

1 **Comparisons of Modeled and Observed Reflectivity and Fall Speeds for Snowfall of**
2 **Varied Riming Degree During Winter Storms on Long Island, NY**

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Abstract

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Derived radar reflectivity and fall speed for four Weather Research and Forecasting model bulk microphysical parameterizations (BMPs) run at 1.33 km grid spacing are compared with ground-based, vertically-pointing Ku-band radar, scanning S-band radar, and in situ measurements at Stony Brook, NY. Simulations were partitioned into periods of observed riming degree as determined manually using a stereo microscope and camera during nine winter storms. Simulations were examined to determine whether the selected BMPs captured the effects of varying riming intensities, provided a reasonable match to the vertical structure of radar reflectivity or fall speed, and whether they produced reasonable surface fall speed distributions. Schemes assuming non-spherical mass-diameter relationships yielded reflectivity distributions closer to observed values. All four schemes examined in this study provided a better match to the observed, vertical structure of reflectivity during moderate riming than light riming periods. The comparison of observed and simulated snow fall speeds had mixed results. One BMP produced episodes of excessive cloud water at times, resulting in fall speeds that were too large. However, most schemes had frequent periods of little or no cloud water during moderate riming periods and thus underpredicted the snow fall speeds at lower levels. Short, 1-4 hour periods with relatively steady snow conditions were used to compare BMP and observed size and fall speed distributions. These limited data suggest the examined BMPs underpredict fall speeds of cold-type snow habits and underrepresent aggregates larger than 4 mm diameter.

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52 **1. Introduction**

53 As operational numerical weather prediction continues a trend towards finer
54 spatial resolution, bulk microphysics schemes (BMPs) are relied upon to capture
55 numerous microphysical processes and characteristics of resulting precipitation. Several
56 assumptions are made within these schemes, including the shape and related parameters
57 of the particle size distribution, various size-fall speed relationships, and mechanisms for
58 the production of dry or rimed snow, and graupel. Several studies have examined the
59 performance of BMPs by comparing characteristics of simulated ice classes against
60 surface, aircraft, and remote sensing acquired during winter storms. Observations on 3-4
61 December 2001 during IMPROVE-II showed that BMPs in the Weather Research and
62 Forecasting (WRF, Skamarock et al. 2008) model available at that time tended to
63 overpredict the snow aloft in the snow growth region (Garvert et al. 2005; Lin and Colle
64 2009). The snow fall speed was found to be too fast in the Purdue Lin (Lin and Colle
65 2009) and WRF six-class, single-moment scheme (WSM6, Hong et al. 2006) when
66 compared to the Thompson et al. (2004) scheme. The revised Thompson et al. (2008,
67 THOM2) scheme incorporated a new mass-diameter relationship and particle size
68 distribution for snow. Lin and Colle (2011) developed the single-moment Stony Brook
69 scheme (SBU-YLIN), which combines the snow and graupel categories into a single
70 precipitating ice class with corresponding riming factor. When compared against the
71 THOM2 and Morrison et al. (2009, MORR) schemes, Lin and Colle (2011) found that
72 the Stony Brook scheme reduced snow amounts aloft, which compared more favorably
73 with in situ observations acquired over the Oregon Cascades.

74 Other examinations of simulated and observed snowfall were performed using
75 observations from the Canadian CloudSat/Cloud-Aerosol Lidar and Infrared Pathfinder
76 Satellite Observations (CALIPSO) Validation Project (C3VP) in Ontario (Skofronick-
77 Jackson et al. 2015, Petersen et al. 2007). Snowfall observed during the 22 January 2007
78 event was comprised primarily of lightly-rimmed dendrites and their aggregates and
79 sampled by ground-based and aircraft observations. Molthan et al. (2010) used C3VP in
80 situ observations and radar observations from 22 January 2007 to evaluate the Goddard
81 six-class scheme with graupel (Tao et al. 2003; Lang et al. 2007) and Molthan et al.
82 (2012) extended the evaluation to include the WSM6, THOM2, MORR, and SBU-YLIN
83 schemes. Their studies generally concluded that schemes providing greater flexibility in
84 size distribution parameters, density, or additional moments improved performance over
85 the use of constant, assumed parameters. Surface measurements of particle fall speeds
86 during the C3VP event suggested that diameter-velocity parameterizations of the
87 THOM2, MORR, and WSM6 overestimated fall speeds for sizes larger than 1 mm, while
88 the SBU-YLIN scheme produced fall speeds closest to observations (Molthan and Colle
89 2012). Meanwhile, the Goddard scheme tended to underestimate fall speeds for all sizes
90 (Molthan et al. 2010). Whereas the Goddard scheme tended to underestimate fall speeds
91 in Molthan et al. (2010), Han et al. (2013) found it provided the best agreement with
92 observations acquired in snowfall located above the melting layer of a broader region of
93 stratiform rainfall, which preceded a cold front affecting western California. Shi et al.
94 (2010) evaluated the performance of the Goddard scheme for lake-effect snow observed
95 during C3VP through comparisons of observed and simulated C- and W-band radar
96 reflectivity and AMSU-B brightness temperatures. Comparisons of simulated and

97 observed radar reflectivity demonstrated that for lake-effect bands, the WRF simulation
98 underestimated the echo top height of the observed band and failed to identify numerous,
99 smaller cores of reflectivity. For broader regions of synoptic-scale precipitation, there
100 was a tendency to overestimate the coverage of reflectivity above 20 dBZ. Additional
101 analysis of reflectivity CFADs revealed an overall ability for their simulation to capture
102 the overall large-scale cloud structures but additional refinements to microphysics and
103 smaller scale features were needed.

104 Regional differences in scheme performance highlighted in the aforementioned
105 studies warrant additional evaluations for other phenomena. Studies have evaluated snow
106 and graupel characteristics within BMPs for events in California, the Pacific Northwest,
107 Appalachians, and southern Ontario, but no known studies to date have examined BMP
108 assumptions in simulations of winter storms in the northeastern United States. Recent
109 studies have documented the evolution of snowfall microphysics in such storms as a
110 precursor to model comparisons. Stark et al. (2013) observed the evolution of ice crystal
111 habits through stereo microscope observations of snow obtained at the surface,
112 corroborated by WSR-88D cross-sections and a vertically-pointing Doppler radar. The
113 degree of riming for ice crystals was assessed from stereo microscope particle images
114 (Mosimann et al. 1994). The degree of riming and prevalence of dendrites increased with
115 snow band maturity and intensity, corresponding to an increase in snow-to-liquid ratio,
116 precipitation, and fall speed. As snow bands passed, weaker ascent and lower relative
117 humidity values corresponded to plate-like crystals, an overall decrease in dendrites, and
118 less riming. Colle et al. (2014) surveyed a dozen winter cyclones across three seasons
119 that impacted the northeastern United States and related snow-to-liquid ratios to

120 predominant crystal habits and degree of riming. Dominant crystal habits and variability
121 in riming were noted in relation to frontal zones and distance from the cyclone center,
122 thus, a single event is likely to be comprised of periods of varying habit, degree of
123 riming, and snow-to-liquid ratio.

