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2	Estimation of Liquid Water Path below the Melting Layer in Stratiform Precipitation Systems
3	using Radar Measurements during MC3E
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#### Abstract

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In this study, the liquid water path (LWP) below the melting layer in stratiform precipitation systems is retrieved, which is a combination of rain liquid water path (RLWP) and cloud liquid water path (CLWP). The retrieval algorithm uses measurements from the vertically pointing radars (VPRs) at 35 GHz and 3 GHz operated by the U.S Department of Energy Atmospheric Radiation Measurement (ARM) and National Oceanic and Atmospheric Administration (NOAA) during the field campaign Midlatitude Continental Convective Clouds Experiment (MC3E). The measured radar reflectivity and mean Doppler velocity from both VPRs and spectrum width from the 35 GHz radar are utilized. With the aid of the cloud base detected by ceilometer, the LWP in the liquid layer is retrieved under two different situations: (I) no cloud exists below the melting base, and (II) cloud exists below the melting base. In (I), LWP is primarily contributed from raindrops only, i.e., RLWP, which is estimated by analyzing the Doppler velocity differences between two VPRs. In (II), cloud particles and raindrops coexist below the melting base. The CLWP is estimated using a modified attenuation-based algorithm. Two stratiform precipitation cases (20 May 2011 and 11 May 2011) during MC3E are illustrated for two situations, respectively. With a total of 13 hours of samples during MC3E, statistical results show that the occurrence of cloud particles below the melting base is low (9%), however, the mean CLWP value can be up to 0.56 kg m<sup>-2</sup>, which is much larger than the RLWP (0.10 kg m<sup>-2</sup>). When only raindrops exist below the melting base, the averaged RLWP value is larger (0.32 kg m<sup>-2</sup>) than the with cloud situation. The overall mean LWP below the melting base is 0.34 kg m<sup>-2</sup> for stratiform systems during MC3E.

# 1. Introduction

Clouds in stratiform precipitation systems are important to the Earth's radiation budget.
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
atmosphere (Feng et al., 2011; 2018). Also, stratiform precipitation systems are responsible for
most tropical and midlatitude precipitation during summer (Xu, 2013). However, the
representation of those systems in global climate and cloud-resolving models is still challenging
(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
precipitation systems. Liquid water path (LWP), defined as an integral of LWC in the atmosphere.
It is a parameter used to provide the characterization of liquid hydrometeors in the vertical column
of atmosphere and study clouds and precipitation. The estimation of LWC/LWP is one of the
critical objectives of the US Department of Energy's (DOE) Atmospheric Radiation Measurement
(ARM) Program (Ackerman and Stokes, 2003).
LWP can be retrieved using the ground-based MicroWave Radiometer (MWR) sensed
downwelling radiant energy at 23.8 and 31.4 GHz (Liljegreen et al., 2001). In last two decades,
ARM has been operating a network of 2-channel (23.8- and 31.4-GHz) ground-based MWR to
provide a time series of LWP at the ARM Southern Great Plains (SGP) site (Cadeddu et al., 2013).
Absorption-based algorithms using multichannels of MWRs have been widely used to retrieve
cloud LWP (e.g., Liljegren et al. 2001; Turner, 2007), and they are known to be accurate methods
to estimate LWP of nonprecipitating clouds with mean LWP error of 15 g m <sup>-2</sup> (Crewell and Löhnert,
2003). However, in precipitating conditions, LWP retrieved from conventional MWR are

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

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), indicating that the LWP consists of both cloud liquid water path (CLWP) and rain liquid water path (RLWP). However, the discrimination between suspended small cloud liquid water droplets and precipitating large raindrops is a very challenging remote sensing problem. Even though the partitioning of LWP into CLWP and RLWP is important in cloud modeling (Wentz and Spencer, 1998; Hillburn and Wentz, 2008), there are few studies retrieved RLWP and CLWP simultaneously and separately (Saavedra et al., 2012; Cadeddu et al., 2017). Battaglia et al. (2009) developed an algorithm to retrieve RLWP and CLWP from the six Advanced Microwave Radiometer for Rain Identification (ADMIRARI) observables under rainy conditions. Saavedra et al (2012) developed an algorithm using both ADMIRARI and a micro rain radar to retrieve and analyze the CLWP and RLWP for midlatitude precipitation during fall. In addition to these 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 (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 Ka-band

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).

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

The goals of this study are to retrieve the LWP below the melting base, which includes both RLWP and CLWP retrievals using radars measurements, and tentatively answer two questions based on observations and retrievals in the stratiform precipitation systems during MC3E: (1) what is the 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 to the LWP. Note that the CLWP and RLWP are constrained in a stratiform precipitation layer below the melting base and above the surface. The LWP estimations in this study are primarily aimed at stratiform precipitating events exhibiting melting-layer features from radar measurements with lower-to-moderate rain rates (RR < 10 mm hr<sup>-1</sup>). The instruments and data used in this study are introduced in section 2. Section 3 describes the methods of retrieving LWP (both RLWP and CLWP). Section 4 illustrates two examples and followed by statistical results from more samples during MC3E. The last section gives the summary and conclusions. Acronyms and abbreviations are listed in Table 1.

#### 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 band" are detected from VPRs. To retrieve LWP associated with stratiform precipitation, this study mainly uses the observations from two co-located VPRs operating at 3-GHz and 35-GHz at DOE ARM SGP Climate Research Facility.

#### 2.1 Vertical Pointing Radars

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 2.6° beamwidth monitoring precipitation overhead. This 3-GHz profiler bridges the gap between cloud radars, which are used to provide the structure of nonprecipitating clouds but are severely attenuated by rainfall, and precipitation radars, which, although unattenuated by rainfall, 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 through the rain. The temporal resolution of the profiles of Doppler velocity spectra is 7 seconds and the vertical resolution is 60 meters. The 3-GHz VPR operated in two modes: a precipitation 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 each profile to provide the "best estimates" of radar moments in the time-height fields. The vertical and

temporal resolutions of KAZR-ARSCL product are 30 meters and 4 seconds, respectively. Since the 3-GHz and 35-GHz VPRs are independent radars with different dwell time and sample volumes (Williams et al., 2016), the radar observations are processed to 1-min temporal and 60-m vertical resolutions in this study.

#### 2.2 Disdrometers

DOE ARM program maintains a suite of surface precipitation disdrometers. Measurements and estimations from the Distromet model RD-80 disdrometer and NASA two-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) measurements at high spectral (20 size bins from 0.3 to 5.4 mm) and temporal resolutions (1 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 is 0.01 mm hr $^{-1}$ . In addition to rain rate, the mean mass-weighted raindrop diameter ( $D_m$ ) is also provided from 2DVD, which is used for evaluating retrieved  $D_m$  from radar measurements.

