



## Abstract

The Atlantic Tradewind Ocean-Atmosphere Mesoscale Interaction Campaign (ATOMIC) took place in January–February 2020. It was designed to understand the relationship between shallow convection and the large-scale environment in the trade-wind regime. Lagrangian large eddy simulations, following the trajectory of a boundary-layer airmass, can reproduce a transition of trade cumulus organization from “sugar” to “flower” clouds with cold pools, observed on February 2–3. The simulations were driven with reanalysis large-scale meteorology and ATOMIC in-situ aerosol data. During the transition, large-scale upward motion deepens the cloud layer. The total water path and optical depth increase, especially in the moist regions where flowers aggregate. Mesoscale circulation leads to a net convergence of total water in the already moist and cloudy regions, strengthening the organization. Stronger large-scale upward motion reinforces the mesoscale circulation and accelerates the organization process by strengthening the cloud-layer mesoscale buoyant turbulence kinetic energy production.

## Plain Language Summary

Fair-weather shallow clouds have different sizes and cloud properties. A field study called ATOMIC and EUREC<sup>4</sup>A was designed to further understand the properties of these clouds. On February 2–3, very small and shallow “sugar” clouds grow into wider and deeper “flower” cloud clusters, no more than 3 km high. The clear spaces between the clouds expand. This study finds that local air circulation is responsible for making the moist and cloudy areas moister, and dry and cloud-free areas drier, enabling a process responsible for this transition. The large-scale vertical winds modulate the rate and strength of this process which occurs locally at smaller scales.

## 1 Introduction

Shallow clouds in a warm boundary layer continue to be a leading source of uncertainty in global climate models (i.e. Bony & Dufresne, 2005; Boucher et al., 2013; Zelinka et al., 2016). Previous studies have used high-resolution simulations and satellite imagery to understand the relationships between shallow cumulus properties and the large-scale atmospheric and oceanic conditions. For example, the Barbados Oceanographic and Meteorological Experiment (BOMEX) examined the turbulent dynamics of shallow cumuli using different large eddy simulation (LES) models (Siebesma et al., 2003). The Cloud

49 Feedback Model Intercomparison Project—Global Atmospheric System Study Intercom-  
50 parison of Large Eddy Models and Single Column Models (CGILS) investigated the mech-  
51 anisms of cloud feedback of shallow cumulus and stratocumulus under idealized climate  
52 change perturbations based on summertime subtropical atmospheric conditions in the  
53 Pacific Ocean (Zhang et al., 2013; Bretherton et al., 2013; Blossey et al., 2013). Bretherton  
54 and Blossey (2017) further explored a mechanism of shallow cumulus organization in dif-  
55 ferent large-scale conditions, including those from BOMEX and one of the CGILS cases.  
56 Other studies have used LES models to explore the relationship between turbulent flux  
57 and cloud amount (Narenpitak & Bretherton, 2019), and processes associated with arc-  
58 shaped organization of shallow cumuli known as cold pools (i.e. Zuidema et al., 2017,  
59 and references therein). In addition, Mieslinger et al. (2019) examined how different me-  
60 teorological conditions affect cloud properties across different oceanic basins using high  
61 resolution satellite imagery. The use of high resolution simulations and remote sensing  
62 tools over the years have enabled studies that lead to better understanding of shallow  
63 cumulus processes.

64 A field campaign designed to study shallow convection in the trade wind region oc-  
65 curred in January–February, 2020 in the Atlantic Ocean east of Barbados. The Atlantic  
66 Tradewind Ocean–Atmosphere Mesoscale Interaction Campaign (ATOMIC) and its Eu-  
67 ropean counterpart, the European field campaign called Elucidating the Role of Clouds-  
68 Circulation Coupling in Climate (EUREC<sup>4</sup>A), formed a field campaign that used instru-  
69 ments on research aircrafts and ships to observe the properties of shallow cumulus clouds  
70 in order to better understand their relationship with the large-scale environment (Quinn  
71 et al., 2020; Pincus et al., 2021; Stevens et al., 2021). Recent studies (i.e. Stevens et al.,  
72 2020; Rasp et al., 2020; Bony et al., 2020) have categorized the mesoscale organization  
73 of shallow cumuli based on the Moderate Resolution Imaging Spectroradiometer (MODIS)  
74 imagery into four types: sugar, gravel, fish, and flowers. Different states of organization  
75 have different cloud properties including boundary layer depth, amount of precipitation,  
76 cloud fraction, and cloud radiative effect.