124 In this study, model simulations of events documented by Stark et al. (2013) and
125 Colle et al. (2014) are categorized by the degree of riming present in surface observations
126 of snowfall. For each 15-minute period, an average degree and range of riming was
127 determined by visual inspection of stereo microscope images. This time series is then
128 used to partition radar observations and model output to represent times when light and
129 moderate riming occurred over the observation site. Model performance is then assessed
130 with respect to observed degree of riming from multiple storms and multiple seasons.
131 Four BMPs are selected based upon their diverse means of characterizing snow size
132 distributions, fall speed relationships, means for graupel production, and simulation of
133 riming characteristics (Tables 1 and 2 of Molthan and Colle 2012). Schemes were
134 selected based upon their frequent use in operational numerical weather prediction and
135 for continuity with the previous study to suggest continued improvements in the
136 simulation of winter weather. Since detailed in situ aircraft observations are not available
137 for this multi-season sampling of storms, evaluations are performed against available
138 ground observations and radar remote sensing of reflectivity and particle fall speed.
139 Model simulations of these quantities and comparison to observations will clarify
140 whether these schemes capture variability in size distribution and fall speed during
141 periods of varying riming degree, necessary to improve simulations of winter

142 precipitation. Comparisons will also identify future opportunities for improvement in the
143 simulation of riming processes.

144 This study is motivated by two key research questions:

- 145 • How realistic are selected single- and double-moment WRF BMPs for simulating
146 snow size distributions, fall speeds, and radar reflectivity for observed periods of light
147 and moderate riming during winter storms over Long Island, New York?
- 148 • How does the WRF BMP performance change for these categories of observed
149 riming, and when large aggregates are present?

150 This paper will be organized as follows: Section 2 will discuss the data and
151 methods used in this study. Section 3 will discuss the model verification results, and the
152 conclusions are in section 4.

153 **2. Data and Methods**

154 This study uses several computed and observed variables related to snow particle fall
155 speed, which we define in Table 1 for clarity. Differences in these variables need to be
156 taken into account when comparing among them. Since the instruments available did not
157 directly observe small-scale turbulence (E) and vertical air motion (w), we can only
158 directly compare among the measured and computed values when we can assume E
159 and/or w are zero. In low horizontal wind conditions, it is often assumed that $w=0$ for
160 surface-based in situ instruments such as disdrometers. Surface observations for events
161 described herein were limited to periods of horizontal wind speeds of 5 m s^{-1} or less. In
162 these environments, small-scale turbulence (E) will be smaller than typical snowfall

163 events, following Schreur and Geertsema (2008), who estimated E as related to half the
164 squared difference of wind gust and average wind speeds.

165 *a. Observations*

166 Microphysical and radar observations for this study were taken during the 2009-
167 2012 winter seasons at Stony Brook, NY (SBNY, see Colle et al. 2014; their Fig. 1),
168 which is on the north shore of Long Island (LI), approximately 93 km east of New York
169 City, New York (NYC). Stark et al. (2013) and Colle et al. (2014) provide details on the
170 experimental setup and location. A vertically-pointing METEK Ku-band micro rain radar
171 (MRR; Peters et al. 2002) was used at SBNY to observe the profile of reflectivity and
172 Doppler velocities to 7750 m above sea level every minute. The MRR has been used to
173 study winter snowstorms in several locations (Cha et al. 2009; Keighton et al. 2009; Prat
174 and Barros 2010; Kneifel et al. 2011a,b; Xie et al. 2012; Maahn and Kollias 2012, Stark
175 et al. 2013; Colle et al. 2014; Maahn et al. 2014; Pokharel et al. 2014a,b; Garrett et al.
176 2015). The radar reflectivities from the short wavelength of the radar (1.25 cm) are
177 subject to attenuation in heavier precipitation (Löffler-Mang et al. 1999) and in
178 conditions when wet snow builds up on the antenna (Stark et al. 2013). The latter is the
179 more relevant for conditions during snowstorms. The MRR data were post-processed to
180 improve sensitivity and data quality using the method of Maahn and Kollias (2012).

181 Observations from the MRR were supplemented by the WSR-88D radar at Upton,
182 NY (KOKX). Vertical profiles of interpolated WSR-88D reflectivity were computed
183 from Level II KOKX data for the vertical column nearest the verification point of each
184 model simulation. The WSR-88D radar data have coarser native sensor spatial resolution,
185 about 500 m in the vertical and horizontal at the 30 km range over the SBU site as

186 compared to the MRR resolution volume size of 250 m in the vertical and ~100 m in the
187 horizontal. For convenience in generating comparisons, the WSR-88D data were
188 interpolated to a Cartesian grid with vertical and horizontal spacing of 250 m and 100 m,
189 respectively.

190 A Particle Size and Velocity (PARSIVEL; Löffler-Mang and Joss 2000; Löffler-
191 Mang and Blahak 2001; Yuter et al. 2006) disdrometer was placed about 1 m above the
192 1-storey roof surface to collect hydrometeor size and fall speed distributions. Battaglia et
193 al. (2010) note that the PARSIVEL measures a “PARSIVEL diameter” based on the
194 maximum shadowed area of the particle as it passes through the disdrometer laser beam.
195 In a limited set of conditions where the snow flake is horizontally aligned, this
196 measurement is equivalent to the widest horizontal dimension of the snow particle,
197 otherwise, the PARSIVEL diameter represents an estimate of widest horizontal diameter
198 with an error less than or equal to 20%. In calm conditions, the PARSIVEL
199 measurement of particle fall speed is equivalent to settling speed (Table 1). Battaglia et
200 al. (2010) determined that the PARSIVEL measured fall speed has a variance less than
201 20% for individual particles and tends to underestimate the mean fall speeds of smaller
202 particles. The larger errors in their reported fall speeds are less relevant here as their
203 instruments were typically operated in windy conditions. Analysis herein was restricted
204 to winter storms with winds less than 5 m s^{-1} to avoid the potential for blowing snow
205 from the surface impacting the results (D. Kingsmill, personal communication, 2011),
206 and to emphasize periods of reduced small-scale turbulence.

207 In order to further characterize precipitation during these events, a stereo
208 microscope and camera were used to observe the snow habit and riming intensity at the

209 SBU site as described in Colle et al. (2014) and Stark et al. (2013). The ice habits were
210 categorized into several main types (needles and columns, dendrites, plates, side planes
211 and bullets), and riming was categorized as light, moderate, or heavy. Heavy riming did
212 occur during short intervals within three sampled storms, but the sample size of heavy
213 riming was insufficient for a comprehensive analysis.

