# Spatial organisation affects the pathway to precipitation in simulated trade-wind convection

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#### Abstract

We investigate whether and how spatial organisation affects the pathway to precipitation in large-domain hectometer simulations of the North Atlantic trades. We decompose the development of surface precipitation (P) in warm shallow trade cumulus into a formation phase, where cloud condensate is converted into rain, and a sedimentation phase, where rain falls towards the ground while some of it evaporates. With strengthened organisation, rain forms in weaker updrafts from smaller cloud droplets so that cloud condensate is less efficiently converted into rain. At the same time, organisation creates a locally moister environment and modulates the microphysical conversion processes that determine the raindrops' size. This reduces evaporation and more of the formed rain reaches the ground. Organisation thus affects how the two phases contribute to P, but only weakly affects the total precipitation efficiency. We conclude that the pathway to precipitation differs with organisation and suggest that organisation buffers rain development.

## Spatial organisation affects the pathway to precipitation in simulated trade-wind convection

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#### 7 Key Points:

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8	• The development of surface precipitation in simulated trade-wind convection is decomposed
9	into a formation and sedimentation phase
10	• As organisation strengthens, less cloud condensate is converted into rain, but more rain
11	reaches the ground as evaporation is suppressed
12	• Organisation affects rain formation by modulating the local moisture environment, cloud
13	vertical motion and microphysical properties

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#### 14 Abstract

We investigate whether and how spatial organisation affects the pathway to precipitation in large-15 domain hectometer simulations of the North Atlantic trades. We decompose the development of 16 surface precipitation (P) in warm shallow trade cumulus into a formation phase, where cloud con-17 densate is converted into rain, and a sedimentation phase, where rain falls towards the ground while 18 some of it evaporates. With strengthened organisation, rain forms in weaker updrafts from smaller 19 cloud droplets so that cloud condensate is less efficiently converted into rain. At the same time, 20 organisation creates a locally moister environment and modulates the microphysical conversion pro-21 cesses that determine the raindrops' size. This reduces evaporation and more of the formed rain 22 reaches the ground. Organisation thus affects how the two phases contribute to P, but only weakly 23 affects the total precipitation efficiency. We conclude that the pathway to precipitation differs with 24 organisation and suggest that organisation buffers rain development. 25

#### <sup>26</sup> Plain Language Summary

Clouds in the trade-wind region organise into a variety of spatial patterns. We investigate how 27 this spatial organisation influences rain development in simulations of trade-wind convection. We 28 divide the formation of surface precipitation into two phases. In the first phase, rain forms from the 29 collision of cloud droplets or the collection of cloud droplets by raindrops. In the second phase, rain 30 falls towards the ground while some of the rain evaporates. Our study shows that as organisation 31 strengthens, rain forms less efficiently, but a larger fraction of that rain reaches the ground as evap-32 oration is reduced. Thus, organisation in the simulations affects the way surface rain is generated. 33 Our analyses suggest that it does so by modulating the cloud vertical motion in which rain forms, 34 the local moisture environment through which rain falls and the microphysical conversion processes. 35

#### 36 1 Introduction

What makes it rain? Precipitation was often neglected in studies of trade-wind convection because 37 it was assumed that the convection is too shallow and short-lived to form precipitation (Siebesma, 38 1998; Stevens, 2005). Although there was already ample evidence of precipitation in the trade-wind 39 region shown by Byers and Hall (1955) or Short and Nakamura (2000), it was not until attention to 40 the trades and its clouds increased due to their large contribution to uncertainty in cloud feedbacks 41 and climate sensitivity (Bony & Dufresne, 2005; Vial et al., 2013) that a more nuanced picture of 42 trade-wind convection settled. The Rain In Cumulus over the Ocean (RICO) campaign (Rauber 43 et al., 2007) was key in substantiating that precipitation is frequent in the trades (Nuijens et al., 44 2009), and highlighted that precipitation was often observed with arc-like cloud structures reminis-45 cent of cold pool outflows (Snodgrass et al., 2009). Subsequent studies confirmed that trade-wind 46 convection organises into a variety of spatial structures — and that this often occurs in conjunc-47 tion with precipitation development (Stevens et al., 2020; Denby, 2020; Bony et al., 2020; Schulz 48 et al., 2021; Vogel et al., 2021; Radtke et al., 2022). How does spatial organisation influence the 49 development of (surface) precipitation in the trades? In this study, we exploit realistic large-domain 50 hectometer(hm)-scale simulations of the North Atlantic trades (Schulz & Stevens, 2023) to investi-51 gate whether and how spatial organisation affects the pathway to trade-cumulus precipitation. 52 53