77 On February 2–3, 2020, a transition from small and shallow clouds called “sugar”  
78 to larger and deeper clouds called “flowers” occurred over the field campaign region (Fig.  
79 1a; animation in Movie S1 in the the Supporting Information (SI)). Backward trajec-  
80 tories following the airmass at 500 m altitude show that these flower clouds originated  
81 from a shallow sugar cloud layer northeast of the NOAA Research Vessel Ronald H. Brown

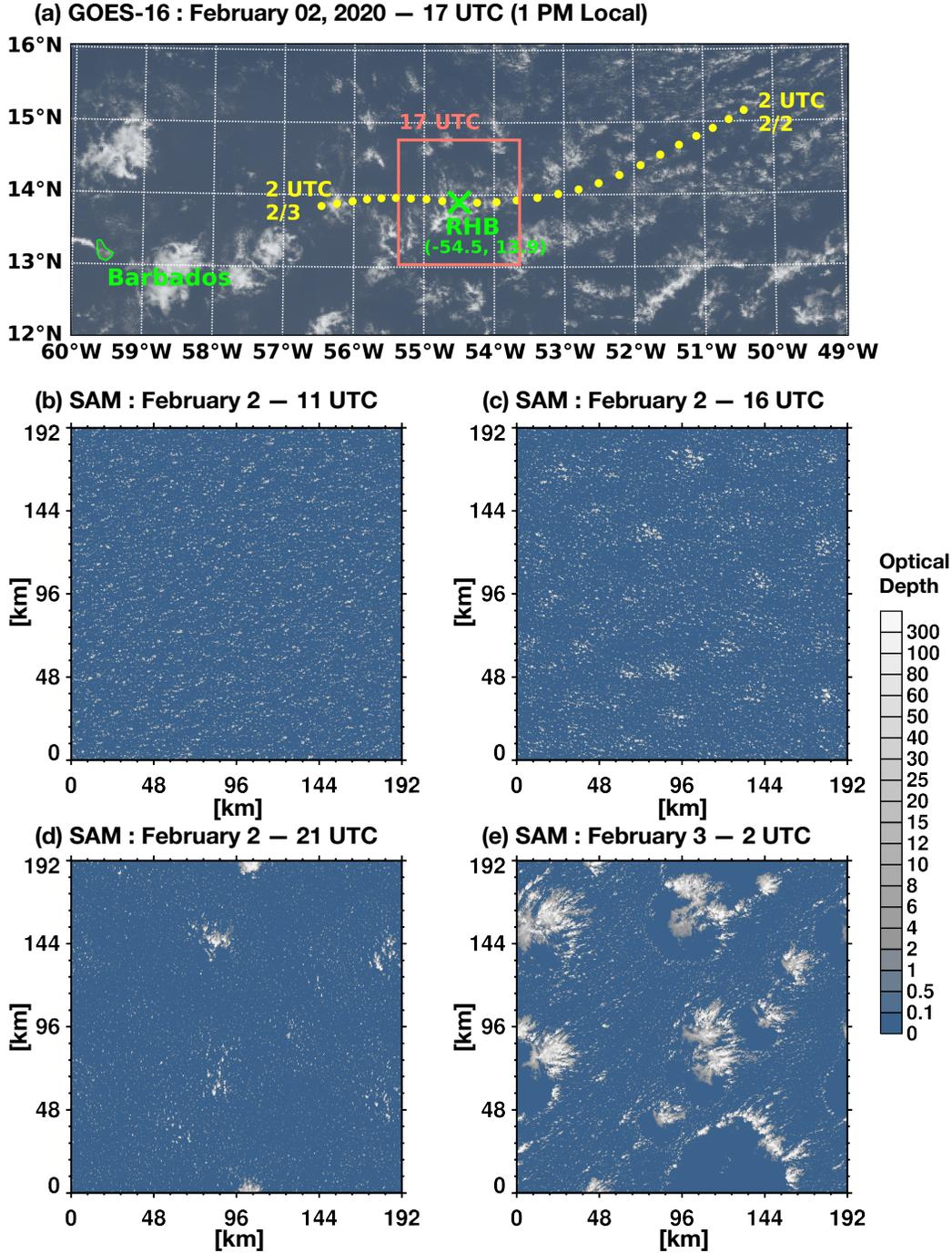
82 (RHB). Larger flowers with cold pools were observed to the southwest, closer to Barba-  
83 dos. This study uses a Lagrangian LES, with the domain following a boundary-layer tra-  
84 jectory (red box and yellow dots in Fig. 1a), to simulate this organization event. To un-  
85 derstand the relationship between the large-scale vertical velocity and the transition of  
86 the mesoscale organization, an additional LES with modified large-scale vertical veloc-  
87 ity is included.

88 The structure of this paper is as follows. Section 2 describes the simulation con-  
89 figurations and the observations used to initialize the simulations. Sections 3 shows the  
90 transition from sugar to flowers represented by the LES. Section 4 discusses the mech-  
91 anisms that are important for the organization. Section 5 identifies the role of large-scale  
92 vertical motion on the sugar-to-flowers transition and the circulation at the mesoscale.  
93 Finally, conclusions are given in Section 6. Figures S1-S11 and Movies S1-S3 are found  
94 in the SI.

