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2	Estimation of Liquid Water Path in Stratiform Precipitation Systems using Radar Measurements
3	during MC3E
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21 Abstract

22 In this study, the liquid water path (LWP) in stratiform precipitation systems is retrieved, which 23 is a combination of rain liquid water path (RLWP) and cloud liquid water path (CLWP). The 24 retrieval algorithm uses measurements from the vertically pointing radars (VPRs) at 35 GHz and 25 3 GHz operated by the U.S Department of Energy Atmospheric Radiation Measurement (ARM) 26 and National Oceanic and Atmospheric Administration (NOAA) during the field campaign 27 Midlatitude Continental Convective Clouds Experiment (MC3E). The measured radar 28 reflectivity and mean Doppler velocity from both VPRs and spectrum width from the 35 GHz 29 radar are utilized. With the aid of the cloud base detected by ceilometer, the LWP in the liquid 30 layer is retrieved under two different situations: (I) no cloud exists below the melting base, and 31 (II) cloud exists below the melting base. In (I), LWP is primarily contributed from raindrops 32 only, i.e., RLWP, which is estimated by analyzing the Doppler velocity differences between two 33 VPRs. In (II), cloud particles and raindrops coexist below the melting base. The CLWP is 34 estimated using a modified attenuation-based algorithm. Two stratiform precipitation cases (20 35 May 2011 and 11 May 2011) during MC3E are illustrated for two situations, respectively. With 36 a total of 14 hours of samples during MC3E, statistical results show that the occurrence of cloud 37 particles below the melting base is low (8%), however, the mean CLWP value can be up to 0.87 38 kg m⁻², which is much larger than the RLWP (0.22 kg m⁻²). When only raindrops exist below the 39 melting base, the averaged RLWP value is larger (0.33 kg m^2) than the with cloud situation. The 40 overall mean LWP below the melting base is 0.39 kg m^2 for stratiform systems during MC3E. 41





42

43 **1. Introduction**

44 Clouds in stratiform precipitation systems are important to the Earth's radiation budget. 45 The vertical distributions of cloud microphysics, ice and liquid water content (IWC/LWC), determine the surface and top-of-the-atmosphere radiation budget and redistribute energy in the 46 47 atmosphere (Feng et al., 2011; 2018). Also, stratiform precipitation systems are responsible for 48 most tropical and midlatitude precipitation during summer (Xu, 2013). However, the 49 representation of those systems in global climate and cloud-resolving models are still challenging 50 (Fan et al., 2015). One of the challenges is due to the lack of comprehensive observations and retrievals of cloud microphysics (e.g. prognostic variables IWC and LWC) in stratiform 51 52 precipitation systems. Liquid water path (LWP), defined as an integral of LWC in the 53 atmosphere. It is a parameter used to provide the characterization of liquid hydrometeors in the 54 vertical column of atmosphere and study clouds and precipitation. The estimation of LWC/LWP 55 is one of the critical objectives of the US Department of Energy's (DOE) Atmospheric Radiation 56 Measurement (ARM) Program (Ackerman and Stokes, 2003).

57 LWP can be retrieved using the ground-based MicroWave Radiometer (MWR) sensed 58 downwelling radiant energy at 23.8 and 31.4 GHz (Liljegreen et al., 2001). In last two decades, 59 ARM has been operating a network of 2-channel (23.8- and 31.4-GHz) ground-based MWR to 60 provide a time series of LWP at the ARM Southern Great Plains (SGP) site (Cadeddu et al., 61 2013). Absorption-based algorithms using multichannels of MWRs have been widely used to 62 retrieve cloud LWP (e.g., Liljegren et al. 2001; Turner, 2007), and it is known to be accurate methods to estimate LWP of nonprecipitating clouds with mean LWP error of 15 g m⁻² (Crewell 63 64 and Löhnert, 2003). However, in precipitating conditions, LWP retrieved from conventional





65 MWR are generally not valid due to the violation of the Rayleigh assumption when large 66 raindrops exist (e.g., Saavedra et al., 2012). In addition, large increase of brightness 67 temperatures is measured as a result of the deposition of raindrops on the MWR's radome. 68 Unfortunately, it is very hard to model and quantify this increase from rain layer on the radome 69 (Cadeddu et al., 2017). This "wet-radome" issue largely inhibits the retrieving of LWPs using 70 ground-based MWR during precipitation. Due to the limitations of retrieving LWP from MWR 71 during precipitation, cloud and precipitation radars were used to simultaneously retrieve LWP 72 (Matrosov, 2010).

73 In the precipitating system, the liquid water cloud droplets and raindrops often coexist in the same atmospheric layer (e.g., Dubrovina, 1982; Mazin, 1989; Matrosov, 2009, 2010), 74 75 indicating that the LWP consists of both cloud liquid water path (CLWP) and rain liquid water 76 path (RLWP). However, the discrimination between suspended small cloud liquid water droplets 77 and precipitating large raindrops is a very challenging remote sensing problem. Even though the 78 partitioning of LWP into CLWP and RLWP is important in cloud modeling (Wentz and Spencer, 79 1998; Hillburn and Wentz, 2008), there are few studies retrieved RLWP and CLWP 80 simultaneously and separately (Saavedra et al., 2012; Cadeddu et al., 2017). Battaglia et al. 81 (2009) developed an algorithm to retrieve RLWP and CLWP from the six Advanced Microwave 82 Radiometer for Rain Identification (ADMIRARI) observables under rainy conditions. Saavedra 83 et al (2012) developed an algorithm using both ADMIRARI and a micro rain radar to retrieve 84 and analyze the CLWP and RLWP for midlatitude precipitation during fall. In addition to these 85 RLWP and CLWP estimations mainly from passive microwave radiometers, there are several studies to estimate the LWP using active radar measurements only. Ellis and Vivekanandan 86 87 (2011) developed an attenuation-based technique to estimate LWC, which is the sum of cloud





water contents (CLWC) and rain liquid water contents (RLWC), using simultaneous S- and Kaband scanning radars measurements. However, it is not always applicable of using these
techniques to retrieve LWC. If raindrop diameters are comparable to at least one of the radars'
wavelength, "Mie effect" will be included in the measured differential reflectivity, however this
"Mie effect" is not very distinguishable from differential attenuation effects (Tridon et al., 2013;
Tridon and Battaglia 2015).

94 Matrosov (2009) developed an algorithm to simultaneously retrieve CLWP and layer-95 mean rain rate using the radar reflectivity measurements from three ground-based W-, Ka-, S-96 bands radars. The CLWP were retrieved based on estimating the attenuation of cloud radar 97 signals compared to S-band radar measurements. Matrosov (2010) developed an algorithm to 98 estimate CLWP using a vertical pointing Ka-band radar and a nearby scanning C-band radar. 99 The layer-mean rain rate was first estimated with the aid of surface disdrometer, and then CLWP 100 was retrieved by subtracting the rain attenuation from total attenuation measured from two radars. 101 For the estimation of RLWP, Williams et al. (2016) developed a retrieval algorithm for rain drop 102 size distribution (DSD) using doppler spectrum moments observed from two collocated vertical 103 pointing radar (VPRs) at frequencies of 3 GHz and 35 GHz. The retrieved air motion and DSD 104 parameters were evaluated using the retrievals from a collocated 448-MHz VPR.

