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2 **Mass-flux characteristics of tropical cumulus clouds from wind profiler observations at**
3 **Darwin, Australia**

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33 **moisture, cloud models**

34

35 **Abstract**

36 Cumulus parameterizations in weather and climate models frequently apply mass-flux
37 schemes in their description of tropical convection. Mass-flux constitutes the product of the
38 fractional area covered by convection in a model grid box and the vertical velocity in the
39 convective clouds. Vertical velocity is difficult to observe making the evaluation of mass-
40 flux schemes difficult. Here, we combine high temporal resolution observations of the in-
41 cloud vertical velocity over two wet-seasons at Darwin derived from a pair of wind profilers
42 with the physical properties (cloud top heights CTH, convective-stratiform classification) of
43 clouds derived from a C-band polarimetric radar to provide estimates of mass-flux and its
44 components. The length of our data set also allows for investigations of the contributions to
45 the overall mass-flux by different convective cloud types and of mass-flux variations with
46 changes in the large-scale conditions. We found cumulus mass-flux was dominated by
47 updrafts and in particular the updraft area fraction, with updraft vertical velocity playing a
48 secondary role. The updraft vertical velocity peaked above 10 km where both the updraft area
49 fraction and air density was minimal, resulting into a marginal effect on mass flux values.
50 Downdraft area fractions are much smaller and velocities much weaker than those in
51 updrafts. Area fraction responds very strongly to changes in mid-level large-scale vertical
52 motion and changes in convective inhibition (CIN). In contrast, changes in the lower-
53 troposphere relative humidity and convective available potential energy (CAPE) strongly
54 modulate in-cloud vertical velocities but have moderate impacts on area fractions. Although
55 average mass-flux is found to increase with increasing cloud top height it is environmental
56 conditions that dictate the magnitude of mass-flux produced by deep convection through a
57 combination of effects on area fraction and velocity.

58

59 1. Introduction

60 Cumulus clouds play an important role in weather and climate by maintaining the large-scale
61 atmospheric circulation (e.g., Fritsch 1975; Emanuel et al. 1994), transporting heat, moisture,
62 and momentum in the atmosphere (Yanai et al. 1973) and producing a multitude of clouds
63 (e.g., Liu and Zipser, 2005). Recent studies indicate the existence of distinct types of cumulus
64 clouds in the tropics (e.g. Johnson et al. 1999; Kumar et al. 2013a). These are shallow
65 cumulus with cloud-top heights (CTH) near the trade inversion layer 1–3 km above the
66 surface, mid-level cumulus congestus clouds with CTH near the 0°C freezing level (FZL),
67 deep cumulonimbus clouds with CTH between FZL and tropopause layer, and overshooting
68 convection with tops extending into the tropopause layer.

69 Individual cumulus clouds, particularly deep and overshooting modes, are generally thought
70 to contain convective-scale (1–10 km) updraft and downdraft cores. Observations reveal that
71 cumulus updraft and downdraft flow characteristics differ in several ways (e.g., Knupp and
72 Cotton 1985). Updrafts are triggered by convergence of environmental airflow and typically
73 start near the cloud base. They dominate in the growing and mature phases of cumulus clouds
74 (Paluch and Knight 1984). Entrainment processes and water loading reduce updraft strength,
75 while latent heating (e.g., Zipser 2003) and precipitation (e.g. Fierro et al. 2009; Heymsfield
76 et al. 2010) enhance the updraft strength. In contrast, downdrafts commonly occur in the
77 mature and decaying phases of cumulus clouds. Mature phase downdrafts are generally
78 forced by cloud edge evaporation cooling, which typically occur in the middle level (5–10
79 km), and entrainment processes near cloud tops. Decay phase downdrafts are forced by
80 precipitation loading, evaporation and melting, occurring below the FZL (May and
81 Rajopadhyaya 1999).

82 In General Circulation Models (GCMs) convection cannot be represented by modelling
83 individual convective clouds. Instead, simple representations of the collective effects of a
84 cumulus cloud ensemble existing within a model grid-box are applied. Amongst the most
85 widespread of these cumulus parameterization approaches is the so-called mass-flux
86 approach (see Arakawa (2004) for an overview). Here, the vertical transport by the cloud
87 ensemble is directly related to the mass-flux through the clouds, itself a product of the air
88 density, fractional area covered by and the vertical velocity within cumulus updrafts and
89 downdrafts. While conceptually simple, the evaluation of mass-flux approaches from
90 observations has proven difficult, as measurements of the area fraction and vertical velocities
91 within up- and downdrafts on the scale of a GCM grid-box are difficult to ascertain. As a
92 result, much of the evaluation of mass-flux schemes has relied on the use of Cloud Resolving
93 Models (e.g., Randall et al. 2003; Derbyshire et al. 2004; Petch et al. 2014).

94 The main motivation of this study is to begin closing this obvious observational gap and to
95 demonstrate the potential of using existing observational data set for evaluating model mass-
96 flux schemes. In particular, we wish to address the following two questions: 1) What is the
97 observed vertical structure of convective mass-flux and which of its components (area or
98 velocity) dominates the overall structure? 2) How sensitive is mass-flux to changes in the
99 environmental conditions?

100 There are previous observational studies that determined direct in-cloud mass-fluxes.
101 Numerous in situ aircraft penetrations conventionally provide the best insights in convective
102 cloud dynamics (e.g. Byers and Braham 1949; Marwitz 1973; LeMone and Zipser 1980;
103 Jorgensen and LeMone 1989; Anderson et al. 2005). However, to facilitate evaluation of
104 mass-flux schemes in GCM, longer temporal length of continuous convective profiling are
105 needed, such as those from advanced remote sensing techniques. Examples of long-term in-

106 cloud mass-flux observations include the works of May and Rajopadhyaya (1999) and
107 Giangrande et al. (2013), where they used wind profiler retrievals from a tropical and
108 subtropical site, respectively. Both studies found the peaks in updraft speeds and updraft core
109 widths associated with deep convection occurred in upper levels, near 10 km altitude. In
110 contrast, downdrafts peaked near the cloud base. In the tropics, updraft cores have smaller
111 speeds, but are wider compared to the subtropics. Heymsfield et al. (2010), who investigated
112 deep convection in both tropics and subtropics using airborne Doppler radars, also reported
113 similar characteristics in vertical velocities for updrafts and downdrafts.

114 To extract mass-flux over a GCM size grid, we need direct measurements of vertical velocity
115 inside every cumulus clouds enclosed by the model grid box. Most commonly, this is
116 achieved using a dual Doppler radar retrieval technique (e.g. Collis et al. 2013). However, the
117 dual Doppler approach requires at least two radars, with the accuracy of retrieved vertical
118 velocity depending on the location within the radar domain. An alternative and more direct
119 approach to determine vertical velocity is to use wind-profiler (May and Rajopadhyaya 1999;
120 Williams, 2012). The current study will be using the latter approach using data collected in
121 Darwin, Australia, for the two wet seasons (Nov-Apr) of 2005/2006 and 2006/2007. The
122 main difficulty in using wind-profiler observations is that they represent a single atmospheric
123 column and temporal aggregation is required to represent larger spatial areas. By comparing
124 the wind-profiler cloud occurrence with volumetric radar data, we demonstrate that the
125 statistical aggregation of the single column profiler measurements over a longer period do
126 depict convection comparable to that which will be observed in a GCM size grid box. We
127 then proceed to determine both the fractional area and in-cloud velocities in convective up-
128 and downdrafts using the profiler information and aggregate them into GCM-equivalent
129 mass-flux profiles.

130 Having determined profile of mass-flux from observations over a GCM size box, we evaluate
131 the sensitivity of the vertical structure and strength of the mass-flux to environmental
132 conditions (lower-troposphere (0–5 km) moisture, CAPE and CIN) and the large-scale
133 vertical motion. The Darwin wet season experiences a wide variety of convective systems
134 due to the presence of two distinct convective regimes - active monsoon/oceanic conditions
135 and build-up/break continental conditions (e.g., McBride and Frank 1999; Pope et al. 2009;
136 Kumar et al. 2013b). This makes Darwin a good location to investigate the sensitivity of
137 mass-flux to varying environmental conditions.

138 Past studies have attempted similar sensitivity tests of mass-flux profiles (or the input
139 parameters used to compute mass-flux) to the synoptic regimes and environmental conditions
140 using both observations and simulations. Cifelli and Rutledge (1994; 1998) using wind
141 profiler observed vertical velocity statistics found significant differences in the mean vertical
142 motion between Darwin break and monsoon storms, with evidence of a bimodal peak in the
143 vertical velocity profile for break cases, while the monsoon cases had a more uniform profile.
144 Here we will extend this study to more details in the environmental conditions. In particular,
145 we will investigate the sensitivity of observed mass-flux to mid- and upper-tropospheric
146 humidity and qualitatively compare the results to those of the idealized CRM simulations in
147 Derbyshire et al. (2004). These simulations implied that in a dry environment, the mass-flux
148 decreases monotonically with height above the cloud base leading to the formation of mostly
149 shallow convection. Moist environments on the other hand led to deep convection with the
150 peak mass-flux located at an elevated height in the mid-troposphere.

151 The paper is structured as follows. Section 2 will introduce the data sets used in the study.
152 Section 3 describes the method to retrieve velocity and area profiles from wind profiler
153 observations and establishes that these point observations when averaged in time provide a

154 good proxy for mass-flux in a GCM-size grid-box. Section 4 presents the main results of the
155 study, including the mean mass-flux profile and its variability, its sensitivity to environmental
156 conditions, and the contributions from different cloud types to the overall mass-flux. This is
157 followed by a summary and discussion in Section 5.

158 **2. Data**

159 The main goal of this study is to provide observational estimates of convective mass-flux and
160 its components at a scale relevant to the parameterisation of convection in GCMs as well as
161 its sensitivity to environmental conditions. This requires the use of a variety of data sets.
162 Specifically, we make use of a pair of wind-profilers embedded in the field-of-view of a
163 scanning C-band dual-polarization radar and combine those with detailed estimates of the
164 large-scale conditions provided by a variational analysis algorithm. Each of these data
165 sources is explained in turn below.