## 95 **2 Data and Simulations**

96 The System for Atmospheric Modeling (SAM) (Khairoutdinov & Randall, 2003)  
97 is employed. The large-scale environment (soundings) and forcings of the simulations are  
98 derived from the European Centre for Medium-Range Weather Forecasts (ECMWF) Re-  
99 analysis 5<sup>th</sup> Generation (ERA5) (Hersbach et al., 2020), following the airmass at 500 m  
100 altitude through the location of the RHB (54.5°W and 13.9°N) at 17 UTC on Febru-  
101 ary 2. The airmass trajectory was calculated by the Hybrid Single-Particle Lagrangian  
102 Integrated Trajectory (HYSPLIT) model (Stein et al., 2015; Rolph et al., 2017) in the  
103 ERA5 reanalysis. Since the trajectory moves approximately with the boundary layer,  
104 large-scale horizontal advection of the temperature and humidity is not included. Instead,  
105 to account for horizontal advection in the free troposphere, the temperature and humid-  
106 ity profiles of the simulation are nudged to the ERA5 soundings above the inversion with  
107 a 30 min relaxation time scale.

108 The control simulation (CTL) is configured with 100 m horizontal grid spacing and  
109 a horizontal domain extent of 192×192 km<sup>2</sup>. The vertical grid spacing is 50 m, increas-  
110 ing geometrically from 5 km to the domain top at 8 km (total of 120 levels). Above that,  
111 the atmospheric profiles from ERA5 are used up to the top of atmosphere for the radi-  
112 ation calculation. The simulation uses a bulk two-moment (bin-emulating) microphysics



**Figure 1.** (a) A satellite image from the Geostationary Operational Environmental Satellite-16 (GOES-16) on February 2. The yellow dots represent hourly coordinates of the air mass following trajectory on which the Lagrangian simulations are based. The red box indicates the simulation’s  $192 \times 192$  km<sup>2</sup> domain extent, centered at the Ronald H. Brown research ship (green ‘x’) at 17 UTC. (b-d) Snapshots of total (cloud+rain) optical depth from the control simulation (CTL) shown at the designated times.

113 scheme (Feingold et al., 1998) and the Rapid Radiative Transfer Model for global climate  
114 model applications (RRTMG) radiation scheme (Mlawer et al., 1997) with time vary-  
115 ing atmospheric profiles above the domain top and the diurnal cycle of solar radiation.  
116 The radiation is computed every 10 seconds. The model’s time step is 2 seconds, and  
117 the duration of the simulation is 24 hours, from 2 UTC on February 2 to 2 UTC on Febru-  
118 ary 3, 2020.

119 An additional simulation called WeakW is performed using the same model con-  
120 figuration as CTL, except with a modified vertical velocity ( $W$ ) in the forcings. The  $W$   
121 profiles for WeakW are 50% weaker than CTL during a period with strong upward mo-  
122 tion, between 11 UTC and 19 UTC. Since SAM linearly extrapolates the hourly  $W$  forc-  
123 ing profiles to the model’s time step,  $W$  in WeakW starts to diverge from CTL at 10 UTC,  
124 and converges again at 20 UTC (Fig. S1a-b).

125 There are two types of aerosol in the simulations: sea salt and mineral dust (Fig.  
126 S1c-f). They are initialized based on the in-situ measurements and are advected within  
127 the domain. The sea-salt particles interact with the cloud microphysics scheme, but not  
128 with radiation. The mineral dust is included and only coupled with the radiation but  
129 not with the microphysics, as it remains in the free troposphere. This is consistent with  
130 the observation, that mineral dust was present above the cloud layer east of Barbados  
131 between January 31 and February 3. See Section 1 of the SI for details on the initial-  
132 ization of the aerosol.

### 133 **3 Transition of Shallow Cumuli: From Sugar to Flowers**

134 Simulation CTL is able to reproduce the transition from sugar to flowers on Febru-  
135 ary 2–3, 2020. Figure 1b-e and Movie S2 show the cloud state evolution. The sugar-to-  
136 flowers transition occurs between 8 UTC and 18 UTC. During this time, the sugar cloud  
137 field forms clusters, which then develop into contiguous aggregates and expand laterally  
138 to mature into flowers. Cumulus clouds interspersed between the flowers are suppressed  
139 compared to the initial sugar cloud field. After 20 UTC, the aggregated flowers produce  
140 precipitation, which partially evaporates before reaching the surface, resulting in cold  
141 downdrafts that produce cold pools adjacent to the flowers. A 3D snapshot of the cloud  
142 field at 2 UTC on February 3 (final time step) is shown in Figure S4.

