contributing to the super-CC behavior of the model, whereas the thermody-  
 199 namic contribution alone is close to the CC rate. It should be noted that the dynamic  
 200 response exhibits a larger dependency on resolution when going from R2B4 to R2B5, than  
 201 to the use of parametrization of convection, and that there is a nearly constant offset be-  
 202 tween the scaling and the actual sensitivities which might indicate either that the pre-  
 203 cipitation efficiency assumption is inaccurate or that other processes influence changes  
 204 in tropical extreme precipitation.

205 **3.3 Are sensitivities related to changes in convective organization?**

206 We showed that tropical extreme precipitation increases with warming at rates higher  
 207 than the CC rate and that strengthening of convective circulations when extremes oc-  
 208 cur leads to those super-CC tendencies in our experiments. As mentioned in Section 1,  
 209 the amplitude of the sensitivity of tropical extreme precipitation to warming might be  
 210 related to changes of convective organization whereby super-CC changes in extremes are  
 211 correlated with increases of convective organization, and so we investigate if this is also  
 212 the case in our simulations. To this end, we quantify the degree of convective organiza-  
 213 tion in the tropics and its change with warming in a variety of ways: using subsiding frac-  
 214 tion prime ( $SF'$ , Noda et al. (2019)) in the entire tropical band, as well as in smaller sub-  
 215 domains of varying sizes centered at the extreme event, and solely in the tropical band  
 216 using organization index with eight point connectivity for resolutions R2B6 and R2B8  
 217 (Iorg, Tompkins and Semie (2017)) and an organization index with zero connectivity to  
 218 include coarse resolutions (Iorg<sub>-0</sub>, Becker and Wing (2020)). A detailed description of  
 219 the metrics is given in supporting information S2.

220 We found both increased and decreased tendencies of convective organization with  
 221 warming (Figure 3). At high resolution (R2B5-8) all indices show a reduction in convec-  
 222 tive organization, whereas at coarse resolution (R2B4) convection tends to self-organise  
 223 with warming in the areas where extremes occurs, while disorganise at large scale (we  
 224 obtained similar results for coarse grained resolutions to R2B4, see Figure S6). Those



**Figure 2.** Sensitivity of tropical extreme precipitation to warming (black), scaling (blue) and contributions to sensitivities from dynamics (yellow) and thermodynamic (red). Values are calculated by averaging the extremes interval (99.9 to 99.99). The dashed grey line indicates the mean fractional change of lower tropospheric saturation specific humidity across simulations and the vertical line separates explicit convection experiments from parametrized convection at R2B4.

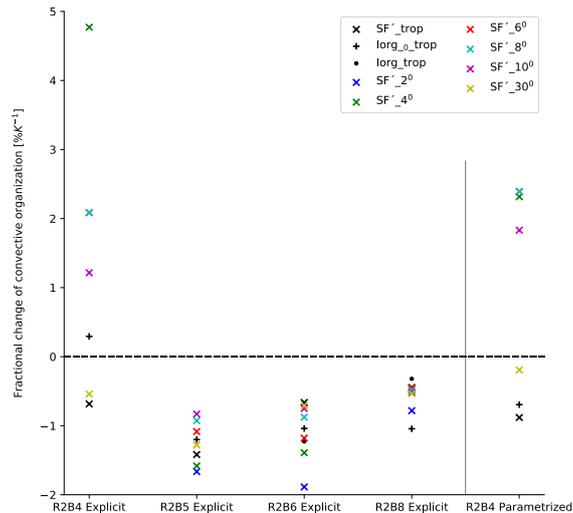
225 results are in agreement with Muller and Held (2012) who found that self-organization  
 226 of convection is favored by coarse resolution but opposite to Bao et al. (2017), Singleton  
 227 and Toumi (2013) and Pendergrass et al. (2016) where increases of extremes correlate  
 228 with self-organization. This discrepancy across resolutions suggests that changes in tropi-  
 229 cal convective organization have a negligible impact on changes of tropical extreme pre-  
 230 cipitation with warming in our simulations.

#### 231 4 Conclusions

232 Aquaplanet simulations with the ICON-A model are performed to explore the sensi-  
 233 tivity of tropical extreme precipitation to warming across a wide range of spatial resolu-  
 234 tions with and without parametrization of convection. We find positive sensitivities with  
 235 amplitudes larger than the increase of lower tropospheric saturation specific humidity,  
 236 or CC rate, at all resolutions. Results from explicit convection simulations converge for  
 237 the strongest precipitation extremes, whereas for parameterized convection simulations,  
 238 the sensitivities strongly vary with horizontal resolution, although results from R2B4 are  
 239 similar to those from the explicit convection simulations. We suggest this occurs since  
 240 the parametrization scheme used in ICON-A was tuned for that particular resolution.

241 We next investigate whether dynamical changes can explain the super-CC behaviour  
 242 of tropical extreme precipitation using a diagnostic framework. In all simulations we find  
 243 positive contributions from dynamics, resulting in stronger updrafts where extreme pre-  
 244 cipitation occurs. Nevertheless, thermodynamical changes resulting from changes in the  
 245 vertical gradient of saturation specific humidity also contribute relative to the simple  
 246 CC-scaling in the higher resolution explicit simulations, but not in the coarse resolution  
 247 simulations (R2B4). Furthermore, it should be noted that there is a considerable resid-  
 248 ual in the diagnosed changes, suggesting that assuming invariant precipitation efficiency  
 249 is inaccurate or that other processes might be involved.

250 Finally, we explore whether convective organization could be involved using an ar-  
 251 ray of indices that in various ways characterise the degree of organization. We find some-  
 252 what surprisingly in most cases organisation decreases with warming: in the explicit con-



**Figure 3.** Fractional changes of organization metrics for large scale ( $SF'_{trop}$ ,  $Iorg_{-0-trop}$  and  $Iorg_{trop}$ ) and for subdomains centered where extremes occur ( $SF'_{-30}$ ,  $SF'_{-10}$ ,  $SF'_{-8}$ ,  $SF'_{-6}$ ,  $SF'_{-4}$  and  $SF'_{-2}$ ). The vertical line separates explicit convection experiments from parametrized convection at R2B4. Note that, given the resolution of R2B4, a subdomain of  $2^\circ \times 2^\circ$  will contain a unique grid point and the fractional change of  $SF'$  will be zero; and subdomains from  $6^\circ \times 6^\circ$  and  $8^\circ \times 8^\circ$  will contain the same amount of grid points, producing equal fractional changes.

253 vection simulations with 10-80 km resolution (R2B5-8) it decreases by about  $1\%K^{-1}$ .  
 254 In the two simulations with 160 km (R2B4) convection disorganizes at large scale; but  
 255 self-organizes in the areas where extremes occur. It is concluded that convective organiza-  
 256 tion played either no or a negligible role in causing the model's super-CC behaviour  
 257 of tropical extreme precipitation.

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 268 scribed in Giorgetta et al. (2018) and its source code and components are available on  
 269 <https://mpimet.mpg.de/en/science/modeling-with-icon/code-availability>. The configu-  
 270 ration scripts to run the model, the data and scripts to reproduce the figures can be found  
 271 in Uribe et al. (2020).

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