166 **2.1 The Darwin wind profiler radar pair**

167 We use data collected by a 50- and 920-MHz wind profiler pair from two wet seasons
168 (October 2005 – April 2006 and October 2006 – April 2007), recorded at 1-min resolution.
169 The main advantage of the use of this data source is that wind profilers provide more accurate
170 estimates of in-cloud vertical velocity than other remote-sensing techniques, including Dual-
171 Doppler radar techniques (e.g., Collis et al. 2013). The disadvantage is that the measurements
172 are taken at a single point, but frequently in time, and a time-space conversion is required to
173 make them useful to study the mass-flux behaviour on scales of a GCM grid-box.

174 Here, vertical velocities are computed by applying the dual-frequency algorithm developed in
175 Williams (2012) to the Doppler returns from the vertical beams of 50- and 920-MHz wind
176 profiler pair. The beam width of the vertical beam is approximately 0.2 km at 1 km height

177 and increases to 2 km by 10 km height. The wind profiler pair was synchronized to begin
178 their vertical beam observations every 1 min. The full description of the Darwin wind profiler
179 setting can be found in Williams (2012).

180 The 50-MHz profiler simultaneously observes both Bragg scatter from ambient air, which
181 provides a direct measurement of the vertical velocity of air parcels (wanted signal), and
182 Rayleigh scatter from hydrometeors (unwanted signal). If signals from the two scattering
183 processing are not properly separated, then the vertical air motion estimates will be biased
184 downwards because of contamination from falling hydrometeors. The Williams (2012)
185 method uses the spectra from the 920-MHz profiler, which are sensitive to mainly
186 hydrometeor returns, to filter out the Rayleigh echo returns from the 50-MHz profiler spectra.
187 The filtered 50-MHz signal is then processed using the standard wind profiling processing
188 technique described in Carter et al. (1995) and is based on the profiler online processing
189 (POP) routine. The POP routine estimates the spectrum noise level, the spectrum signal start
190 and end integration points, and the first three moments—power, mean reflectivity-weighted
191 Doppler velocity, and the spectrum width (equal to twice the spectrum standard deviation).
192 The mean Doppler velocity corresponds to the vertical air motion. The accuracy of the
193 vertical velocity retrieval by the Darwin wind profiler pair is estimated to be between 0.05 to
194 0.25 m s^{-1} using a Monte Carlo simulation design (Williams 2012). Further comparisons
195 between the Darwin wind profiler and statistical techniques for the separation of terminal fall
196 velocity and vertical air velocity also yielded an agreement to within $0.1\text{-}0.15 \text{ ms}^{-1}$ (Protat
197 and Williams 2011).

198 The profiler vertical velocity measurements are interpolated onto a vertical grid of 100 m
199 resolution over an altitude range of 1.7 – 17 km. However, the highest quality data is thought

200 to be limited to heights below 11 km (May and Rajopadhyaya 1999), because of the reduction
201 in profiler sensitivity with height.

202 Finally, the vertical velocity data from the wind profiler was further filtered to keep only
203 measurements that were from within cumulus clouds (see Section 3 for more detail). To
204 achieve this, we need to know; i) if a cumulus cloud occurred at the profiler site, and ii) what
205 the estimated cloud top height of the cumulus cloud at the time of the profiler measurement
206 might be. This information on the physical properties of the clouds is extracted from the
207 collocated C-band scanning radar, which contains the wind profilers within its field-of-view,
208 roughly 24 km southwest from the radar location (see Fig. 1 of May et al. 2002). The CPOL
209 measurements are introduced in more detail in the following subsection. Vertical velocities
210 outside cumulus clouds are not considered here.

211 **2.2 Darwin CPOL radar**

212 We use measurements of reflectivity from the C-band polarimetric radar (CPOL; Keenan et
213 al. (1998)), which have been sampled onto a cubic grid with a horizontal grid size of 2.5 km x
214 2.5 km, and vertical resolution of 0.5 km. The horizontal scanning area of CPOL is
215 approximately 70,000 km², sufficient enough to contain few GCM size grid boxes.

216 The CPOL measurements are used to identify different cloud types that are present over the
217 wind profiler. Here we make use of earlier studies (e.g., Kumar et al. 2013a-b; 2014) that
218 demonstrated the utility of the CPOL measurements to identify convective cloud cells and
219 determine their cloud top height. We make use specifically of two physical characteristics
220 derived from the CPOL data in this study:

- 221 i) We apply a convective vs stratiform classification and use only those clouds identified
222 as convective over the wind profiler to extract their mass-flux characteristics.

223 Specifically we apply the widely used Steiner algorithm to identify convective clouds
224 (Steiner et al. 1995) at the CPOL pixel collocated with the wind profiler location. This
225 algorithm has been successfully employed in several previous studies to differentiate
226 between the convective and stratiform cloud types (e.g. Kumar et al. 2013a-b; Penide et
227 al. 2013a). As the CPOL radar takes 10 minutes to complete a full volume scan, all 1-
228 min scans of the wind profiler falling into a 10-minute interval of convective cloud
229 occurrence over the profilers are used as valid measurements of vertical velocity.

230 ii) As our focus is on convective mass-flux, we would like to filter out any vertical
231 velocity measurements taken in cirrus anvils and/or in clear air above active convective
232 drafts. To do so we make use of the 0-dBZ echo top height extracted from the CPOL
233 reflectivity profile over the profiler site. Previous studies have shown that the 0-dBZ
234 echo tops from C-band radar observations are usually within 1 km of cloud top heights
235 estimated by millimetre cloud radars such as that on CloudSat (Casey et al. 2012) or on
236 the ground at Darwin (Kumar et al. 2013a). To ensure that we study continuous up- or
237 downdrafts we require that there is vertically continuous reflectivity signal between the
238 lowest CPOL level of 2.5 km height and the 0-dBZ echo top. We also apply the echo
239 top height to classify the observed cumulus clouds as either congestus, deep and
240 overshooting (Kumar et al. 2013a; 2014), allowing us to investigate the contribution to
241 total mass-flux from the various cumulus modes.

242 2.3 Background environmental conditions

243 Apart from providing overall mass-flux estimates we also aim to examine the effects of the
244 environmental conditions on the mass-flux behaviour. To do so requires reliable
245 observational estimates of key environmental parameters. Here we use 6-hourly information
246 on lower-troposphere (0–5) km relative humidity (RH_{0-5}), CAPE (Convective Available

247 Potential Energy), CIN (Convective inhibition), and the large-scale vertical motion at 500
248 hPa (ω_{500}). We use two main sources to derive these parameters.

249 The RH_{0-5} is extracted from the Darwin airport operational radiosoundings. We simply
250 average the relative humidity measurements between 0 and 5 km. The remaining three
251 parameters, CAPE, CIN and ω_{500} , were from a large-scale data set derived for the Darwin
252 region by Davies et al. (2013) by applying the variational budget analysis technique of Zhang
253 and Lin (1997) using NWP analysis data as “pseudo-radiosondes” and radar and satellite
254 observations at the surface and top of the atmosphere, as suggested by Xie et al. (2004). By
255 comparing their approach to results from the Tropical Warm Pool International Cloud
256 Experiment field study (May et al. 2008), Davies et al. (2013) showed that this technique
257 provides much better estimates of the large-scale state of the atmosphere than the direct use
258 of analyses or reanalyses from Numerical Weather Prediction Centres. The median over the
259 two wet seasons for CAPE, CIN, ω_{500} , respectively, were 548 J kg^{-1} , 43 J kg^{-1} and -0.38 hPa
260 Hour^{-1} . Note that a negative value for vertical motion represents upward motions.

261 **3.0 Method**

262 The main motivation of this study is to provide a statistical picture of mass-flux profiles using
263 observations, which will then be useful to evaluate existing cumulus mass-flux scheme in
264 models and assess the respective contributions of convective area fraction and vertical
265 velocity to updraft mass flux. Ideally, this would require high resolution observations of
266 vertical velocity both in time and space over a volume of $100 \text{ km} \times 100 \text{ km}$ in the horizontal
267 (typical GCM grid box) and 20 km in the vertical. No such measurements exist. As outlined
268 in the introduction, in this study we will make use of vertical velocity retrievals as derived
269 from dual-frequency wind profiler observations. However, we will combine the wind profiler

270 information with that from the scanning CPOL radar to investigate the representativeness of
271 the single site measurements for convection over a GCM size grid.

272 While our overall goal is to provide a statistical study of several hundred cumulus cells
273 occurring over time in a GCM box we first illustrate our methodology to derive vertical
274 motion and area fraction profiles using a snapshot of a deep convective case observed
275 concurrently by both radar types shown in Figure 1. Figure 1a shows the time-height cross-
276 section of reflectivity from the CPOL radar at the profiler site, which is available in 10 min
277 time intervals and 0.5 km resolution in height. The remaining panels of Fig. 1 show the wind
278 profiler measurements. The profiler observations are available at a much finer resolution of 1
279 min in time and 0.1 km in height. The red circles in Fig. 1a depict the 0-dBz ETH locations at
280 those times where the Steiner classification finds a convective cloud over the profiler site
281 (also indicated by the black line).

282 The differences between the CPOL reflectivities (Fig. 1a) and the 50 MHz (Fig. 1b) and 920
283 MHz (Fig. 1c) wind profiler reflectivities are found to be quite large, with the CPOL
284 reflectivities in better agreement with the 920 MHz wind profiler reflectivities than with the
285 50 MHz wind profiler reflectivities. This is not surprising, as the 50 MHz wind profiler
286 reflectivities are a mixture of echoes from clear air and hydrometeors, while CPOL is only
287 sensitive to hydrometeors. The differences between CPOL and the 920 MHz reflectivities
288 likely reflect the high temporal evolution of the convective event within the sampling
289 resolution of CPOL (10 minutes), which is captured by the 920 MHz observations at 1-
290 minute resolution.

291 Examinations of CPOL radar loops for the event described in Fig. 1 revealed that the
292 overshooting convective system sampled in Figure 1 was embedded in widespread stratiform
293 clouds and the whole system was moving across the profiler from the southwest. The time-

294 height sections of vertical velocity (Fig. 1d) indicate that the storm was present over the
295 profiler location for approximately one hour. The regions with vertical motion exceeding 1.5
296 m s^{-1} (strong updrafts) and below -1.5 m s^{-1} (strong downdrafts) are shown by the black
297 contours. The upward motions first occur at the low-levels around 0440 LT, coinciding with
298 the arrival of the storm. Within 30 min the region of strong upward motion shifted rapidly
299 from low-levels to above the freezing level (approximately 5 km). From between 0520–0540
300 LT, the updrafts remain constantly strong between 5 and 15 km. After 0550 LT, there is a
301 secondary increase in upward motions at around 7 km. By this time, the main convective cell
302 had passed over the profiler sites and the profiler is now sampling the stratiform anvils of the
303 storm as indicated by the absence of convective clouds in Steiner classification applied to
304 CPOL (Fig 1a).

