How do changes in warm-phase microphysics affect deep convective clouds?

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Abstract. Understanding aerosol effects on deep convective clouds and the derived effects on the radiation budget and rain patterns can largely contribute to estimations of climate uncertainties. The challenge is difficult in part because key microphysical processes in the mixed and cold phases are still not well understood. For deep convective clouds with a warm base, understanding aerosol effects on the warm processes is extremely important as they set the initial and boundary conditions for the cold processes. Therefore, the focus of this study is the warm phase, which can be better resolved. The main question is: “How do aerosol-derived changes in the warm phase affect the properties of deep convective cloud systems?” To explore this question, we used a weather research and forecasting (WRF) model with spectral bin microphysics to simulate a deep convective cloud system over the Marshall Islands during the Kwajalein Experiment (KWJAE). The model results were validated against observations, showing similarities in the vertical profile of radar reflectivity and the surface rain rate. Simulations with larger aerosol loading resulted in a larger total cloud mass, a larger cloud fraction in the upper levels, and a larger frequency of strong updrafts and rain rates. Enlarged mass both below and above the zero temperature level (ZTL) contributed to the increase in cloud total mass (water and ice) in the polluted runs. Increased condensation efficiency of cloud droplets governed the gain in mass below the ZTL, while both enhanced condensational and depositional growth led to increased mass above it. The enhanced mass loading above the ZTL acted to reduce the cloud buoyancy, while the thermal buoyancy (driven by the enhanced latent heat release) increased in the polluted runs. The overall effect showed an increased upward transport (across the ZTL) of liquid water driven by both larger updrafts and larger droplet mobility.

These aerosol effects were reflected in the larger ratio between the masses located above and below the ZTL in the polluted runs. When comparing the net mass flux crossing the ZTL in the clean and polluted runs, the difference was small. However, when comparing the upward and downward fluxes separately, the increase in aerosol concentration was seen to dramatically increase the fluxes in both directions, indicating the aerosol amplification effect of the convection and the affected cloud system properties, such as cloud fraction and rain rate.

1 Introduction

The overall effect of aerosol on clouds is one of the most challenging questions in climate research (IPCC, 2013). Within this domain, aerosol interactions with convective clouds and the derived effects on rain patterns are especially difficult to determine due to the tight coupling of dynamic and microphysical processes (Altaratz et al., 2014; Fan et al., 2016; Tao et al., 2012). Environmental thermodynamic conditions determine the overall potential for cloud and precipitation formation, whereas aerosol properties dictate how efficiently the cloud will develop within this given environmental potential (Dagan et al., 2015a, b). Aerosols act as cloud condensation nuclei (CCN) and ice nuclei, thus affect-
ing the initial size distributions of cloud droplets and ice crystals, respectively. Higher aerosol loading means an increased amount of CCN and therefore the activation of more, albeit smaller, droplets with a narrower size distribution (Squires and Twomey, 1961; Twomey, 1977; Warner and Twomey, 1967). More activated droplets at the initial stage of a cloud lifetime will provide a larger collective surface area for condensation and therefore more efficient consumption of the available supersaturation (Dagan et al., 2015a; Koren et al., 2014; Pinsky et al., 2013; Seiki et al., 2014). Moreover, efficient condensation in polluted cases is prolonged because the onset of the collision–coalescence process is delayed (Albrecht, 1989; Saleeby et al., 2010; Squires, 1958). This is also reflected in delayed rain onset (Berg et al., 2008; Dagan et al., 2015b; Jin and Shepherd, 2008; Suzuki et al., 2008; Xue et al., 2008).

Deep convective cloud invigoration by aerosols has been shown in observational (Andreae et al., 2004; Koren et al., 2005; Storer et al., 2014) and modeling (Khain et al., 2005; Lee et al., 2008) studies. Special attention has been given to the mixed and cold processes, for which smaller supercooled droplets are likely to freeze at lower temperatures (Rosenfeld and Woodley, 2000). Therefore, the freezing latent heat will be released higher in the atmosphere, further enhancing convection (Han et al., 2012; Koren et al., 2008; Lynn et al., 2007; Rosenfeld et al., 2008; Tao et al., 2007). The intensified convection in the polluted environment increases cloud top height and enlarges the cloud cover in the upper troposphere due to a larger number of small-size ice crystals at those altitudes that settle slowly (Fan et al., 2013; Koren et al., 2010; Lee and Feingold, 2013; Morrison and Grabowski, 2011; Storer et al., 2014). The larger amount of supercooled droplets has been suggested as one of the critical elements in the cloud invigoration chain of events, and it depends on the upward transport of liquid mass across the zero temperature level (ZTL).