### 143 3.1 Multiscale Partitioning

144 Although the simulations are run at 100 m grid spacing, it is helpful to coarse-grain  
 145 the outputs into larger tiles. This approach partitions the results into contributions from  
 146 the large-scale, mesoscale, and cumulus-scale processes. Coarse-graining filters out the  
 147 details at the smaller scales that may be associated with shallow convection but are not  
 148 relevant to the organization. This method was first introduced by Bretherton and Blossey  
 149 (2017). The tile size of  $16 \times 16 \text{ km}^2$  is chosen for this study as it represents the horizon-  
 150 tal variability of flower shallow cumuli in the simulations. See Section 2 of the SI for de-  
 151 tails on how the tile size is determined.

152 The partitioning of total water mixing ratio ( $q_t$ ) is given by:

$$153 \quad q_t = \overline{q_t} + q_t'' + q_t''' \quad . \quad (1)$$

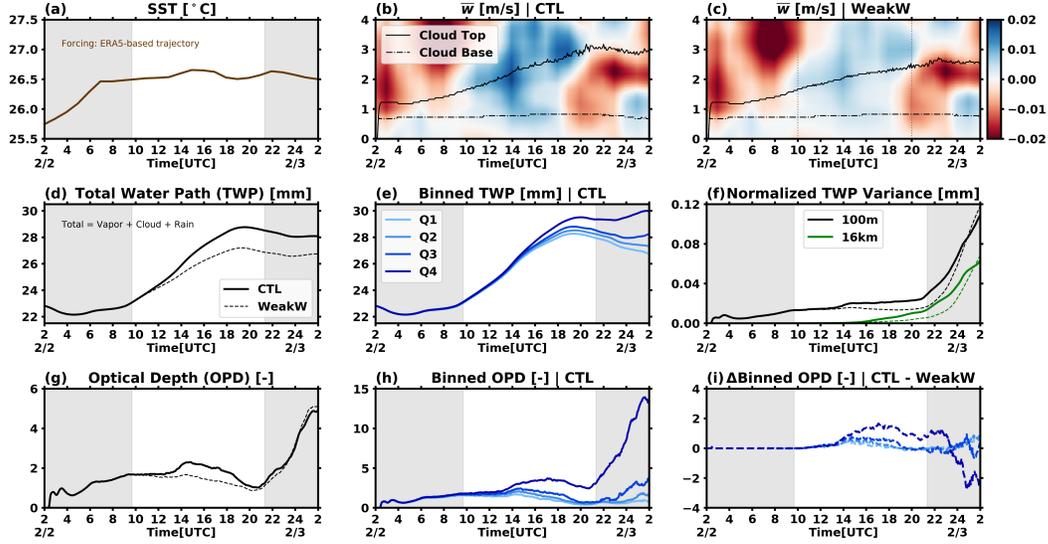
154 The overline is the domain-mean, the double prime is the perturbation coarse-grained  
 155 to  $16 \times 16 \text{ km}^2$  tiles, representing variability associated with the mesoscale ( $\geq 16 \text{ km}$ ). The  
 156 triple prime represents variability associated with cumulus-scale processes ( $< 16 \text{ km}$ ). The  
 157 partitioning is detailed in Section 3 of the SI.

158 The coarse-grained outputs are sorted by total water path (TWP, a sum of verti-  
 159 cally integrated water vapor, cloud, and rain) and binned into quartiles. Quartile 1 (Q1)  
 160 represents the driest and cloud-free areas while Quartile 4 (Q4) represents the moistest  
 161 and cloudiest areas of the simulation. The  $16 \times 16 \text{ km}^2$  tiles in each quartile are not nec-  
 162 essarily adjacent to one another.

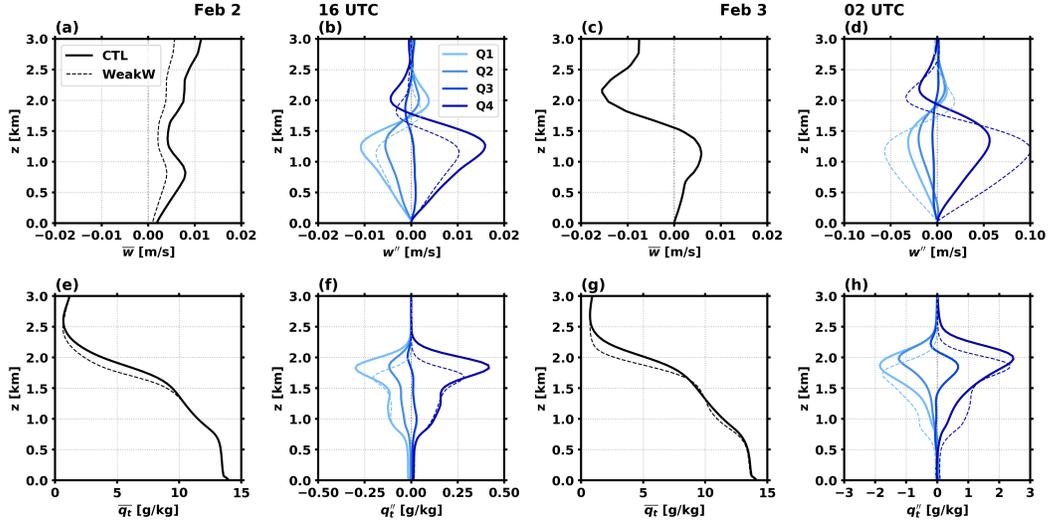
### 163 3.2 Shallow Convection Organization