305 While present in Figure 1 it is evident that downdrafts occur much less frequently and with
306 much weaker magnitudes than updrafts. This is well known and has been illustrated in other
307 studies using radar profiler measurements (e.g. see May et al. 2002; Heymsfield et al. 2010;
308 Giangrande et al. 2013). The observed regions of downdrafts, although short-lived (so
309 smaller spatial width), are consistent with the different downdraft types known to exist (e.g.,
310 Knupp and Cotton 1985). Downdrafts forming at low levels, which are more frequent than
311 downdrafts in higher levels, are likely associated with precipitation loading, evaporation and
312 melting and can be seen throughout the active storm phase. Several downdrafts can be found
313 at mid levels, such as the observed strongest downdraft around 0540 LT between 7 and 10
314 km. They are thought to be forced by cloud edge evaporation cooling. Note that we
315 frequently observe downdrafts in the stratiform regions of the cloud system sampled in our
316 data set. As we focus on convective situations only, these will not form part of our analysis.
317 Downdrafts are also observed between 10 and 15 km near the cloud tops. These are thought
318 to be penetrative downdrafts, which are short-lived (i.e., of small size) and are likely the

319 result of the entrainment of sub-saturated air into the cloud. It is clear from the case study
320 illustrated in Figure 1 that vertical motions vary significantly over the storm lifetime, height
321 and also between convective and stratiform structures. We do not attempt to study the
322 evolution of vertical velocities as function of storm lifetime because the profiler may be
323 sampling only a section of individual storms.

324 To be of use for model evaluation, the derived mass-flux profiles must be representative for
325 an area the size of a GCM grid-box. To account for all cumulus clouds over the model size
326 grid requires computation of convective area fraction. The area fraction is typically defined as
327 the ratio of the size of all convective cells in the domain over the total domain size. Scanning
328 radars, such as CPOL, are the most suitable to calculate area fraction using this spatially
329 sampling approach. Since we wish to compute mass-flux from a vertically pointing wind
330 profiler, which takes measurements over a column with a small cross-section area, the area
331 fraction cannot be directly estimated using these measurements. . Instead convective area
332 fraction is determined as the ratio of the time CPOL identifies convection above the profiler
333 over the total sampling time. We use a long total sampling time of two wet season with the
334 rationale that the convection, at a point, derived from this long time series is a good sample of
335 that occurring in the entire domain over the same sampling time.

336 To evaluate this approach, the area fractions were derived as described above using both the
337 scanning CPOL and vertically pointing wind profiler, respectively (Fig. 2). Recall that only
338 convective cloud columns from CPOL are used to calculate both the spatial statistics from
339 CPOL and the temporal statistics at the profiler site. The convective area fraction from CPOL
340 was calculated for various circular regions of radius ranging from 10 to 100 km centered on
341 the wind profiler site. The CPOL area fractions for three selected domain sizes shown in Fig.
342 2 are remarkably similar. This suggests that convection experienced at the profiler site is a

343 good approximation for convection experienced in a GCM size grid box centered on the wind
344 profiler location. Importantly, the convective area fraction derived from the wind profiler for
345 the whole time period (solid curve, $|v| \geq 0 \text{ m s}^{-1}$) shows a similar structure as the area fraction
346 from CPOL in the lower and middle troposphere but drops off more rapidly above 8 km. The
347 CPOL radar takes 10 mins to complete each volumetric scan, so when present it is assumed
348 that the convection will last for the entire 10 mins. The example discussed in Fig. 1 shows
349 that the temporal variability is high within 10 minutes, with large differences observed
350 between CPOL and 920 MHz reflectivities. In contrast, the wind profiler samples every 1
351 min, so even though a 10-min window is classified as convective by CPOL, the individual ten
352 1-min profiles from the wind profiler does not always contain valid vertical velocity
353 measurements. Inevitable instrumental problems may have further contributed to this. Also,
354 at higher altitude, the profiler area fraction begins to drop relatively rapidly compared to the
355 CPOL fractions due to the drop in profiler sensitivity with altitude. The CPOL sensitivity
356 does not change much with height.

357 We further evaluate the area fraction estimates from the profiler by applying consecutively
358 larger thresholds to the vertical velocity measurements. The thresholds of $|v|$ greater 0.5, 1.0
359 and 1.5 m s^{-1} are chosen as they have been employed by previous investigators to identify
360 updraft/downdraft cores in cumulus clouds (e.g., LeMone and Zipser, 1980; May et al. 2002;
361 Giangrande et al. 2013). Changing the velocity threshold leads to significant differences in
362 area fraction from the two radars, in particular below the freezing level. Above the height of
363 10 km, the fractions for different thresholds are similar because the profiler is detecting only
364 highly reflective regions from the high altitudes. From hereon we will use all vertical velocity
365 data points from the identified convective periods at the profiler without applying any further
366 thresholds since the profiler-derived area fraction with no threshold was closest to the CPOL

367 area fractions and therefore likely captures best the behaviour of the entire domain more
368 robustly.

369 Equipped with estimates of area fraction and in-cloud vertical velocity from the profiler
370 measurements we can now calculate the mass-flux M_c ($\text{kg s}^{-1} \text{m}^{-2}$). Here, M_c is defined
371 using the traditional GCM-type definition for mass-flux by considering all cumulus cloud
372 occurring over a large area:

$$373 \quad M_c = \rho\sigma_u v_u + \rho\sigma_d v_d \quad (1)$$

374 where: ρ is the air density (kg m^{-3});

375 σ_u , is the area fraction of updraft cores in the grid box and is a dimensionless
376 quantity. The σ_u can be further subdivided into the numbers of cores and the width of cores;

377 v_u is the mean velocity of updraft (m s^{-1});

378 and σ_d and v_d is the area fraction and mean velocity of the downdraft cores,
379 respectively.

380 The vertical profile of air density is computed using standard textbook formulae, with input
381 temperature and pressure fields extracted from the Darwin radiosonde. The mean profiles of
382 all remaining variables in Equation (1) are computed using the profiler vertical velocity data
383 from the convective intervals. We found that unlike the area fraction, the mean mass-flux
384 profile was largely independent of the different velocity threshold (result not shown). This is
385 so since larger vertical velocity thresholds lead to smaller area fraction (Fig. 2) but much
386 larger mean vertical velocities with the two effects compensating and leading to similar mean
387 mass-flux values.

388 4. Results

389 4.1 Overall characteristics of convective mass-flux and its components

390 a) Mean mass-flux profile

391 Convective clouds were identified by the CPOL radar over the profiler site for a total of 283
392 10-minute scans during the two wet-seasons analysed here. This corresponds to a convective
393 area fraction near the surface of approximately 0.5 %. Note that this represents an average
394 including many instances with no convection present in the domain for significant periods of
395 time. It is therefore not comparable to convective area fractions found in previous studies
396 (e.g., Davies et al. 2013), which reach values up to 10 % but reflect instantaneous conditions
397 rather than long temporal averages. Going back to the overall time average, Table 1
398 summaries the contributions to the total convective area fraction from congestus (CTH < 7
399 km), deep (CTH between 7 and 15 km) and overshooting clouds (CTH > 15). It also shows
400 the variability of convective cloud frequency as function of different environment and large-
401 scale terciles. The results shown in Table 1 are discussed further in sections 4.2 and 4.3.

402 Mean profiles of the overall mass-flux as well as upward and downward mass-flux profiles
403 are shown in Fig. 3. Here, the lower x-axis represents the overall mean over the entire two
404 seasons including the very frequent times (99.5%) of no convective clouds present over the
405 profiler site. To provide at least a rough estimate of the values of mass-flux “when present”, a
406 value more useful to modellers, we average mass-fluxes over 3-hour windows and discard all
407 windows with no presence of convective clouds (~93 %). The results are indicated by the
408 upper x-axis in Figure 3. A 3-hour window translates to a grid size of roughly 60 km;
409 calculations based on 5 m s^{-1} mean average propagating speed of convective cells [Kumar et
410 al. 2013b]. Note that removing zeros will not affect the profile shape but only its magnitude.

411 The overall mean mass-flux (thick curve) increases steadily from near cloud base to peak at 6
412 km just above the freezing level, and thereafter decreases gradually with height. At all levels,
413 except at very high altitudes, the mass-flux totals are dominated by updrafts (thin curve).
414 Importantly, these observational results also validate those reported in many studies using
415 cloud-resolving models (e.g. Derbyshire et al. 2004; Kuang and Bretherton, 2006) and are
416 also in good agreement with previous attempts to retrieve mass-fluxes from profiler
417 observations (e.g., May and Rajopadhyaya, 1999).

418 **b) Mean area fraction and vertical velocity**

419 Equation 1 indicates that updraft and downdraft mass-fluxes are affected by three
420 fundamental factors; the number of cores, the size of the cores and the vertical velocity in the
421 cores. The product of the number and size terms divided by domain size gives the area
422 fraction. We now examine the characteristics of these three fundamental factors with the aim
423 to understand the relative contributions of these factors to the mass-flux totals.

424 We begin by examining the variations in convective area fraction (thick solid line in Fig. 4a)
425 divided into upward area fraction (thin solid line) and downward area fraction (dashed line).
426 Once again we show the overall period averages with the lower x-axis and those for 3-hour
427 windows that contain convection with the upper x-axis. At low levels, updraft and downdraft
428 area fractions are nearly equal. The updraft fraction remains more or less constant from near
429 the surface to 8 kilometre and then decreases steadily at higher levels. Starting from the top
430 the small downdraft fraction increases slightly to just above the freezing level, where a
431 significant increase in downdraft fraction occurs, indicating the potential importance of this
432 level in downdraft formation. In Fig. 4c and 4d, the upward and downward area fractions
433 (shaded) are subdivided into the number of cores (solid lines) and their size (dashed line).
434 The core width is measured in minutes, and represents the number of consecutive 1-min

435 periods with vertical motion $> 0 \text{ m s}^{-1}$ for an updraft core. Downdraft cores are defined
436 analogously using downward motion.