In this study, the CLWP retrieval algorithms in Matrosov (2009 and 2010) have been modified given the available radar measurements, vertical pointing Ka- and S-band radars, during the Midlatitude Continental Convective Clouds Experiment (MC3E) field campaign. For the estimation of RLWP, we will basically follow the idea described in Williams et al. (2016) to retrieve microphysical properties for raindrops, however instead of retrieving vertical air motion and rain DSDs (Williams et al., 2016), this study aims at retrieving RLWCs, and then integrating





RLWCs over the liquid layer to estimate RLWP. Overall, in this study, algorithms from three former publications are modified and combined to estimate the LWP in the stratiform precipitating systems.

114 The goals of this study are to retrieve the LWP, which includes both RLWP and CLWP 115 retrievals using radars measurements, and tentatively answer two questions based on 116 observations and retrievals in the stratiform precipitation systems during MC3E: (1) what is the 117 occurrence of cloud below the melting base in the stratiform precipitation systems; (2) what are the values of simultaneous CLWP, RLWP and LWP, and how does CLWP or RLWP contribute 118 119 to the LWP. Note that the CLWP and RLWP are constrained in a stratiform precipitation layer 120 below the melting base and above the surface. The LWP estimations in this study are primarily 121 aimed at stratiform precipitating events exhibiting melting-layer features from radar 122 measurements with lower-to-moderate rain rates ($RR < 10 \text{ mm hr}^{-1}$). The instruments and data 123 used in this study are introduced in section 2. Section 3 describes the methods of retrieving LWP 124 (both RLWP and CLWP). Section 4 illustrates two examples and followed by statistical results 125 from more samples during MC3E. The last section gives the summary and conclusions. 126 Acronyms and abbreviations are listed in Table 1.

127

128 **2. Data**

The MC3E field campaign, co-sponsored by the NASA Global Precipitation Measurement and the U.S. DOE ARM programs, was conducted at the ARM SGP (northern Oklahoma) during April-June 2011 to study convective clouds and improve model parametrization (Jensen et al., 2015). MC3E provided an opportunity to develop new retrieval methods to estimate cloud microphysics and precipitation properties in precipitation systems





- (Giangrande et al., 2014; Williams, 2016; Tian et al., 2016; Tian et al, 2018). Several stratiform
 rain cases were observed by the VPRs during MC3E (as shown in Fig. 1). Distinct signatures of
 "bright banding" are detected from VPRs. To retrieve LWP associated with stratiform
 precipitation, this study mainly uses the observations from two co-located VPRs operating at 3GHz and 35-GHz at DOE ARM SGP Climate Research Facility.
- 139 2.1 Vertical Pointing Radars

140 The 3-GHz (S-band) VPR was deployed by NOAA Earth System Research Laboratory for the six-weeks during the MC3E. The NOAA 3-GHz VPR is a vertical pointing radar with 141 142 2.6° beamwidth monitoring precipitation overhead. This 3-GHz profiler bridges the gap between 143 cloud radars, which are used to provide the structure of nonprecipitating clouds but are severely 144 attenuated by rainfall, and precipitation radars, which, although unattenuated by rainfall, 145 generally lack the sensitivity to detect more detailed cloud structure. The 3-GHz VPR observes the raindrops within the Rayleigh scattering regime and its signal attenuation are negligible 146 147 through the rain. The temporal resolution of the profiles of Doppler velocity spectra is 7 seconds 148 and the vertical resolution is 60 meters. The 3-GHz VPR operated in two modes: a precipitation 149 mode and a low-sensitivity mode. The precipitation mode observations are used in this study.

The Ka-band ARM zenith radar (KAZR) is also a vertical pointing radar, operating at 35 GHz permanently deployed by DOE ARM at the SGP site. The KAZR measurements include reflectivity, vertical velocity, and spectral width from near-ground to 20 km. The KAZR data used in this study are the KAZR Active Remote Sensing of Clouds (ARSCL) product produced by the ARM (www.arm.gov). The KAZR-ARSCL corrects for atmospheric gases attenuation and velocity aliasing. By selecting the mode with the highest signal-to-noise ratio at a given point, data from two simultaneous operating modes (general and cirrus mode) are combined for





157 each profile to provide the "best estimates" of radar moments in the time-height fields. The 158 vertical and temporal resolutions of KAZR-ARSCL product are 30 meters and 4 seconds, 159 respectively. Since the 3-GHz and 35-GHz VPRs are independent radars with different dwell 160 time and sample volumes (Williams et al., 2016), the radar observations are processed to 1-min 161 temporal and 60-m vertical resolutions in this study.

162 2.2 Disdrometers

163 DOE ARM program maintains a suite of surface precipitation disdrometers. 164 Measurements and estimations from the Distromet model RD-80 disdrometer and NASA two-165 dimensional video disdrometers (2DVD) deployed at the ARM SGP site are used in this study. The RD-80 disdrometer provides the most continuous raindrop size distribution (DSD) 166 167 measurements at high spectral (20 size bins from 0.3 to 5.4 mm) and temporal resolutions (1 168 minute), and its minimal detectable precipitation amount is 0.006 mm hr⁻¹. From 2DVD, the rain DSDs are observed from 41 bins (0.1 - 10 mm), and its minimal detectable precipitation amount 169 170 is 0.01 mm hr⁻¹. In addition to rain rate, the mean mass-weighted raindrop diameter (D_m) is also 171 provided from 2DVD, which is used for evaluating retrieved D_m from radar measurements.

172 2.3 Ceilometer

A Vaisala laser ceilometer (CEIL) operates at the SGP Central Facility, sensing cloud presence up to a height of 7700m with 10-m vertical resolution. The laser ceilometer transmits near-infrared pulses of light, and the receiver detects the light scattered back by clouds and precipitation. It is designed to measure cloud-base height.

177

178 **3. The Methodology of Liquid Water Path Estimation**

As mentioned earlier, both RLWP and CLWP contribute to the LWP. With aid of thecloud base height detected by ceilometer, LWP is retrieved under two different situations: (I) the





181 cloud base is higher than the melting base and (II) the cloud base is lower than the melting base. 182 For situation (I), there are almost no cloud droplets below melting base (CLWP = 0), and thus 183 the LWP below the melting base is solely from raindrops. The LWP is calculated by integrating 184 RLWCs over this layer. The RLWCs could be retrieved by analyzing the measured Doppler 185 Velocity Differences ("DVD Algorithm") from two collocated VPRs. In situation (II), the small 186 cloud droplets and large raindrops coexist below the melting base. Both raindrops and cloud 187 particles contribute to LWP. RLWP will be still estimated using "DVD Algorithm". CLWP will 188 be retrieved using an attenuation-based algorithm named as "Attenuation Algorithm". The 189 algorithms for LWP estimation are summarized in a flowchart (Fig. 2).

190 **3.1 Situation I (no cloud droplets exist below the melting base)**

191 The algorithm from Williams et al. (2016) was developed based on an assumption that 192 the 3-GHz VPR operates within the Rayleigh scattering regime for all raindrops, while the 35-193 GHz VPR operates within the Rayleigh scattering regime for small raindrops (diameters < -1.3194 mm) and non-Rayleigh scattering regime for larger raindrops (diameters ≥ -1.3 mm). The 195 different scattering regimes for the two operating frequencies result in different estimated radar 196 moments. These estimated radar moments are in functions of rain microphysics. Thus, the rain 197 microphysics could be retrieved with given measured radar moments. The details of this "DVD 198 Algorithm" and uncertainty estimation are introduced in Appendix A.