#### 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.

### 3. The Methodology of Liquid Water Path Estimation

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

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

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

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

#### 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 consists

of rain attenuation, liquid clouds attenuation, and gaseous attenuation. The total attenuation (A) are expressed as:

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$$A = 2 C R_m \Delta H + 2 B CLWP + G.$$
 (1)

 $R_{\rm m}$  is layer-mean rain rate, and  $\Delta H$  (km) is the thickness of the layer (Matrosov, 2009). G is the two-way attenuation/absorption from atmospheric gases, which is relatively small, and the absorption by gases has been already corrected in the KAZR ARSCL dataset and is assumed to be zero in our retrieval.

C and B are the coefficients for rainfall and cloud liquid water attenuation.

$$B=0.0026\pi\lambda^{-1}Im[-(m^2-1)(m^2+2)^{-1}], \qquad (2)$$

where  $\lambda$  is the wavelength of Ka-band radar, and m is the complex refractive index of water. The unit of B is dB/g m<sup>-2</sup>.

$$C = 0.27 \text{ b},$$
 (3)

213 where b is the correction factor considering raindrop fall velocities with changing air density.

$$b = (\rho_{am}/\rho_{a0})^{0.45}, \tag{4}$$

- where  $\rho_{am}$  and  $\rho_{a0}$  are the mean air density in the rain layer and the density at normal atmospheric conditions.
- Based on (1), CLWP can be written as:

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$$CLWP = \frac{A-2 C R_m \Delta H - G}{2 B}$$
 (5)

The attenuation (A) is estimated by comparing the drop in Ka-band reflectivity with the un-attenuated S-band reflectivity through the cloud. Assuming the changes in reflectivity with altitude due to changes in raindrop size distributions with altitude are similar for Ka- and S-band reflectivities, then the difference in reflectivities through the cloud is a proxy for attenuation. 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 limiting the accuracy of radar retrievals.

The  $R_m$  is estimated as:

$$R_m = \frac{\sum_{h_0}^{MB} RR(h) \times \Delta h}{\Delta H},\tag{7}$$

where  $\Delta h$  equals 60 meters and MB is the melting base and  $h_0$  is the height of the lowest unsaturated KAZR rang gate (Matrosov, 2010). 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 is on the order of 1 dB/kg m<sup>-2</sup>. The uncertainty of retrieved CLWP would be  $\sim 0.25$  kg m<sup>-2</sup> with 0.5 dB uncertainty from measured radar reflectivity difference or  $\sim 0.5$  kg m<sup>-2</sup> for 1.0 mm hr<sup>-1</sup> uncertainty from estimated layer-mean rain rate. Compared to the typical mean rain rate observed in the stratiform system ( $\sim 2$  - 4 mm hr<sup>-1</sup>), 1.0 mm hr<sup>-1</sup> represents a  $\sim 30\%$  uncertainty. The uncertainty for CLWP retrievals is roughly estimated as  $\sim 0.56$  kg m<sup>-2</sup> (sqrt ( $0.25^2+0.5^2$ )) in this study. For reference, the expected uncertainty is reported as  $\sim 0.25$  kg m<sup>-2</sup> for typical rainfall rates ( $\sim 3$  - 4 mm hr<sup>-1</sup>) in Matrosov (2009) retrieval method. More details of the estimation of CLWP uncertainties are in Appendix B.

#### 4. Retrieval Results and Discussions

#### **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 May 2011

and 11 May 2011) are given as examples to demonstrate the estimation of LWP in stratiform precipitation system for two different situations.

#### **4.1.1 Case A**

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

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

To evaluate the rain property retrievals, we compare the retrieved rain microphysical properties, the  $D_m$ , and rain rate at the surface, with the surface disdrometers measurements (Fig. 4). The  $D_m$  values range from 1.0 to 2.5 mm during a 3.5-hr period with nearly identical mean values of 1.79 mm and 1.81 mm from both retrievals and 2DVD measurements. There are large variations for rain rates, ranging from 0 to 8 mm hr<sup>-1</sup>, with means of 3.19, 3.17 and 2.88 mm hr<sup>-1</sup>, respectively, from 2DVD, RD-80 and radar retrieval. The mean rain rates from 2DVD and RD-80 measurements are almost the same although there are relatively large differences during 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% confidence intervals of mean differences and root mean square errors) can be found in Table 2.

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 0.55 kg m<sup>-2</sup> for this case, which is also the LWP below the melting base.

#### **4.1.2** Case B

On 11 May 2011, a surface cold front moved across the Oklahoma-Texas area and then convections were initiated. At 1600 UTC, a mesoscale convective system organized with a parallel stratiform precipitation region. Two-three hours later (~1830 UTC), the mesoscale convective system was transitioned to a trailing stratiform mode passed over the ARM SGP site. The large stratiform regions are observed by two VPRs and disdrometers as shown in Figs 1d-1f. 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 below the melting base.

Firstly, the surface rain microphysics (D<sub>m</sub>, RLWC, rain rate and RLWP) are retrieved using "DVD Algorithm". These rain property retrievals are compared with the surface disdrometers measurements (Fig. 5). The D<sub>m</sub> values at the surface range from 0.90 to 2.30 mm during a 4.5-hr period with the mean values of 1.41 mm and 1.52 mm, respectively, from both retrievals and 2DVD measurements. The difference between the retrieval and 2DVD measurement may be due to different sampling volumes between radar and the surface disdrometer, as well as wind shear. The rain rates, in this case, vary quite large, ranging from 0.02 to 4.78 mm hr<sup>-1</sup> with means of 1.36, 1.26 and 1.66 mm hr<sup>-1</sup>, respectively from single 2DVD, RD-80, and our retrieval. It is found that, from both Case A and Case B, the mean value from RD-80 is smaller than that from 2DVD. This may be due to the different ranges of measurable drop sizes from two types of disdrometers (0.3 -

5.4 mm for RD 80, while 0.1 to 10 mm for 2DVD). More statistics can be also found in Table 2. Overall, the mean differences are still within the retrieval uncertainties for this case.

Secondly, the CLWP is retrieved using "Attenuation Algorithm" introduced in section 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 column. When cloud droplets and raindrops coexist below the melting base, the mean values are 0.11 kg m<sup>-2</sup> and 1.64 kg m<sup>-2</sup> for RLWP and CLWP, and the corresponding LWP below the melting layer is 0.76 kg m<sup>-2</sup>. While when only raindrops exist below the melting base, there is no CLWP (CLWP =0), and the RLWP and LWP are the same (with average of 0.34 kg m<sup>-2</sup>). It is noticed that even though the occurrence of CLWP is low (11%) 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 mean value of LWP is 0.37 kg m<sup>-2</sup> for all the samples in Fig. 5c. The blue uncertainty bars in Figure 5c show the retrieved CLWP uncertainty with assuming both of the uncertainties of attenuation and total rain rate are 30%. Due to the variations of the attenuation and total rain rate with time, the estimated uncertainties of CLWP varies point to point. More details about the estimation of CLWP are in Appendix B.