Precipitation formation depends on dynamic, thermodynamic and microphysical interactions on 54 different spatial and temporal scales. Due to the broad range of scales and processes involved, an 55 understanding of rain formation and contributing processes remains challenging, even for warm, 56 shallow trade cumulus. The representation of trade-cumulus precipitation among large-eddy simu-57 lations (LES) differs largely (vanZanten et al., 2011). An understanding of how spatial organisation 58 relates to warm rain development could help interpret and reduce these differences. Organisation 59 may affect how efficient rain forms and how much evaporates through modulating mesoscale circu-60 lations or the local moisture environment (Seifert & Heus, 2013; Narenpitak et al., 2021). Moreover, 61 understanding the relationship between spatial organisation and precipitation may also be key to 62 disentangle the mechanisms of organisation and explain its influence on the total cloud cover in the 63 trades, a prerequisite to further constrain the climate feedback of the trades (Nuijens & Siebesma, 64 2019; Bony et al., 2020). 65

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Analysing rain radar measurements upstream of Barbados taken during the EUREC<sup>4</sup>A field cam-67 paign (Hagen et al., 2021; Stevens et al., 2021), Radtke et al. (2022) show that while the occurrence 68 of trade-wind precipitation is related to organised cells, the mean rain rate varies largely indepen-69 dently of the cells' *degree* of organisation. However, scenes with similar precipitation but different 70 degrees of spatial organisation also differed in the moisture environment. Similarly, Yamaguchi et 71 al. (2019) find that in idealized LES, shallow cumulus precipitation varies little, but the sizes and 72 spatial distribution of clouds differ in response to large changes in the aerosol environment. Could 73 spatial organisation be a mechanism to maintain precipitation in different environments, enabling 74 or creating a different pathway to precipitation? 75

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To answer this question on a process-level, we make use of large-domain hm-scale simulations of the 77 North Atlantic trades that were run for the period January to February 2020 during the EUREC<sup>4</sup>A 78 campaign (Bony et al., 2017; Stevens et al., 2021; Schulz & Stevens, 2023) and were designed to 79 explore spatial organisation on the mesoscale (20-200km). We follow the method of Langhans et 80 al. (2015) and Lutsko and Cronin (2018) and decompose the development of surface precipitation 81 into two phases, (i) a formation phase, in which cloud condensate is converted into rain water, and 82 (ii) a sedimentation phase, in which the formed rain water falls towards the ground while part of 83 it evaporates. Sect. 2 describes the setup and microphysical scheme of the simulations and our 84 analysis method. Sect. 3.1 shows that spatial organisation in scenes of  $\mathcal{O}(100 \,\mathrm{km})$  influences how 85 these two phases contribute to the development of trade-wind precipitation. Sect. 3.2 explains such 86 behaviour and interprets it as a form of buffering, before we conclude in Sect. 4. 87

#### <sup>88</sup> 2 Methods

#### <sup>89</sup> 2.1 EUREC<sup>4</sup>A large-domain ICON hm-scale simulations

The simulations are conducted with the LES configuration of the ICOsahedral Non-hydrostatic (ICON) model (Dipankar et al., 2015). ICON solves the compressible Navier–Stokes equations on an unstructured grid as detailed in Zängl et al. (2015) and Dipankar et al. (2015). To explore mesoscale convective variability, the simulations are performed with a relatively fine hectometrescale mesh over a large domain of  $\mathcal{O}(1000 \text{km})$  for an extended EUREC<sup>4</sup>A campaign period from 9 January to 19 February 2022. They are realistically forced with initial and boundary data from a storm resolving simulation at 1.25-km grid spacing, which is in turn initialised and nudged at its lateral boundaries to the atmospheric analysis of the European Centre for Medium-Range Weather Forecasts (similar to Klocke et al., 2017). Here, we analyse a simulation with 625 m gridspacing that covers the western tropical Atlantic from 60.25-45.0 °W and 7.5-17.0 °N, spanning about 1650 km in the east-west direction, and 1050 km in the north-south direction. In the vertical, 150 levels are used, resulting in 70 m and 85 m vertical resolution at 1000 m and 2000 m height, respectively.

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Schulz and Stevens (2023) show that this simulation reproduces differences in the mesoscale struc-103 ture underlying the canonical forms of trade cumulus organisation of Stevens et al. (2020), as well as 104 variability in precipitation, which makes them a good starting point to investigate how the process 105 of precipitation may vary with spatial organisation. A nested 312 m simulation does not show a sub-106 stantially greater skill in representing different cloud organisations or rain rates (Schulz & Stevens, 107 2023) and shows the same qualitative behaviour in our analysis (not shown). We refer the reader 108 to Schulz and Stevens (2023) for an in-depth observational evaluation and comparison of the simu-109 lations. In the supplementary material we summarise and discuss the evaluation of the simulation 110 properties relevant for our work. 111