437 The mean core width associated with upward motion (dashed line in Fig. 4c) increases
438 gradually from an average of ~ 2 min at cloud base to a maximum average width of ~ 6 min at
439 a height of 8 km. Assuming a propagation speed of 5 ms^{-1} , this translates into a width of ~ 600
440 m near cloud base and ~ 1.8 km at mid-levels. Above 8km the updraft core width decreases
441 sharply. In contrast, the core frequency associated with updrafts is highest near cloud base,
442 decreasing monotonically with increasing height. The net effect of this pattern in updraft
443 width and frequency is that the upward area fraction is highest and constant between cloud
444 base and 8 km. Downdraft number increases downwards with a particularly sharp increase
445 near the freezing level. The average width of downdraft cores is ~ 3 min and remains fairly
446 constant with height. Once again assuming a 5 ms^{-1} propagation speed, this translates into a
447 size of ~ 900 m.

448 The mean vertical velocity (thick curve in Fig. 4b) increases gradually with height and peaks
449 at $\sim 4 \text{ m s}^{-1}$ at 12 km. The mean profile of vertical velocity is the sum of the velocity in
450 updrafts (thin curve in Fig. 4b) and downdrafts (dashed curve) weighted by the fractional area
451 of up- and downdraft cores. The updraft velocity evolution with height is very similar to the
452 mean with a drop between 2 and 3 km followed by a steady increase to values of $\sim 5 \text{ m s}^{-1}$ at
453 high levels. In contrast, the downdrafts show much weaker velocities of $\sim 1 \text{ m s}^{-1}$ which are
454 almost constant throughout the cloud layer with slightly large values near the tops of very
455 deep clouds.

456 Next, we reconcile the vertical structure of the mass-flux (Fig. 3) with area fraction (Fig. 4a,
457 Fig.4c–4d) and vertical velocity (Fig. 4b). As it is difficult to mentally sum all contributing
458 factors to the total mass-flux we compare the updraft and downdraft terms separately. The

459 increase in updraft mass-flux between 2 and 5 km is largely a reflection of the vertical
460 velocity increase combined with a small increase in area fraction. The large reduction in
461 updraft mass-flux above 8 km is due to the strong decrease in area fraction, which is slightly
462 offset by an increase in vertical velocity. Note that the decrease in density with height also
463 affects the overall mass-flux profile so that constant velocity and area fraction would still
464 imply a reduction of mass-flux with height. As the downdraft velocities are small and
465 relatively constant with height, the strong increase in downdraft mass-flux below 6 km (Fig.
466 3) is to first order driven by the corresponding increase in downdraft area fraction.

467 Overall, perhaps with the exception of the low-level increase of updraft mass-flux, the total
468 mass-flux is governed to first order by the area fraction. If confirmed at other locations, this
469 would provide the opportunity of estimating the first order characteristics of mass-flux from
470 area fraction alone, a quantity that is much more easily measured using instruments both on
471 the ground and in space than vertical motion.

472 **4.2 Sensitivity of mass-flux to environmental and large-scale conditions**

473 Of key relevance to cumulus parameterisation is the connection of mass-flux with the
474 environmental conditions in which the convection is embedded. In this section, we examine
475 the relationship between $RH_{0.5}$, CAPE, CIN and ω_{500} with the updraft mass-flux, upward
476 area fraction and upward velocity. As the downdraft contribution to overall mass-flux is
477 relatively small we focus on updraft behaviour only.

478 For the analysis shown in this section, the environmental conditions are grouped into terciles
479 of their respective probability density functions. This ensures that the wind profiler sampling
480 time in each tercile is identical. Note though that the amount of convective clouds observed in
481 each tercile can still vary significantly depending on how favourable the conditions in each

482 tercile are for convection. The tercile boundaries for each variable, the amount of time with
483 which convective clouds occur in each tercile and their sub-division into congestus, deep and
484 overshooting modes are shown in Table 1.

485 a) Effect of 0-5 km Relative Humidity (RH_{0-5})

486 A moist environment, which is represented by the upper tercile of RH_{0-5} , is thought to be
487 important to support the formation of deep convection over its shallower counterparts (e.g.,
488 Redelsperger et al. 2002; Takemi and Liu 2004). The results shown in top panels of Fig. 5
489 reveal several interesting differences between dry (solid curves, $RH_{0-5} < 68\%$) and moist
490 (dashed curve, $RH_{0-5} > 82\%$) conditions updraft mass-flux (left), area fraction (middle) and
491 velocity (right). The updraft mass-flux (Fig. 5a) in dry conditions exhibits a sharp peak at the
492 height of 6 km with a strong drop-off in mass-flux above that level, while in moist conditions
493 a smoother and deeper mass-flux profile is evident. The behaviour in dry conditions likely
494 indicates the prevalence of shallower clouds (see Section 4.3). The updraft area fraction is
495 much smaller in dry conditions, indicative of a less frequent occurrence of convection (see
496 also Table 1). As seen before for the overall means (Fig. 3 and 4), area fraction increases
497 from cloud base to mid-levels, followed by a decrease higher up. Vertical velocity increases
498 with height in both states of RH_{0-5} . Perhaps surprisingly, the velocities are stronger in dry
499 conditions than in moist conditions, partly compensating the lower mass-flux strength
500 induced by the lower area fractions in that state. The higher velocities can be understood by
501 the need to produce stronger updrafts to penetrate through the dry atmosphere, while in moist
502 conditions weaker updrafts occur more frequently and can penetrate higher into the moist
503 troposphere more easily.

504 b) Effect of CAPE

505 We next study the relationship of mass-flux to CAPE (Fig. 5d–f). The differences in the
506 upper ($> 747 \text{ J kg}^{-1}$; dashed) and lower tercile ($< 365 \text{ J kg}^{-1}$; solid) CAPE conditions are much
507 smaller than those for $\text{RH}_{0.5}$. The mass-flux is slightly weaker in low CAPE conditions and it
508 reaches higher levels in high CAPE conditions. Somewhat paradoxically, low CAPE
509 conditions give rise to higher area fractions. This is consistent with the findings of Kumar et
510 al. (2013b) who showed that low CAPE conditions are associated with more frequent but
511 shallower convective clouds over Darwin. The air parcels in the convective clouds are less
512 buoyant in low CAPE conditions, leading to weaker updraft speed (Fig. 5f) and often
513 shallower cloud. In contrast, in high CAPE condition, convection is much deeper because the
514 air parcels have greater growth momentum. While less frequent in high CAPE conditions,
515 convection that occurs does show significantly larger vertical velocity. The net effect is that
516 the updraft mass-flux at all heights, except near cloud base, is higher in high CAPE
517 conditions compared to low CAPE.

518 c) Effect of CIN

519 In general, when the convective inhibition (CIN) of the atmosphere is low, more convective
520 cloud systems are likely to form. This is confirmed by our analysis of mass-flux in the lowest
521 ($< 30 \text{ J kg}^{-1}$; solid) and highest ($> 62 \text{ J kg}^{-1}$; dashed) CIN terciles (Fig. 5g–i). There is a large
522 difference in mass-flux between high and low CIN conditions, which is entirely caused by
523 differences in area fraction, which is synonymous with the frequency of occurrence of
524 convection. The vertical velocity profiles are largely unaffected by the state of CIN,
525 indicating that CIN is more likely a predictor for the existence of convection than its strength.

526 d) Effect of large-scale upward motion at 500 hPa (ω_{500})

527 Similar to CIN, large-scale vertical motion is strongly related to the existence of convection
528 (Fig. 5j-l). Almost all convective events occur in the “lower” tercile, which comprises upward
529 motion ($\omega_{500} \leq -1.82$ hPa Hour⁻¹, solid), while the upper tercile of large-scale downward
530 motion ($\omega_{500} \geq 1.24$ hPa Hour⁻¹, dashed) is more or less void of convection. The very small
531 fraction (7%) of convective systems that do form when there is large-scale downward motion
532 tend to have very high vertical velocities in the upper part of the updrafts, although the poor
533 sampling in this class prevents us from drawing any firm conclusions.

534 e) Summary of effects of environmental conditions on mass-flux

535 In its entirety Fig. 5 provides an important set of lessons about convective behaviour that can
536 potentially be used in the construction of cumulus parameterisations. It is clear that different
537 environmental parameters, many of which have been used in constructing elements of
538 existing cumulus schemes, have different effects on the mass-flux, mainly because they affect
539 its two components, area and velocity, in different ways. Large-scale vertical motion and CIN
540 are strongly related to area fraction. These conditions strongly influence the existence and
541 prevalence of convection and through the area fraction exerts a strong control on the
542 convective mass-flux. In addition, RH_{0.5} is also strongly related to vertical motion in the
543 clouds, although it is likely that there is no direct causality in that relationship. Instead, we
544 speculate that the higher velocities in dry conditions are a result of weaker updrafts not being
545 able to penetrate the dry atmosphere. Changes in CAPE have the least impact on the
546 convective area fraction but instead show a strong relationship with cloud growth dynamics.
547 In low CAPE conditions, the convective systems tend to be moderately more frequent but
548 with weak updraft speed while high CAPE conditions support stronger vertical motion,
549 leading to overall slightly higher mass-fluxes in those conditions. In summary, there is some
550 evidence from Fig. 5, that the components of mass-flux are responding differently to different

551 environmental conditions, making it difficult to relate mass-flux itself to only one of them.
552 This may indicate a potential benefit from treating area and velocity separately in future
553 cumulus parameterisation approaches.

554 **4.3 Contributions of each cumulus cloud type to the total mass-flux in different** 555 **environmental conditions**

556 Having investigated the overall mass-flux properties and their relationship to the state of the
557 environment the convection is embedded in we now investigate the contributions of
558 individual cumulus cloud modes, namely congestus, deep and overshooting clouds, to the
559 overall cumulus mass-flux. The three cloud modes are defined by tracking convective cells
560 and identifying their maximum echo top height (ETH, Kumar et al. 2013a, 2014). Cells that
561 never exceed a 7 km ETH are classified as congestus, those that exceed 15 km ETH are
562 classified as overshooting and the rest as deep convection. Kumar et al. (2013a) noted that
563 these three modes have remarkably different rainfall and drop size characteristics, and thus, it
564 will be worthwhile to examine the vertical velocity and mass-flux characteristics of these
565 cumulus modes separately as well as quantify their overall effect.