214 *b. WRF simulations*

215 The Weather Research and Forecast Model (WRF; Skamarock et al. 2008)
216 version 3.3 was utilized for simulations of several of the observed winter storms. The
217 North American Mesoscale (NAM) model analysis data at 12 km grid spacing (NAM 218
218 hereafter) and six-hourly time increments were used as initial and boundary conditions in
219 the majority of the simulations, though in a limited number of events, the Global
220 Forecasting System (GFS) analysis data at 0.5° grid spacing were used because
221 simulations with the NAM 218 data were too dry, or precipitation placement was not in
222 agreement with observations. Sea surface temperature and snow cover data were included
223 in these initial and boundary condition datasets at model initialization. The WRF was run
224 using an outermost 36 km resolution domain with one-way nesting for three inner
225 domains at 12 km, 4 km, and 1.33 km grid spacing as illustrated in Figure 1. The 1.33 km
226 domain was used in the analysis for this paper. Thirty-nine vertical levels were used, and
227 the top of the model was 100 hPa. Model physics included the Betts-Miller-Janjic
228 cumulus scheme (Betts and Miller 1993; Janjic 1994) on the 36 km and 12 km domains,
229 Yonsei University (YSU, Hong et al. 2006) planetary boundary layer scheme, and the
230 Unified Noah land surface physics (Ek et al. 2003). Within the 4 and 1.33 km resolution
231 domains, a convective parameterization was not used, and all cloud or precipitation

232 processes were simulated with the WSM6 (Hong et al. 2006), THOM2 (Thompson et al.
233 2008), SBU-YLIN (Lin and Colle 2011), or MORR (Morrison et al. 2009) bulk
234 microphysics schemes. Molthan and Colle (2012; their Table 1) provide a detailed
235 overview of the characteristics of snowfall within the WRF v3.3 schemes used in this
236 study. The WRF model and BMPs were specially configured to output the particle size
237 distribution intercept (N_{os}) and slope parameter (λ_s) of snowfall size distributions, along
238 with parameters necessary to obtain the radar reflectivity and radar reflectivity-weighted
239 fall speeds of precipitating species in each scheme. Molthan and Colle (2012) provide
240 details on the derivation of model reflectivity, size distribution parameters, and fall
241 speeds.

242 A list of cases simulated, their respective initialization times, and initial
243 conditions are given in Table 2, which represents a subset of a larger number of storms
244 evaluated by Colle et al. (2014). The verification point in the WRF model was obtained
245 through a bilinear interpolation of 1.33 km resolution grid boxes nearest to SBNY in each
246 simulation. For the simulations of 19-20 December 2009, the simulated heavy snow
247 band was approximately 58 km southwest of the actual location. In this case, a
248 representative point for SBNY was selected relative to the simulated snow band. The
249 verification points for each simulated case and BMP are shown in Table 3. With the
250 exception of the 7 January 2011 event (~4 hours), each simulation included at least six
251 hours of spin-up time to generate precipitation prior to verification. Other simulations of
252 the 7 January 2011 event with a longer start-up time did not capture the precipitation that
253 occurred over SBNY.

254 *c. Comparisons of volumetric characteristics*

255 Derivations of the model-derived reflectivities and fall velocities are
256 straightforward and computed using assumptions consistent with each of the BMPs
257 (Molthan and Colle 2012). Model-simulated properties were calculated for WRF grid
258 boxes with at least 0.001 g kg^{-1} of hydrometeor mixing ratio, thus, reflectivity and fall
259 speed distributions correspond to model volumes with at least a trace of snow, graupel, or
260 rain. Comparison of model output to observations is more complex as there are several
261 limitations of the observations that preclude direct comparison. As noted previously,
262 MRR observed reflectivity is subject to attenuation when snow accumulates on the MRR
263 antenna. The differences in sensor spatial resolution between the MRR and WSR-88D
264 will manifest most strongly when the storm structure is more spatially heterogeneous and
265 non-uniform beam filling is present (Rinehart 1991). Though many schemes represent
266 sub-grid variability in clouds through a cloud fraction defined in both microphysics and
267 radiation schemes, their representation is not sufficient to account for the same effects of
268 a non-uniformly filled or partially-filled radar resolution volume. In addition, the model-
269 derived reflectivities are not subject to instrument sensitivity constraints and can be
270 computed for lower precipitation ice concentrations than can be detected by either of the
271 two radars. The cm-wavelength MRR and WSR-88D radars do not have sufficient
272 sensitivity to observe the non-precipitating portions of cloud.

273 We compare the radar reflectivity in the vertical column from the WRF model
274 directly over the SBU measurement site to the radar reflectivity observed by the MRR
275 and to the vertical column of WSR-88D data from KOKX taken over the site.
276 Simulations from the innermost, 1.33 km domain are separated into whether there was
277 light, moderate or heavy riming observed at SBNY (Colle et al. 2014). The set of these

278 profiles are accumulated into an asynchronous volume of data from which joint
279 frequency distributions of reflectivity and height using contoured frequency by altitude
280 diagrams (CFADs; Yuter and Houze 1995). We truncate the observed CFADs for the
281 MRR and WSR-88D at the altitude where the number of samples is less than 20% of the
282 maximum number of samples at one level in the volume (Yuter and Houze 1995). Some
283 differences existed in the specific timing between the simulated and observed
284 precipitation as shown in Figure 2 .

285 Similarly, we compare joint frequency distributions of measures of snow fall
286 speed with height between the model column over the measurement site and the MRR.
287 The MRR Doppler velocity is not directly equivalent to the model's computed mean fall
288 speed. Errors in any combination of model vertical air motion, size distribution, size-fall
289 speed relation, and/or particle density would yield errors in the model-computed mean
290 reflectivity-weighted fall speed (V_{cf} , Table 1).

291 As a net result of these differences, we do not expect close quantitative matches
292 between the model and observed Z or fall speed variables. Rather, we focus on the degree
293 of agreement in the trends of the modes of the distributions with height and changes in
294 width of the distribution with height. We also note large differences in trends in
295 maximum Z values with height. The interplay among the model kinematics, microphysics
296 and latent heat release is such that it is not possible to attribute differences solely to
297 individual components within the microphysics parameterizations such as size
298 distributions and particle densities.

299 **3. Model Microphysical Evaluation**

300 *a. Evaluation of simulations during observed light riming events*

301 We first examine combined statistics from 21 occurrences of light riming within nine
302 events (Fig. 2). For these time periods with the light riming designation, surface stereo
303 microscope observations indicate that less than 1% of particles were graupel (Colle et al.
304 2014; their Figure 6). Colle et al. (2014) showed that cold type crystals (side planes and
305 bullets, plates, and needles) were dominant (~80%) during observed, light riming periods
306 (~80%), whereas the schemes examined assume slower-falling dendritic habits. Mean
307 profiles and CFADs of simulated hydrometeor categories are shown in Figures 3 and 4,
308 respectively, for simulations sampled during observed periods of light riming. Mean
309 profiles of cloud ice for the SBU-YLIN and MORR schemes are similar (Figs. 3a, 3d),
310 and frequently less than 0.05 g kg^{-1} within an altitude range of 4-9 km (Figs. 4b, 4n). The
311 WSM6 produces cloud ice throughout the column (Fig. 3b), in sharp contrast to the
312 THOM2 scheme, which produces the smallest amount of cloud ice, confined to 6-10 km
313 (Fig. 3c). In the THOM2 simulations, cloud ice mixing ratios were smaller than 0.05 g kg^{-1}
314 kg^{-1} , or a single CFAD joint histogram cell size (Fig. 4j). Though the four schemes differ
315 in their partitioning of ice mass into cloud ice, snow, or precipitating ice, they produce a
316 similar vertical distribution of total snow and ice mixing ratios. Partitioning of mixing
317 ratios among these categories exaggerates some of their differences. Rather than
318 simulating small crystals through production of cloud ice mixing ratio, the THOM2
319 applies a bi-modal size distribution within the simulated snow category. The MORR,
320 WSM6, and THOM2 schemes produce mean cloud water profiles of 0.02 g kg^{-1} or less
321 within the lowest 4 km (Figs. 3a-c), where mean temperatures range from -15 to 0°C (Fig.
322 5a). The SBU-YLIN scheme increases mean cloud water throughout the column, to 0.06
323 g kg^{-1} at 4 km. Increased cloud water mixing ratio continues through 8 km, inconsistent