164 Figure 2 shows the evolution of the simulations. The thick solid lines represent the  
 165 results from CTL. Sea surface temperature (SST) increases as the trajectory moves south-  
 166 westward, and remains constant as the trajectory moves westward. The deepening of the  
 167 cloud layer in CTL occurs after 6 UTC and becomes more obvious after 10 UTC, when  
 168 the domain-mean vertical velocity ( $\overline{w}$ ) shifts from negative to positive, helping the cloud  
 169 layer to deepen (Fig 2b). After 20 UTC, the cloud depth remains constant as the bound-  
 170 ary layer encounters large-scale subsidence.

171 The domain-mean TWP increases as the cloud layer deepens during the transition  
 172 process (Fig. 2d). As the organization strengthens, the TWP distribution becomes more



**Figure 2.** Time series of: (a) domain-mean sea surface temperature; (b) domain-mean vertical velocity, and cloud top and base heights, from CTL; (c) as in Panel (b) but for the weaker vertical velocity simulation (WeakW); (d) domain-mean total water path (TWP) of both CTL (solid) and WeakW (dash); (e) TWP sorted into quartiles from CTL; (f) variances of TWP computed at the full resolution (black) and the 16 km coarse-grained resolution (green), from both CTL (solid) and WeakW (dash); (g) domain-mean optical depth (OPD) from both simulations; (h) OPD, binned by TWP, from CTL; (i) the change in OPD, binned by TWP, between CTL and WeakW. Grey shading is applied between the daylight hours of 5:48 am and 17:23 pm (local time), when the top-of-atmosphere incoming shortwave radiation exceeds zero in SAM.



**Figure 3.** Vertical profiles of various variables at 16 UTC on February 2 (two left columns) and 2 UTC on February 3 (two right columns) of both CTL (solid) and WeakW (dash): (a,c) domain-mean vertical velocity ( $\bar{w}$ ); (b,d) mesoscale perturbations of vertical velocity binned by TWP quartiles ( $w''$ ); (e,g) domain-mean total water mixing ratio ( $\bar{q}_t$ ); and (f,h) mesoscale perturbations of total water ( $q''_t$ ), binned by TWP quartiles. For the binned profiles, only Q1 and Q4 from WeakW are shown.

173 asymmetrical; the moist areas become moister while the dry areas become drier (Fig.  
 174 2e). The variance of TWP normalized by the mean can be used as a proxy for the or-  
 175 ganization (Fig. 2f). The coarse-grained normalized TWP variance (green) increases sev-  
 176 eral hours after the full 100 m resolution normalized variance (black), evidence that the  
 177 moist patches are initially smaller than 16 km and later grow during the transition. The  
 178 domain-mean optical depth (OPD) also increases, except for a dip around 21 UTC (Fig.  
 179 2g), when the small isolated sugar clouds disappear while the larger cloud clusters have  
 180 yet to aggregate and grow (Fig. 1e). After 20 UTC, both the normalized TWP variance  
 181 and the OPD increase rapidly as the organization strengthens.

182 Figure 3 shows the vertical profiles at two different times, during and after the tran-  
 183 sition. At both times, regardless of  $\bar{w}$ , the binned mesoscale vertical velocity perturba-  
 184 tions ( $w''$ ) are positive in the cloud and subcloud layers and negative in the inversion layer  
 185 of the moistest quartile (Q4). In the drier quartiles (Q1-Q2), the signs of  $w''$  are oppo-  
 186 site. The moist quartiles also have positive mesoscale total water perturbations ( $q''_t$ ). Mass  
 187 continuity requires that in the moist and cloudy regions, where  $w''$  is positive (negative)

188 in the subcloud (inversion) layer, there is a local convergence (divergence) below (aloft),  
 189 consistent with the findings in Bretherton and Blossey (2017). The following section will  
 190 show that this local circulation is key for redistributing the total water, leading to mesoscale  
 191 organization.