#### **4.2 Statistical Results**

Box and whisker plots of retrieved RLWP, CLWP and LWP for situations (I), (II) and all samples during MC3E are shown in Fig. 6. The horizontal orange and red dashed lines indicate the median and mean, boundaries of the box represent the first and third quartiles, and the whiskers are the 10<sup>th</sup>- and 90<sup>th</sup> -percentiles. During MC3E, a total of 13 hours of stratiform rain were observed by VPRs at the ARM SGP Climate Research Facility, in which 91% and 9% the samples are categorized into the situations (I) and (II), respectively. The mean RLWPs are 0.32 kg m<sup>-2</sup> and

0.10 kg m<sup>-2</sup> for the situations (I) and (II). There are a substantial amount of small cloud droplets sustaining in the rain layer and having not yet converted to larger raindrops, which may partially explain smaller RLWP in the situation (II). The mean value of surface rain rate is 2.06 mm hr<sup>-1</sup> when cloud droplets exist, which is also smaller than the mean value (2.38 mm hr<sup>-1</sup>) in the rainonly situation. The mean CLWP in the situation (II) is as large as ~0.56 kg m<sup>-2</sup> even though their occurrence is very low (9%), which is much larger than mean RLWP in the liquid layer. The LWP from the situation (II) (0.66 kg m<sup>-2</sup>) is much larger than the mean LWP from the situation (I) (0.32 kg m<sup>-2</sup>), which is primarily contributed by cloud droplets. The overall mean LWP for stratiform rain during MC3E is 0.34 kg m<sup>-2</sup>.

We also processed the ARM MWR retrieved LWPs during MC3E and compared them with our retrievals as illustrated in Fig. 7a. The corresponding LWP uncertainties are also provided as the grey error bar for each retrieval with rain rate indicated by colors. It is noticed that the MWR has no LWP estimation when the rain rate is large. The MWR-retrieved LWPs increase with increased rain rate, and the retrievals from MWR are much larger than the new LWP retrievals at high rate rates. The newly retrieved LWPs weakly correlate with rain rates, and most values are less than 1.0 kg m<sup>-2</sup>, especially at high rain rates. The MWR retrieved LWPs increase with rain rate generally. The increase of retrieved LWP with rain rate from MWR is possibly due to the "wet radome" effect (Cadeddu et al., 2017). In addition to the issue from standing water on the radome, the extinctions due to raindrops also affect MWR retrievals. The extinction for rain is much larger than that for cloud (Sheppard, 1996), and thus, the small amount of rain water could enhance the measured brightness temperature significantly. More details of extinctions and brightness temperature calculations are shown in Appendix B. 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<sup>1</sup>). The differences of LWPs from MWR and this study are shown in Fig. 7b. The LWP differences increase with increased rain rate. The LWP differences between MWR retrieval and this study could be caused by the following reasons. 1) MWR-retrieved LWP represents the entire 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 represents the LWP below the melting base. As Battaglia et al (2003) pointed out the brightness temperature generally increases if mixed-phase precipitation is included. 2) The MWR radome was wet during the raining periods and the deposition of raindrops on the radome can cause a large increase in the measured brightness temperatures (Cadeddu et al., 2017). 3) Large extinctions due to rain drops would affect MWR retrievals. 4) Uncertainties exist in the retrieved LWP from this study.

### **5. Summary and Conclusions**

LWP is a critical parameter for studying clouds, precipitation, and their life cycles. LWP can be retrieved from microwave radiometer measured brightness temperatures during cloudy and light precipitation conditions. However, MWR-retrieved LWPs are questionable under moderate and heavy precipitation conditions due to the "wet radome" and the large extinction in the unit volume caused by large raindrops. LWPs below the melting base in stratiform precipitation systems are estimated, which include both RLWP and CLWP. The measurements used in this study are mainly from two VPRs, 35-GHz from ARM and 3-GHz from NOAA 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 13 hours samples during MC3E show that the occurrence of cloud droplets below the melting base is low (9%), while the CLWP value can be up to 0.56 kg m<sup>-2</sup>, which is much larger than the RLWP (0.10 kg m<sup>-2</sup>). When only raindrops exist below the melting base, the averaged RLWP value is 0.32 kg m<sup>-2</sup>, which is much larger than the mean RLWP in the cloud droplets and raindrops coexisted situation. Our retrievals are also compared with ARM MWR retrieved LWPs. It is noticed that the MWR has no LWP estimation when the rain rate is large. The MWR-retrieved LWPs increase with increased rain rate, and the retrievals from MWR are much larger than our LWP retrievals at high rate rates. The LWP differences are fully discussed.

Reliable retrievals of RLWC and RLWP are critical for model evaluation and improvement, as RLWC (rain mixing ratio) is an important prognostic variable in weather and climate models. Furthermore, the retrievals in the whole rain layer would be useful to understand the microphysical processes (i.e., condensation, evaporation, autoconversion, and accretion etc.) and have great potential to improve model parametrizations in the future. Overall, the LWP (CLWP and RLWP) retrievals derived in this study can be used to evaluate the models that separately predict cloud and precipitation and contribute comprehensive information to study cloud-to-precipitation transitions. Note that the attenuation by liquid is more profound at 94 GHz and the ratio of attenuations by

liquid clouds and by rain is larger at 94 GHz compared to 35 GHz (Matrosov, 2009). Thus, using 94 GHz (W-band) radar measurements to develop a retrieval algorithm may be promising if the W-band radar signals are not fully attenuated. In addition, analyzing collocated multiple-frequency VPRs would also better define the uncertainties of retrievals made with collocated radar operating at different frequency pairs.

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

Retrieving RLWC and other rain microphysical properties (i.e., drop size and rain rate) is based on the mathematics of DSD radar reflectivity-weighted velocity spectral density  $S_{DSD}^{\lambda}$ [(mm<sup>6</sup> m<sup>-3</sup>) (m s<sup>-1</sup>)<sup>-1</sup>], which is a product of radar raindrop backscattering cross section  $\sigma_b^{\lambda}$ (D) (mm<sup>2</sup>) and DSD number concentration N<sub>DSD</sub>(D) (mm<sup>-1</sup> m<sup>-3</sup>):

418 
$$S_{DSD}^{\lambda}(v_z) = \left[\frac{\lambda^4}{\pi^5 |K_w|^2} \sigma_b^{\lambda}\right] N_{DSD}(D) \frac{dD}{dv_z}. \tag{A1}$$

- The  $\frac{dD}{dv_z}$  [mm (m s<sup>-1</sup>)<sup>-1</sup>] is used as a coordinate transformation from diameter to velocity, where
- $v_z$  (m s<sup>-1</sup>) is the raindrop terminal velocity of diameter D (mm) at altitude z.  $\lambda$  is the wavelength
- 421 of radar.  $|K_w|^2$  equals 0.93 and it is the dielectric factor.
- The  $N_{DSD}(D)$  can be expressed as a normalized gamma shape distribution with three parameters
- 423 (Leinonen et al., 2012):

$$N_{DSD}(D; N_{w}, D_{m}, \mu) = N_{w}f(D; D_{m}, \mu), \tag{A2}$$

425 where

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

- N<sub>w</sub> is the scaling parameter,  $\mu$  is a shape parameter,  $\Gamma(x)$  is the Euler gamma function, and D<sub>m</sub> is
- 428 a mean mass-weighted raindrop diameter estimated from the ratio of the fourth to third DSD
- 429 moments:

430 
$$D_{m} = \frac{M_{4}}{M_{3}} = \frac{\int_{D_{min}}^{D_{max}} N_{DSD}(D)D^{4}dD}{\int_{D_{min}}^{D_{max}} N_{DSD}(D)D^{3}dD}.$$
 (A4)

- 431 where  $D_{\text{min}}$  and  $D_{\text{max}}$  represent the minimum and maximum diameters in the distribution,
- 432 respectively.
- The intrinsic (non-attenuation) reflectivity factor and the mean velocity and the spectrum
- variance are the zeroth, first, and second reflectivity-weighted velocity spectrum moments:

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

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

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

- 438 where  $v_i$  is the discrete velocities and  $\Delta v$  is velocity resolution in the integration.
- The Doppler Velocity Difference (DVD) is defined as

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

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

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

444 
$$RR(mm hr^{-1}) = \frac{6\pi}{10^4} \sum_{D_{min}}^{D_{max}} N_{DSD}(D, N_w, D_m, \mu) D_i^3 v_z(D_i) \Delta D. \tag{A10}$$

- In addition, there are two newly defined radar-related parameters ( $\alpha$  and  $\beta$ ), which are also
- 446 dependent on  $D_m$  and  $\mu$  only:

$$\alpha=10 \log_{10}(Z_{DSD}^{3GHz}/RLWC) \tag{A11}$$

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

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

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

The variabilities of 3-GHz and 35-GHz VPR observations within each 1-minute/60-meter bin are regarded as the measurement uncertainties and will be propagated through the retrieval to

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

The uncertainties of RLWP are estimated based on the uncertainties of RLWC. More specifically, we first estimated the RLWC uncertainties at each height level, and then we calculated the ratios of RLWC uncertainties to the mean retrieved RLWCs at each height level, which represent percentage values of retrieval uncertainties. Finally, we calculated the mean ratio of the uncertainties in the whole liquid layer below melting base and regarded this mean ratio as the uncertainty of RLWP.

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. However, it is very difficult to quantify this type of retrieval uncertainty. In this study, we further compared our retrievals with independent surface disdrometers measurements to estimate the uncertainties of retrievals. Also, when both radars are observing at Rayleigh scattering for small raindrops, the reflectivity-weighted radial velocities for these particles should be the same. In order to have a difference in radial velocity during the retrieval, large droplets must exist. The

maximum diameters in drop size distribution measured from disdrometer for all the stratiform cases during MC3E are investigated. It is found that the occurrence of small-droplets-only (maximum diameter <1.3 mm) is very low (less than 3%). Thus, it will not have a significant impact on the retrieval results. Notice that this algorithm is not suitable for strong convective rain due to the wind shear and strong turbulence as well as severe attenuation and extinction of the Kaband radar signal.

### **Appendix B: CLWP Uncertainty**

CLWP can be simplified and estimated as following equation:

$$CLWP = \frac{A - 2 C R_{total}}{2 B}. \tag{B1}$$

The attenuation (A) is estimated by comparing the drop in Ka-band reflectivity with the unattenuated S-band reflectivity. The rain attenuation is estimated by the rain attenuation coefficient (C) multiplied by the total rain rate ( $R_{total}$ ). C and B are the coefficients of rain and cloud water attenuation with values of  $\sim 0.26$  dB /km /mm hr<sup>-1</sup> and  $\sim 0.87$  dB / kg m<sup>-2</sup>, respectively. The influence of temperature uncertainty in B on the retrieval error is minor compared to the uncertainties of the total attenuation (A) and total rain rate ( $R_{total}$ ) (Matrosov 2010). The uncertainty of CLWP is calculated as

$$\Delta CLWP = \sqrt{\left(\frac{\partial CLWP}{\partial A} \times \Delta A\right)^2 + \left(\frac{\partial CLWP}{\partial R_{total}} \times \Delta R_{total}\right)^2}$$
 (B2)

512 
$$\Delta CLWP = \sqrt{\left(\frac{1}{2B} \times A \times \Delta a\right)^2 + \left(-\frac{c}{B} \times R_{total} \times \Delta r\right)^2}$$
 (B3)

For given uncertainties of attenuation ( $\Delta a$ ) and total rain rate ( $\Delta r$ ), the uncertainty of CLWP can be calculated based on equation (B3).

# Appendix C: Calculations of Extinction and Brightness Temperature at Microwave

#### **Radiometer Channels**

To better explain the "overestimation" issue of retrieved LWP from microwave radiometer, several examples are given in this appendix. Firstly, we calculated the extinction cross section per volume as a function of the drop equivolume diameter for the two frequencies in MWR (23.8 GHz and 31.4 GHZ) with a T-matrix method (Figure B). It is clearly shown that the extinction cross section is increasing with the diameter when the diameter is smaller than 3 mm. This indicates the extinction (cross section) for rain drops (diameter  $> \sim 50$  um) is much larger than that for cloud droplets (diameter  $< \sim 50$  um). Secondly, we calculated the extinction coefficient as a function of RLWC for populations with three different drop size distributions (DSDs). The DSDs are modeled according to the exponential Marshall and Palmer (MP) distribution  $N(D) = N_0 e^{AD}$ , where  $N_0$ =8000 m<sup>-3</sup> mm<sup>-1</sup>.  $N_0$  is changed to 4000 and 32000 m<sup>-3</sup> mm<sup>-1</sup> to represent thunderstorm and drizzle DSDs. More details of the DSDs please see Battaglia et al. (2009). Figure C clearly shows the extinctions of cloud and rain also is DSD-dependent. For example, at 31.4 GHz, even though the RLWC is the same, the extinctions are much larger from the precipitation with the thunderstorms and MP DSDs than the extinctions from light precipitation with the drizzle DSD.