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In the simulations, turbulence is parameterised with the Smagorinsky scheme, microphysics with the 113 two-moment mixed-phase bulk microphysics scheme of Seifert and Beheng (2006) and a cloud con-114 densation nuclei concentration of  $130 \text{ cm}^{-3}$  is prescribed. Warm rain is produced by autoconversion 115 and accretion, defined following Seifert and Beheng (2001) as  $\frac{\partial L_r}{\partial t}\Big|_{au} \sim L_c^2 \overline{x}_c^2$  and  $\frac{\partial L_r}{\partial t}\Big|_{acc} \sim L_c L_r$ , 116 where  $L_{\rm r}$  is rain water content,  $L_{\rm c}$  cloud water content and  $\overline{x}_{\rm c} = \frac{L_{\rm c}}{N_{\rm c}}$  mean mass of cloud droplets 117 with cloud droplet number concentration  $N_{\rm c}$ . To quantify the production of rain, we recalculate the 118 autoconversion and accretion rates from the instantaneous 3D model output of cloud water, rain 119 water and cloud effective radius  $r_{\rm eff}$ , from which the volume radius  $r_{\rm v}$  is derived by  $(r_{\rm v}/r_{\rm eff})^3 = 0.8$ 120 (Freud & Rosenfeld, 2012) to calculate the cloud droplets' mean mass. The 3D output is available 121 every 3h. 122

#### 123 **2.2** Investigating spatial organisation and the pathway to precipitation

We investigate spatial organisation in scenes of  $4 \times 4^{\circ}$  (about  $450 \times 450 \text{ km}$ ), an area extent similar to previous studies (Radtke et al., 2022; George et al., 2022). Figure 1a shows three scenes with different degrees of organisation as an example. To mask high ice clouds, scenes with outgoing longwave radiation  $< 275 \text{ W m}^{-2}$  are excluded (changes in the threshold do not affect our qualitative results), as well as scenes with little precipitation  $P < 0.01 \text{ mm h}^{-1}$ . Here and if not indicated otherwise, we refer to domain mean values. In total about 2000 scenes are used in the analyses (about 7 scenes across the domain every 3 hours).

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Following Bretherton and Blossey (2017) and Narenpitak et al. (2021), we assess the degree of spatial organisation as mesoscale variability in the moisture field, which is closely connected to the cloud structure. First, the total water path,  $W_{\rm T}$ , defined as the sum of vertically integrated water vapor, cloud condensate, and rain, is coarse-grained into tile sizes of  $20 \times 20 \,{\rm km}^2$ , representing variability associated with the mesoscale (Orlanski, 1975). Subsequently, the coarse-grained  $W_{\rm T}$  is binned into quartiles and the difference between the fourth and first quartile is calculated as organisation metric  $O_{\Delta W_{\rm T}}$ . This metric classifies the three example scenes from weakly organised (low

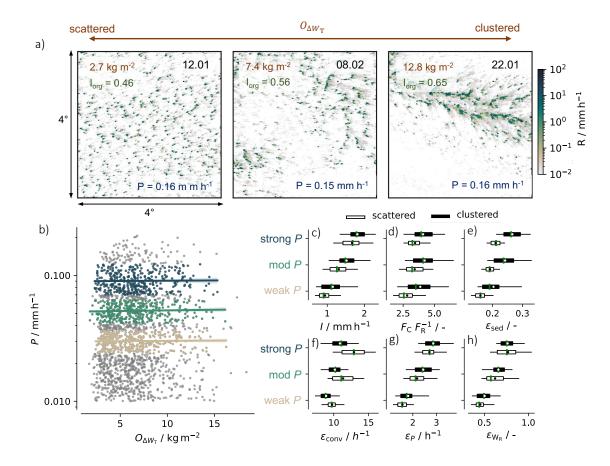


Figure 1. a) Three example scenes with similar scene-averaged precipitation P (i.e. rain amount, blue) but different degrees of organisation characterised by  $O_{\Delta W_{\rm T}}$  (orange) and the  $I_{\rm ORG}$  (green). Colour shading denotes rain rate R. Grey shading denotes cloud albedo calculated from simulated cloud liquid water path. b) P as a function of  $O_{\Delta W_{\rm T}}$ . Three different rain regimes with weak P =(0.024, 0.037], mod P = (0.042, 0.064] and high P = (0.07, 0.12] are distinguished. c) Rain intensity I, d) cold pool fraction,  $F_{\rm C}$ , per rain fraction,  $F_{\rm R}$ , e) sedimentation efficiency  $\epsilon_{\rm sed}$ , f) conversion efficiency  $\epsilon_{\rm conv}$ , g) precipitation efficiency  $\epsilon_{\rm P}$ , and h) rain water loading efficiency  $\epsilon_{\rm W_R} = \frac{W_{\rm R}}{W_{\rm L}}$ , where  $W_{\rm R}$  rain water path, shown for the three precipitation regimes, separated into a clustered ( $O_{\Delta W_{\rm T}} > 70^{\rm th}$  percentile, filled bars) and scattered sample ( $O_{\Delta W_{\rm T}} < 30^{\rm th}$ percentile, empty bars). The green line denotes the median, the dot the mean, the box the interquartile range and the whiskers denote the 5<sup>th</sup> and 95<sup>th</sup> percentile.