Koren et al. (2015) recently showed that an important component of the aerosol effect on clouds can be captured by the effective terminal velocity (η) properties; η is calculated as the mass-weighted mean hydrometeor terminal velocity within a given volume in the cloud. As such, it describes the terminal velocity of the given volume’s center of gravity (COG). In other words, η describes the COG velocity (always down) relative to the surrounding air velocity. Smaller droplets with narrow variance will have significantly smaller |η| and therefore, in a given updraft, will move higher in the atmosphere than larger droplets (larger |η|). The sum of the surrounding air velocity (w) and η describe the hydrometeor COG velocity with respect to the surface. We define this velocity as \( V_{\text{COG}} = w + \eta \), and it captures both the aerosol effect on the condensation efficiency (as it controls the latent-heat release that fuels the cloud updraft, w) and the previously described effect on η. In the early stages of cloud evolution, when condensation is the dominant process (the collection process is not yet significant), higher condensation efficiency and smaller |η| act together to push the hydrometeors higher in the polluted clouds. Such effects have only been shown and discussed for warm clouds (Dagan et al., 2015a, 2017; Heilblum et al., 2016; Koren et al., 2015, 2014).

A big part of the challenge in understanding aerosol effects on deep convective clouds is attributed to the large uncertainties related to ice nucleation processes (DeMott et al., 2015). These processes have been shown to be extremely sensitive to the aerosol surface properties in a manner that is not yet understood (Vali, 2014), and many freezing schemes are based on empirical relationships that do not create one comprehensive theory.

Here we focus on warm-phase processes within a deep convective system. Warm-phase processes (which are less complex) dictate much of the boundary and initial conditions with regard to hydrometeors, moisture, and heat fluxes to the mixed and cold phases. More specifically, the objectives of this study are to explore the aerosol effects on the condensate mass fluxes crossing the ZTL to understand the role of warm microphysical processes in deep convective clouds. We also analyze processes in the mixed and cold phases as well as downward fluxes crossing the ZTL from the mixed to warm phase with the caveat of less certainty in the distribution of the specific ice particles.

2 Model and simulation setup

The simulations were conducted using the weather research and forecasting (WRF) model, version 3.6.1, including a fast version of spectral bin microphysics (Fast-SBM; Khain et al., 2009; Skamarock et al., 2008). Three size distributions of hydrometeors were used to describe water drops, ice crystals and snow, and graupel. Each one was defined by 33 mass doubling bins; i.e., the mass of the hydrometeors in bin \( k \) was double the particle mass in bin \( \left(k - 1\right) \). CCN were described by a separate size distribution containing 33 bins with minimum and maximum sizes of 1.23 nm and 2 µm, respectively. More details about the warm and cold processes considered in Fast-SBM can be found in Khain et al. (2004, 2009) and Lynn et al. (2007).

A deep convective cloud system was simulated over the Marshall Islands during the Kwajalein Experiment (KWJEX; 23 July–25 September 1999). The simulations were conducted with three nested grids (G01, G02, and G03). The domain configuration is shown in Fig. 1a; the grid sizes were \( 1320 \times 1080 \) km, \( 600 \times 480 \) km, and \( 300 \times 300 \) km with a horizontal grid spacing of 12,000, 2400, and 480 m and time steps of 40, 8, and 2 s for G01, G02, and G03, respectively. There were 60 vertical levels for each grid with stretched spacing between 70 m near the ground to 400 m above 2000 m of height. The G01 run was driven by the Climate Forecast System Reanalysis (CFSR) data (every 6 h: https://climatedataguide.ucar.edu/climate-data/climate-forecast-system-reanalysis-cfsr), and it provided boundary conditions for the nested grids.
values for the G02 run using the two-way nested method. The innermost grid (G03) was driven by the G02 run data every 10 min using the one-way nested method (nest-down) to maintain similar initial and boundary conditions in different aerosol scenario simulations. The G01 and G02 runs were initiated on 19 August 1999 at 00:00 UTC (12:00 LT) with a spin-up time of 12 h and a total run time of 24 h. The G03 run was initiated 12 h later at 12:00 UTC (00:00 LT) on 19 August and ended at 00:00 UTC (12:00 LT) on 20 August. Taking a spin-up time of 4 h, the simulation results for G03 were analyzed between 4 and 12 h of the simulation. The same configuration of physical schemes was used for all three nested grids, including the Fast-SBM microphysical scheme, the RRTMG longwave and shortwave radiation schemes (Iacono et al., 2008), the Yonsei University (YSU) planetary boundary layer scheme (Hong et al., 2006), and the Noah land surface scheme (Chen and Dudhia, 2001).