#### 192 **4 The Mechanism of Transition**

193 This section analyzes the budget of mesoscale total water perturbations  $q_t''$  in the  
 194 four TWP quartiles to determine a mechanism responsible for the transition. Based on  
 195 Equation 12 of Bretherton and Blossey (2017) and the derivation in Section 3 of the SI,  
 196 the budget of  $q_t''$  at each level can be written as:

$$197 \quad \frac{\partial q_t''}{\partial t} = A + B + C + S_q'' \quad . \quad (2)$$

198 Each term on the right hand side of Equation (2) is described as follows: The first term  
 199 is the advection of mesoscale variability due to trajectory-relative large-scale wind ( $\bar{\mathbf{v}}$ )  
 200 and mesoscale perturbations of the wind velocity ( $\mathbf{v}''$ ):

$$201 \quad A = -(\bar{\mathbf{v}} + \mathbf{v}'') \cdot \nabla q_t'' \quad . \quad (3)$$

202 Let [ ] denote coarse-graining of the cumulus-scale field inside the brackets to a mesoscale  
 203 region of  $16 \times 16$  km<sup>2</sup>, and let  $\rho$  denote the reference density profile. The second term rep-  
 204 represents the vertical and horizontal gradients of the cumulus-scale  $q_t$  flux coarse-grained  
 205 to  $16 \times 16$  km<sup>2</sup>:

$$206 \quad B = B_v + B_h = -\frac{1}{\rho} \frac{\partial}{\partial z} [\rho w''' q_t'''] - \nabla_h \cdot [\mathbf{v}''' q_t'''] \quad . \quad (4)$$

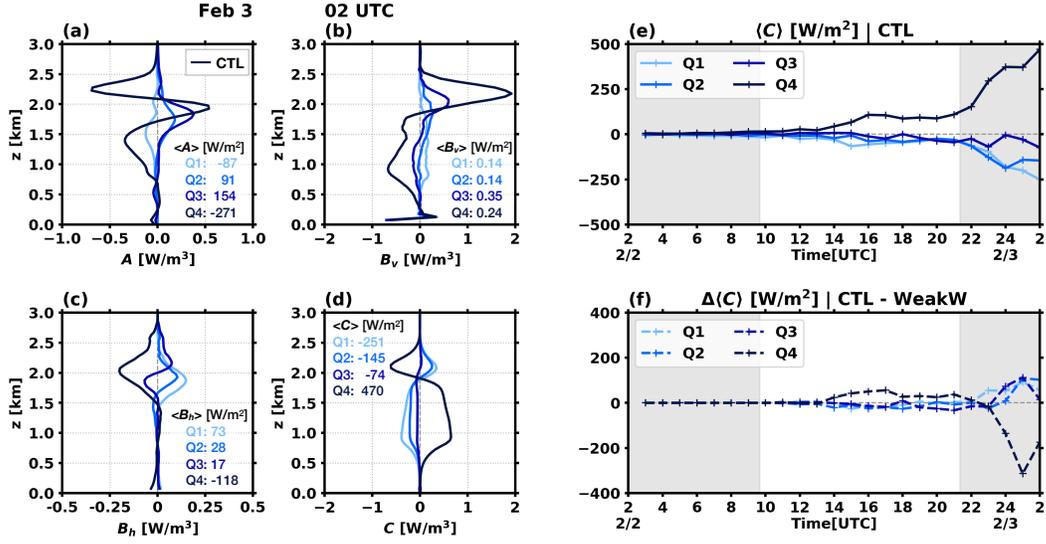
207 Eq. (4) was derived with the anelastic approximation used in SAM. The third term is  
 208 the mesoscale vertical advection of large-scale  $q_t$ :

$$209 \quad C = -w'' \frac{\partial \bar{q}_t}{\partial z} \quad . \quad (5)$$

210 Finally, the fourth term (Fig. 4d) is the source term of  $q_t''$  which represents the diver-  
 211 gence of precipitation mass flux ( $F_p$ ):

$$212 \quad S_q'' = \left( -\frac{1}{\rho} \frac{\partial F_p}{\partial z} \right)'' \quad . \quad (6)$$

213 Figure 4(a-d) shows vertical profiles of  $A$ ,  $B_v$ ,  $B_h$ , and  $C$  binned by TWP quar-  
 214 tiles at the end of CTL, and the vertically integrated values between 0 and 3 km (de-  
 215 noted by  $\langle \rangle$ ). (The  $S_q''$  profiles and their vertically integrated values are much smaller



**Figure 4.** Vertical profiles of: (a) large-scale and mesoscale advection of  $q_t''$  ( $A$ ); (b) vertical gradient of the cumulus-scale vertical  $q_t$  flux ( $B_v$ ); (c) horizontal gradient of the cumulus-scale horizontal  $q_t$  flux ( $B_h$ ); and (d) mesoscale vertical advection of the large-scale  $q_t$  ( $C$ ), at 2 UTC on February 3 from CTL, all coarse-grained to  $16 \times 16$  km<sup>2</sup> and binned by TWP. The vertically integrated values between 0 and 3 km are also shown, denoted by  $\langle \rangle$ . (g) Hourly time series of  $\langle C \rangle$  binned by TWP quartiles from CTL. (h) The change in  $\langle C \rangle$  time series between CTL and WeakW.