 $O_{\Delta W_{\rm T}}$ ) on the left, to more strongly organised (high  $O_{\Delta W_{\rm T}}$ ) on the right. This is consistent with a visual subjective classification of the cloud field, the nearest neighbour clustering technique  $I_{\rm ORG}$ (Tompkins & Semie, 2017), and the cloud pattern classification of Stevens et al. (2020). According to this classification, the left scene depicts a Gravel pattern, characterised by scattered convection (low  $I_{\rm ORG}$ ), and the right scene a Fish pattern, characterised by more clustered convection (higher  $I_{\rm ORG}$ ). An overview and more in-depth discussion of different organisation metrics has recently been given by e.g. Janssens et al. (2021).

<sup>147</sup> To investigate the pathway to trade-wind precipitation, we decompose the development of surface

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rain following Langhans et al. (2015) and Lutsko and Cronin (2018) into (i) a formation phase and 148 (ii) a sedimentation phase. In phase (i), warm rain initially forms by the merging of small cloud 149 droplets, parameterised with the autoconversion rate. Additionally, rain is produced as falling rain-150 drops collect cloud droplets, parameterised with the accretion rate. Autoconversion dominates the 151 formation of rain especially for young or short-lived clouds, while accretion contributes more to the 152 formation of rain as clouds live longer and there is more time available for the collision-coalesence 153 process to take place (Feingold et al., 2013). To quantify how efficient the formation of rain water 154 is, we define a conversion efficiency 155

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$$\epsilon_{\rm conv} = \frac{C_{\rm R}}{W_{\rm L}},\tag{1}$$

where  $C_{\rm R} = C_{\rm Auto} + C_{\rm Acc}$  with  $C_{\rm Auto}$  and  $C_{\rm Acc}$  denoting the vertically integrated autoconversion and accretion rates and  $W_{\rm L}$  the cloud liquid water path. In phase (ii), the rain formed by autoconversion and accretion sediments towards the ground. During this process, some rain evaporates. The rain that does not evaporate reaches the ground as surface precipitation, P, so that we call

$$\epsilon_{\rm sed} = \frac{P}{C_{\rm R}} = 1 - \epsilon_{\rm evap} \tag{2}$$

the sedimentation efficiency with  $\epsilon_{\text{evap}}$  the evaporation efficiency. Please note that we do not refer to in-cloud sedimentation here, but, following Langhans et al. (2015) and Lutsko and Cronin (2018), we use sedimentation efficiency to refer to how much rain reaches the ground instead of evaporating.

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The product of the conversion and sedimentation efficiencies describes how much cloud water in a given time interval is returned to the surface as precipitation, representing an overall precipitation efficiency  $\epsilon_{\rm P}$ , e.g. as used in Lau and Wu (2003):

$$\underbrace{\frac{P}{W_{\rm L}}}_{\epsilon_{\rm P}} = \underbrace{\frac{C_{\rm R}}{W_{\rm L}}}_{\epsilon_{\rm conv}} \cdot \underbrace{\frac{P}{C_{\rm R}}}_{\epsilon_{\rm sed}}.$$
(3)

Said differently, the inverse of  $\epsilon_{\rm P}$  is the time it takes to remove all cloud water at the given precipita-170 tion rate, thus describing a typical residence time. It is to note that precipitation efficiency itself has 171 no unique definition (e.g. Sui et al., 2020). Different results may emerge for different definitions and 172 also depend on local versus domain-mean views. However, using an approximation of the condensa-173 tion rate following Muller and Takayabu (2020) instead of liquid water path in (3) results in the same 174 qualitative behaviour. Here, we mainly exploit precipitation efficiency and its decomposition into 175 conversion and sedimentation efficiencies as a proxy for the pathway that precipitation development 176 takes. 177

#### 178 **3 Results**

#### <sup>179</sup> 3.1 The pathway to precipitation varies with organisation

The hm-scale simulations reproduce EUREC<sup>4</sup>A observations in that scene precipitation in the trades varies mainly independently of organisation (Radtke et al., 2022). This is depicted in the example scenes in Fig. 1a, which display a similar rain rate but vastly different degrees of organisation, and is more quantitatively shown in Fig. 1b. In the simulations, scene rain rates vary up to  $0.2 \text{ mm h}^{-1}$  as shown in Fig. 1b, which compares well to rain rates observed in the RICO (Nuijens et al., 2009) and EUREC<sup>4</sup>A campaign (Radtke et al., 2022). In the following, we will show whether also the pathway to these rain rates is similar or in how far organisation affects how these rain rates are generated, and could thus be a process to maintain precipitation in different environments.

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To investigate this, we group our sample of scenes into three precipitation regimes, a weak, a moderate, and a strong precipitation regime, as visualised in Fig. 1b. In each regime, we define more organised scenes as  $O_{\Delta W_{\rm T}} > 70^{\rm th}$  percentile and less organised scenes as  $O_{\Delta W_{\rm T}} < 30^{\rm th}$  percentile, which we refer to as clustered and scattered sample, respectively. The conclusions are insensitive to the exact choice of threshold. This relative way of differentiating between more and less organised scenes is based on the simulated distribution of organisation, which will vary depending on how variable the large-scale conditions are and possibly which specific model is used.