199 **3.2** Situation II (cloud particles and rain droplets coexist below the melting base)

In situation (II), substantial cloud particles exist below melting base, and both RLWP and CLWP retrievals are needed to estimate the LWP. The total two-way attenuation of 35-GHz VPR signals, A (in decibels, dB), in a layer between the melting base and the cloud base, mainly





203	consists of rain attenuation, liquid clouds attenuation, and gaseous attenuation. The total
204	attenuation (A) are expressed as:
205	$A=2 C R_m \Delta H + 2 B CLWP + G. $ (1)
206	R_m is layer-mean rain rate, and ΔH (km) is the thickness of the layer (Matrosov, 2009). G is the
207	two-way attenuation/absorption from atmospheric gases, which is relatively small, and the
208	absorption by gases has been already corrected in the KAZR ARSCL dataset and is assumed to
209	be zero in our retrieval.
210	C and B are the coefficients for rainfall and cloud liquid water attenuation.
211	$B=0.0026\pi\lambda^{-1}Im[-(m^2-1)(m^2+2)^{-1}],$ (2)
212	where λ is the wavelength of Ka-band radar, and m is the complex refractive index of water.
213	The unit of B is $dB/g m^{-2}$.
214	C = 0.27 b, (3)
215	where b is the correction factor considering raindrop fall velocities with changing air density.
216	$b = (\rho_{am} / \rho_{a0})^{0.45}, \tag{4}$
217	where ρ_{am} and ρ_{a0} are the mean air density in the rain layer and the density at normal atmospheric
218	conditions.
219	Based on (1), CLWP can be written as:
220	$CLWP = \frac{A-2 C R_{m} \Delta H - G}{2 B} $ (5)
221	The attenuation (A) is estimated by comparing the drop in Ka-band reflectivity with the
222	un-attenuated S-band reflectivity through the cloud. Assuming the changes in reflectivity with
223	altitude due to changes in raindrop size distributions with altitude are similar for Ka- and S-band
224	reflectivities, then the difference in reflectivities through the cloud is a proxy for attenuation.
225	This can be expressed using





$A \cong [Z_{Ka}(cloud \ base) - Z_{Ka}(melting \ base)] - [Z_S(cloud \ base) - Z_S(melting \ base)] \ (6)$
Notice that the absolute calibration of the radar was not important to the retrieval results since
the retrieval of CLWP used S-Ka differential attenuation. This avoids the radar calibration
(Tridon et al., 2015 and 2017), which is a serious issue limits the accuracy of radar retrievals.
The R _m is estimated as:
$R_m = \frac{\sum_{CB}^{MB} RR(h) \times \Delta h}{\Delta H},\tag{7}$
where Δh equals 60 meters and MB and CB are the melting base and the cloud base. RRs in the
layer between the melting base and the cloud base are calculated from the "DVD algorithm".
The uncertainties of retrieved CLWP are mainly due to the uncertainties of estimated R_m
and observed total attenuation from VPRs. The value of B_k is on the order of 1 dB/kg m ⁻² . The
uncertainty of retrieved CLWP would be $\sim 0.25~kg~m^2$ with 0.5 dB uncertainty from measured
radar reflectivity difference or $\sim 0.5~kg~m^2$ for 1.0 mm $hr^{\rm -1}$ uncertainty from estimated layer-
mean rain rate. Compared to the typical mean rain rate observed in the stratiform system (~ 2 - 4
mm hr ⁻¹), 1.0 mm hr ⁻¹ represents a ~ 30% uncertainty. The uncertainty for CLWP retrievals is
roughly estimated as ~ 0.56 kg m ⁻² (sqrt (0.25 ² +0.5 ²)) in this study. For reference, the expected
uncertainty is reported as ~ 0.25 kg m ⁻² for typical rainfall rates (~ 3 - 4 mm hr ⁻¹) in Matrosov
(2009) retrieval method.

243

244 **4. Retrieval Results and Discussions**

245 **4.1 Case Studies**

Even though situation (I) is dominated (Fig. 1), especially in Case A, the ceilometer cloud base estimates can be lower than the melting base (Cases B to D). Two case studies (20





248 May 2011 and 11 May 2011) are given as examples to demonstrate the estimation of LWP in

- 249 stratiform precipitation system for two different situations.
- 250 4.1.1 Case A

251 On 20 May 2011, an upper level low-pressure system at central Great Basin moved into 252 the central and northern Plains, while a surface low pressure at southeastern Colorado brought 253 the warm and moist air from the southern Plains to a warm front over Kansas. and a dry line 254 extended southward from the Texas-Oklahoma. With those favorable conditions, a strong north-255 south oriented squall line developed over Great Plains and propagated eastward. The convection 256 along the leading edge of this intense squall line exited the ARM SGP network around 11 UTC 257 20 May leaving behind a large area of stratiform rain (Case A in Fig. 1). This stratiform system 258 passed over the ARM SGP site and observed by two VPRs, and disdrometers as shown in 259 Figures 1a-1c. It clearly shows the 3-GHz radar echo tops are much lower than those from the 260 35 GHz VPR. Even though there is attenuation at 35-GHz by the raindrops and melting 261 hydrometeors, the 35-GHz radar can still detect more small ice particles at near the cloud top. 262 The "bright band", which occurs in a uniform stratiform rain region, is clearly seen from the 3-263 GHz VPR (a sudden increase and then decrease in radar reflectivity) but is not obvious from the 264 35-GHz VPR due to the non-Rayleigh scattering effects at 35 GHz (Sassen et al., 2005; 265 Matrosov, 2008).

Figures 1a-1b clearly show that the ceilometer detected cloud base is in the middle of the melting layer, indicating almost no cloud particles below the melting layer and the LWP in the liquid layer equals to RLWP. The RLWP is retrieved using the "*DVD Algorithm*" introduced in section 3.1 and Appendix A. Figure 3 shows an example of the DVD retrieval algorithm at 13:40 UTC on May 20, 2011. Radar reflectivity from 3 GHz, Doppler velocities from 3 GHz





271 and 35 GHz, and spectrum variance from 35 GHz are the inputs of DVD algorithm. The 272 Doppler velocity differences (3 GHz - 35 GHz) from the surface to 4 km are also plotted in Fig. 273 3d. The melting base is defined as the height of maximum curvature in the radar reflectivity 274 profile at 3 GHz (Fabry and Zawadzki, 1995), which is clearly seen at 2.5 km in Fig. 3. Below 275 2.5 km, the Doppler velocity differences between the two VPRs become relatively uniform, 276 indicating that the process of melting snow/ice particles into raindrops is completed. Retrieved 277 profiles of rain microphysical properties and their corresponding uncertainties (horizontal bars at 278 different levels) in the rain layer (0 - 2.5 km) are shown in Figs 3f-3h. In general, the retrieved 279 D_m values from the surface to 2.5 km are nearly a constant of ~2 mm (Fig. 3f), while the 280 retrieved RLWC and rain rate values slightly decrease from 2.5 km to the surface. One of the 281 highlights of this study is, in addition to the surface rain rate, which can usually be observed 282 using surface disdrometers, the vertical profiles of rain microphysical properties are retrieved. 283 These retrieved rain microphysical properties will shed light on the understanding of liquid cloud 284 and rain microphysical processes (like condensation, evaporation, autoconversion and accretion 285 etc.) in the models.