216 and hence negligible, as shown in Figure S6.) A positive quantity means the respective  
 217 term is responsible for moistening the region.

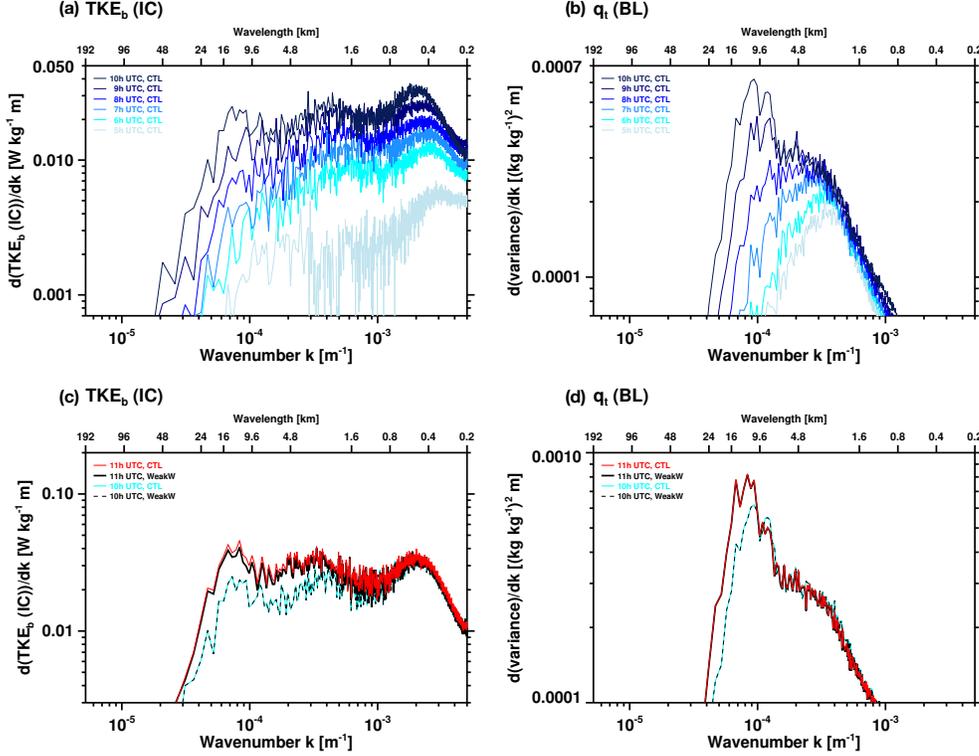
218 In Q4,  $A$  is small and tends to dry out the boundary layer. Although  $B_v$  is large,  
 219  $\langle B_v \rangle$  is negligible in all quartiles. This is expected because the vertical cumulus-scale flux  
 220 transfers total water vertically from the cloud layer to the inversion layer but not hor-  
 221 zontally. When coarse-grained within  $16 \times 16 \text{ km}^2$  regions,  $B_h$  is small, but  $\langle B_h \rangle$  is non-  
 222 negligible and results in drying in Q4, albeit secondary to  $\langle A \rangle$ . The magnitude of  $C$  is  
 223 larger than that of  $A$  and  $B_h$ , and  $\langle C \rangle$  is the only term that moistens the cloud layer in  
 224 Q4, in which flower clouds aggregate. Because  $\frac{\partial \bar{q}_t}{\partial z}$  is always negative (Fig. 3e,g), the sign  
 225 of  $C$  always follows the sign of  $w''$ . Due to mass continuity, a positive  $C$  in the cloud layer  
 226 of Q4 is associated with a horizontal total water convergence below the cloud plumes,  
 227 and divergence in the inversion. A positive  $\langle C \rangle$  indicates a net total water convergence  
 228 in the lower troposphere of the moistest quartile.

229 To demonstrate that  $\langle C \rangle$  drives moistening in Q4 and drying in Q1 through Q3,  
 230 Figure 4(e) shows the hourly time series of  $\langle C \rangle$  binned by TWP quartiles from CTL. This  
 231 provides the evidence that the net convergence and divergence of total water due to mesoscale  
 232 circulation renders the moist and cloudy patches moister, and the dry and cloud-free patches  
 233 drier.

## 234 5 The Role of Large-Scale Vertical Motion

235 To examine the role of large-scale vertical velocity for the sugar-to-flower transi-  
 236 tion, an additional simulation is performed and analyzed. Simulation WeakW has a 50 %  
 237 weaker  $\bar{w}$  during the period of strong upward motion, i.e., 10 UTC and 20 UTC (Fig.  
 238 2c). It produces a shallower cloud layer and lower TWP than CTL. Movie S3 shows the  
 239 cloud field evolution in WeakW. Simulation CTL exhibits a more rapid transition from  
 240 the sugar to the flower cloud state (Fig. S7). It has greater normalized TWP variance  
 241 and optical depth, especially in Q4 where flowers aggregate (Fig. 2f,i).