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Fig. 1c shows that, instead of the mean rain rate, organisation tends to increase the rain inten-197 sity, which is again in line with observational studies of trade-wind (Radtke et al., 2022) and deep 198 convection (Louf et al., 2019). That is, clustered convection produces the same amount of scene 199 precipitation than scattered convection with more intense rain covering a smaller area. Detecting 200 cold pools based on the calculation and criterion of a mixed layer height smaller than 400 m follow-201 ing Touzé-Peiffer et al. (2022), clustered scenes are also populated by more cold pools as shown in 202 Fig. 1d, possibly associated with this increase in rain intensity. In clustered scenes, the cold-pool 203 fraction is about four times greater than the rain fraction, whereas in scattered scenes it is about 204 three times greater. These findings may already hint to an altered precipitation process in more 205 organised compared to less organised scenes. 206

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We investigate the relationship between organisation and the conversion, sedimentation and to-208 tal precipitation efficiency (eq. 3), shown in Fig. 2a. Organisation maximises towards the lower right 209 of the phase space, at low conversion and high sedimentation efficiencies. An increase in the degree 210 of organisation is thus related to a decrease in how efficient cloud water is converted into rain and 211 an increase in how efficient rain sediments as a greater fraction of rain reaches the ground instead 212 of evaporating. The sedimentation efficiency varies between 0.1 and 0.3, emphasising that much of 213 the rain evaporates, as reported by Naumann and Seifert (2016) or Sarkar et al. (2022). Fig. 2b 214 shows that precipitation maximises towards the upper right of the same phase space, that is, at high 215 sedimentation and conversion efficiencies. Within a precipitation regime, as shown in Fig. 1e, f, rain 216 thus sediments more efficiently but forms less efficiently in clustered compared to scattered scenes. 217 This behaviour is slightly enhanced in regimes with stronger precipitation. 218

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The product of the conversion and sedimentation efficiencies gives the overall precipitation efficiency, 220 denoted in the contour lines in Fig. 2. Precipitation efficiency varies closely with precipitation and 221 lies mostly between  $1 h^{-1}$  and  $3 h^{-1}$ . That one to three times the cloud liquid water path precipitates 222 per hour emphasises the rapid turnover and rain formation in trade-wind clouds, which with tops 223 greater than 2500 m "usually rain within half an hour" (Squires 1958). Because conversion efficiency 224 decreases but sedimentation efficiency increases with organisation, contours of precipitation effi-225 ciency and organisation lie perpendicular to each other in Fig. 2a. This means that organisation and 226 precipitation efficiency, like precipitation, vary mainly independently of each other. Composited 227 on the three different precipitation regimes, Fig. 1g shows that precipitation efficiency compared 228 to the conversion and sedimentation efficiency varies only weakly with organisation with a slight 229

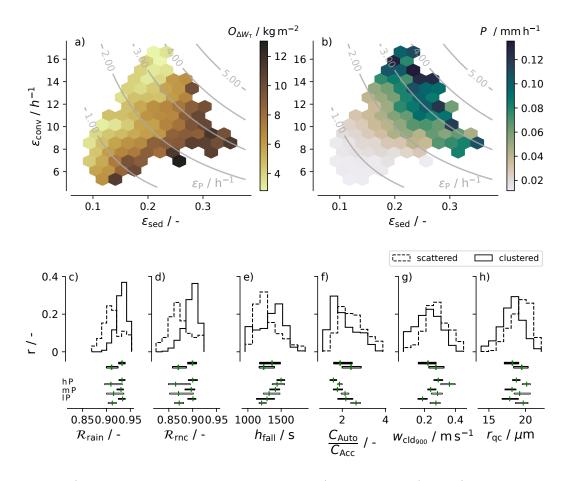


Figure 2. a) Degree of mesoscale organisation  $O_{\Delta W_{\rm T}}$  and b) precipitation P (shading) as a function of conversion efficiency  $\epsilon_{\rm conv}$  and sedimentation efficiency  $\epsilon_{\rm sed}$ . Contour lines denote precipitation efficiency from eq. (3). Relative frequency of c) rain-conditioned relative humidity  $\mathcal{R}_{\rm rain}$ , d) rain-and-no-cloud-conditioned relative humidity  $\mathcal{R}_{\rm rac}$ , e) fall height  $h_{\rm fall}$ , f) ratio of autoconversion  $C_{\rm Auto}$  to accretion  $C_{\rm Acc}$ , g) cloud-conditioned vertical velocity at 900 hPa  $w_{\rm cld_{900}}$  and h) mean cloud droplet radius  $r_{\rm qc}$  for the scenes separated into scattered (dashed line, empty bars) and clustered (solid line, filled bars) convection and divided into three precipitation regimes (as in Fig. 1, hP denoting the high P regime, mP the mod P regime and lP the low P regime). Horizontal boxes denote the interquartile range, vertical lines the median.

tendency to increase with organisation. Analysing the ratio of rain water path to cloud water path instead of the ratio between precipitation and cloud liquid water path gives the same result (Fig. 1h).