324 with surface observations of lightly rimed particles (Fig. 3d). Cloud water CFADs
325 capture infrequent amounts of cloud water content greater than mean values for each
326 scheme, predominately in the lowest 4 km (Fig. 4). Each scheme shows the greatest
327 increases in mean snow (or precipitating ice) mixing ratios between echo top and 4 km
328 where the rate of increase slows and then decreases toward the surface. The MORR and
329 WSM6 schemes produce graupel with mean profile amounts greater than 0.01 g kg^{-1}
330 confined to the lowest 4 km (Figs. 3a-b), with infrequent occurrence of amounts
331 exceeding 0.25 g kg^{-1} (Figs. 4d, 4h). The THOM2 simulations produced very small
332 amounts of graupel with mean values less than 0.001 g kg^{-1} . Graupel is produced by
333 several and different processes within the MORR, WSM6, and THOM2 schemes,
334 however, detailed microphysical process budgets for each simulation are beyond the
335 scope of this study. The SBU-YLIN scheme represents snow and graupel through a rimed
336 precipitating ice category but produces excessively high riming intensities as a result of
337 excessive cloud water (Fig. 6a). The THOM2 scheme produces a larger mean profile of
338 snow and occasional, larger amounts of snow mixing ratio than the MORR and WSM6
339 schemes that produce graupel. Median, liquid-equivalent precipitation from these
340 schemes ranges from $0.1\text{-}0.2 \text{ mm h}^{-1}$ with the highest amounts resulting from the WSM6
341 scheme (Fig. 7).

342 Figure 8 shows the frequency distributions of observed MRR and WSR-88D
343 reflectivity (Z) along with values derived from WRF simulations during light riming
344 periods. In the lowest 2 km, the most frequently occurring (modal) values of MRR
345 reflectivity are around 16-20 dBZ (Fig. 8a), comparable to the WSR-88D reflectivity
346 within the same altitude range (Fig. 8b). As compared to the MRR, the WSR-88D has a

347 broader range of Z values at each height, with small occurrences of values that exceeded
348 those observed by the MRR. The cause for the lack of Z values greater than 24-28 dBZ in
349 the MRR data is not clear.

350 Simulated reflectivity is highly variable and lacks the distinct modes observed by the
351 MRR and WSR-88D, suggesting that observed precipitation structures are more uniform
352 at various altitudes than the corresponding model simulations during light riming events
353 (Fig. 8c-f). Both the modal and maximum Z values observed by the WSR-88D increased
354 between 6 km and 1 km altitude consistent with increase in particle sizes via depositional
355 growth and aggregation. The increase was from 8 to 18 dBZ for the modal values and 18
356 to 34 dBZ for the maximum values. The THOM2 scheme has a similar trend of
357 maximum values with height while the other three schemes have maximum reflectivities
358 that are too high from 4-6 km. The THOM2 and SBU-YLIN schemes produce reflectivity
359 distributions comparable to observations in the lowest 4 km while the WSM6 and MORR
360 schemes exceed the observed reflectivity distribution from the WSR-88D. The higher
361 reflectivity values in the WSM6 and MORR simulations than observed likely result from
362 the prediction of graupel, representing moderate to heavily rimed particles in contrast to
363 observed, light riming.

364 Figure 9 shows the distributions of observed Doppler velocity and simulated fall
365 speed variables. Throughout the vertical column, there is a fairly consistent range in
366 Doppler velocities observed by the MRR, from 0.3-2.0 m s⁻¹ and above 4 km altitude, the
367 mode in MRR Doppler velocity is less distinct than lower levels (Fig. 9a). The most
368 frequently observed Doppler velocities of around 1.0 m s⁻¹ are consistent through an
369 altitude of 4 km. There is broadening of the observed fall speed distribution to nearly

370 1.75 m s^{-1} below 2 km associated with the increased particle growth and reflectivity
371 increase in this layer. The MORR and THOM2 fall speed distributions are narrower than
372 observed below 2 km altitude (Figs. 9b, 9d), while the SBU-YLIN (Fig. 9e) has a second
373 mode below 5 km altitude at high fall speed values, inconsistent with observations.
374 Schemes incorporating a temperature-dependent size distribution, such as the WSM6,
375 THOM2, and SBU-YLIN have trends of increasing fall speed with decreasing height
376 between 3 and 7 km. The MORR scheme has less of a change in fall speed with height
377 and is more consistent with observations. Unfortunately, lack of data from the MRR in
378 the lowest 1 km precludes validation of the $\sim 0.2 \text{ m s}^{-1}$ increase in fall speed peak
379 frequency for the MORR. At 1-2 km, where the MRR provides observations, the MORR
380 and WSM6 simulations provide the best match to MRR fall speeds ($\sim 1.0 \text{ m s}^{-1}$) while the
381 THOM2 and SBU-YLIN slightly underestimated fall speeds by around 0.25 m s^{-1} .

382 *b. Evaluation of simulations during observed moderate riming events*

383 During moderate riming periods (21 occurrences within seven events; Fig. 2) the
384 observed snow contained about 50% dendrites and plates, 20-25% needles, less than 10%
385 cold type crystals, and small amounts ($< 4\%$) of graupel (Colle et al. 2014). All schemes
386 increase their predicted mean snow and combined cloud ice and snow mass (Fig. 10),
387 shown as higher frequency of larger mixing ratios, particularly in the lowest 3-4 km (Fig.
388 11). Cloud water also increases in all schemes, particularly in the lowest 2 km, along
389 with increases in the mean profile through 6 km. Increases occur at a range of
390 temperatures from -15°C to 0°C (Fig. 5a), where the simulations for these moderate
391 riming events average as much as 1.5°C warmer than light riming cases (Fig. 5b). The
392 MORR scheme exhibits an increase in mean cloud water and frequency from 1-4 km

393 where mean temperatures range from -15 to -5°C, up to 1.5°C warmer than light riming
394 cases. Overall, the most frequent mixing ratio of cloud water remains less than 0.1 g kg⁻¹,
395 and the mean value is 0.05 g kg⁻¹ or less for all schemes except the SBU-YLIN. The
396 simulated cloud water amount is less than expected for periods of moderate riming. For
397 example, Lin and Colle (2011) and Lin et al. (2011) showed for two cases over the
398 Washington Cascades that observed and simulated cloud water was 0.1 to 0.3 g kg⁻¹ for
399 moderate riming periods. Some of the cloud water in the SBU-YLIN scheme appears to
400 be erroneously high (> 0.3 g kg⁻¹), resulting in heavily-rimed precipitating ice, increased
401 precipitation, and decrease in snow mass by fallout in the lowest 3 km. Excess
402 production of cloud water may be related to issues with the saturation adjustment process
403 in the SBU-YLIN scheme (Molthan and Colle 2012). Although the aforementioned
404 schemes produce some additional snow and cloud water, the MORR and WSM6
405 simulations produce amounts and frequencies of graupel comparable to light riming
406 simulations with similar maximum values and frequencies (Figs. 11d, 11h). Mean
407 profiles of graupel in moderate riming events are similar to light riming events for the
408 MORR, WSM6, and THOM2 profiles (Fig. 10a-c).