286 To evaluate the rain property retrievals, we compare the retrieved rain microphysical 287 properties, the D_m , and rain rate at the surface, with the surface disdrometers measurements (Fig. 288 4). The D_m values range from 1.0 to 2.5 mm during a 3.5-hr period with nearly identical mean 289 values of 1.79 mm and 1.81 mm from both retrievals and 2DVD measurements. There are large 290 variations for rain rates, ranging from 0 to 8 mm hr⁻¹, with means of 3.19, 3.17 and 2.88 mm hr⁻¹, 291 respectively, from 2DVD, RD-80 and radar retrieval. The mean rain rates from 2DVD and RD-292 80 measurements are almost the same although there are relatedly large differences during 293 certain time periods, while the retrievals from this study, on average, underestimate the rain rate





- by ~10% compared to the disdrometer measurements. More statistics (mean differences, their 95%
- 295 confidence intervals of mean differences and root mean square errors) can be found in Table 2.
- 296 Overall, the mean differences are within the retrieval uncertainties. The variation of RLWP (Fig.
- 4c) mimics the variation of retrieved rain rate in Fig. 4d. The mean value of RLWP is 0.56 kg m⁻¹
- ² for this case, which is also the LWP below the melting base.
- 299 **4.1.2 Case B**

300 On 11 May 2011, a surface cold front moved across the Oklahoma-Texas area and then 301 convections were initiated. At 1600 UTC, a mesoscale convective system organized with a 302 parallel stratiform precipitation region. Two-three hours later (~1830 UTC), the mesoscale 303 convective system was transitioned to a trailing stratiform mode passed over the ARM SGP site. 304 The large stratiform regions are observed by two VPRs and disdrometers as shown in Figs 1d-1f. 305 Figures 1d-1f clearly show that the ceilometer detected cloud bases are lower than the melting bases occasionally. Under this situation, both RLWP and CLWP could contribute to the LWP 306 307 below the melting base.

308 Firstly, the surface rain microphysics (D_m, RLWC, rain rate and RLWP) are retrieved 309 using "DVD Algorithm". These rain property retrievals are compared with the surface 310 disdrometers measurements (Fig. 5). The D_m values at the surface range from 0.8 to 2.2 mm 311 during a 4.5-hr period with the mean values of 1.46 mm and 1.57 mm, respectively, from both 312 retrievals and 2DVD measurements. The difference between the retrieval and 2DVD 313 measurement may be due to different sampling volumes between radar and the surface 314 disdrometer, as well as wind shear. To further investigate the difference, the measurements from five NASA 2DVDs located within 5 km away from VPRs are collected and processed. The 315 316 almost same mean values and slight variation from 5 NASA 2DVDs measurements suggest that





317 the difference between radar retrievals and the surface disdrometer measurements may be true,

318 while averaging from more measurements can only smooth the variation.

319 The mean rain rate values from five NASA 2DVDs and the surface disdrometer are very 320 comparable, with a mean difference of $0.3 \text{ mm } \text{hr}^{-1}$. The almost same mean values between the 321 surface disdrometer and 5 NASA 2DVDs measurements suggest that the DVDs apart within 5 322 km can capture very similar rain properties during a longer time period, such as 4.5 hours in this 323 case, although there are some large differences from their point-to-point measurements. The rain 324 rates, in this case, vary quite large, ranging from 0 to 9 mm hr^{-1} with means of 1.81, 1.64 and 325 1.98 mm hr⁻¹, respectively from single 2DVD, RD-80, and our retrieval. It is found that, from 326 both Case A and Case B, the mean value from RD-80 is smaller than that from 2DVD. This may 327 be due to the different ranges of measurable drop sizes from two types of disdrometers (0.3 - 5.4)328 mm for RD 80, while 0.1 to 10 mm for 2DVD). More statistics can be also found in Table 2. 329 Overall, the mean differences are still within the retrieval uncertainties for this case.

330 Secondly, the CLWP is retrieved using "Attenuation Algorithm" introduced in section 331 3.2. Figure 5c shows the time series of RLWP, CLWP and LWP retrievals. It is found that the CLWP values (when they exist) are usually larger than RLWP values in the same vertical 332 333 column. When cloud droplets and raindrops coexist below the melting base, the mean values are 334 0.31 kg m⁻² and 1.00 kg m⁻² for RLWP and CLWP, and the corresponding LWP below the 335 melting layer is 1.31 kg m⁻². While when only raindrops exist below the melting base, there is no 336 CLWP (CLWP =0), and the RLWP and LWP are the same (with average of 0.33 kg m^2). It is 337 noticed that even though the occurrence of CLWP is low (12%) in this case, the value of CLWP can be very large when it exists, and it is about two times larger than the mean RLWP. The 338 mean value of LWP is 0.45 kg m⁻² for all the sample in Fig. 5c. 339





340 4.2 Statistical Results

Box and whisker plots of retrieved RLWP, CLWP and LWP for situations (I), (II) and all 341 342 samples during MC3E are shown in Fig. 6. The horizontal orange and red dashed lines indicate 343 the median and mean, boundaries of the box represent the first and third quartiles, and the whiskers are the 10th- and 90th -percentiles. During MC3E, a total of 14 hours of stratiform rain 344 345 were observed by VPRs at the ARM SGP Climate Research Facility, in which 92% and 8% the 346 samples are categorized into the situations (I) and (II), respectively. The mean RLWPs are 0.33 kg m⁻² and 0.22 kg m⁻² for the situations (I) and (II). There are a substantial amount of small 347 348 cloud droplets sustaining in the rain layer and have not yet converted to larger raindrops, which 349 may partially explain smaller RLWP in the situation (II). The mean value of surface rain rate is 350 1.78 mm hr^{-1} when cloud droplets exist, which is also smaller than the mean value (2.06 mm hr⁻¹) 351 in the rain-only situation. The mean CLWP in the situation (II) is as large as ~ 0.87 kg m² even 352 though their occurrence is very low (8%), which is much larger than mean RLWP in the liquid layer. The ratio of RLWP and CLWP ranges from 4:1 to 2:1 for precipitating shallow marine 353 354 clouds reported at Lebsock et al. (2011), while our results from MC3E do not seem to have a 355 clear linear relationship between CLWP and RLWP (figure is not shown). The LWP from the 356 situation (II) is much larger than the mean LWP from the situation (I), which is primarily 357 contributed by cloud droplets. The overall mean LWP for stratiform rain during MC3E is 0.39 358 kg m⁻².

We also processed the ARM MWR retrieved LWPs during MC3E and compared with our retrievals as illustrated in Fig. 7. Statistical results of the retrieved LWPs from this study and MWR are averaged for each measured rain rate bins (bin size = 0.25 mm hr^{-1}). When the rain rate is greater than ~ 6mm hr⁻¹, there are no MWR LWP retrievals. Fig. 7b shows that the MWR retrieved LWPs, as expected, monotonically increase with rain rate, which is possible due to the





364 "wet radome" effect (Cadeddu et al., 2017). "Wet radome" is a particularly complicated 365 situation because the standing water often looks physically like a layer and less like a collection 366 of drops, making the MWR overestimate LWPs (personal communication with Dave Turner, 367 2018), and so far, no effective method was found to solve this problem (Cadeddu et al., 2017).