In addition, the brightness temperatures at 23.8 and 31.4 GHz channels are calculated using the MicroWave Radiative Transfer (MWRT) model. Five different sensitivity tests are generated with five combinations of CLWP and RLWP values (Table A). Table A lists the results and clearly demonstrates that the brightness temperatures in channels increase with increased cloud water amount (larger CLWP) and the rain water amount (larger RLWP). Comparing the results from

539	test #2 and #3, it is clearly seen that the brightness temperatures contributed by rain drops are 31
540	and 51 K more than that contributed by cloud droplets at the frequencies of 23.8 and 31.4 GHz,
541	even though their LWPs are the same (1 kg m <sup>-2</sup> ) in these two tests.
542	
543	Data availability. NOAA vertical profile radar datasets are publicly available in the DOE archives
544	(http://iop.archive.arm.gov/arm-iop/2011/sgp/mc3e/williams-s_band/). KAZR ARSCL data are
545	available on the DOE ARM archive (https://www.arm.gov/data).
546	
547	Author contributions. JT and XD conceptualized this study. JT developed the retrievals, analyzed
548	the data and drafted the manuscript. XD contributed the interpretation of the results and the editing
549	of the paper. CW contributed the discussions to the methodology in rain and cloud microphysics
550	retrieval and the editing of the paper. BX and PW contributed the discussions of the results.
551	
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553	
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561	

62	References
563	Ackerman, T. P., and Stokes, G. M: The Atmospheric Radiation Measurement Program. Phys
564	Today, 56,38–44, doi:10.1063/1.1554135, 2003
565	Battaglia, A., Saavedra, P., T. Rose, and Simmer, C.: Characterization of precipitating clouds by
566	ground-based measurements with the triple-frequency polarized microwave radiometer
567	ADMIRARI, J. Appl. Meteorol., 49(3), 394–414, 2009
568	Battaglia, A., C. Kummerow, D. Shin, and C. Williams, 2003: Constraining Microwave
569	Brightness Temperatures by Radar Brightband Observations. J. Atmos. Oceanic
570	Technol., 20, 856-871, https://doi.org/10.1175/15200426(2003)020<0856:CMBTBR>2.0
571	.CO;2
572	Cadeddu, M. P., Liljegren, J. C., and Turner, D. D.: The Atmospheric radiation measurement
573	(ARM) program network of microwave radiometers: instrumentation, data, and retrievals
574	Atmos. Meas. Tech., 6, 2359-2372, https://doi.org/10.5194/amt-6-2359-2013, 2013
575	Cadeddu, M. P., Marchand, R., Orlandi, E., Turner, D. D. and Mech, M. (2017). Microwave
576	Passive Ground-Based Retrievals of Cloud and Rain Liquid Water Path in Drizzling
577	Clouds: Challenges and Possibilities, IEEE Trans. Geosci. Remote Sens., vol. 55, no. 11
578	pp. 6468-6481, doi: 10.1109/TGRS.2017.2728699
579	Crewell, S., and Löhnert, U. (2003). Accuracy of cloud liquid water path from ground-based
580	microwave radiometry 2. Sensor accuracy and synergy, Radio Sci., 38, 8042,
581	doi:10.1029/2002RS002634, 3.
582	Dubrovina, L. S.: Cloudness and precipitation according to the data of airplane soundings
583	Gidrometeoizdat, Leningrad (in Russian), 218 pp,1982

584 Ellis, S. M., and Vivekanandan, J.: Liquid water content estimates using simultaneous S and Ka 585 band radar measurements, Radio Sci., 46, RS2021, doi:10.1029/2010RS004361, 2011 586 Fabry, F. and Zawadzki, I.: Long-Term Radar Observations of the Melting Layer of Precipitation 587 and Their Interpretation. J. Atmos. Sci., 52, 838–851, https://doi.org/10.1175/1520-588 0469(1995)052<0838:LTROOT>2.0.CO;2, 1995 589 Fan, J., Liu, Y.-C., Xu, K.-M., North, K., Collis, S., Dong, X, and Ghan, S. J.: Improving 590 representation of convective transport for scale-aware parameterization: 1. Convection and 591 cloud properties simulated with spectral bin and bulk microphysics, Journal of Geophysical 592 Research: Atmosphere, 120, 3485–3509, https://doi.org/10.1002/2014JD022142, 2015 593 Feng, Z., Dong, X. Q., Xi, B. K., Schumacher, C., Minnis, P., and Khaiyer, M.: Top-of-atmosphere 594 radiation budget of convective core/stratiform rain and anvil clouds from deep convective 595 Geophysical Research, 116, D23202. systems. Journal of https://doi.org/ 596 10.1029/2011JD016451, 2011 597 Feng, Z., Leung, L. R., Houze, R. A., Jr., Hagos, S., Hardin, J., Yang, Q., Han, B. and Fan, J.: 598 Structure and evolution of mesoscale convective systems: Sensitivity to cloud 599 microphysics in convection-permitting simulations over the United States. Journal of 600 10, Advances in Modeling Earth Systems, 1470–1494. 601 https://doi.org/10.1029/2018MS001305, 2018 602 Giangrande, S. E., Collis, S., Theisen, A. K., and Tokay, A.: Precipitation estimation from the 603 ARM distributed radar network during the MC3E campaign, J. Appl. Meteor. Climatol., 604 doi:10.1175/JAMC-D-13-0321.1, 2014 605 Jensen, M.P., Petersen, W. A., Bansemer, A., Bharadwaj, N., Carey, L. D., Cecil, D. J, and Zipser, 606 E. J.: The Midlatitude Continental Convective Clouds Experiment (MC3E), Bulletin of the

607	American Meteorological Society. 151221073208006. https://doi.org/10.1175/BAMS-Details.				
608	14-00228.1, 2015				
609	Leinonen, J., Moisseev, D., M. Leskinen, M., and W.A. Petersen, W.A.: A Climatology of				
610	Disdrometer Measurements of Rainfall in Finland over Five Years with Implications for				
611	Global Radar Observations. J. Appl. Meteor. Climatol., 51, 392-404,				
612	https://doi.org/10.1175/JAMC-D-11-056.1, 2012				
613	Leinonen, J.: High-level interface to T-matrix scattering calculations: architecture, capabilities and				
614	limitations, Opt. Express, vol. 22, issue 2, 1655-1660 doi: <u>10.1364/OE.22.001655</u> , 2014				
615	Liljegren, J. C., Clothiaux, E. E., Mace, G. G., Kato, S., and Dong, X.: A new retrieval for cloud				
616	liquid water path using a ground-based microwave radiometer and measurements of cloud				
617	temperature, J. Geophys. Res., 106(D13), 14485-14500, doi:10.1029/2000JD900817				
618	2001				
619	Lebsock, M.D., L'Ecuyer, T.S. and Stephens, G.L.: Detecting the Ratio of Rain and Cloud Water				
620	in Low-Latitude Shallow Marine Clouds. J. Appl. Meteor. Climatol., 50, 419-432				
621	https://doi.org/10.1175/2010JAMC2494.1, 2011				
622	Matrosov, S. Y.: Assessment of radar signal attenuation caused by the melting hydrometeor layer				
623	IEEE Trans. Geosci. Remote Sens., 46,1039–1047 doi: 10.1109/TGRS.2008.915757, 2008				
624	Matrosov, S. Y.: A method to estimate vertically integrated amounts of cloud ice and liquid and				
625	mean rain rate in stratiform precipitation from radar and auxiliary data, J. Appl. Meteor				
626	Climatol., 48, 1398–1410, doi:10.1175/2009JAMC2196.1, 2009				
627	Matrosov, S. Y.: Synergetic use of millimeter- and centimeter-wavelength radars for retrievals of				
628	cloud and rainfall parameters, Atmos. Chem. Phys., 10, 3321-3331,				
629	https://doi.org/10.5194/acp-10-3321-2010, 2010				