566 The breakdown of the total time for which the three cumulus convective modes are found at
567 the profiler site is shown in Table 1. We find that the most frequent type of convection
568 sampled by the profilers is deep convection, with just over half of all cases in this category.
569 The other two types contribute roughly one quarter each to the overall sample.

570 The mean profile of upward mass-flux associated with the three cumulus modes and the
571 components of these mass-flux profiles are displayed in Fig. 6. Given its high frequency the
572 highest contribution to the upward mass-flux in the lower 8 km of the troposphere is from the
573 deep mode. The mean vertical velocity intensity of this mode shows intermediate strength

574 updraft velocities of 2 to 4 m s⁻¹ with a bimodal structure with peaks at 6 km and above 10
575 km. The congestus mode contributes about one quarter of the area fraction below 4 km, but
576 due to its relatively weak upward motion on the order of only 1 m s⁻¹ makes a relatively small
577 contribution to overall mass-flux. The overshooting mode contributes around one quarter to
578 the area fraction below 10 km and dominates the area fraction above that level. It shows the
579 strongest vertical motion of the three modes with average values increasing from around 4 m
580 s⁻¹ at 5 km to 6 m s⁻¹ above 10 km.

581 As the mass-fluxes were shown to be sensitive to the environmental conditions we next
582 investigate how the relative contribution from the three cloud modes may change with the
583 state of the environment. It was evident from Fig. 5 that ω_{500} and CIN mostly determined the
584 existence of convection, while RH_{0.5} and CAPE had a more direct influence on its structure.
585 We therefore focus on the latter two parameters.

586 The total time of each cumulus mode during the different environmental conditions are given
587 in Table 1. The most notable change in total time of individual cumulus modes with respect
588 to different environment conditions occurs for the overshooting mode when sorted with
589 respect to CAPE. While constant in overall terms (Fig. 7h and k), overshooting cloud forms
590 17% of all convection in low CAPE conditions but 37% in high CAPE conditions. This is a
591 result of the occurrence of both the congestus and deep mode decreasing as CAPE increases
592 (Fig. 7 h and k). As expected, the vertical velocities for the deep and in particular for the
593 overshooting mode increase with CAPE (Fig. 7i and l), leading to the overall larger mass-
594 fluxes in high CAPE conditions discussed earlier (Fig. 5). We now see that this increase is
595 predominantly driven by an increase in the velocities in the overshooting mode.

596 Changes in RH_{0.5} (Fig. 7a–f) also strongly affect the overall mix of the occurrence of
597 convective modes. In dry conditions, 60% of the time convection is present is associated with

598 either the congestus or overshooting mode. In contrast, in moist conditions the deep mode
599 becomes the dominant mode occurring 54% of time. The area of all three convective modes
600 increases significantly in moist conditions (Fig. 7b and e), while the velocities in the deep
601 modes decrease by about half with little change in the congestus mode. This once again
602 highlights that deep convection of both types is stronger but less frequent in dry conditions.

603 **4.4 Variability in mass-flux measurements**

604 The results shown so far have focused entirely on the mean behaviour of mass-flux and its
605 components, although some indication of variability is revealed by the breakdown into cloud
606 modes. In this section we aim to investigate the variability of mass-flux at the typical scale of
607 a GCM grid box across different events, as this is more readily comparable to what the mass
608 flux parameterization produces. To enable this investigation we need to compute the mass-
609 flux over some discrete time window rather than averaging over long periods of time. This
610 once again requires finding a compromise between representing the size of a GCM grid-box
611 and the results being affected by the time evolution of the convective systems over the time
612 window. We choose a 3-hour time averaging window (~60 km), but we will also contrast our
613 results to those found using a longer, 6-hour, window (~100 km). As most time-windows will
614 have no convection at all in them, we focus our investigation on the 95th, 99th and 99.5th
615 percentile of the respective distribution functions. Figure 8 shows these percentiles for area
616 fraction (top) and mass-flux (bottom) for both the 3-hour (green) and 6-hour (red) time
617 windows. For comparison, we include the area fractions measured by CPOL in a 50km radius
618 around the profiler site in Figure 8a.

619 While the length of the time window does not affect the mean profile of area fractions, it does
620 affect the variability. Shorter time windows will produce larger variability because there will
621 be increases in incidence of both very large and very small area fraction. Of the 2300 (1150)

622 available 3-hour (6-hour) time blocks 93% (88%) had a convective area fraction of 0. As
623 expected, the upper percentiles of the area fraction distribution yields larger (smaller) values
624 for the 3-hour (6-hour) window ranging from 0.05 (0.03) for the 95th percentile to 0.1 (0.08)
625 for the 99th percentile. The 6-hour window is in closer agreement with the CPOLE area
626 fractions.

627 The upper percentiles of the mass-flux distribution associated with the 3- and 6-hour
628 windows are shown in Fig. 8b. This figure is in the same format as Fig. 8a, except the 98th,
629 not the 95th percentile is shown, as the 95th percentile mass-fluxes were too small to be seen
630 clearly. The 98th percentile mass-fluxes have the same shape as the mean updraft and
631 downdraft mass-flux profile (Fig. 3), with peak updraft and downdraft mass-flux just above
632 the freezing level and close to cloud base, respectively. At higher percentiles very large
633 updraft mass-flux values can be seen at higher altitude and are associated with large vertical
634 velocity events associated with deep and overshooting convection.

635 **5. Summary and Discussion**

636 The aim of this study was to derive convective mass-fluxes and their components on the scale
637 of a GCM grid-box from wind-profiler observations and thereby to provide a zeroth-order
638 observational reference for the evaluation of cumulus mass-flux schemes. The analysis
639 conducted characterised the updrafts and downdrafts of convective clouds with continuous
640 dual-frequency wind profiler observations taken over two wet-seasons near Darwin,
641 Australia. We found the net mass-flux over the entire measurement period to be positive
642 (upwards) between 2 and 14 km height with a peak at ~6 km. The downdraft cumulus mass-
643 flux was shown to be strongest close to cloud base associated with precipitation processes,
644 with vertical motion values of less than half of that seen with the updrafts.

645 The separation of mass-flux into velocity and area fraction, the latter itself a product of core
646 width and frequency, showed that the mass-flux was most strongly regulated by area fraction
647 compared to the vertical velocity. While of secondary importance to overall mass-flux
648 magnitude, the vertical velocity intensities revealed some crucial properties related to the
649 cloud dynamics. The convective updraft velocity exhibited a dominant peak in the upper-
650 levels (>10 km), and a small secondary peak in lower level at 6 km particularly associated
651 with the deep convective cloud mode. The observed structures in vertical velocity intensities
652 associated with the deep convection (Fig. 6) matched well with the updraft profiles reported
653 in Heymsfield et al (2010). The overshooting convective mode had more intense vertical
654 velocity magnitudes than the deep mode at all height levels, increasing monotonically with
655 height.

656 By separating the mass-flux into contributions from different cloud types, we demonstrated
657 that wide variety of vertical velocity intensities and cumulus sizes contribute to the mean
658 mass-flux profile. This was shown to be due to a complex interplay of the frequency, size and
659 strength of cumulus clouds with the environment. The analysis revealed that $\sim 80\%$ of the
660 cumulus population over the two seasons formed when the large-scale vertical motions were
661 strongly upwards (≤ -1.82 hPa Hour $^{-1}$) and/or when CIN was small (≤ 30 J kg $^{-1}$). Both low-
662 level relative humidity ($RH_{0.5}$) and CAPE had a more moderate effect on the existence of
663 cumulus clouds but these parameters had a significant impact on the vertical velocity and
664 hence growth dynamics of the clouds. Higher mean velocities were mainly associated with
665 deeper convection that formed in dry ($RH_{0.5} < 68\%$) and high CAPE conditions (Fig. 5 and
666 Fig. 7). While the latter is easily explained by energetic arguments, the former is a less
667 obvious result. We interpret this result as driven by the effects of the entrainment of dry air
668 into the clouds limiting the vertical growth of clouds (e.g., Redelsperger et al. 2002). The

669 very few deep cumulus clouds that do succeed to grow in unfavourable dry conditions need
670 very strong vertical growth momentum and hence display very large vertical velocities.

671 The downdraft vertical velocities and frequencies were significantly less than those for
672 updrafts at all height levels, except at cloud base and near cloud top, where they were similar.
673 This is consistent with the conceptual picture that a convective cloud is generally made up of
674 one or more dominant updraft cores, which are partly compensated by small and short-lived
675 downdrafts driven by evaporative cooling from both cloud and precipitation hydrometeors
676 (see Fig 1).

677 Our study has extended previous investigations, such as studies of May and Rajopadhyaya
678 (1999) for the tropical Darwin region and Giangrande et al. (2013) for mid-latitude central
679 plain of United States by examining not only the overall mass-flux but its components at
680 scales relevant to GCM evaluation. Unlike these studies, we accepted all values of vertical
681 motion in our statistical analysis rather than setting a threshold value. This led to better
682 agreement with convective area fraction profile shapes derived from the CPOL scanning
683 radar (Fig 2), likely making our sample more representative. The mean updraft and downdraft
684 vertical velocity profiles found here are nevertheless in good agreement with earlier studies
685 (e.g., Heymsfield et al. 2010). The sensitivity of mass-flux to the environmental moisture
686 conditions is in broad agreement with the modelling study of Derbyshire et al. (2004). Both
687 the observational and model results show that during the moist conditions, the mass-flux has
688 a broad peak at mid levels, while in dry conditions, the mass-flux decreases monotonically
689 with height albeit this decrease starts at higher levels in the observations (4 km) than in the
690 model simulations (cloud base).

