# The Impact of Raindrop Collisional Processes on the Polarimetric Radar Variables



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### Introduction

We investigate how the raindrop collisional processes in warm rain (coalescence, breakup) affect the radar reflectivity factor at horizontal polarization  $Z_{\mu\nu}$  differential reflectivity  $Z_{DR\nu}$  and specific differential phase  $K_{DP}$ .

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The fingerprint of each microphysical process is quantified individually and in combination for a variety of DSD shapes and nominal rainfall rates using a spectral bin model and electromagnetic scattering calculations. These fingerprints are compared to disdrometer and radar observations.

### Methods

The 1-D version of the explicit bin microphysical model of Prat et al. (2012) is initialized with various DSDs and rainfall rates at the top of the model domain.

The DSD is allowed to evolve under the influence of selected microphysical processes: size sorting/settling, coalescence, collisional breakup, aerodynamic breakup.

The predicted DSDs are converted into vertical profiles of  $Z_{\mu}, Z_{DR},$  and  $K_{DP}$  using T-Matrix scattering calculations (details of the parameters can be found in Kumjian and Ryzhkov 2012).



Fig. 1: Evolution of the vertical profiles of  $(a) Z_{\mu\nu} (b) Z_{\mu\mu\nu} (c) K_{\mu\nu}$  and (d)rainfall rate RR for a full-physics simulation over the first 5 minutes (gray curves) and after one hour (black curve). The initial impact of **size sorting** is evident by the large decreases in  $Z_{\mu\nu} K_{\mu\nu\nu}$  and RR coincident with a large increase in  $Z_{\mu\nu}$ .

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Fig. 2: Fingerprints of individual microphysical processes (first three rows) and full physics (bottom row) in vertical profiles of ( $1^{44}$  column)  $Z_{\mu}$  ( $2^{ad}$  column)  $Z_{\mu}$  ( $3^{ad}$  colum

### The impact of radar wavelength:

 $Changes in \ Z_{\mu} \ are \ largest \ in \ magnitude \ at \ X \ band; \ changes \ in \ Z_{DR} \ are \ largest \ at \ C \ band. \ Note \ the \ nonmonotonic behavior of the \ K_{DP} \ profiles \ at \ all \ 3 \ wavelengths; \ K_{DP} \ increases \ to \ larger \ than \ the \ value \ aloft \ at \ X \ band.$ 



Fig. 3: Normalized vertical profiles of (a)  $Z_{\mu}$  (b)  $Z_{D\nu}$ and (c)  $K_{D\nu}$  for a full-physics simulation initialized with a Marshall-Palmer DSD with nominal rainfall rate of 20 mm h<sup>-1</sup>. Profiles are shown for **S band**, C **band**, and **X band**.

#### The impact of different initial DSDs:

Full-physics simulations of the collisional processes show that the polarimetric fingerprints can occupy various regions in the  $\Delta Z_{n} c \Delta Z_{0n}$  and  $\Delta Z_{0n} c \Delta K_{0p}$  parameter space (Fig. 4). For example,  $\Delta Z_{min} \Delta Z_{0m}$ , and  $\Delta K_{0p}$  are positive when coalescence dominates and negative when breakup dominates. However, for the whole range of initial DSD shapes and RR, the fingerprint identified for evaporation ( $\Delta Z_{man} a \Delta A_{mp} < 0, \Delta Z_{0m} > 0$ ; Kumjian and Ryzhkov 2010) is distinct from those produced by the collisional processes.



Fig. 4: (a) Changes in  $Z_{\rm in}$  vs. changes in  $Z_{\rm on}$  (b) changes in  $Z_{\rm on}$  vs. changes in  $K_{\rm on}$  over the 3-km domain for full-physics simulations initialized with a variety of DSDs of varying rainfall rates. Markers indicate S band, C band, and X band.

## Comparison with Observations



Fig. 5: Evolution of the  $Z_{rr}Z_{rs}$  pairs for the DSD at each time and each height level for the full **physics** simulations, and (a) breakup only, (b) coalescence only. The numbers represent the rainfall rate olot. Comparison with disdrometer observations collected in Oklahoma (blue solid curve: Cao et al. 2008), Florida (blue dashed curve: Zhang et al. 2006), and the envelope of observations in Florida (dated blue curves: Brandes et al. 2004).

The full-physics simulations for large RR produce negatively biased  $Z_{DR'}$  indicating overaggressive breakup of drops and production of too many small drops. In contrast, simulations where only coalescence is permitted better match the disdrometer observations.



Fig. 6: (left) Field of  $Z_{\mu}$  in an RHI taken by the CSAPR at the DOE ARM Southern Great Plains site on 20 May 2011. The rectangle shows the window used for the averaging in the right panel. (right) Average vertical profiles of  $Z_{\mu\nu} Z_{\mu\nu}$  and  $\rho_{\mu\nu}$  (black curves) with ±1 standard deviation shown (gray curves). The linear fit to the mean profiles is shown in areen.

Fig. 7: (a) Change in  $Z_{\rm H}$  vs. initial  $Z_{\rm H}$  in simulations (black dots) and CSAPR observations (grav squares). (b) as in (a), except showing change in  $Z_{\rm DR}$  vs. initial  $Z_{\rm DR}$ . Note the strong signal for evaporation in the observations (i.e., negative changes in  $Z_{\rm ag}$ ).



### Conclusions

-Each individual microphysical process (size sorting, breakup, coalescence, evaporation) produces a distinct fingerprint in vertical profiles of  $Z_{\mu\nu}$   $Z_{DR\nu}$  and  $K_{DP}$ . This can allow for identifying the dominant process in rainfall.

-These polarimetric fingerprints are dependent on radar wavelength.

-Comparisons with disdrometer and radar observations suggest that the accepted parameterizations of drop breakup are too aggressive for the largest rainfall rates, resulting in very "tropical" DSDs heavily skewed towards smaller drops.

-Polarimetric radar observations in rain may be used to improve such parameterizations via inverse modeling techniques.