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233 Our analysis thus suggests that organisation weakly affects precipitation efficiency in terms of how

much cloud water precipitates on average, but changes the pathway to precipitation in terms of how

the formation versus sedimentation phases contribute to the development of surface precipitation.

 $_{236}$  Next, we investigate physical mechanisms behind this behaviour.

#### <sup>237</sup> 3.2 How does organisation affect the pathway to precipitation?

#### 3.2.1 Sedimentation efficiency

The sedimentation efficiency describes how much rain reaches the ground instead of evaporating. Following Lutsko and Cronin (2018), we suggest that  $\epsilon_{sed}$  should scale to a first approximation with the moisture environment through which the rain falls, or more explicitly with the saturation deficit, and the time it takes the rain to fall:

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$$\epsilon_{\text{evap}} = 1 - \epsilon_{\text{sed}} \sim (1 - \mathcal{R}_{\text{rain}}) \cdot t_{\text{fall}} = (1 - \mathcal{R}_{\text{rain}}) \cdot \frac{h}{v}, \tag{4}$$

where  $\mathcal{R}_{rain}$  is the averaged relative humidity the falling rain experiences, i.e. conditioned on pixels with rain water  $q_r > 0.001 \,\mathrm{g \, kg^{-1}}$  (vanZanten et al., 2011), and  $t_{fall}$  the average fall time, which depends on the average fall height h and fall velocity v. The higher the saturation deficit or the longer the rain falls and thus has time to evaporate, the higher evaporation and the lower the amount of rain reaching the ground.

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We hypothesise that organisation influences the moisture environment through which rain falls, 250 since it manifests itself in an uneven (horizontal) distribution of moisture, as also used in our metric 251 of organisation. Figure 2c shows that in the simulations, rain in clustered scenes indeed typically falls 252 through a more humid environment with a lower saturation deficit than in scattered scenes. This is 253 true for all precipitation regimes, with little variations in  $\mathcal{R}_{rain}$  between precipitation regimes. We 254 find that rain falls through a moister environment because the environment outside of or beneath 255 clouds is closer to saturation (about 3%, Fig. 2d), not just because more rain may falls within than 256 outside of clouds, e.g. due to different wind shears and cloud tilts. This is in line with the idea 257 that clouds in more organised scenes develop preferentially in the parts of the domain with moister, 258 more favourable thermodynamic conditions, e.g. preconditioned by former clouds (sometimes called 259 mutual-protection hypothesis, Seifert & Heus, 2013) or established through enhanced moisture trans-260 port into anomalously moist patches by mesoscale circulations (Narenpitak et al., 2021; George et 261 al., 2022). That way, clusters may form, clouds may be better protected from updraft buoyancy 262 reduction through entrainment (Mapes & Neale, 2011; Becker et al., 2018), and less rain evaporates. 263 264

Besides the moisture environment, organisation could also influence the fall time of the rain drops by 265 modulating the fall height or fall velocity (eq. 4). We define the fall height as average height where 266 rain is produced by autoconversion and accretion,  $h_A$ . Analysing  $h_A$  shows that rain in clustered 267 convection falls on average from slightly higher heights than in scattered convection (Fig. 2e), re-268 lated to a tendency of clouds to grow deeper and inversion heights to increase with organisation (not 269 shown). If the fall velocity stays unchanged, this would suggest that organisation slightly increases 270 the time it takes for rain to fall to the ground, which would act to enhance, not to reduce evaporation 271 in more strongly organised scenes. 272

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For the mean fall velocity, multiple factors, e.g. the strength of up- and downdrafts and the raindrops' size is decisive. The raindrop size was not included in the model output but the way rain is produced, i.e. in how far autoconversion versus accretion dominate the formation of rain, may serve as a proxy for the raindrops' size. Figure 3a shows that in how far autoconversion versus accretion contributes to rain formation explains 79% of the variations in sedimentation efficiency. Because autoconversion produces initial "embryo" raindrops when cloud droplets merge, whereas accretion

is responsible for the growth of raindrops through further collection of cloud droplets, an increased 280 contribution of accretion to rain formation indicates that raindrops have grown larger. Fig. 2f shows 281 that in more organised scenes the contribution of accretion to rain formation is increased. Raindrops 282 in more organised scenes are thus likely larger. Larger rain drops fall faster, reducing the fall time 283 and hence evaporation. 284

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When including, in addition to the relative importance of autoconversion and accretion,  $\mathcal{R}_{rain}$  as additional predictor, 85% of the variations in sedimentation efficiency can be explained. Additionally including  $h_A$  does not explain further variations. Our analysis thus suggests, as illustrated in Fig. 4, that organisation reduces evaporation and increases the sedimentation efficiency because rain in more organised scenes is increasingly produced by accretion so that raindrops are larger and fall 290 faster, through an environment that is moister.