691 Despite the availability of two wet seasons of observations, perhaps the biggest limitation of
692 our study remains the relatively small sample size. This once again highlights the difficulty of

693 supporting the development of cumulus parameterizations with the relevant measurements.
694 An obvious way to alleviate this problem is to use data from scanning radar systems. Such
695 systems can provide frequent measurements of convective area fractions at GCM grid-box
696 scale (e.g., Davies et al. 2013) but the challenge is to derive long time series of reliable
697 retrievals of in-cloud vertical velocity from them. This will be the next step of this work. We
698 will use the computationally-efficient dual-Doppler retrieval technique from Protat and
699 Zawadzki (1999), which will be evaluated first using the wind profiler vertical velocities as in
700 Collis et al. (2013), but applied to a much longer dataset over Darwin. Our finding that mass-
701 flux profiles tend to be dominated by the convective area fraction and that in-cloud velocities
702 vary with cloud depth may also enable us to derive mass-flux estimates from scanning
703 systems by statistically modelling, rather than measuring, vertical motion and combining
704 those with more easily observed area fractions. This will be the subject of a further study that
705 will extend the first useful foray into supporting cumulus parameterization development more
706 directly with long-term observations presented in this paper.

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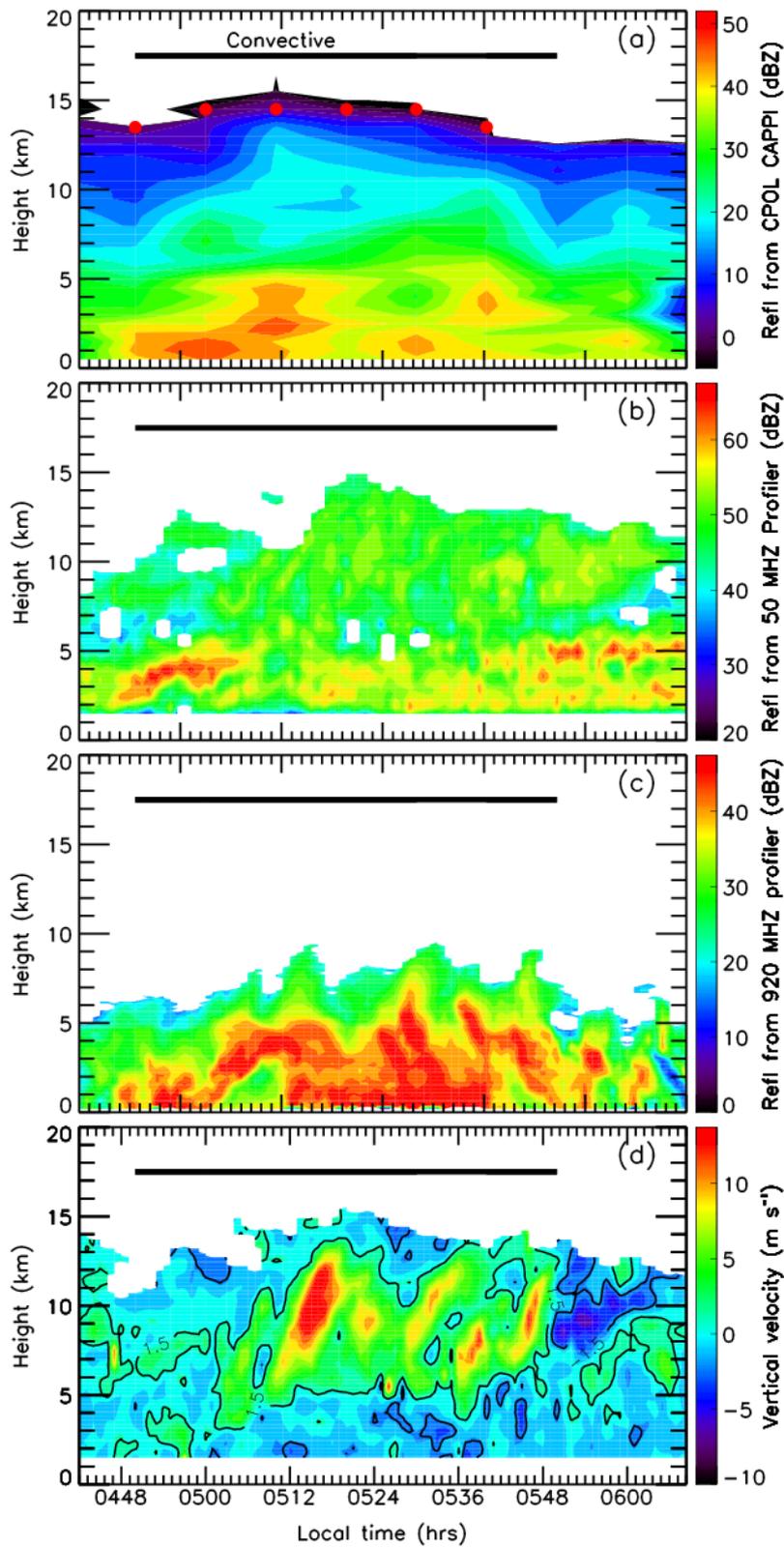
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870 Figure 1: Overshooting convection captured by the wind profiler around 0500 LT on 21st
 871 March 2006. a) Time-height section of the Darwin C-band polarimetric (CPOL) radar
 872 reflectivity collocated with wind profiler site. CPOL data used here is from volumetric mode,

873 where the data was collected over 10 mins intervals. The CPOL data was gridded in height
874 steps of 0.5 km. The red line is the 0-dBZ ETH of Steiner classified convective columns;
875 ETH of stratiform columns are not important to this study. The length of convective interval
876 in all panels is highlighted by the thick horizontal line. b)-d) Reflectivity from the 50 MHz
877 wind profiler, reflectivity from the 920 MHz wind profiler, vertical velocity (v) obtained
878 using the combination of the 50 and 920 MHz Doppler velocities, respectively. The profiler
879 data were displayed in using its primary resolution in bins of 1 min in time and 0.1 km bins in
880 height.

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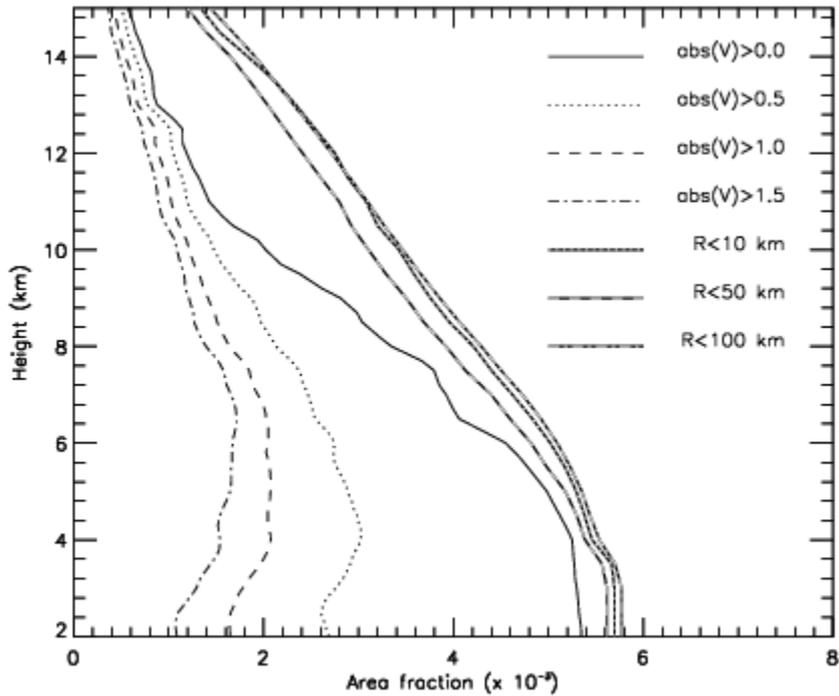
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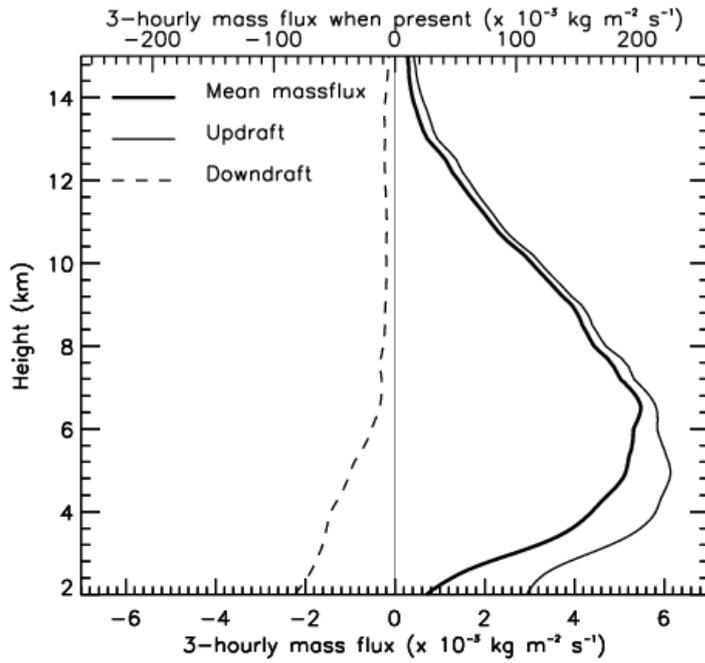
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Figure 2: The two wet-season mean profile of convective area fractions using vertically pointing observations from the wind profiler and volumetric observation from CPOL. As explained in the text, wind profiler area fractions were extracted by applying the “time approach” for different absolute vertical velocity threshold ranging from 0 to 1.5 m s⁻¹ in steps of 0.5 m s⁻¹. The area fractions from CPOL is extracted using the “space approach” for different circular region of radius ranging from 10 to 100 km centred over the wind profiler site.



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912 Figure 3: Two wet-season mean vertical profile of mass flux (thick curve), updraft (thin
 913 curve) and downdraft (dashed curve). Mass flux values were extracted using the GCM-type
 914 definition for mass flux considering all cumulus clouds occurring over a large area.
 915 Secondary x axis represents mass flux values provided there was at least one cumulus cloud
 916 in the 3-hour window. A 3-hour bin corresponds to nearly 60 km in distance, assuming the
 917 atmospheric flow speed is 5 m s⁻¹.