The initial CCN distribution in SBM was calculated using the empirical equation 

\[ N_{CCN} = N_0 S^k, \]

where \( N_{CCN} \) is the numerical concentration of activated aerosol particles at supersaturation (S) with respect to water (%). \( N_0 \) and k are constants for determining the aerosol concentrations in different aerosol scenarios. The calculation method was as detailed by Khain et al. (2000). Observations showed that the KWAJEX clouds developed in a pristine environment with an average drop number of less than 100 cm\(^{-3}\) (Rangno and Hobbs, 2005). Therefore, in the G01 and G02 simulations that produced the meteorological conditions for the G03 run, we used \( N_0 = 100 \) cm\(^{-3}\) and \( k = 0.5 \). The G03 simulations were carried out for clean (\( N_0 = 100 \) cm\(^{-3}\) and \( k = 0.5 \)), semi-polluted (\( N_0 = 500 \) cm\(^{-3}\) and \( k = 0.5 \)), and polluted (\( N_0 = 2000 \) cm\(^{-3}\) and \( k = 0.5 \)) conditions. The aerosols were initially assumed to be homogeneously distributed in the lowest 2 km layer of the domain, with an exponential decrease with height above this layer. The initial domain mean profiles of temperature and dew-point temperature for the G03 run for all three cases are shown in Fig. 1b, revealing a warm and humid environment that supports the formation and development of deep convection with a maximum relative humidity of 88% at around 500 m above sea surface level. As shown in Fig. 1b, the wind field was dominated by easterly winds; therefore, we excluded a belt of \( \sim 50 \) km near the eastern boundary of G03 from the analyzed region to minimize the impact of cloud formation at the boundary. A narrow belt near the western side (\( \sim 5 \) km) of G03 was also excluded from the analyzed data, as were \( \sim 30 \) km belts near the southern and northern sides of the domain. Results for all three runs were collected every 2 min of simulation time.

In the analysis, we selected 0°C as our reference level for separating the warm- and mixed-phase parts in the clouds. Although cloud droplets freeze at temperatures colder than 0°C, above the ZTL there is potential for freezing.

3 Results

3.1 Comparison with observations

For validation purposes, ground-based Kwajalein S-band radar measurements were used (located as denoted by the black dot in Fig. 1a) with a 10.71 cm wavelength and a coverage radius of 150 km. A detailed description of the radar measurements in KWAJEX can be found in Yuter et al. (2005). Figure 2a and b show a comparison of the contoured frequency by altitude diagrams (CFADs) of the reflectivity measured by the radar vs. that simulated by the clean run (for a 10 cm wavelength). CFADs are probability distribution functions per height level presented as percentages (Yuter and Houze, 1995). The comparison shows that the clean run captured the vertical structure and magnitude of...
the observed CFAD reasonably well. The highest probability in the clean run CFAD is located around values of 18–28 dBZ from the surface up to 4.8 km; above this (5–8 km) it is 12–22 dBZ, and at the upper levels (9 to 13 km) it is between 7 and 17 dBZ, which is in agreement with the observed radar reflectivity. There is an overestimation of the modeled reflectivity above the ZTL (4.8 km) compared to the observed one. It can be explained by an overestimation of large ice hydrometeors (mostly graupel, but snow particles as well) above the ZTL. This is due to feedbacks caused by the simple melting scheme used by the model (see Sect. 3.3).

Figure 2c shows the normalized frequency of rain rates for 16:00–24:00 UTC from observations (radar-based estimation as described by Houze et al., 2004) and from the small grids in the three simulations (clean, semi-polluted, and polluted). The distributions show a peak in light rain rates (< 0.5 mm h^{-1}; zoomed-in view in the lower left corner of Fig. 2c) and a long tail at the heavy rainfall end. The radar-observed rain-rate distribution is located between the clean and semi-polluted simulation results. Given the uncertainties in rain estimations from radar observations and the model limitations, this shows that the model captured the general rain-rate distribution.