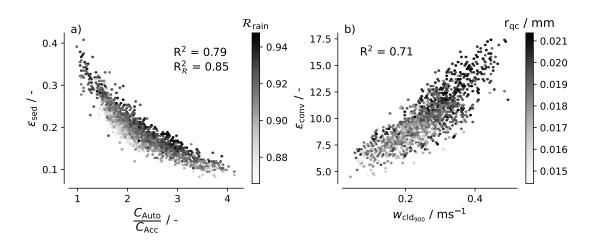


Figure 3. a) Sedimentation efficiency  $\epsilon_{sed}$  as a function of the relative importance of autoconversion  $C_{\text{Auto}}$  and accretion  $C_{\text{Acc}}$ . Shading denotes the rain-conditioned relative humidity  $\mathcal{R}_{\text{rain}}$ . b) Conversion efficiency  $\epsilon_{\rm conv}$  as a function of cloud-conditioned vertical velocity at 900 hPa  $w_{\rm cld_{900}}$ . Shading denotes the mean cloud droplet radius  $r_{\rm qc}$ .

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#### 3.2.2 Rain formation efficiency 292

Rain starts to form when sufficient cloud water has been produced and cloud droplets have grown to 293 raindrop size (e.g. Seifert & Stevens, 2010). To initiate and grow cloud particles the air's saturation 294 is important and influenced by thermodynamic conditions as well as vertical motions (Rogers & 295 Yau, 1996). 296

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In the simulations, organisation influences the clouds' vertical motion. Figure 2g shows that in 298 clustered scenes the mean in-cloud vertical motion near cloud base,  $w_{\rm cld_{900}}$  (cloud-conditioned, i.e 299 where cloud water  $q_c > 0.01 \,\mathrm{g \, kg^{-1}}$ , and at 900 hPa), is weaker than in more scattered scenes. This 300 initially appears surprising. It can be attributed, in part, to the presence of stronger downdrafts, 301 e.g. the 25th percentile of  $w_{cld_{900}}$  is lower, and in part to weaker updrafts as the mean and median 302 cloud upward motion is reduced (not shown). Bao and Windmiller (2021) found a similar decrease in 303

vertical motions with organisation in deep convection. Because organisation creates more favourable 304 thermodynamic conditions for cloud and rain formation with a local increase in humidity (Fig. 2 c, d), 305 this may allow clouds and rain to develop in less favourable dynamic conditions, i.e. at weaker mean 306 upward motions. Additionally, the cloud population in more organised scenes could also consist of 307 more long-lived, older clouds as indicated by the increased contribution of accretion to rain forma-308 tion. In these, updrafts may have already started to weaken. Future analysis of the cloud lifecycle 309 with respect to organisation may therefore potentially reconcile the apparent contradiction between 310 the expected larger raindrops resulting from increased accretion and the presence of weaker updrafts. 311

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More organised scenes also differ from less organised scenes in the mean cloud droplet radius. Fig. 2h 313 shows that in clustered scenes, the mean cloud droplet radius is smaller by about 1.3 µm than in 314 scattered scenes. From moderate to high precipitation, this difference increases, which is in line 315 with the strong decrease in conversion efficiency at high precipitation. The smaller cloud droplet 316 size in more organised scenes agrees with the weaker vertical motions. Besides, Cooper et al. (2013) 317 showed that mixing and entrainment affects cloud droplet growth and the onset of precipitation. By 318 changing the mixing characteristics of clouds, organisation might also influences the cloud droplets' 319 size. 320

Figure 3b shows that 70% of the variations in conversion efficiency are explained by the mean vertical motion at cloud base, to which the mean cloud droplet size is correlated. To conclude and as illustrated in Fig. 4, our analyses suggest that organisation reduces the efficiency with which cloud water is converted into rain water because rain in clustered scenes forms in weaker updrafts that correlate with smaller mean cloud droplets. We hypothesise that this is because favourable thermodynamic conditions may compensate for weaker dynamic conditions and organisation may affect the lifetime as well as the mixing characteristics of the cloud population.

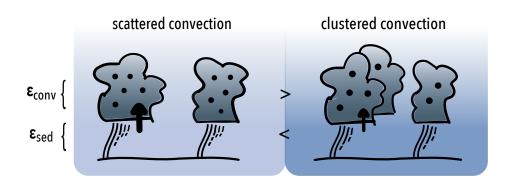


Figure 4. Conceptual illustration of the simulated pathway to surface precipitation in weakly (left) versus strongly (right) organised convection: For similar surface precipitation, as organisation strengthens, rain forms in a locally moister environment (shading) in weaker updrafts (arrow) and increasingly from accretion indicating larger raindrops (dots). As a consequence, evaporation is reduced, so that more rain reaches the ground compared to scattered convection ( $\epsilon_{sed}$  is larger), but rain forms less efficiently ( $\epsilon_{conv}$  is smaller), with both changes having a compensating, i.e. buffering effect on surface rain development.