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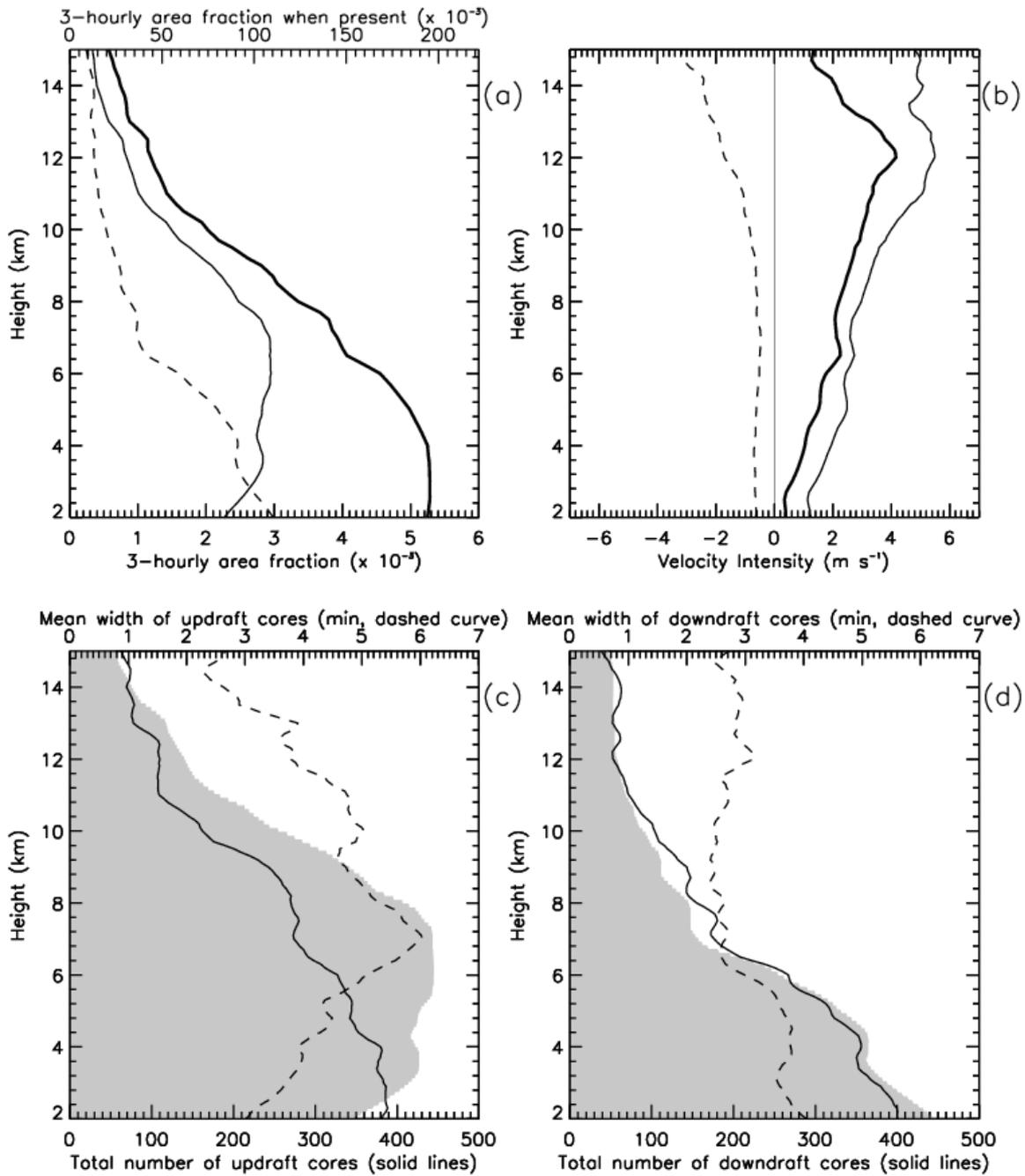
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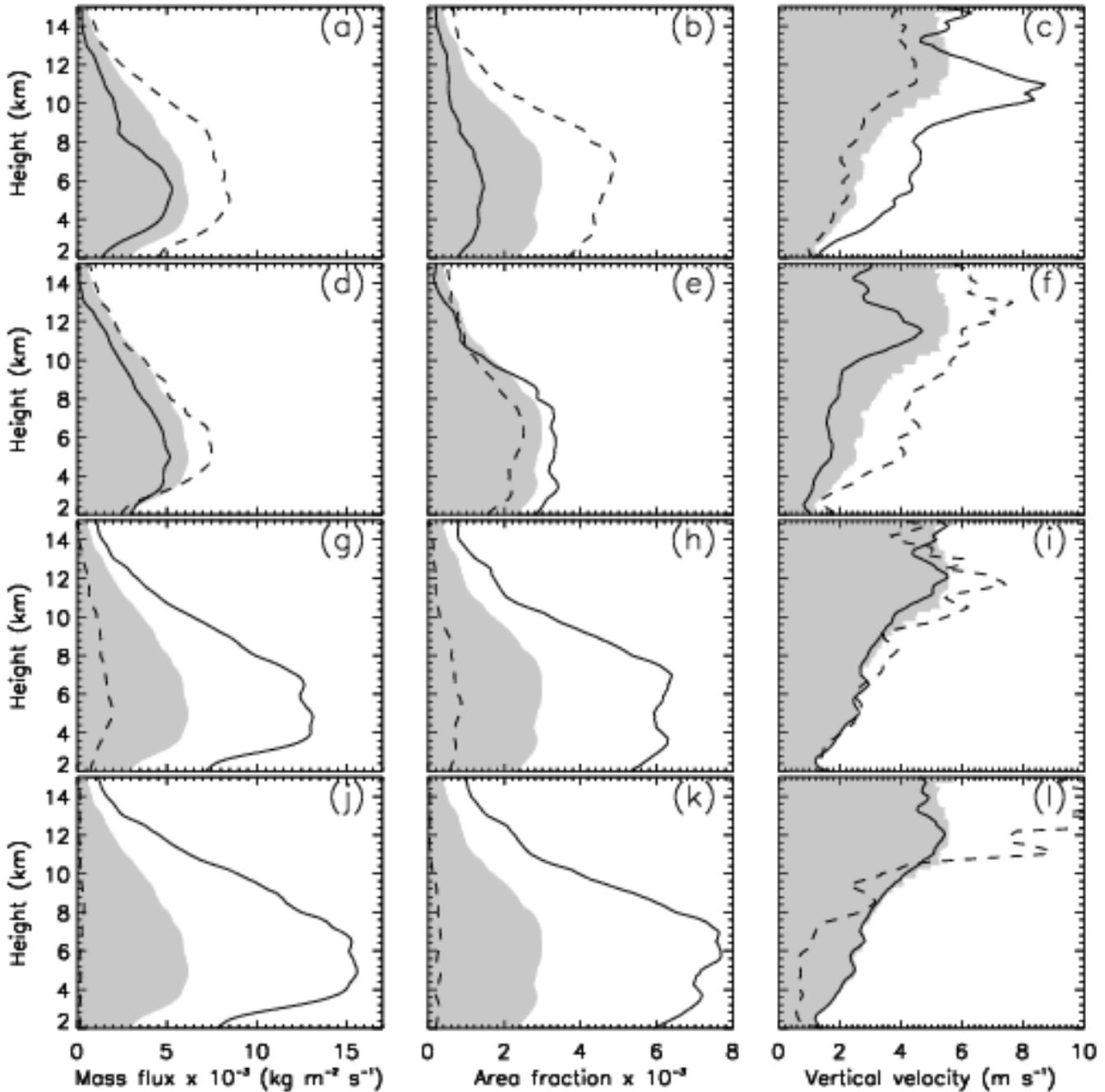
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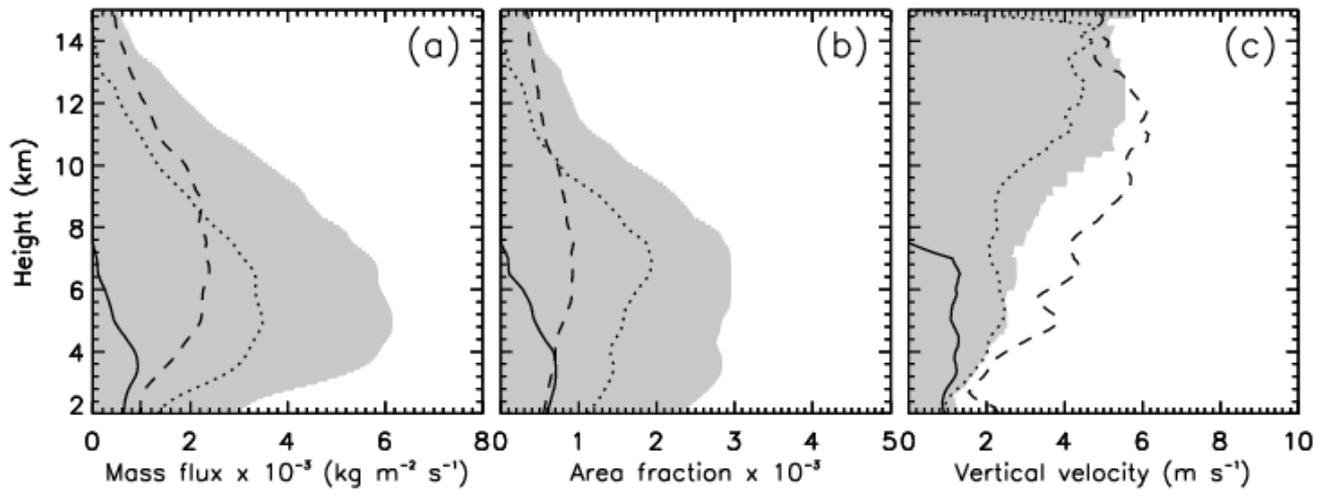
Figure 4: The same format as Fig.3 and shows mean profiles of the (a) mean area fractions, (b) vertical velocity, (c) cumulative count of updraft cores (solid curve) and mean width (dashed curve) - product of these two factors equates to upward area fraction (shaded) and (d) is the same as (c) and shows the characteristics of downdrafts.



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 935 Figure 5: Effect of 0–5 km relative humidity ($RH_{0.5}$, top panels), CAPE (second panels), CIN
 936 (third panels) and 500 hPa large-scale vertical velocity (ω_{500}) on updraft mass flux (left
 937 column), upward area fraction (middle column) and upward vertical velocity intensities (right
 938 column). The shaded region is the overall updraft means without applying any environmental
 939 sorting. The solid and dotted line in each figure corresponds to lower and upper terciles of the
 940 background condition. The tercile boundaries are in table 1.