### 3.2 Aerosol effects on cloud macrophysical properties

An examination of cloud properties in the three different runs revealed differences in both macrophysical and microphysical properties. Figure 3a, c, and e show the evolution of the vertical profiles of cloud fraction for the three runs (calculated as the ratio between the number of cloudy volume pixels (voxels), i.e., the total condensate exceeding 0.01 g kg^{-1}, at each vertical level and the total horizontal number of voxels). There is a significantly larger cloud fraction in the middle and upper levels in the semi-polluted and polluted cases compared to the clean run. This trend can be recognized above ~4.8 km, and it is very pronounced at high levels (above 10 km) after the third hour of the simulation. This figure also indicates a higher cloud top height under polluted conditions. An examination of the low-altitude levels (be-
Figure 3. (a) Time series of the vertical profiles of a cloud fraction. (b) Frequency of air vertical velocity per altitude from the clean simulation results. Panels (c, d) and (e, f) are similar to panels (a, b) for the semi-polluted and polluted simulations, respectively. Panels (g) and (h) are vertical profiles of the mean cloud fraction and mass-weighted mean vertical velocity from 4 to 12 h of the simulation in all cases.

low 4.8 km) also shows a larger cloud fraction in the polluted case, which is probably the combined effect of mass created in this layer and downward sedimentation of the cloud mass. We note that the cloud fraction below 1 km includes precipitating regions. Figure 3g displays the mean vertical profiles of the cloud fraction from 4 to 12 h of the simulation. It summarizes the trends discussed above.

Aside from the larger cloud fraction, Fig. 3 also shows stronger updrafts under higher aerosol loading conditions. Figure 3b, d, and f display the number of voxels (at each altitude) per given interval of vertical velocities for the clean, semi-polluted, and polluted simulations, respectively. Most of the cloudy voxels are characterized by vertical velocities in the range of $-1$ to $1 \text{ m s}^{-1}$ in all three runs. The peak updrafts show a bimodal structure with peaks at 2–3 km and 10–12 km, which is consistent with the findings of Heymsfield et al. (2010).

The enhanced aerosol loading leads to enlarged areas of both strong updrafts and strong downdrafts from the surface to the upper troposphere. The number of voxels with updrafts exceeding $5 \text{ m s}^{-1}$ increases above 4.8 km of altitude with an up to 24- and 76-fold increase in the semi-polluted and polluted runs, respectively, compared to the clean run. Figure 3h shows an increasing trend in the mean vertical velocity profiles for more polluted runs, indicating the potential to promote more water vapor and condensate rise to higher altitudes and hence enhance the growth of hydrometeors at those levels, as shown further on.

Our results agree with previous numerical studies that reported an aerosol invigoration effect of tropical deep convective clouds (Cui et al., 2011; Fan et al., 2013; Khain et al., 2008; Li et al., 2013; Tao and Li, 2016; Tao et al., 2007). However, other numerical studies showed no clear evidence of this effect or even an opposite effect (Lee and Feingold, 2010; Morrison and Grabowski, 2011, 2013). The reasons behind those differences were examined in previous studies that showed the lower sensitivity of cloud and rain processes in bulk schemes to aerosol concentration (Khain et al., 2009, 2015; Lebo and Seinfeld, 2011; Lebo et al., 2012; Heiblum et al., 2016).

Figure 4a shows the time series of the mean total water path (TWP; averaged over cloudy columns) in the clean, semi-polluted, and polluted runs (solid curves). The TWPs in all three runs started increasing around 4 h into the simulation and peaked between 6 and 7 h of the simulation. The semi-polluted curve shows the highest values (between $\sim 5.5$ and 7 h of the simulation) compared with the other two simulations. On the other hand, the total cloud mass (dashed) curves show the highest values for the polluted run and the lowest for the clean run throughout the simulation. This difference can be explained by the partitioning of the total mass into clouds in the three runs. Even though the total mass was higher in the polluted run, it was distributed over a larger area (higher cloud fraction, larger anvils; see Fig. 3c, e), and therefore the mean TWP was lower than in the semi-polluted run.

Figure 4b illustrates the normalized occurrence frequency distribution of the TWP in the three runs. It shows a higher probability for high TWPs as the aerosol amount increases, suggesting that the fraction of deeper clouds (with more integrated mass) was larger in the more polluted runs. The clean run shows a higher frequency of occurrence for the range of small TWPs ($<0.02 \text{ kg m}^{-2}$, as denoted in the lower left corner of Fig. 4b).

To further explore aerosol effects on the vertical distribution of cloud mass, we examined the domain partitioning of total mass in the layers above (in height units, meaning at colder temperatures) and below the ZTL (Fig. 4c). It was seen that bigger mass both below and above the ZTL contributes to the largest total mass in the more polluted runs.

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The cloud mass above the ZTL was exceptionally large in the semi-polluted run between 6.5 and 7 h of the simulation, which is consistent with the largest mean TWP in this run during that time (Fig. 4a). For the clean case, the cloud mass above the ZTL was always smaller than the one below that level, but for the semi-polluted and polluted cases, the mass above the ZTL was larger for short periods (Fig. 4c). As shown in Fig. 4d, the ratios between the cloud masses above and below the ZTL were larger in the more polluted cases, suggesting much more efficient upward transport of cloud mass from the warm environment to the subzero temperature region and the subsequent productive growth of condensate in those levels.