630 Mazin, I. P. (Ed.): Clouds and the Cloudy Atmosphere. Gidrometeoizdat, Leningrad, 648 pp, 1989. 631 Saavedra, P., Battaglia, A., and Simmer, C.: Partitioning of cloud water and rainwater content by 632 ground-based observations with the Advanced Microwave Radiometer for Rain 633 Identification (ADMIRARI) in synergy with a micro rain radar, J. Geophys. Res., 117, 634 D05203, doi:10.1029/2011JD016579, 2012 635 Sassen, K., Campbell, J. R., Zhu, J., Kollias, P., Shupe, M., and Williams, C.: Lidar and Triple-636 Wavelength Doppler Radar Measurements of the Melting Layer: A Revised Model for and 637 Dark-Brightband Phenomena. J. Appl. Meteor., 44, 301–312, 638 https://doi.org/10.1175/JAM-2197.1, 2005 639 Sheppard, B.E.: Effect of Rain on Ground-Based Microwave Radiometric Measurements in the 640 20-90-GHz Range. J. Atmos. *Oceanic* Technol., 13, 1139–1151, 641 https://doi.org/10.1175/1520-0426(1996)013<1139:EOROGB>2.0.CO;2, 1996 642 Tian, J., Dong, X., Xi, B., Wang, J., Homeyer, C. R., McFarquhar, G. M., and Fan J.: Retrievals 643 of ice cloud microphysical properties of deep convective systems using radar 644 measurements, Journal of Geophysical Research: Atmosphere., 121,10,820–10,839, 645 https://doi.org/10.1002/2015JD024686, 2016 646 Tian, J., Dong, X., Xi, B., Minnis, P., Smith, W. L., Jr, Sun-Mack, S., Thieman, M., Wang, J.: 647 Comparisons of ice water path in deep convective systems among ground-based, GOES, 648 and CERES-MODIS retrievals. Journal of Geophysical Research: Atmospheres, 123, 649 1708–1723. https://doi.org/10.1002/2017JD027498, 2018 650 Tridon, F., and Battaglia, A.: Dual-frequency radar Doppler spectral retrieval of rain drop size 651 distributions and entangled dynamics variables, J. Geophys. Res. Atmos., 120, 5585–5601, 652 doi:10.1002/2014JD023023, 2015

653	Iridon, F., Battaglia, A., and Kollias, P.: Disentangling Mie and attenuation effects in rain using a
654	Ka-W dual-wavelength Doppler spectral ratio technique, Geophys. Res. Lett., 40, 5548
655	5552, doi:10.1002/2013GL057454, 2013
656	Tridon, F., Battaglia, A., Luke, E., and Kollias, P.: Rain retrieval from dual-frequency radar
657	Doppler spectra: validation and potential for a 25 midlatitude precipitating case-study,
658	Quarterly Journal of the Royal Meteorological Society, 143, 1364-1380, 2017.
659	Turner, D. D., Clough, S. A., Liljegren, J. C., Clothiaux, E. E., Cady-Pereira, K. E., and Gaustad,
660	K. L.: Retrieving liquid water path and precipitable water vapor from the Atmospheric
661	Radiation Measurement (ARM) microwave radiometers. IEEE Trans. Geosci. Remote
662	Sens., 45, 3680–3690, 2007
663	Wentz, F.J. and Spencer, R.W.: SSM/I Rain Retrievals within a Unified All-Weather Ocean
664	Algorithm. J. Atmos. Sci., 55, 1613–1627, https://doi.org/10.1175/1520-
665	0469(1998)055<1613:SIRRWA>2.0.CO;2, 1998
666	Williams, C. R.: Reflectivity and liquid water content vertical decomposition diagrams to diagnose
667	vertical evolution of raindrop size distributions, J. Atmos. Oceanic Technol.,
668	doi:10.1175/JTECH-D-15-0208.1, 2016
669	Williams, C. R., Beauchamp, R. M., and Chandrasekar, V.: Vertical air motions and raindrop size
670	distributions estimated from mean Doppler velocity difference from 3- and 35-GHz
671	vertically pointing radars. IEEE Trans. Geosci. Remote Sens., 54, 6048-6060,
672	https://doi.org/10.1109/ TGRS.2016.2580526, 2016
673	Xu, W.: Precipitation and convective characteristics of summer deep convection over east Asia
674	observed by TRMM, Monthly Weather Review., 141, 1577-1592.
675	https://doi.org/10.1175/MWR-D-12-001

**Table 1**. Acronyms and Abbreviations Used in This Study

Acronyms and Abbreviations	Full Name
2DVD	Two-dimensional video disdrometer
A	Total two-way attenuation of 35-GHz VPR signals
ARSCL	Active remote sensing of clouds
ARM	Atmospheric Radiation Measurement
В	coefficients for cloud water attenuation
C	coefficients for rainfall attenuation
CLWP	Cloud liquid water path
D	Raindrop diameter
$D_{m}$	Mean mass-weighted raindrop diameter
$D_{max}$	Maximum diameters in the size distribution
$D_{\min}$	Minimum diameters in the size distribution
DOE	Department of Energy
DSD	Drop size distribution
DVD	Doppler velocity difference
G	Two-way gaseous absorption
IWC	Ice water content
KAZR	Ka-band ARM zenith radar
LUT	Looking up table
LWP	Liquid water path
MB	Base of melting layer
MC3E	Mid-latitude continental convective clouds experiment
MMCR	Millimeter-wavelength cloud radar
MWR	Microwave radiometer
$N_{ m DSD}$	Number concentration
$N_0$	Intercept of ice particle size distribution
NOAA	National Oceanic and Atmospheric Administration
$N_{ m w}$	Scaling parameter in the drop size distribution
RLWP	Rain liquid water path
$R_{m}$	Layer-mean rain rate
RR	Rain rate

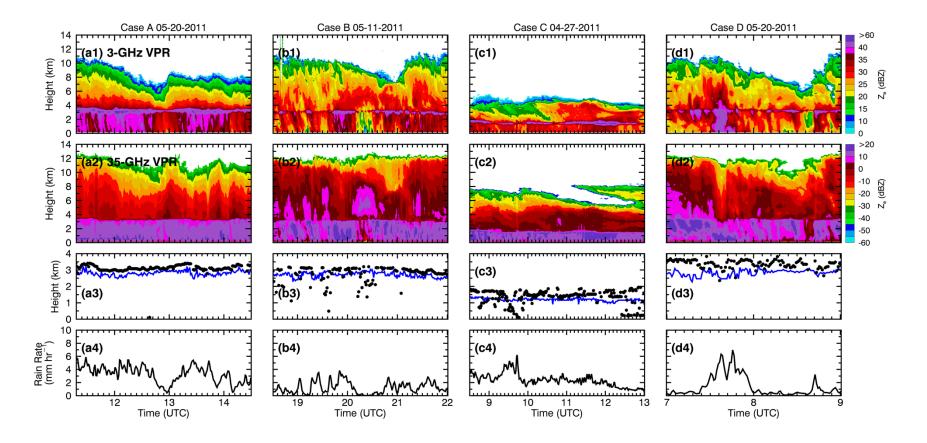
$S_{ m DSD}^{\lambda} \ v_{ m DSD}^{\lambda}$	Radar reflectivity-weighted velocity spectral density
νηςη	First reflectivity-weighted velocity spectrum moments
030	represent the mean velocity
٧z	Raindrop terminal velocity
$\mathbf{Z}_{ extsf{DSD}}^{\lambda}$	Zeroth reflectivity-weighted velocity spectrum moments
030	represent the intrinsic (non-attenuation) reflectivity factor
$\Gamma(\mathbf{x})$	Euler gamma function
	Radar wavelength
$\sigma_{ m b}^{\lambda}$	Raindrop backscattering cross section
- U	Shape parameter