242 Although mesoscale organization forms more rapidly in CTL compared to WeakW,  
 243 the same mechanisms take place in both simulations; moist areas become moister and  
 244 dry areas become drier. Figure 3b,f shows that with stronger upward motion, the  $w''$  and  
 245  $q_t''$  profiles of CTL during the transition period have the same structure as those in WeakW,  
 246 except with larger magnitudes. After 23 UTC, when the organization in WeakW catches



**Figure 5.** Spectra of (a) buoyant turbulence kinetic energy production in the cloud layer ( $TKE_b(IC)$ ), expressed in units of  $W \text{ kg}^{-1}$  of boundary-layer mass, and of (b) total water mixing ratio in the boundary layer ( $q_t(BL)$ ), from CTL, plotted hourly from 5 UTC to 10 UTC on February 2. Spectra of (c)  $TKE_b(IC)$  and (d)  $q_t(BL)$ , from CTL and WeakW at 10 UTC and 11 UTC on February 2.

247 up with CTL,  $w''$  becomes stronger in WeakW compared to CTL (Fig. 3h). This is con-  
 248 sistent with the change in  $\langle C \rangle$ , which is greater in Q4 of CTL compared to WeakW be-  
 249 tween 10 UTC and 23 UTC (Fig. 4f), and smaller thereafter. In other words, the stronger  
 250 upward motion assists the aggregation of total water on the mesoscale, accelerating or-  
 251 ganization.

252 Figure 5 shows spectra of buoyant turbulence kinetic energy (TKE) production in  
 253 the cloud layer ( $TKE_b(IC)$ ) and of boundary-layer total water ( $q_t(BL)$ ). Circulation on  
 254 the mesoscale and aggregation of moisture emerge in the form of peaks between 9.6 and  
 255 16 km that are clearly discernible by 10 UTC (Fig. 5a,b). Up to 10 UTC, CTL and WeakW  
 256 have the same  $\bar{w}$ , hence their TKE production spectra are identical, but at 11 UTC, CTL  
 257 exhibits a strengthening of cloud-level TKE production on the mesoscale (Fig. 5c). In

258 the sub-cloud layer, TKE production exhibits a much weaker response (Fig. S9b). This  
259 disproportionate strengthening of cloud-level mesoscale TKE production relative to other  
260 scales, due to the more positive  $\bar{w}$  in CTL compared to WeakW, increases and persists  
261 over the period during which  $\bar{w}$  differs between the simulations (Fig. S10). No discernible  
262 difference exists in the spectra of total water at 11 UTC (Fig. 5d), and only after sev-  
263 eral hours does a stronger mesoscale peak emerge in CTL compared to WeakW (Fig. S11).  
264 It is hence the strengthening of cloud-level mesoscale TKE production that strengthens  
265 aggregation of moisture on the mesoscale and accelerates the sugar-to-flower transition  
266 in response to a more positive  $\bar{w}$ .

## 267 6 Conclusions

268 The ATOMIC and EUREC<sup>4</sup>A field campaign took place in the Atlantic Ocean east  
269 of Barbados in January–February 2020, with a goal to better understand the relation-  
270 ship between shallow cumuli and large-scale meteorological and oceanic conditions. On  
271 February 2–3, a transition of trade cumulus organization from sugar to flowers was ob-  
272 served. This study shows that a Lagrangian LES following a boundary-layer airmass tra-  
273 jectory can reproduce the transition. During the sugar-to-flowers transition, the clouds  
274 become organized, and the cloud layer deepens and moistens.

275 Although the large-scale vertical wind helps deepen the cloud layer, the mesoscale  
276 wind drives the sugar-to-flowers transition. The mesoscale circulation, driven by a lo-  
277 cal ascending (descending) air inside (above) the shallow cumulus plumes, leads to a net  
278 moisture convergence in the moist patches, in which the clouds aggregate. This renders  
279 the moist patches moister and dry patches drier.

280 It is shown that large-scale vertical velocity regulates the sugar-to-flower transi-  
281 tion by modulating cloud-layer buoyant TKE production at the mesoscale, and the mesoscale  
282 circulation by which moisture aggregates. In the considered case, stronger large-scale up-  
283 ward motion accelerates the sugar-to-flower transition by strengthening cloud-layer mesoscale  
284 TKE production. Given the broad interest in the vertical structure of subsidence engen-  
285 dered by ATOMIC and EUREC<sup>4</sup>A, a follow-on study examining how the structure of  
286 the large-scale vertical velocity impacts the mesoscale organization is warranted.

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