#### 329 **3.2.3** Buffering

Organisation is associated with an increase in the sedimentation efficiency, but a decrease in the 330 formation efficiency and thus influences rain development in an opposing or stabilising way. This 331 may be interpreted as a form of buffering. Buffering as introduced in Stevens and Feingold (2009) 332 denotes that if there are different pathways to reach the same final state, these buffer or stabilise the 333 system against disruptions to any particular pathway. For example, cloud deepening as a dynami-334 cal response to increased droplet number concentration has been shown to buffer the microphysical 335 suppression of precipitation (Seifert et al., 2015). Our analyses suggest that organisation has a sta-336 bilising, i.e. buffering, relationship to rain development, due to opposing effects on rain formation 337 and sedimentation efficiency, as illustrated in Fig. 4. While in more scattered convection, rain de-338 velopment is characterised by efficient conversion of cloud water into rain water but also subsequent 339 evaporation is strong, in more clustered convection, increased sedimentation efficiency compensates 340 for a decreased conversion efficiency. These variations on the pathway to precipitation by organisa-341 tion may be an additional explanation for why rain development is so common in the trades. They 342 also offer an explanation for the observed (Radtke et al., 2022) and simulated independence between 343 rain amount and organisation rather than a reinforcement. 344

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Based on our results, we hypothesise that such buffering effect of organisation on rain develop-346 ment is related to (i) an interplay of thermodynamic and dynamic conditions and (ii) (life)time 347 effects. Regarding (i), more favourable thermodynamic conditions may allow clouds to develop un-348 der less favourable dynamic conditions. Regarding (ii), the increased contribution of accretion to rain 349 formation with organisation indicates longer-lived, older clouds, in agreement with the expectation 350 of more sustained convection with organisation. In older, more mature clouds, the rain formation 351 process had time to evolve (Feingold et al., 2013), so that raindrops may grow larger and fall out 352 more efficiently, while updrafts may have already weakened so that further rain forms less efficiently. 353

#### 354

#### 4 Summary and conclusions

We exploit realistic large-domain hm-scale simulations of the North Atlantic trades to investigate whether and how organisation affects the pathway to trade-cumulus precipitation. We decompose the development of surface precipitation following Langhans et al. (2015) into a formation phase, where cloud condensate is converted to rain, and a sedimentation phase, where the formed rain falls to the ground while some of it evaporates. In the simulations, organisation affects how these two phases contribute to rain development, summarised schematically in Fig. 4.

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With strengthened organisation, rain in the hm-scale simulations forms in and falls through a lo-362 cally more humid environment. Additionally, rain is increasingly produced by accretion rather than 363 autoconversion, which indicates that clouds live longer and raindrops grow larger. Larger raindrops, 364 that fall through a more humid environment experience less evaporation, leading to an increase in 365 the sedimentation efficiency. The relative importance of accretion and autoconversion explains 79%366 of the variations in sedimentation efficiency, increasing to 85% when including the rain-conditioned 367 relative humidity as additional predictor. A locally more humid environment is in line with the 368 idea that an increase in organisation is related to more humid patches in which clouds develop and 369 which protect clouds from dilution and raindrops from evaporation. It may suggest that organisa-370

tion also increases the efficiency with which cloud condensate is converted to rain. However, in more 371 organised scenes rain forms in weaker updrafts (as in Bao & Windmiller, 2021), and from smaller 372 cloud droplets. This leads to cloud water being less efficiently converted to rain, in agreement with 373 radiative-convective equilibrium simulations by Lutsko and Cronin (2018). 71% of the variations in 374 conversion efficiency are explained by the in-cloud vertical motion at cloud base, to which the cloud 375 droplet size is correlated. Possibly because the thermodynamic environment is more favourable with 376 organisation, less favourable dynamic conditions already allow for rain formation, or lifetime effects 377 may play a role here. Both effects, the increase in sedimentation efficiency and the decrease in 378 conversion efficiency, largely compensate, so that organisation does not substantially affect the total 379 precipitation efficiency. 380

381

Our analyses suggest that organisation does not reinforce, but rather buffers rain development 382 via opposing effects on the rain formation and sedimentation efficiencies. They offer an explana-383 tion for the observed and simulated independence between rain amount and organisation. While in 384 less organised scenes rain development is characterised by efficient conversion of cloud condensate 385 into rain, in more organised scenes more efficient sedimentation, as evaporation is suppressed, in-386 creasingly contributes to surface rain development. It remains to be shown in how far these results 387 carry over to other models and observations. In our simulations, we conclude that the pathway to 388 precipitation differs with spatial organisation. 389

#### <sup>390</sup> Open Research Section

The simulation output is freely available and can be easily accessed via the EUREC<sup>4</sup>A-Intake catalog at https://github.com/eurec4a/eurec4a-intake as described at https://howto.eurec4a.eu/

simulations. Detailed information about the simulations is given in Schulz and Stevens (2023).

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