**Table 2.** Statistics (mean differences, 95% confidence interval of mean differences, RMSEs) of  $D_m$ , RR between this study (RET) and disdrometers (2DVD, RD-80) for Case A and Case B

	Mean Differences (95% confidence interval)	RMSE
Case A: D <sub>m</sub> (RET, 2DVD) (mm)	0.04 (-0.07, -0.01)	0.24
Case A: RR (RET, RD-80) (mm hr <sup>-1</sup> )	-0.45 (-0.57, -0.33)	0.96
Case A: RR (RET, 2DVD) (mm hr <sup>-1</sup> )	-0.61( -0.77, -0.43)	1.38
Case B: D <sub>m</sub> (RET, 2DVD) (mm)	0.10 (-0.14, -0.07)	0.27
Case B: RR (RET, RD-80) (mm hr <sup>-1</sup> )	0.40 (0.19, 0.60)	1.51
Case B: RR (RET, 2DVD) (mm hr <sup>-1</sup> )	0.30(0.09, 0.52)	1.58

Table A. The brightness temperatures (TB) at 23.8 and 31.4 GHz for different assumptions of CLWP and RLWP values.

Sensitivity Test	CLWP (kg m <sup>-2</sup> )	RLWP (kg m <sup>-2</sup> )	TB at 23.8 GHz	TB at 31.4 GHz
#1	2	0	197.20	196.28
#2	1	0	186.34	177.49
#3	0	1	217.28	228.20
#4	0	2	254.51	272.09
#5	1	1	225.37	239.88



**Figure 1.** Time series of (a1) radar reflectivity ( $Z_e$ ) from NOAA 3-GHz vertical pointing radar (VPR), (b1) radar reflectivity from ARM 35-GHz VPR, (c1) melting base (blue lines) and cloud base (black dots), and (d1) rain rates from RD-80 surface disdrometer measurement for Case A (20 May 2011, 11:20 – 14:30 UTC); (b1)-(b4) for Case B (11 May 201, 18:30 – 22:00 UTC); (c1)-(c4) for Case C (27 April 2011, 8:30 – 13:00 UTC); (d1)-(d4) for Case D (20 May 2011, 7:00 – 9:00 UTC). Note that the ranges of radar dBZ values are different in 3-GHz and 35-GHz radars.

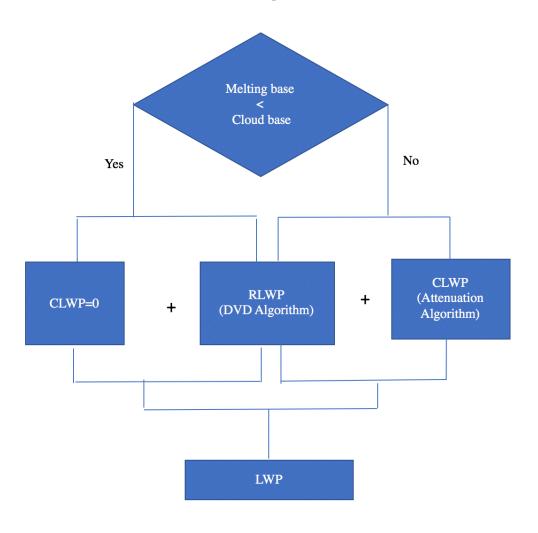
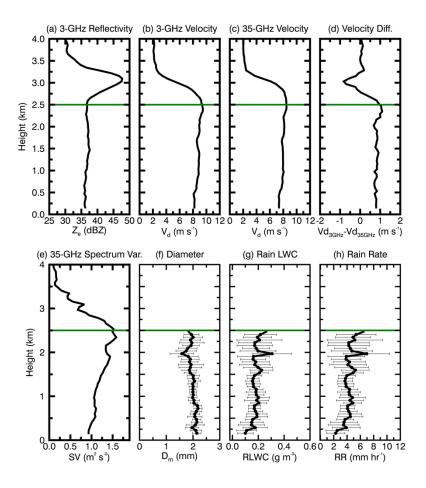
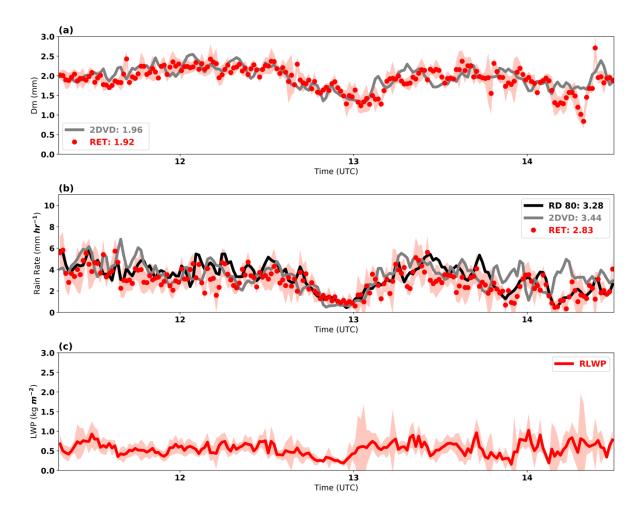


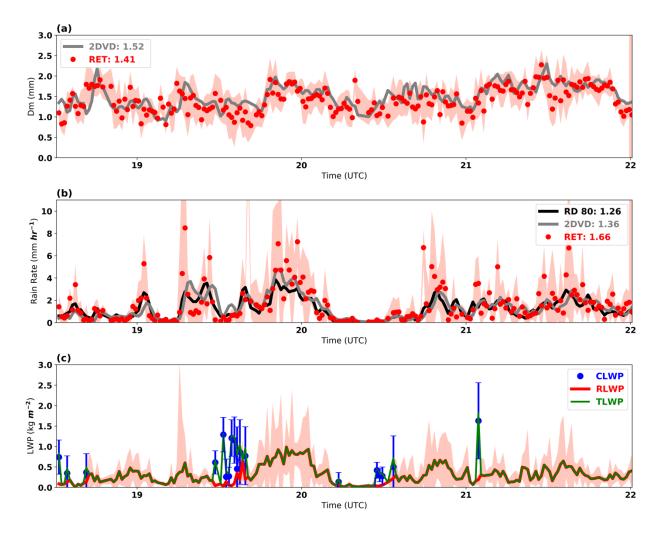
Figure 2. Algorithm flowchart to retrieve liquid water path (LWP) below melting base.