368 In addition to the issue from standing water on the radome, the scattering effects due to 369 raindrops also affect MWR retrievals. Two general retrieval methods are commonly used to 370 retrieve LWP from the observed brightness temperatures: statistical methods (Liljegren et al., 371 2001) and physical retrievals (Turner et al, 2007). No matter which retrieval is used, the 372 radiative transfer code usually only models the absorptions from atmospheric gases and cloud 373 liquid water. The scattering effect is not taken into consideration during the retrieval, that is, it is 374 under the assumption that the brightness temperature is primarily due to the emission of cloud 375 droplets in the MWR retrieval. Even small drizzle particles still have a scattering effect, which 376 could contribute higher brightness temperature measured by MWR and result in larger retrieved 377 LWPs than the "true" LWPs. Therefore, the MWR retrieved LWPs are most likely 378 overestimated for precipitating clouds.

379 In this study, we mainly focus on the stratiform rain systems with mean rain rates of 2-4 380 mm hr¹. The scattering effect for large raindrops is more significant than drizzles. Sheppard 381 (1996) examined the effect of raindrops on MWR brightness temperature measurements at 31 382 GHz and found that cloud absorption coefficient is only $\sim 2/3$ of rain absorption coefficient, 383 however, the scattering effect of raindrops is not insignificant where its scattering coefficient is 384 about half of cloud absorption coefficient. Thus, MWR measured brightness temperatures for 385 precipitating clouds would be higher, due to the scattering by raindrops, than those for non-386 precipitating clouds, and then result in higher LWPs than the 'true" LWPs. The differences of





LWPs from MWR and this study are shown in Fig. 7c. The LWP differences increase almost linearly with increased rain rate. The differences could be due to (1) MWR retrieved LWP represents the whole vertical column (RWLP and CLWP below melting layer, large water coated ice particles in the melting layer and supercooled LWCs above the melting layer), while our retrieval only represent the LWP below the melting base; (2) existing uncertainty in retrieved LWP from this study (~0.6 kg m⁻² when includeing CLWP estimates).

393

394 5. Summary and Conclusions

395 LWP is a critical parameter for studying clouds, precipitation, and their life cycles. LWP 396 can be retrieved from microwave radiometer measured brightness temperatures during cloudy 397 and light precipitation conditions. However, MWR-retrieved LWPs are questionable under 398 moderate and heavy precipitation conditions due to the "wet radome" and non-Rayleigh 399 scattering effects caused by large raindrops. LWPs below the melting base in stratiform 400 precipitation systems are estimated, which include both RLWP and CLWP. The measurements 401 used in this study are mainly from two VPRs, 35-GHz from ARM and 3-GHz from NOAA 402 during the MC3E field campaign.

In this study, the microphysical properties of raindrops, such as D_m , RLWC (and RLWP), and RR, are estimated following the method described in Williams et al. (2016) using measurements from co-located Ka- and S-band radars VPRs. The retrieved rain microphysical properties are validated by the surface disdrometer measurements. Instead of retrieving vertical air motion and rain DSDs (Williams et al., 2016), this study aims at retrieving RLWCs and then integrating RLWCs over the liquid layer to estimate RLWP. The CLWP is retrieved based on





the modifications of the methods in Matrosov (2009 and 2010) with available radar
measurements, vertical pointing Ka- and S-band VPRs, during the MC3E field campaign.

The applicability of retrieval methods is illustrated for two stratiform precipitation cases (20 May 2011 and 11 May 2011) observed during MC3E. Statistical results from a total of 14 hours samples during MC3E show that the occurrence of cloud droplets below the melting base is low (8%), while the CLWP value can be up to 0.87 kg m⁻², which is much larger than the RLWP (0.22 kg m⁻²). When only raindrops exist below the melting base, the averaged RLWP value is 0.33 kg m⁻², which is much larger than the mean RLWP in the cloud droplets and raindrops coexisted situation.

418 Reliable retrievals of RLWC and RLWP are critical for model evaluation and 419 improvement, as RLWC (rain mixing ratio) is an important prognostic variable in weather and 420 climate models. Furthermore, the retrievals in the whole rain layer would be useful to 421 understand the microphysical processes (i.e., condensation, evaporation, autoconversion, and 422 accretion etc.) and have great potential to improve model parametrizations in the future. Overall, 423 the LWP (CLWP and RLWP) retrievals derived in this study can be used to evaluate the models 424 that separately predict cloud and precipitation separately, and contribute comprehensive 425 information to study cloud-to-precipitation transitions.

426

427 Appendix A: Doppler Velocity Differences Algorithm ("DVD Algorithm")

428 Retrieving RLWC and other rain microphysical properties (i.e., drop size and rain rate) is 429 based on the mathematics of DSD radar reflectivity-weighted velocity spectral density S_{DSD}^{λ} 430 [(mm⁶ m⁻³) (m s⁻¹)⁻¹], which is a product of radar raindrop backscattering cross section $\sigma_b^{\lambda}(D)$ 431 (mm²) and DSD number concentration N_{DSD}(D) (mm⁻¹ m⁻³):





432
$$S_{DSD}^{\lambda}(v_z) = \left[\frac{\lambda^4}{\pi^5 |K_w|^2} \sigma_b^{\lambda}\right] N_{DSD}(D) \frac{dD}{dv_z}.$$
(A1)

433 The $\frac{dD}{dv_z}$ [mm (m s⁻¹)⁻¹] is used as a coordinate transformation from diameter to velocity,

- 434 where v_z (m s⁻¹) is the raindrop terminal velocity of diameter D (mm) at altitude z. λ is the
- 435 wavelength of radar. $|K_w|^2$ equals 0.93 and it is the dielectric factor.

436 The N_{DSD}(D) can be expressed as a normalized gamma shape distribution with a three

437 parameters (Leinonen et al., 2012):

438
$$N_{DSD}(D; N_w, D_m, \mu) = N_w f(D; D_m, \mu),$$
 (A2)

439 where

440
$$f(D; D_m, \mu) = \frac{6}{4^4} \frac{(\mu+4)^{(\mu+4)}}{\Gamma(\mu+4)} (\frac{D}{D_m})^{\mu} \exp\left[-(\mu+4)\frac{D}{D_m}\right].$$
(A3)

441 N_w is the scaling parameter, μ is a shape parameter, $\Gamma(x)$ is the Euler gamma function, and D_m is 442 a mean mass-weighted raindrop diameter estimated from the ratio of the fourth to third DSD 443 moments:

444
$$D_{m} = \frac{M_{4}}{M_{3}} = \frac{\int_{D_{min}}^{D_{max}} N_{DSD}(D) D^{4} dD}{\int_{D_{min}}^{D_{max}} N_{DSD}(D) D^{3} dD}.$$
 (A4)

445 where D_{min} and D_{max} represent the minimum and maximum diameters in the distribution, 446 respectively.