### 3.3 Aerosol effects on cloud microphysical properties

To gain a better process-level understanding of the aerosol effect on the cloud mass spatial distribution, we evaluated the magnitudes of the phase transition processes above and below the ZTL. Figure 5 shows the changes in cloud mass (summed for the whole domain between 4 and 12 h of the simulation) driven by the following processes: condensation and evaporation of liquid drops, deposition and sublimation of ice particles, drop freezing, and riming by ice particles. Figure 6a–f complete the picture by presenting the total mass per height of the different types of cloud particles.

The results show that an increase in aerosol loading enhanced all of the examined processes, including the gain of water and ice mass (condensation and deposition), loss of mass (evaporation and sublimation), and transition from the liquid to ice phase (freezing and riming). Since increased aerosol concentration enhanced both types of processes, i.e., both source and sink for the hydrometeor mass, the overall effect was somewhat reduced due to mild cancellation of the gained mass by the enhancement of evaporation and sublimation. Hence, the released and absorbed latent heat were both higher in the more polluted runs, a fact that can explain the enhanced aspect of the dynamics, i.e., stronger updrafts and downdrafts (Fig. 3b, d, f). The contribution of the freezing and riming processes to the total latent heat release increased from 4.5 % in the clean run to 10.5 % in the polluted run, suggesting a larger mass of supercooled water above the ZTL (Fig. 6f). The supercooled water reached higher altitudes in the polluted runs (compared to the clean run) as inferred from the higher maximal altitude at which the condensation and freezing processes took place (Fig. 5a, c). This is consistent with previous studies showing a higher probability of droplets reaching subzero temperatures under polluted conditions (Carrió and Cotton, 2011; Khain et al., 2001). The freezing and riming of a larger amount of supercooled water and the enhanced depositional growth of ice particles both contributed to the larger ice mass above the ZTL (Fig. 6c–e). Note that the simple melting scheme used by the model al-
Figure 5. Total changes in cloud mass (kg) over 4 to 12 h of the simulation due to (a) condensation/evaporation (solid/dashed curves), (b) deposition/sublimation (solid/dashed curves), and (c) freezing for clean (blue curves), semi-polluted (green curves), and polluted (red curves) experiments. The freezing mass includes both the freezing of liquid drops and riming by ice particles. Note that the x-axis scales are different. The dashed black lines denote the zero temperature level (ZTL; ∼4.8 km).

Figure 6. Vertical profile of (a–f) total mass over the domain (kg) and (g–l) mean effective terminal velocity ($\eta$; m s$^{-1}$) over cloudy regions during 4–12 h of the simulation for cloud droplet (first column), raindrop (second column), ice crystal (third column), snow (fourth column), graupel (fifth column), and all hydrometeors (sixth column). The dashed black lines denote the ZTL (∼4.8 km).

The enhanced gain of water and ice mass in the polluted runs yields higher mass loading that acted to reduce the cloud buoyancy. The upper panel of Fig. 7 shows the vertical pro-

towed for the immediate melting of ice particles while cross-
ing the ZTL. The resulting drops, formed by the melting of big graupel (and snow) particles, broke up immediately into smaller drops. Parts were carried up again by the updraft and frozen by riming, so there may be an overestimation of big graupel particles above the ZTL (as shown in Fig. 2b).
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Figure 7. Vertical profiles of (a) mean buoyancy averaged over cloudy regions during 4–12 h of the simulation and its components: (b) thermal term (BT), (c) water vapor term (BV), and (d) water loading term (WL). (e–i) Vertical profiles of WL for cloud droplet (e), raindrop (f), ice crystal (g), snow (h), and graupel (i). The black dashed lines denote the ZTL (∼4.8 km).

files of the mean buoyancy (total B and the following components: BT is thermal, BV is vapor, and WL is water loading) for the domain cloudy voxels (between 4 and 12 h of the simulation). Indeed, water loading played an important role and as expected there was a “payment” once the polluted clouds became thicker with more gained liquid and ice mass that was transported higher in the atmosphere; the added water loading acted to counterbalance the overall buoyancy. Figure 7a shows that the total B profiles switch between smaller values (more negative) for the polluted runs compared to the clean run in low altitudes (from 2.4 to 6.7 km) to larger values at higher levels (above 6.7 km). In all levels, BT was larger in the polluted case (or equal near the freezing level) and the water loading was the smallest (a less-negative WL buoyancy component) in the clean run from 1.5 to 8.5 km. The lower panel of Fig. 7 shows the vertical profiles of the mean waterloading components for the different types of cloud particles. Note that it is different from the profiles of total mass shown in Fig. 6 as it presents the mean B values, and hence it is influenced by the cloud coverage in each level. It shows a significant increase in water loading (smaller WL buoyancy component) for the polluted runs in all but the snow hydrometeors. Moreover, it reveals that larger rain content likely to originate higher in the clouds can explain about half of the extra water loading in the polluted cases in the lower part of the clouds.