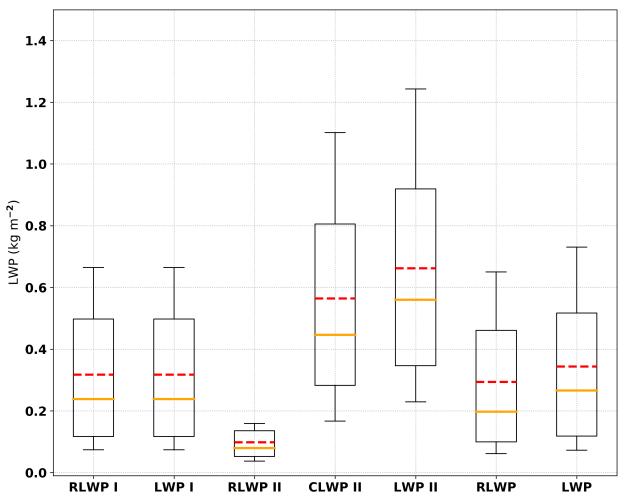
**Figure 3**. An example of illustrating the Doppler Velocity Differences (DVD) retrieval algorithm at 13:40 UTC on May 20, 2011. The inputs of the DVD retrieval algorithm are: (a) 3-GHz vertical pointing radar reflectivity factor ( $Z_e$ ), (b) 3-GHz radar Doppler velocities ( $V_d$ ), (c) 35-GHz radar 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.



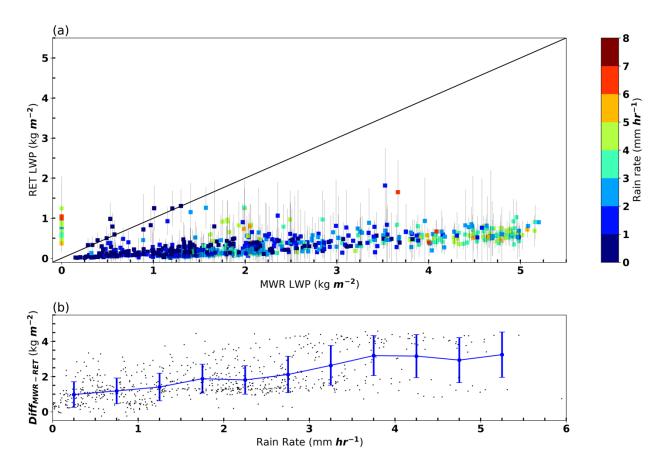
**Figure 4.** Time series of (a) retrieved (RET) (red dots) and 2DVD surface disdrometer estimated (grey line)  $D_m$ , (b) RET (red line), 2DVD (grey line) and RD-80 (black line) surface disdrometer rain rate estimates, and (c) retrieved rain liquid water path (RLWP, red dots) for Case A (May 20, 2011.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



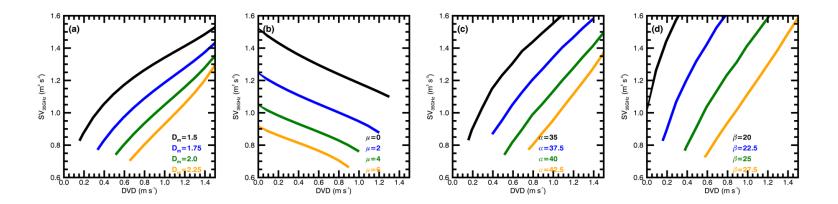
**Figure 5.** Time series of (a) retrieved (RET) (red dots) and 2DVD surface disdrometer estimated (grey lines)  $D_m$ , (b) RET (red dots), 2DVD (grey line) and RD-80 (black line) surface disdrometer rain rate estimates, and (c) rain liquid water path (RLWP, red line), cloud liquid water path (CLWP,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



**Figure 6**. Box and whisker plots of retrieved RLWP, CLWP and LWP for situation (I), (II) and all samples. The horizontal orange line within the box indicates the median, boundaries of the box represent the 25<sup>th</sup>- and 75<sup>th</sup> -percentile, and the whiskers indicate the 10<sup>th</sup>- and 90<sup>th</sup> -percentile values of the results. The red dash lines represent the mean values.



**Figure 7**. (a) Comparisons between LWP from microwave radiometer (MWR, in x-axis) and LWP retrievals from this study (RET, in y-axis, with estimated uncertainty in gray lines). The rain rates are indicated by colors. (b) the LWP differences between two estimations (MWR- RET), shown as a function of rain rate.



**Figure A.** Comparisons of (a) mass-weighted mean diameter  $D_m$  (mm), (b) shape parameter  $\mu$ , (c) parameter  $\alpha$ = 10 log( $Z_{3GHz}/RLWC$ ), and (d) parameter  $\beta$  = 10 log( $Z_{3GHz}/RR$ ) calculated as functions of Doppler velocity difference (DVD) and spectrum variance at 35 GHz (SV<sub>35GHz</sub>). Note that the units of RLWC and RR are g m<sup>-3</sup> and mm hr<sup>-1</sup>.

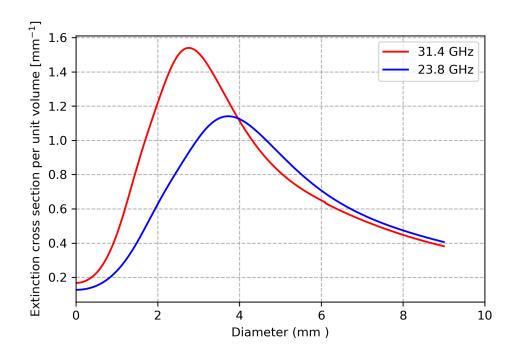
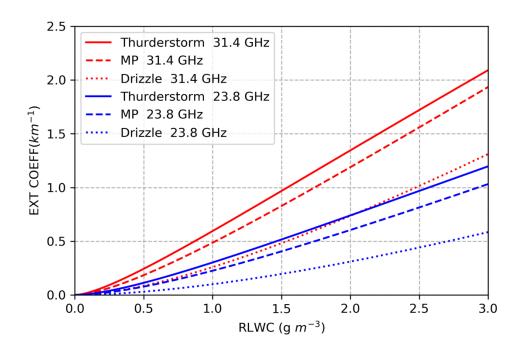


Figure B. The extinction cross section per unit volume as a function of the drop equivolume diameter for the two frequencies in MWR (23.8 GHz and 31.4 GHZ).



**Figure C.** The extinction coefficient as a function of RLWC for precipitations with three different drop size distributions (DSDs), which are for heavy precipitation (thunderstorm), moderate precipitation (MP) and drizzle precipitation (drizzle).