409 The CFAD of MRR reflectivity has a distribution mode below 2 km altitude that is
410 sharper and greater than light riming periods (20 dBZ, Fig. 12a). The modal value of
411 WSR-88D reflectivity in the lowest 2 km is similar at 20-24 dBZ (Fig. 12b). Both MRR
412 and WSR-88D distributions of Z indicate a steady increase in the modal value of
413 reflectivity with decreasing altitude from 6-2 km. Such “diagonalization” of the
414 reflectivity CFAD indicates growth of particles as they descend (Yuter and Houze 1995).
415 As with light riming cases, the CFAD of WSR-88D reflectivity included small

416 frequencies of higher reflectivity near the surface, as high as 32-34 dBZ. Similarly, the
417 MRR data included small frequencies of reflectivity from 24-28 dBZ. All four models
418 exhibit clear modes in the reflectivity distribution that increased with decreasing height,
419 similar to the diagonalization seen in observations, though their modal values and ranges
420 differ from the MRR and WSR-88D data. The MORR and WSM6 schemes (Figs. 12c-d)
421 produce modal and maximum reflectivity exceeding WSR-88D and MRR observations
422 near the surface (Fig. 12c-d). The THOM2 and YLIN schemes are a better match to
423 observed reflectivity values and trends with altitude in terms of modal and maximum
424 values (Figs. 12e-f).

425 The MRR Doppler velocity distribution produces a mode that increases with
426 decreasing altitude from to 1.0 m s^{-1} at 4 km altitude to 1.25 m s^{-1} at the surface (Fig.
427 13a). The mode in the MRR near-surface Doppler velocity increases by 0.25 m s^{-1} versus
428 the light riming periods. In the WSM6 simulations, an increase in predicted snow mass
429 and larger particles inferred from radar reflectivity contributes to an overall increase in
430 mean fall speeds. Excessive fall speeds above 5 km in WSM6 likely result from
431 erroneously large particles associated with simulated reflectivity greater than observed by
432 the MRR and WSR-88D.

433 The MORR and THOM2 schemes produce vertical profiles of modal fall speeds
434 for the moderate riming periods similar to their performance during light riming periods
435 despite increases in snow and graupel content (Figs. 9 and 13). In WSM6, the modal fall
436 speeds increase between 6 km and 2 km but do not change much above or below that
437 layer. The inference is that increases in snow content from additional riming did not
438 translate to increases in fall speeds through changes in their diameter-fall speed

439 relationships. In addition, cloud water is likely underpredicted during moderate riming
440 events, contributing to an underprediction of fall speeds at lower levels even if the
441 schemes accounted for varied riming conditions. In SBU-YLIN, riming effects are
442 allowed to influence fall speeds, but excessive cloud water contributed to high riming
443 factors and exacerbated fall speed errors previously observed in light riming events (Fig.
444 6b). These errors resulted in isolated occurrences of fall speeds of $1.5\text{-}3\text{ m s}^{-1}$,
445 comprising as much as 15% of fall speeds in the lowest 1-2 km (Fig. 13e).

446 *c. Surface size distribution and fall speed*

447 In Figures 14-16, we compare observed and simulated surface size distributions
448 and fall speed measures for four short cases from 1 to 4 hours in duration. These cases
449 highlight some details of the representation of particles as a function of degree of riming
450 and whether aggregation is present. The BMP scheme size distributions are average
451 values for the set of 15-minute intervals in each case and are compared to the observed
452 distribution of PARSIVEL diameter (Section 2a). For context, we also show average
453 mixing ratio profiles for snow, cloud water and graupel in the lowest 3 km. We compare
454 distributions of PARSIVEL fall speed (or settling speed, $w=0$ and $E\neq 0$, Table 1), to a
455 computed mean mass-weighted fall speed ($w=0$, $E=0$, Table 1) for each 15 min model-
456 simulated period that includes contributions from snow, graupel, and rain.

457 *i. No riming*

458 A period of mainly cold type crystals (51% side planes and 20% bullets) occurred
459 from 0145 to 0500 UTC on 16 February 2010, with little or no riming observed. A small
460 amount of plates ($\sim 10\%$) and columns ($\sim 10\%$) were also observed with light riming. The

461 observed and simulated size distributions for this period are illustrated in Figure 14a. All
462 four BMPs slightly underestimate the number concentrations of aggregated snow
463 particles with $D < 8$ mm, with the MORR and SBU-YLIN closest to the observed for
464 diameters from 6 to 8 mm. Each scheme simulates between 0.1 and 0.15 g kg^{-1} of snow
465 above 1 km, but decreases the amount to around 0.05 g kg^{-1} at the surface (Fig. 14b). This
466 reduction results from sublimation, since these cold type habits are mainly observed near
467 the outer edges of the comma head (Colle et al. 2014), where the low-levels are still
468 moistening. The reduction in snow mixing ratio near the surface may contribute to an
469 overall reduction in particle number concentrations as shown by BMPs that predict lower
470 number concentrations of particles across all observed size bins.

471 The observed fall speed distribution is generally between 1 and 1.5 m s^{-1} (Fig.
472 14c). The MORR and SBU-YLIN schemes have particle fall speeds clustered around 1 m
473 s^{-1} , with a few values at ~ 1.25 m s^{-1} (Fig. 14d), while the WSM6 and the THOM2
474 produces fall speeds slower than the peak in the observations, ranging from 0.5 to 0.75 m
475 s^{-1} . These schemes simulate a small amount of cloud water (< 0.05 g kg^{-1}) above 1 km
476 (Fig. 14b). There was little riming observed during this event as well, so the observed fall
477 speeds (tail > 1.5 m s^{-1}) are likely related to faster falling cold type crystals compared to
478 the conventional plates and dendrites used in these schemes. Underestimation of surface
479 fall speeds in this sample of observed crystals is comparable to the underestimate of fall
480 speeds in the broader sampling of light riming simulations (Fig. 9), therefore, schemes
481 may not be accounting for faster fall speeds for cold type crystals. The SBU-YLIN
482 includes a temperature dependence term for fall speeds, but it is based on the local

483 temperature and not necessarily where the snow particles are formed, and the observed
484 side planes and bullets are likely formed in the middle and upper levels of the cloud.

485 ii. Light to moderate riming

486 A mix of 70% plates and 16% side planes was observed from 1000 to 1200 UTC
487 on 21 February 2011 with observed riming intensities that range from none to moderate,
488 with the peak riming intensity occurring at 1115 UTC. The WSM6 scheme slightly
489 underestimates number concentrations of particles across all sizes, and the SBU-YLIN
490 simulates a greater number concentration of particles than observed for all sizes, with the
491 exception of particles around 2 mm (Fig. 15a). The MORR and THOM2 schemes
492 produce number concentrations of particles similar to observations for diameters 1 mm or
493 greater (Fig. 15a).

494 Each scheme simulates snow mixing ratios of 0.15 to 0.25 g kg⁻¹ near the surface
495 (Fig. 15b) and the SBU-YLIN scheme has a small amount (< 0.05 g kg⁻¹) of cloud water
496 between 1.5 and 2 km. However, in general, all schemes have very little cloud water,
497 which likely contributes to the lack of fall speeds greater than 1.5 m s⁻¹ (Fig. 15d). Most
498 of the observed fall speeds in this two-hour period were between 1 and 1.5 m s⁻¹ (Fig.
499 15c). All BMPs concentrate their fall speeds around 1 m s⁻¹ (Fig. 15d), while the
500 observed peak was slightly greater at ~1.1 m s⁻¹ (Fig. 15c). The small amount of cloud
501 water within the SBU-YLIN scheme resulted in an increase in the diagnosed riming
502 factor for the precipitating ice class and some of the greater fall speeds. This also likely
503 contributes to larger standard deviations in the particle size distribution when the scheme
504 transitioned from between periods of rimed and unrimed precipitation.