It is important to note that among all of the processes, condensation contributed most to the mass gain. It peaked at 2.4 km for all three runs. Moreover, even above the ZTL, condensation still dominated the cloud mass gain compared to depositional growth for all three runs (Fig. 5a vs. 5b). Since the droplet nucleation process was negligible above the ZTL in the present study (accounting for ∼5% of the total nucleated drop mass), most of the cloud drops that grew by condensation above the ZTL originated in the warm part of the clouds. The mass gained by the enhanced condensation in the polluted runs was further boosted by the stronger updrafts that increased the supersaturation. Better droplet mobility (see Fig. 6g) further implies that a significant part of the enlarged liquid mass generated below the ZTL in the more polluted cases (as shown in Fig. 6a, b) was transported upward by the stronger updrafts (Fig. 3h) and continued growing via condensation at altitudes above the ZTL.

3.4 Aerosol effects on upward transport of condensate mass

As noted in the previous section, an important part of the aerosol effect on deep convective clouds has to do with its influence on the transport of mass from the warm part to the upper levels. To evaluate the hydrometeor fluxes, their terminal velocities (which are correlated with size) must be considered. As described above, the effective terminal velocity...
(\eta) of a given volume within a cloud is a measure of the terminal velocity of the volume hydrometeor COG (Koren et al., 2015). As such, it can be linearly added to the mean air updraft velocity weighted by hydrometeor mass to infer the hydrometeor \( V_{\text{COG}} \), which is the COG velocity relative to the surface.

Since upward motion is considered positive, the sign of \( \eta \) is always negative. To avoid confusion, we will discuss the magnitude (absolute values) of \( \eta \) hereafter. Figure 6g–l show the vertical profiles of the mean \(|\eta|\) for all types of hydrometeors. The polluted case has the smallest \(|\eta|\) for the cloud droplets (Fig. 6g), allowing the water droplets to be pushed higher in the atmosphere by the enhanced updrafts. Raindrops, however, are larger for the polluted case (Fig. 6h). We note that the model uses a simple melting scheme in which there is immediate melting when crossing the ZTL downward. Therefore, all of the hydrometeors below this level are liquid only. The \(|\eta|\) value of all liquid drops below the ZTL was larger in more polluted cases (Fig. 6i); that of graupel particles increased with more aerosols, similar to raindrops (Fig. 6k), and that of ice crystals and snow decreased with aerosol loading above 10 km (Fig. 6i). This suggests that the larger cloud fraction of the polluted runs in the upper troposphere contains smaller ice particles that exhibit slower sedimentation (Figs. 3g and 6l).

What roles do the two characteristic velocities play in the overall aerosol effect? The COG perspective allows for a calculation of the vertical movement of the hydrometeors as a superposition of the two characteristic velocities of the system, i.e., the air vertical velocity (\( w \)) and \( \eta \). As explained earlier, the sum of the two (weighted by mass) velocities (\( V_{\text{COG}} \)) is the COG velocity relative to the surface (Koren et al., 2015). Using the definition of COG velocity, we calculated the condensate mass fluxes (kg s\(^{-1}\)) as the product of the total condensate mass and \( V_{\text{COG}} \) (m s\(^{-1}\)). Figure 8 shows the temporal evolution of the two average characteristic velocities (\( V_{\text{COG}} \), the air vertical velocity (\( w \)), and \( \eta \)). We separate pixels of upward and downward mass flux motion by the sign of \( V_{\text{COG}} \).

We note that despite the fact that the \(|\eta|\) value of the cloud droplets is significantly smaller for the polluted clouds (as shown in Fig. 6g), allowing them to be pushed higher in the atmosphere by the enhanced updrafts, at the ZTL (~4.8 km), the \(|\eta|\) values of the polluted hydrometeors are larger than those of the clean ones. This indicates that the contribution of the larger raindrops in the polluted cases controls the overall \(|\eta|\) values at this level. For all cloudy area grid boxes (Fig. 8a–c), the overall \( V_{\text{COG}} \) is negative despite the fact that the averaged updrafts are positive, indicating that sedimentation measured by \(|\eta|\) values is larger than the updrafts. The positive upward-flux areas (Fig. 8d–f) are defined as \( V_{\text{COG}} > 0 \), which implies that the average updrafts have to be larger than \(|\eta|\). In contrast, in the negative flux area (Fig. 8g–i), the updrafts are fairly weak and \( V_{\text{COG}} \) is controlled by \(|\eta|\).