447 The intrinsic (non-attenuation) reflectivity factor and the mean velocity and the spectrum
448 variance are the zeroth, first, and second reflectivity-weighted velocity spectrum moments :

449
$$Z_{DSD}^{\lambda} = \sum_{v_{\min}}^{v_{\max}} S_{DSD}^{\lambda}(v_i) \Delta v$$
 (A5)

450
$$v_{DSD}^{\lambda} = \frac{\sum_{v_{min}}^{v_{max}} s_{DSD}^{\lambda}(v_i) v_i \Delta v}{z_{DSD}^{\lambda}}$$
(A6)

451
$$SV_{DSD}^{\lambda} = \frac{\Sigma_{v_{min}}^{v_{max}}(v_i - v_{DSD}^{\lambda})^2 S_{DSD}^{\lambda}(v_i) \Delta v}{Z_{DSD}^{\lambda}}.$$
 (A7)





452 where v_i is the discrete velocities and Δv is velocity resolution in the integration.

453 The Doppler Velocity Difference (DVD) is defined as

$$DVD = v_{DSD}^{3 \text{ GHz}} - v_{DSD}^{35 \text{ GHz}}.$$
 (A8)

- 455 Note that both DVD and SV are dependent on DSD parameters (D_m and μ) only.
- 456 The RLWC and rain rate (RR) can also be described using the DSD:

457
$$RLWC(g m^{-3}) = \frac{\pi}{6} 10^{-3} \sum_{D_{min}}^{D_{max}} N_{DSD}(D, N_w, D_m, \mu) D_i^3 \Delta D$$
(A9)

458
$$RR(mm hr^{-1}) = \frac{6\pi}{10^4} \sum_{D_{min}}^{D_{max}} N_{DSD}(D, N_w, D_m, \mu) D_i^3 v_z(D_i) \Delta D.$$
(A10)

459 In addition, there are two newly defined radar-related parameters ($Z_{3GHZ}LWC$ and $Z_{3GHZ}RR$), 460 which are also dependent on D_m and μ only:

461
$$Z_{3GHZ}LWC=10 \log_{10}(Z_{DSD}^{3GHz}/RLWC)$$
(A11)

462
$$Z_{3GHZ}RR=10 \log_{10}(Z_{DSD}^{3GHz}/RR)$$
(A12)

463 In this study, four variables, DVD, SV at 35 GHz (SV_{35GHz}), Z_{3GHZ}LWC and Z_{3GHZ}RR, are 464 pre-calculated using different groups of D_m and μ values, and then these values are stored in 465 look-up tables (LUTs). Raindrop backscattering cross sections are calculated using the T-matrix 466 with different temperatures and oblate raindrop axis ratios (Leinonen, 2014). LUT examples are 467 illustrated in Fig. A as functions of DVD and SV_{35GHz}. If we assume that the observed radar 468 Doppler velocity difference and spectrum variance from the 35-GHz radar is equal to the DSD 469 velocity difference and variance (DVD and SV_{35GHz}), the measured Doppler velocity difference 470 and spectrum variance at 35-GHz can determine a solution for D_m from the LUT (Fig. A(a)). 471 Similarly, a value of $Z_{3GHZ}LWC$ (or $Z_{3GHZ}RR$) can be found with measured DVD and SV_{35GHZ} 472 using the LUT in Fig. A(b) (or Fig. A(c)). Then RLWC (or RR) can be estimated using (A11) 473 (or (A12)) with measured reflectivity at 3-GHz (Z_{3GHZ}).





474 The observed radar Doppler velocity difference can be assumed to be equal to the DSD 475 velocity difference for two reasons: (1) even though the radar observed Doppler velocity 476 spectrum can be broaden by the air motion, this spectrum broadening variance is small (within 477 2%) relative to the DSD velocity spectrum because of the narrow beamwidth (0.2°) of KAZR 478 and (2) spectrum broadening is symmetric, which does not affect the first spectrum moment and 479 the DSD mean Doppler velocity only shifts due to the air motion. Therefore, the measured 480 differences of Doppler velocity between the 3-GHz and 35- GHz radars vertical pointing 481 observations are independent of air motion and can be assumed to be the same as DVD from 482 (A8). The validity of such an assumption is fully discussed in Williams et al. (2016).

483 The variabilities of 3-GHz and 35-GHz VPR observations within each 1-minute/60-meter 484 bin are regarded as the measurement uncertainties and will be propagated through the retrieval to 485 produce retrieval uncertainties. The retrieval uncertainties are estimated follow two steps: (1) 486 construct a distribution of input radar measurements. For example, the temporal resolution for 3-487 GHz VPR is seven seconds, thus there are about nine radar reflectivities observed for one minute. 488 A normal distribution is generated first using the mean and standard deviations of these nine 489 observed radar reflectivities for this 1-min/60-m resolution/bin. (2) repeat the DVD retrievals 490 using samplings from distributions of all input measurements. We randomly select 100 groups 491 of members from those (DVD, SV_{35GHz}, Z_{3GHz}) normal distributions to form 100 realizations, and 492 then produce 100 separate output estimates. The mean and standard deviation of the 100 493 solutions are regarded as the final retrieval and the retrieval uncertainty.

It is noted that the uncertainty here only considers estimates of instrument noise, not the uncertainties associated with assumptions used in the retrieval. For example, the gamma size distribution used in (A2) is an approximation which may introduce error into the retrieval.





497 However, it is very difficult to quantify this type of retrieval uncertainty. In this study, we 498 further compared our retrievals with independent surface disdrometers measurements to estimate 499 the uncertainties of retrievals. Also, when both radars are observing at Rayleigh scattering for 500 small raindrops, the reflectivity-weighted radial velocities for these particles should be the same. 501 In order to have a difference in radial velocity during the retrieval, large droplets must exist. The 502 maximum diameters in drop size distribution measured from disdrometer for all the stratiform 503 cases during MC3E are investigated. It is found that the occurrence of small-droplets-only 504 (maximum diameter <1.3 mm) is very low (less than 3%). Thus, it will not have a significant 505 impact on the retrieval results. Notice that this algorithm is not suitable for strong convective 506 rain due to the wind shear and strong turbulence as well as severe attenuation and extinction of 507 the Ka-band radar signal.

508

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520 References

521 Ackerman, T. P., and Stokes, G. M: The Atmospheric Radiation Measurement Program. Phys.

522 Today, 56,38–44, doi:10.1063/1.1554135, 2003

- 523 Battaglia, A., Saavedra, P., T. Rose, and Simmer, C.: Characterization of precipitating clouds by
- 524 ground-based measurements with the triple-frequency polarized microwave radiometer
- 525 ADMIRARI, J. Appl. Meteorol., 49(3), 394–414, 2009
- 526 Cadeddu, M. P., Liljegren, J. C., and Turner, D. D.: The Atmospheric radiation measurement
- (ARM) program network of microwave radiometers: instrumentation, data, and retrievals,
 Atmos. Meas. Tech., 6, 2359-2372, https://doi.org/10.5194/amt-6-2359-2013, 2013
- 529 Cadeddu, M. P., Marchand, R., Orlandi, E., Turner, D. D. and Mech, M. (2017). Microwave
- 530 Passive Ground-Based Retrievals of Cloud and Rain Liquid Water Path in Drizzling
- 531 Clouds: Challenges and Possibilities, IEEE Transactions on Geoscience and Remote