505 iii. Light to moderate riming and many aggregates

506 Figure 16a shows the observed and simulated size distributions for a time period
507 with 65% dendrites and 20% plates observed from 1530 UTC 26 February 2010 to 0000
508 UTC 27 February 2010. During this period the observed degree of riming is light to
509 moderate (0.5 to 2.5) and many aggregates of dendrites are also observed (not shown).
510 The particle size distributions from the examined BMPs are similar to the observed size
511 distributions for particles smaller than 4 mm, but the BMPs underestimate the number
512 concentrations of particles larger than 4 mm. The particle size distribution from the
513 MORR scheme produces slightly larger particles than the other BMPs, and thus a better
514 fit to observations. One hypothesis for the underestimation of the number concentrations
515 of larger ($D > 4$ mm) particles is a poor representation of snow aggregation, or the shift to
516 particle size distributions comprised of larger particles at the expense of smaller crystals
517 (Fig. 16b). Each scheme produces 0.05 to 0.1 g kg⁻¹ of snow below 1 km but varied in
518 their production of cloud water, ranging from 0.05 to 0.15 g kg⁻¹ of cloud water between
519 1 and 2 km. Riming of snow is implied by the collocation of snow and cloud water in the
520 models. An increase in snow content rather than cloud water might have contributed to
521 larger numbers of larger particle sizes in modeled size distributions.

522 Unlike the above cases and the combined CFAD results, the THOM2, MORR,
523 and SBU-YLIN schemes are able to produce more cloud water (0.05 to 0.15 g kg⁻¹). The
524 distribution of observed fall speeds peak at ~ 0.75 m s⁻¹ and ~ 1.1 m s⁻¹ (Fig. 16c), with a
525 tail to fall speeds exceeding 2 m s⁻¹. Model-simulated fall speeds are clustered between
526 0.8 and 1.1 m s⁻¹ (Fig. 16d). The better fall speed prediction in the model, including
527 some fall speeds greater than 1 m s⁻¹ is likely the result of better simulation of the cloud

528 water. The scheme with the least amount of cloud water (WSM6) has the worst fall
529 speed prediction near the surface (peaking around 0.8 m s^{-1}). The THOM2 has several
530 periods with fall speeds from ~ 2 to 3 m s^{-1} for mixing ratios below 0.1 g kg^{-1} (Fig. 16d).
531 These faster fall speeds result from trace amounts of faster-falling graupel, or brief
532 production of drizzle with 0.005 to 0.01 g kg^{-1} of rain water simulated at the surface
533 between 2215 and 0000 UTC (not shown). Aggregates and lightly rimed snow likely
534 comprised the smaller peak ($\sim 0.75 \text{ m s}^{-1}$), while the moderately rimed habits likely fell
535 within the second peak ($\sim 1.1 \text{ m s}^{-1}$) and the tail of higher fall speeds. As observed in the
536 CFAD analysis, larger riming factors in the SBU-YLIN scheme contribute to faster fall
537 speeds of 1.25 to 1.75 m s^{-1} for predicted snow mixing ratios less than 0.1 g kg^{-1} (Fig.
538 16d).

539 **4. Conclusions**

540 Reflectivity and fall speed from four BMPs (MORR, WSM6, THOM2, and SBU-
541 YLIN) run down to 1.33 km grid spacing within the WRF model were compared to
542 vertically-pointing radar observations at Stony Brook, NY for nine snow events that were
543 partitioned into periods of observed riming intensity. Comparisons of observed and
544 modeled particle size distributions and fall speeds at the surface were made for selected
545 periods with distinct sets of crystal habits. Motivating research questions sought to
546 examine whether the selected schemes were able to reproduce key characteristics of the
547 observed distributions of reflectivity and fall speed within various categories of observed
548 riming. In light riming periods, the WSM6 and MORR schemes produced larger
549 reflectivities (Z) than observed, particularly in the lowest 4 km where they produced
550 higher-density graupel particles inconsistent with the light degree of riming observed at

551 the surface. The THOM2 scheme only produced trace, insignificant amounts of graupel
552 and the SBU-YLIN scheme limited the occurrence of higher riming factors, with a better
553 representation of observed WSR-88D reflectivity in the surface to 4 km altitude range.
554 These results encourage a more detailed examination of graupel sources within the
555 WSM6 and MORR simulations to reduce the generation of higher-density graupel
556 particles in periods of observed, light riming. For fall speed variables, MRR observations
557 captured a consistent mode around 1 m s^{-1} in the lowest 1-2 km. The THOM2 and SBU-
558 YLIN schemes produced a consistent mode in the same range, but underestimated fall
559 speeds by around 0.25 m s^{-1} . An analysis of the 16 February 2010 event suggests that
560 this may be related to the underpredicted fall speeds of cold-type snow habits.

561 During moderate riming, the THOM2 and SBU-YLIN were both closer to Z
562 observations in terms of the most frequent and maximum values at varying altitudes. The
563 MORR and WSM6 simulations produced modal and maximum values of Z that exceeded
564 observations. As in light riming cases, the MORR and WSM6 schemes produced higher-
565 density graupel within the lowest 4 km, which may contribute to their much higher than
566 observed values of radar reflectivity. The MORR, WSM6, and THOM2 schemes
567 increased the amount of cloud water between 1 and 4 km as the observed riming intensity
568 increased from light to moderate categories, but the amounts were less than expected for
569 moderate riming, based on previous field studies. The SBU-YLIN scheme produced the
570 largest mean cloud water profile and infrequent occurrences of higher cloud water
571 amounts from 4-6 km, contributing to an increase in particle density and slight
572 overestimation of radar reflectivity.

573 There was relatively little change in the near-surface fall speeds with increasing
574 riming category among the MORR, WSM6, and THOM2 simulations, and as a result
575 modeled fall speeds underestimated MRR-obtained surface fall speeds by 0.25 to 0.5 m s⁻¹
576 ¹. These schemes may not produce enough cloud water during these winter storms to
577 capture the observed riming processes (e.g., there is little cloud water in the THOM2
578 scheme), or the snow fall speeds do not account for increases in particle fall speeds
579 during observed moderate riming periods, likely because they assume dry snow fall
580 speeds unless the scheme contributes mass to the graupel category. The THOM2 and
581 WSM6 schemes incorporate temperature-dependent particle size distribution
582 characteristics for snowfall that provided an increase in fall speeds approaching the
583 surface but were ~0.25 to 0.50 m s⁻¹ slower than surface observations. The MORR and
584 THOM2 simulations produce similar fall speed distributions with height regardless of
585 riming category.