Finally, we link mass and velocity trends to see the aerosol effect on the total mass flux. Figure 9 shows the temporal evolution of the mass fluxes crossing the ZTL. Similar to the velocity analyses, we separated pixels of upward and downward mass flux motion by the sign of \( V_{\text{COG}} \). The domain average total (net) upward and downward fluxes (left column) and the corresponding fraction of the area that represents these fluxes (right column) are shown.

Similar to the \( V_{\text{COG}} \) trend, the total (net) mass flux crossing the ZTL (Fig. 9a) shows all negative flux values, indicating net downward movement of condensate, which is a result of larger mass production above the freezing level and thus sedimentation down. While the temporal evolution of the overall fluxes crossing the ZTL is not dramatically different for the three runs (mean values of 5.02, 7.47, and 3.46 x 10\(^{6}\) kg s\(^{-1}\) for the polluted, semi-polluted, and clean run, respectively), when the fluxes are observed separately according to their sign, a different view emerges. All flux trends (up and down) were dramatically enhanced in the polluted cases. The upward liquid mass fluxes in the semi-polluted and polluted cases were ~4–9 and ~8–21 times larger than those in the clean simulation (with mean values of 3.9, 1.8, and 0.32 x 10\(^{6}\) kg s\(^{-1}\) for the polluted, semi-polluted, and clean run). These trends are controlled by the enlarged upward motion area (as shown in Fig. 9d), the larger COG velocity (Fig. 8d), and the increased liquid water loading. These findings clearly demonstrate the enhanced cloud mass transport to upper levels under more polluted conditions. What is the average ratio between the mass that was transported up from below the ZTL and the mass that was produced locally above the ZTL in the different runs? This mass ratio is defined as

\[
\mu = \frac{\text{mass flux upward}}{\text{net mass flux}}
\]

We note that for the upward liquid mass transport, the ZTL crossing took place in a relatively small area (an order of magnitude less than the downward motion area; see Fig. 9d, f), affecting the mass partitioning below and above the ZTL and thus the formation and growth of ice particles. Moreover, the differences in area for the upward motion between the clean and polluted conditions were the most significant. Figure 9g shows that the upward mass flux area of the polluted case is 1.8–3.9 times larger than for the clean one. This impacted the variance of the mass flux, which was larger in the more polluted cases (Fig. 9c). The increased variance is driven by the enhancement of the field dynamics by aerosol, as shown throughout this study. Polluted clouds exhibited larger updrafts with larger variance (as shown in Fig. 8),
larger updraft area (Fig. 9d), and larger mass fluxes, all of which tend to increase the variance in the upward mass flux.

Similar to the upward transport, the downward transport of cloud mass from subzero temperature levels to the warm environment (Fig. 9e), the mean surface rain rate, and accumulated precipitation (Fig. 9g) were also larger in the more polluted cases.

4 Discussion and summary

A major fraction of the deep convective clouds around the globe have a warm base. This is more obvious over the tropical belt, where the freezing-level height is located at around 5 km (and the cloud base is at a much lower altitude). This is also true for many of the midlatitude frontal systems (particularly in the summer). Exceptions to this rule are likely to occur either in a very cold atmosphere or in the case of orographic clouds for which the lifting condensation level is high.

As such, the warm-phase properties can be considered as the initial and boundary conditions for the mixed and cold phases. Therefore, changes in aerosol loading can affect the mixed- and cold-phase properties, not only directly by serving as ice nuclei but also indirectly by affecting the warm-phase properties and the fluxes between the phases. In this work, we focused on the interface between the phases. Although cloud droplets freeze at temperatures colder than 0 °C, we selected the ZTL as our reference level between the warm and mixed phases because above it there is a potential for freezing.