532 Sensing, vol. 55, no. 11, pp. 6468-6481, doi: 10.1109/TGRS.2017.2728699

- Crewell, S., and Löhnert, U. (2003). Accuracy of cloud liquid water path from ground-based
 microwave radiometry 2. Sensor accuracy and synergy, Radio Sci., 38, 8042,
 doi:10.1029/2002RS002634, 3.
- 536 Dubrovina, L. S.: Cloudness and precipitation according to the data of airplane soundings,
- 537 Gidrometeoizdat, Leningrad (in Russian), 218 pp,1982
- 538 Ellis, S. M., and Vivekanandan, J.: Liquid water content estimates using simultaneous S and K_a
- band radar measurements, Radio Sci., 46, RS2021, doi:10.1029/2010RS004361, 2011
- 540 Fabry, F. and Zawadzki, I.: Long-Term Radar Observations of the Melting Layer of Precipitation
- 541 and Their Interpretation. J. Atmos. Sci., 52, 838-851, https://doi.org/10.1175/1520-
- 542 0469(1995)052<0838:LTROOT>2.0.CO;2, 1995





- Fan, J., Liu, Y.-C., Xu, K.-M., North, K., Collis, S., Dong, X, and Ghan, S. J.: Improving
 representation of convective transport for scale-aware parameterization:1. Convection
 and cloud properties simulated with spectral bin and bulk microphysics, Journal of
 Geophysical Research: Atmosphere, 120, 3485–3509,
 https://doi.org/10.1002/2014JD022142, 2015
- Feng, Z., Dong, X. Q., Xi, B. K., Schumacher, C., Minnis, P., and Khaiyer, M.: Top-ofatmosphere radiation budget of convective core/stratiform rain and anvil clouds from
 deep convective systems. Journal of Geophysical Research, 116, D23202. https://doi.org/
 10.1029/2011JD016451, 2011
- Feng, Z., Leung, L. R., Houze, R. A., Jr., Hagos, S., Hardin, J., Yang, O., Han, B. and Fan, J.: 552 553 Structure and evolution of mesoscale convective systems: Sensitivity to cloud 554 microphysics in convection-permitting simulations over the United States. Journal of 10, 555 Advances in Modeling Earth Systems, 1470-1494. 556 https://doi.org/10.1029/2018MS001305, 2018
- Giangrande, S. E., Collis, S., Theisen, A. K., and Tokay, A.: Precipitation estimation from the
 ARM distributed radar network during the MC3E campaign, J. Appl. Meteorol. Climatol.,
 doi:10.1175/JAMC-D-13-0321.1, 2014
- Jensen, M.P., Petersen, W. A., Bansemer, A., Bharadwaj, N., Carey, L. D., Cecil, D. J, and
 Zipser, E. J.: The Midlatitude Continental Convective Clouds Experiment (MC3E),
 Bulletin of the American Meteorological Society. 151221073208006.
 https://doi.org/10.1175/BAMS-D-14-00228.1, 2015
- Leinonen, J., Moisseev, D., M. Leskinen, M., and W.A. Petersen, W.A.: A Climatology of
 Disdrometer Measurements of Rainfall in Finland over Five Years with Implications for





Global Radar Observations. J. Appl. Meteor. Climatol., 51, 392-404,
https://doi.org/10.1175/JAMC-D-11-056.1, 2012
Leinonen, J.: High-level interface to T-matrix scattering calculations: architecture, capabilities
and limitations, Opt. Express, vol. 22, issue 2, 1655-1660 doi: 10.1364/OE.22.001655,
2014
Liljegren, J. C., Clothiaux, E. E., Mace, G. G., Kato, S., and Dong, X.: A new retrieval for cloud
liquid water path using a ground-based microwave radiometer and measurements of
cloud temperature, J. Geophys. Res., 106(D13), 14485-14500,
doi:10.1029/2000JD900817, 2001
Lebsock, M.D., L'Ecuyer, T.S. and Stephens, G.L.: Detecting the Ratio of Rain and Cloud
Water in Low-Latitude Shallow Marine Clouds. J. Appl. Meteor. Climatol., 50, 419–432,
https://doi.org/10.1175/2010JAMC2494.1, 2011
Matrosov, S. Y.: Assessment of radar signal attenuation caused by the melting hydrometeor layer.
IEEE Trans. Geo Sci. Remote Sens.,46,1039-1047 doi: 10.1109/TGRS.2008.915757,
2008
Matrosov, S. Y.: A method to estimate vertically integrated amounts of cloud ice and liquid and
mean rain rate in stratiform precipitation from radar and auxiliary data, J. Appl. Meteorol.,
48, 1398–1410, doi:10.1175/2009JAMC2196.1, 2009
Matrosov, S. Y.: Synergetic use of millimeter- and centimeter-wavelength radars for retrievals of
cloud and rainfall parameters, Atmos. Chem. Phys., 10, 3321-3331,
https://doi.org/10.5194/acp-10-3321-2010, 2010
Mazin, I. P. (Ed.): Clouds and the Cloudy Atmosphere. Gidrometeoizdat, Leningrad, 648 pp,
1989.





- 589 Saavedra, P., Battaglia, A., and Simmer, C.: Partitioning of cloud water and rainwater content by
- 590ground-based observations with the Advanced Microwave Radiometer for Rain591Identification (ADMIRARI) in synergy with a micro rain radar, J. Geophys. Res., 117,
- 592 D05203, doi:10.1029/2011JD016579, 2012
- 593 Sassen, K., Campbell, J. R., Zhu, J., Kollias, P., Shupe, M., and Williams, C.: Lidar and Triple-594 Wavelength Doppler Radar Measurements of the Melting Layer: A Revised Model for 595 Dark-J. and Brightband Phenomena. Appl. Meteor.. 44. 301-312. 596 https://doi.org/10.1175/JAM-2197.1, 2005
- 597 Sheppard, B.E.: Effect of Rain on Ground-Based Microwave Radiometric Measurements in the
- 598
 20–90-GHz
 Range.
 J.
 Atmos.
 Oceanic
 Technol.,
 13,
 1139–1151,

 599
 https://doi.org/10.1175/1520-0426(1996)013<1139:EOROGB>2.0.CO;2,
 1996
- Tian, J., Dong, X., Xi, B., Wang, J., Homeyer, C. R., McFarquhar, G. M., and Fan J.: Retrievals
 of ice cloud microphysical properties of deep convective systems using radar
 measurements, Journal of Geophysical Research: Atmosphere., 121,10,820–10,839,
 https://doi.org/10.1002/2015JD024686, 2016
- Tian, J., Dong, X., Xi, B., Minnis, P., Smith, W. L., Jr, Sun-Mack, S., ... Wang, J.: Comparisons
 of ice water path in deep convective systems among ground-based, GOES, and CERESMODIS retrievals. Journal of Geophysical Research: Atmospheres, 123, 1708–1723.
 https://doi.org/10.1002/2017JD027498, 2018
- Tridon, F., and Battaglia, A.: Dual-frequency radar Doppler spectral retrieval of rain drop size
 distributions and entangled dynamics variables, J. Geophys. Res. Atmos., 120, 5585–
- 610 5601, doi:10.1002/2014JD023023, 2015