586 Comparisons between surface and simulated particle size distributions suggest
587 that while BMPs simulate reasonable number concentrations of particles smaller than 4
588 mm, there was an underestimate of the number concentrations particles larger than 4 mm.
589 It is hypothesized that underestimation of the number of larger particles results from
590 BMPs not adequately simulating the aggregation process, or by allowing for particle size
591 distribution parameters to evolve to smaller intercept and slope parameters as aggregation
592 occurs. Fixed parameters of density and spherical shape may be a limiting factor, though
593 schemes such as THOM2 and SBU-YLIN that incorporate variable bulk density did not
594 markedly improve performance. Lower density aggregates cannot be well represented in
595 a BMP that assumes a fixed density for snow, such as used in the WSM6 and MORR

596 schemes. However, the double moment MORR scheme seemed to provide a better
597 comparison to observations than other schemes during periods of aggregation, perhaps
598 benefitting from greater flexibility in the determination of size distribution parameters by
599 predicting both mass and total number concentration. Schemes that predicted higher-
600 density and more heavily rimed graupel particles during light and moderate riming
601 periods resulted in excessive radar reflectivity contrary to radar observations and the lack
602 of these heavily rimed particles at the surface. However, schemes that produced more
603 unrimed snow were not able to capture increases in fall speed during observed moderate
604 riming periods, suggesting that they were unable to predict the observed changes in
605 riming degree.

606 For schemes that favor the production of higher-density graupel rather than
607 unrimed snow, future work should examine opportunities for a smoother transition
608 between the dry and heavily rimed ice categories to improve the representation of a
609 broader range of riming categories. Additional vertical levels should be included to
610 better capture convective-scale processes contributing to the development of cloud water
611 or small-scale microphysical variability. Such an examination would be best achieved
612 with detailed in situ observations from multiple riming regimes to guide improvements,
613 supplemented by additional radar remote sensing comparisons where in situ data are
614 unavailable.

615

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811 **List of Tables:**

812 1 Snow particle fall speed variables. Vertical air motion is w and turbulence is E . In
813 the third and fourth equations below, w is assumed to be constant within a radar
814 resolution volume and a WRF model grid box.

815 2 Initialization time, total run time, and initial and boundary conditions used in the
816 cases simulated with the WRF V3.3 model.

817 3 Verification points used for the WRF model validation results.

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832 Table 1. Snow particle fall speed variables. Vertical air motion is w and turbulence is E .
 833 In the third and fourth equations below, w is assumed to be constant within a radar
 834 resolution volume and a WRF model grid box.

Name		Description	Reference
Terminal velocity	V_t	Velocity of hydrometeor in still air ($w=0, E=0$) usually measured for individual particles	Locatelli and Hobbs (1974)
Settling speed	V_s	$V_s = V_t + E$ ($w=0$). Usually measured for individual particles	Wang and Maxey 1993; Nielsen 1993
Mean Doppler velocity	V_r	$V_r = \frac{\int_D (V_t(D) + E) Z(D) dD}{\int_D Z(D) dD} + w,$ within a vertically-pointing radar resolution volume.	Doviak and Zrnich (1993)
Computed mean reflectivity-weighted fall speed	V_{cf}	$V_{cf} = \frac{\int_D V_t(D) Z(D) dD}{\int_D Z(D) dD} + w,$ within a WRF model grid box. ($E=0$)	Calculated from scheme assumptions
Computed mean mass-weighted fall speed	V_m	$V_m = \frac{\int_D V_t(D) m(D) dD}{\int_D m(D) dD},$ within a WRF model surface grid box ($w=0, E=0$).	Calculated from scheme assumptions

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838 Table 2. Initialization time, total run time, and initial and boundary conditions used in the
 839 cases simulated with the WRF V3.3 model.

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Case	Initialization Time	Total Run Time (Hours)	Initial and Boundary Conditions
19-20 December 2009	19 December 2009 1200 UTC	24	NAM 218
8 January 2010	8 January 2010 0000 UTC	18	NAM 218
28 January 2010	28 January 2010 0000 UTC	18	NAM 218
16 February 2010	15 February 2010 1800 UTC	30	NAM 218
26 February 2010	26 February 2010 0000 UTC	24	NAM 218
7 January 2011	7 January 2011 1200 UTC	12	NAM 218
21 February 2011	21 February 2011 0000 UTC	18	0.5° GFS
21 January 2012	21 January 2012 0000 UTC	24	NAM 218
11 February 2012	11 February 2012 0000 UTC	18	0.5° GFS

841

842 Table 3. Verification points used for the WRF model validation results.

843

19-20 December 2009		
BMP	Verification Point	Location
WSM6	40.7220, -73.7655	Queens, NY
THOM2	40.7720, -73.8754	La Guardia, NY
SBU-YLIN	40.6910, -73.9757	Brooklyn, NY
MORR	40.7428, -73.9908	Manhattan, NY
Remaining Simulated Events		
All	40.9044, -73.1184	SBNY

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845 **List of Figures**

- 846 1 WRF model domains used in this study.
- 847 2 Observed light or moderate riming periods and corresponding simulation
848 times when model profiles were extracted for performing comparisons.
- 849 3 Mean profiles of hydrometeor content for selected microphysics schemes and
850 simulations sampled during observed periods of light riming shown in Figure
851 2. Combined mixing ratios of cloud ice and snow are shown as a black
852 dashed line.
- 853 4 Contoured frequency with altitude diagrams (CFADs) for hydrometeor
854 species obtained from surveyed microphysics schemes during the light riming
855 periods shown in Figure 2. Mean profiles of each hydrometeor type are
856 provided as a solid black line. Trace amounts of cloud ice and graupel in the
857 THOM2 scheme are shown as mean profiles in Figure 3. Note that the x-axis
858 scale for the snow or precipitating ice column differs from the remaining
859 panels.
- 860 5 (left) Mean temperature profiles and standard deviations of temperature
861 during periods of light and moderate riming shown in Figure 2, and (right)
862 difference of the mean temperature profile between moderate and light riming
863 periods.
- 864 6 CFADs (shaded) and mean profiles (black line) for the unitless riming factor
865 used to parameterize the mass-diameter and diameter-fall speed relationships
866 for the precipitating ice category within the SBU-YLIN scheme, partitioned
867 into model simulations of observed a) light and b) moderate riming periods
868 shown in Figure 2.

869 7 Box and whisker plots of liquid equivalent precipitation from various
870 microphysics schemes accumulated for light (left, light shading) and moderate
871 (right, heavier shading) precipitation events. Shaded regions bound the first
872 and third quartiles with median values inset. Extended, dashed lines represent
873 the 10th and 90th percentiles.

874 8 CFADs of observed reflectivity (dBZ) for light riming periods shown in
875 Figure 2. (a) from the MRR and (b) from the WSR-88D at OKX. (c)
876 Simulated reflectivity (dBZ) from the MORR scheme, (d) from the WSM6
877 scheme, (e), from the THOM2 scheme, and (f) from the SBU-YLIN scheme.
878 The dashed line at 5.25 km in panel a) and 6 km in panel b) indicates the
879 altitude at which point the observed CFADs were truncated aloft due to the
880 limited number of observations above these altitudes, as described in Section
881 2b.

882 9 CFADs of fall speed variables (positive downward, m s⁻¹) for light riming
883 periods shown in Figure 2. (a) Doppler velocity observed from the MRR, and
884 (b) computed mean reflectivity-weighted fall speed simulated from the
885 MORR scheme, (c) from the WSM6 scheme, (d), from the THOM2 scheme,
886 and (e) from the SBU-YLIN scheme. The dashed line at 5.25 km in panel a)
887 indicates the altitude at which point the observed CFADs were truncated aloft
888 due to the limited number of observations above these altitudes, as described
889 in Section 2b.