We used the WRF model with Fast-SBM to simulate a case study of a deep convective cloud system over the Marshall Islands during KWJEX on 19 August 1999 (12:00–24:00 UTC, 19 August; 00:00–12:00 LT, 20 August). As a sanity check, we compared the simulated vertical distribution of the radar reflectivity and the normalized occurrence frequencies of radar-estimated rain rates to observations and showed that the model results are similar to the observed values (Fig. 2). We analyzed how changes in cloud field properties are related to changes in aerosol concentration at the macro level, showing a notable increase in the total cloud mass, a larger cloud fraction in the upper levels, a higher cloud top, and a larger frequency of strong updrafts and heavy rain rates (Figs. 3 and 4). Larger mass both below and above the ZTL was shown to contribute to the larger total cloud mass in the polluted runs. Examining processes on a finer scale revealed that increasing aerosol concentration is related to the enhancement of mass (water and ice) source and sink processes (Fig. 5). The aerosol effect on the cloud warm-phase processes could be divided into two main branches, one linked to the enhancement of diffusion processes (condensation and evaporation) and the other to greater droplet mobility. Per given volume within the cloud, we refer here to droplet mobility as the way in which the COG of the total hydrometeor mass moves with the surrounding air (w, updraft; Koren et al., 2015). The effective...
terminal velocity ($\eta$) is inversely proportional to the droplet mobility and is the measure for the water mass COG velocity. More aerosols yield more activated droplets. This implies enhancement of the overall condensation rate that drives more latent heat release (Fig. 5a; condensation). At the cloud edges under subsaturated conditions, the evaporation is enhanced following the same line of reasoning (Fig. 5a; evaporation). All of this is in agreement with Lee and Feingold (2013) for deep convective clouds and several studies of warm convection (Dagan et al., 2015a, b; Koren et al., 2014; Pinsky et al., 2013; Seiki and Nakajima, 2014). The condensation process still played a major role, even above the ZTL, and it made a larger contribution to the gain in mass than depositional growth. Since droplet activation was negligible above the ZTL, we suggest that the liquid drops participating in the condensational growth come from below the ZTL.

\[ V_{\text{COG}} = \omega + \eta \]

$V_{\text{COG}}$ captures both the aerosol effect on the condensation efficiency (as it controls the latent heat release that fuels the cloud updrafts; $\omega$) and the mobility effect as captured by $\eta$ (Koren et al., 2015). We calculated condensate mass fluxes across the ZTL as a product of cloud mass and $V_{\text{COG}}$. Again, on the same microscale, an increase in aerosol concentration was related to flux enhancement in all directions. Although the net mass fluxes changed only slightly, the condensate mass fluxes up and down were dramatically amplified. Larger $V_{\text{COG}}$ led to the enhanced upward transport of liquid mass from warm to mixed parts under polluted conditions (Fig. 9c, d). The overall aerosol effects are reflected by a larger ratio between the cloud mass above and below the ZTL. We show that there is a “payment” for the enhanced mass production and transport in the polluted runs. The larger mass loading acts to reduce the buoyancy, while the enhanced latent heat release acts to increase it. The enhanced mass flux upward and the enhanced mass production above the ZTL yield the enhancement of the mass flux down (partly manifested as enhanced rain that explains much of the increase in mass loading below the ZTL).

Our study highlights the importance of aerosol effects on the warm processes in deep convective clouds using condensate mass flux as a measure of hydrometeor transport in clouds between the warm, mixed, and cold domains. Such effects enhance the thermodynamic and dynamic (vertical

\[ \text{Rain rate (left Y) and accum. surf. prec. (right Y)} \]

\[ \text{Rain area fraction} \]

\[ \text{Net mass fluxes} \]

\[ \text{Total fraction} \]

\[ \text{Upward mass fluxes} \]

\[ \text{Downward mass fluxes} \]

\[ \text{Upward motion area fraction} \]

\[ \text{Downward motion area fraction} \]

\[ \text{Time series of (a) net cloud mass fluxes and (b) area fraction over the entire cloudy area crossing the ZTL for clean (blue curve), semi-polluted (green curve), and polluted (red curve) experiments. Panels (c, d) and (e, f) are similar to panels (a, b), but for upward ($V_{\text{COG}} > 0$) and downward ($V_{\text{COG}} < 0$) motion areas, respectively. Panel (g) is a time evolution of domain-averaged rain rate (solid lines) and accumulated surface rain (dashed lines). Panel (h) is a time evolution of the rainy area fraction at the surface.} \]
winds) processes as well as changes in the overall structure and properties of the field (demonstrated here as cloud fraction per height or changes in rain rate).

Data availability. The simulations were conducted using the WRF model (version 3.6.1, handled by the Mesoscale and Microscale Meteorology Laboratory of NCAR) with spectral bin microphysics (Khain et al., 2009). The model is described in detail and can be downloaded from https://www.mmm.ucar.edu/weather-research-and-forecasting-model (Skamarock et al., 2008). Additional technical information about the simulations and initialization is available upon request from the contact author. The KWAJEX radar data used for validation of the model results can be downloaded from the Goddard Earth Sciences Data and Information Services Center (GES DISC): https://disc2.gesdisc.eosdis.nasa.gov/data/TRMM_GV_L2/TRMM_2A55UW.7/1999/231/ (Yuter et al., 2005).

Competing interests. The authors declare that they have no conflict of interest.

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