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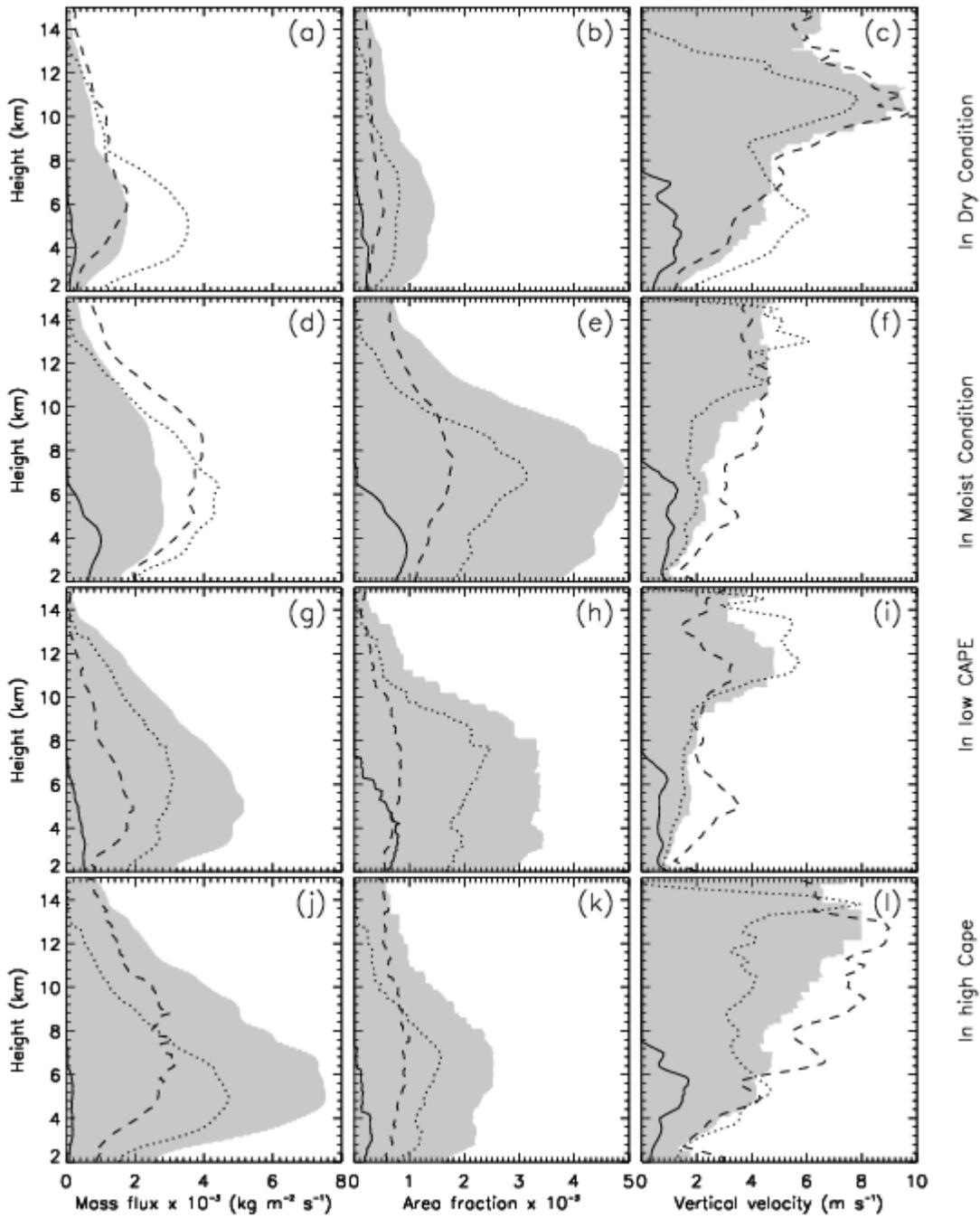
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948 Figure 6: Contribution to (a) updraft mass flux, (b) upward area fraction, and (c) vertical
949 velocity from congestus (solid), deep (dotted) and overshooting (dashed) cumulus clouds.
950 The shaded region represents is due to all the cumulus cloud modes.

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957 Figure 7: The same format as Fig. 6 and shows the mean response of the three cumulus
 958 modes in dry conditions (top panels), moist condition (second panels), low CAPE condition
 959 (third panels) and high CAPE condition (bottom panels).

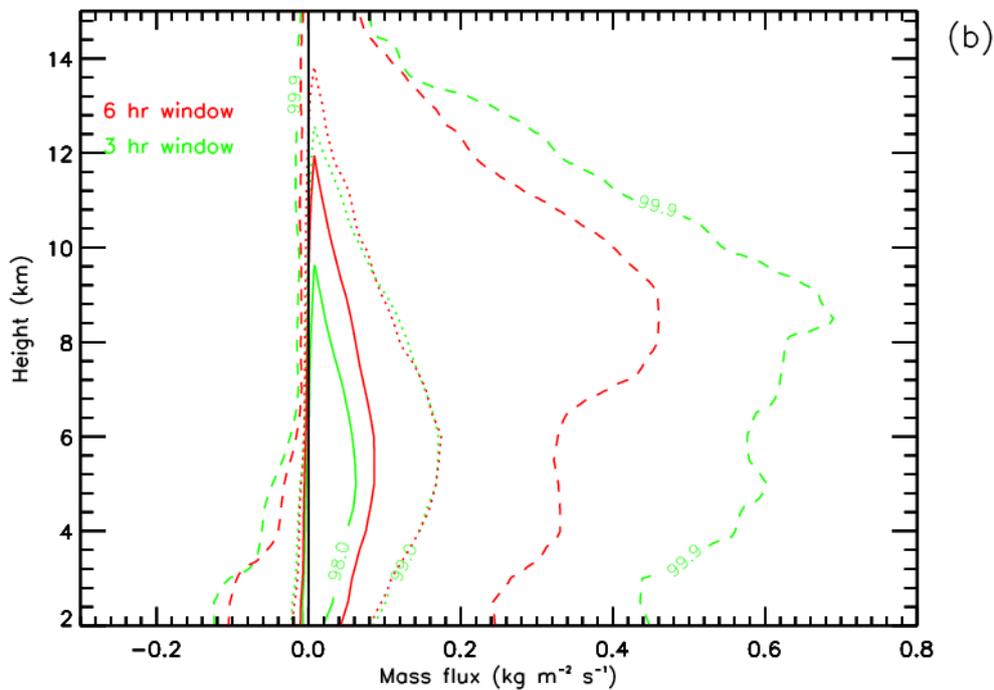
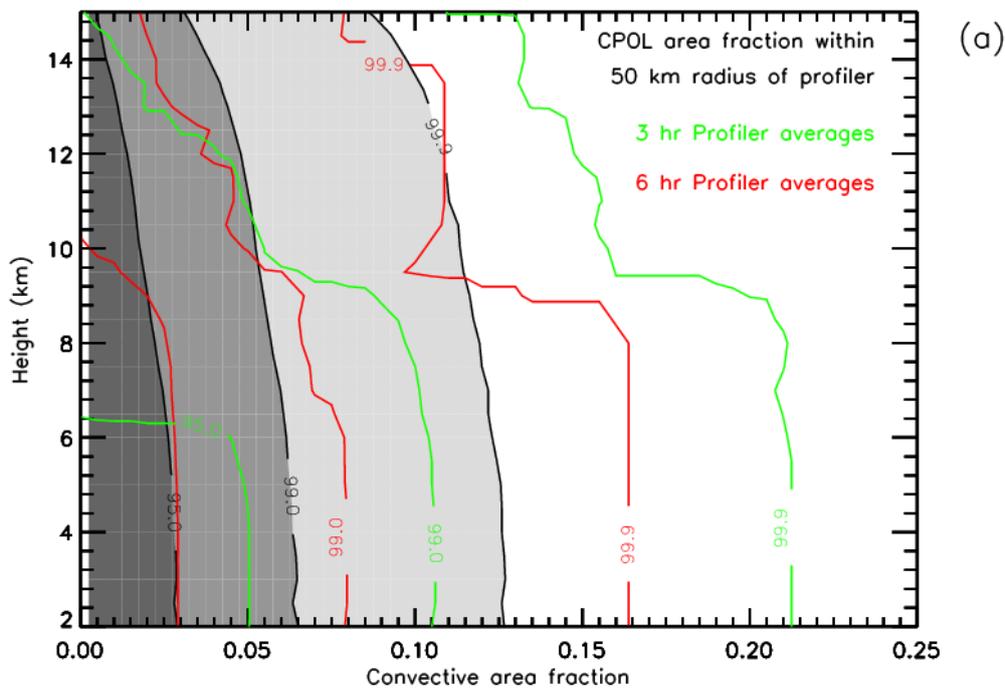
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1003 Figure 8: a) 2D cumulative probability distribution of convective area fraction from CPOL
 1004 over the circular region of radius 50 km centred at the profiler site (shaded) and convective
 1005 area fraction from wind profiler observations over discrete time intervals of 3 hours (green)
 1006 and 6 hours (red). b) The same as a) and shows the mass flux from wind profiler.

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1011 Table 1: The two wet-season occurrence frequency of congestus, deep and over-shooting
 1012 clouds, and separately for four environment background conditions
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Environmental Condition	Total of the 1-min wind profiler scans			
	Congestus (CTH < 7 km)	Deep (7 km < CTH < 15 km)	Over- shooting (CTH > 15 km)	All cumulus clouds
0-5 km RH \leq 68%	145	196	148	489
0-5 km RH 68%-82%	228	557	184	969
0-5 km RH \geq 82%	292	741	337	1370
CAPE \leq 365 J kg ⁻¹	251	620	184	1055
CAPE 365 – 747 J kg ⁻¹	305	540	222	1067
CAPE \geq 747 J kg ⁻¹	109	334	263	706
CIN \leq 30 J kg ⁻¹	527	1030	512	2069
CIN 30 – 62 J kg ⁻¹	78	346	127	551
CIN \geq 62 J kg ⁻¹	60	118	30	208
$\omega_{500} \leq -1.82$ hPa Hour ⁻¹	485	1232	649	2366
$\omega_{500} -1.82$ to 1.24 hPa Hour ⁻¹	109	176	10	295
$\omega_{500} \geq 1.24$ hPa Hour ⁻¹	71	86	10	167
All	665	1494	669	2828

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