- 611 Tridon, F., Battaglia, A., and Kollias, P.: Disentangling Mie and attenuation effects in rain using
- 612 a Ka-W dual-wavelength Doppler spectral ratio technique, Geophys. Res. Lett., 40, 5548
- 613 5552, doi:10.1002/2013GL057454, 2013
- Tridon, F., Battaglia, A., Luke, E., and Kollias, P.: Rain retrieval from dual-frequency radar
 Doppler spectra: validation and potential for a 25 midlatitude precipitating case-study,
 Quarterly Journal of the Royal Meteorological Society, 143, 1364–1380, 2017.
- 617 Turner, D. D., Clough, S. A., Liljegren, J. C., Clothiaux, E. E., Cady-Pereira, K. E., and Gaustad,
- 618 K. L.: Retrieving liquid water path and precipitable water vapor from the Atmospheric
- Radiation Measurement (ARM) microwave radiometers. IEEE Trans. Geosci. Remote
 Sens., 45, 3680–3690, 2007
- Wentz, F.J. and Spencer, R.W.: SSM/I Rain Retrievals within a Unified All-Weather Ocean
 Algorithm. J. Atmos. Sci., 55, 1613–1627, https://doi.org/10.1175/15200469(1998)055<1613:SIRRWA>2.0.CO;2, 1998
- Williams, C. R.: Reflectivity and liquid water content vertical decomposition diagrams to
 diagnose vertical evolution of raindrop size distributions, J. Atmos. Oceanic Technol.,
 doi:10.1175/JTECH-D-15-0208.1, 2016
- Williams, C. R., Beauchamp, R. M., and Chandrasekar, V.: Vertical air motions and raindrop
 size distributions estimated from mean Doppler velocity difference from 3- and 35-GHz
- 629 vertically pointing radars. IEEE Transactions on Geoscience and Remote Sensing, 54,
- 630 6048–6060, https://doi.org/10.1109/ TGRS.2016.2580526, 2016
- Ku, W.: Precipitation and convective characteristics of summer deep convection over east Asia
 observed by TRMM, Monthly Weather Review., 141, 1577-1592.
 https://doi.org/10.1175/MWR-D-12-001



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Mid-latitude continental convective clouds experiment National Oceanic and Atmospheric Administration Fotal two-way attenuation of 35-GHz VPR signals Scaling parameter in the drop size distribution Maximum diameters in the size distribution Mean mass-weighted raindrop diameter Minimum diameters in the size distribution ntercept of ice particle size distribution **Atmospheric Radiation Measurement** coefficients for cloud water attenuation Two-dimensional video disdrometer Millimeter-wavelength cloud radar coefficients for rainfall attenuation Active remote sensing of clouds **Fwo-way** gaseous absorption Ka-band ARM zenith radar Doppler velocity difference Cloud liquid water path Microwave radiometer Number concentration Drop size distribution Department of Energy Rain liquid water path Base of melting layer Jayer-mean rain rate Liquid water path ce water content Raindrop diameter Looking up table Full Name Rain rate **Fable 1.** Acronyms and Abbreviations Used in This Study Acronyms and Abbreviations ARSCL 2DVD CLWP NOAA RLWP $\mathbf{N}_{\mathrm{DSD}}$ ARM D_{m} $\mathbf{R}_{\mathbf{n}}$ Ξ







	Spsp	Radar reflectivity-weighted velocity spectral density
	v_{DSD}^{λ}	First reflectivity-weighted velocity spectrum moments
		represent the mean velocity
Vz		Raindrop terminal velocity
	Znsn	Zeroth reflectivity-weighted velocity spectrum moments
		represent the intrinsic (non-attenuation) reflectivity factor
$\Gamma(\mathbf{x})$		Euler gamma function
۲		Radar wavelength
	σ ^λ	Raindrop backscattering cross section
ц	2	Shape parameter







	Mean Differences (95% confidence interval)	RMSE
Case A: D_m (RET, 2DVD) (mm)	-0.02 (-0.05, 0.01)	0.24
Case A: RR (RET, RD-80) (mm hr ⁻¹)	-0.29 (-0.40, -0.19)	0.98
Case A: RR (RET, 2DVD) (mm hr ¹)	-0.31(-0.48, -0.15)	1.45
Case B: D _m (RET, 2DVD) (mm)	-0.11 (-0.14, -0.07)	0.29
Case B: D _m (RET, 2DVD-all) (mm)	-0.09 (-0.13, -0.05)	0.34
Case B: RR (RET, RD-80) (mm hr ¹)	0.34 (0.16,0.53)	1.63
Case B: RR (RET, 2DVD) (mm hr ⁻¹)	0.17(-0.01,0.36)	1.61
Case B: RR (RET, 2DVD-all) (mm hr ¹)	0.14 (-0.08,0.36)	1.89

31







35-GHz VPR, and (c) rain rates from RD-80 surface disdrometer measurement for Case A (20 May 2011, 11:00 – 15:30 UTC); (d)-(f) Figure 1. Time series of (a) radar reflectivity (Z_e) from NOAA 3-GHz vertical pointing radar (VPR), (b) radar reflectivity from ARM for Case B (11 May 201, 18:30 – 23:00 UTC); (g)-(i) for Case C (27 April 2011, 8:00 – 12:00 UTC); (j)-(l) for Case D (20 May 2011 659 660 661 662 663 663

7:00 – 9:00 UTC). Ceilometer cloud base height estimates are shown with black dots at 1-minute resolution. Note that the ranges of

radar dBZ values are different in 3-GHz and 35-GHz radars.







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Figure 3. An example of illustrating the Doppler Velocity Differences (DVD) retrieval algorithm at 13:40 UTC on May 20, 2011. The inputs of the DVD retrieval algorithm are: (a) 3-GHz vertical pointing radar reflectivity factor (Z_e), (b) 3-GHz radar Doppler velocities (V_d), (c) 35-GHz radar Doppler velocities (V_d), and (e) 35-GHz radar spectrum variances (SV). The Doppler velocity difference between 3-GHz and 35 GHz is shown in (d). The outputs of the DVD retrieval algorithm are: (f) mass-weighted mean diameter D_m , (g) rain liquid water content (RLWC), and (h) rain rate (RR). Retrieval uncertainties are shown as horizontal thin black lines.



35





Figure 4. Time series of (a) retrieved (RET) (red dots) and 2DVD surface disdrometer estimated
(grey line) D_m, (b) RET (red dots), 2DVD (grey line) and RD-80 (black line) surface disdrometer
rain rate estimates, and (c) retrieved rain liquid water path (RLWP, red dots) for Case A (May 20,
2011.







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Figure 5. Time series of (a) retrieved (RET) (red dots) and 2DVD surface disdrometer estimated (grey lines) D_m , (b) RET (red dots), 2DVD (grey line) and RD-80 (black line) surface disdrometer rain rate estimates, and (c) rain liquid water path (RLWP, red dots), cloud liquid water path (CLWP, blue dots) and liquid water path (LWP = RLWP+CLWP, green lines) for Case B (May 11, 2011).







693RLWP ILWP IRLWP IICLWP IILWP IIRLWPLWP694Figure 6. Box and whisker plots of retrieved RLWP, CLWP and LWP for situation (I), (II) and695all samples. The horizontal orange line within the box indicates the median, boundaries of the696box indicate the 25th- and 75th -percentile, and the whiskers indicate the 10th- and 90th -percentile697values of the results. The red dash lines indicate the mean values.









Figure 7. (b) Statistic comparisons between LWP retrievals from this study (RET, dots with one standard deviation bars in green) and microwave radiometer (MWR, black dots with one standard deviation bars in black), (a) corresponding sample numbers (blue bars) in each rain rate bin (0.25 mm hr⁻¹), and (c) the LWP differences between two estimations, shown as a function of rain rate for all cases.









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