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891 11 As in Figure 4 but for moderate riming periods shown in Figure 2. Note that
892 the x-axis scale for the snow or precipitating ice column differs from the
893 remaining panels.

894 12 As in Figure 8, but for moderate riming periods shown in Figure 2.

895 13 As in Figure 9 but for moderate riming periods shown in Figure 2.

896 14 Period of 51% side planes and 20% bullets observed from 0145 UTC to 0500
897 UTC on 16 February 2010. (a) Observed and simulated surface size
898 distribution, (b) mean mixing ratio for snow, cloud water, and graupel (g kg^{-1}),
899 (c) the distribution of PARSIVEL settling speeds, V_s (m s^{-1}), normalized to
900 the number of particles every 15 minutes, and (d) mean mass-weighted fall
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902 g kg^{-1}). Error bars represent one standard deviation above and below the
903 simulated size distribution. The diameter* for panel a) notes that the panel
904 compares the “PARSIVEL diameter” for observations discussed in Section 2,
905 and the diameter of assumed, spherical and frozen hydrometeors within the
906 model, where schemes assume a single crystal habit.

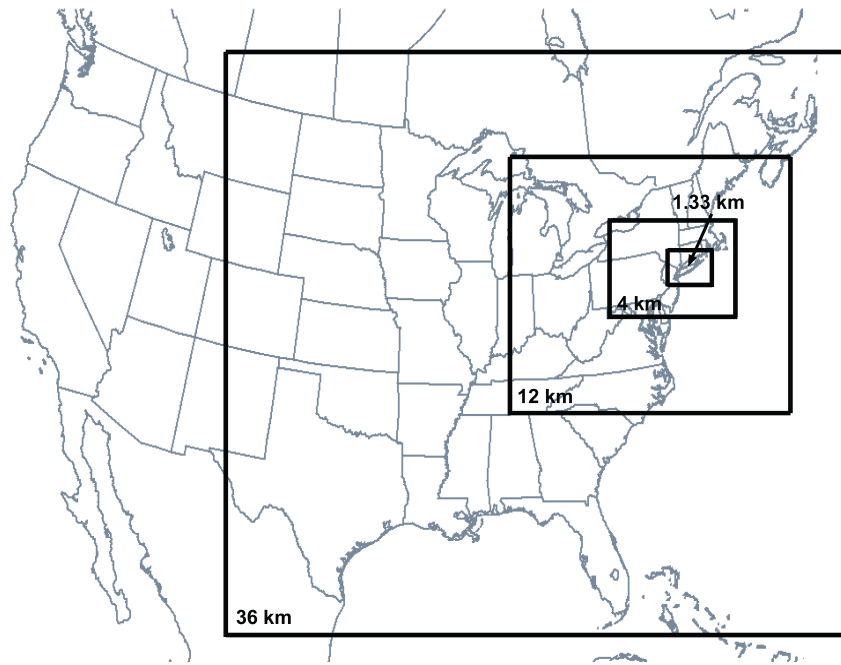
907 15 As in Fig. 8 except during a period of 70% plates and 16% side planes
908 observed from 1000 to 1200 UTC 21 February 2011.

909 16 As in Fig. 8 except during a period of 65% dendrites and 20% plates observed
910 from 1530 UTC on 26 February 2010 to 0000 UTC on 27 February 2010.

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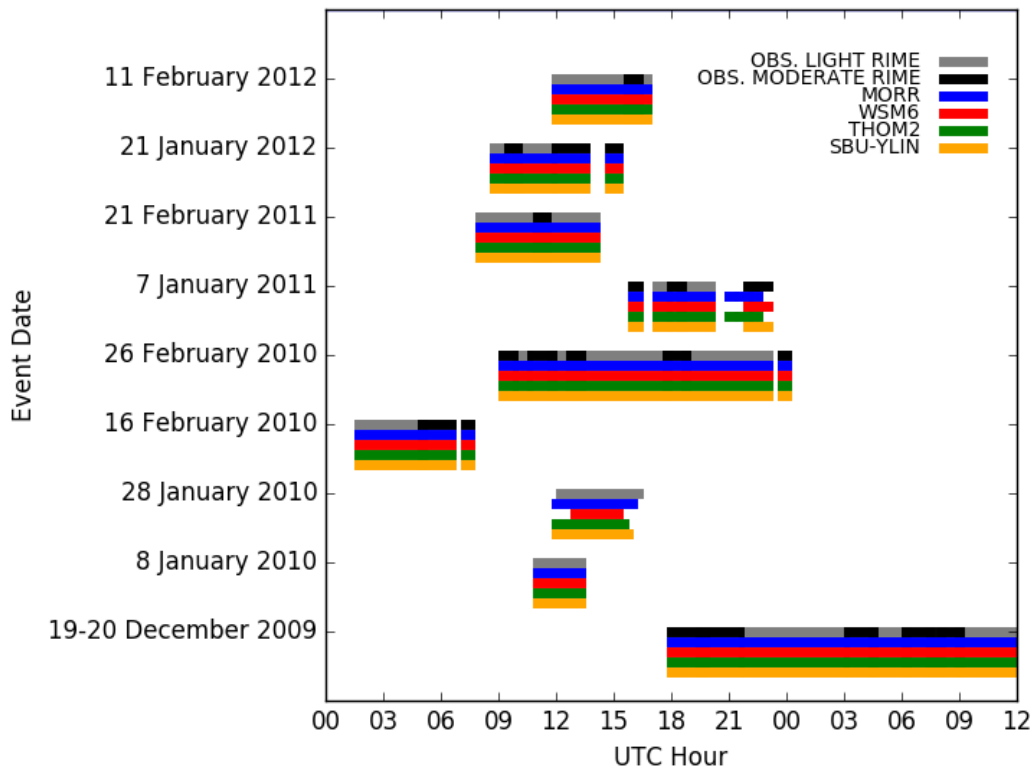
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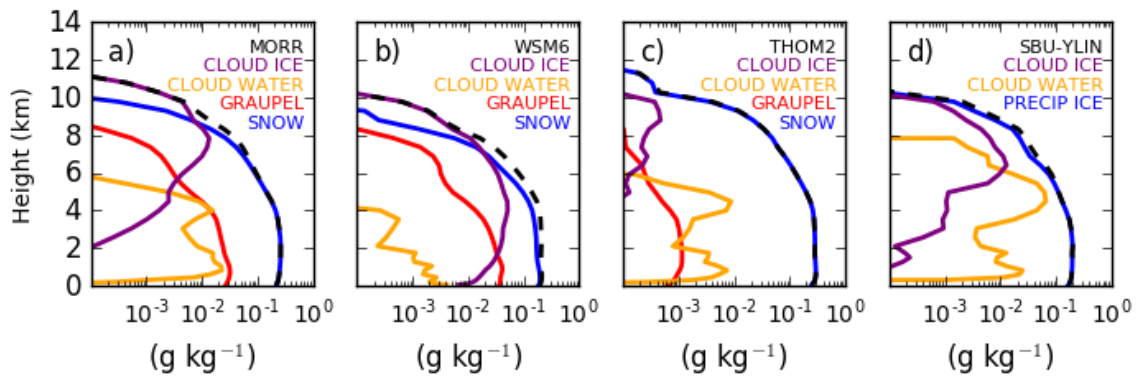
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915 Figure 1. WRF model domains used in this study.



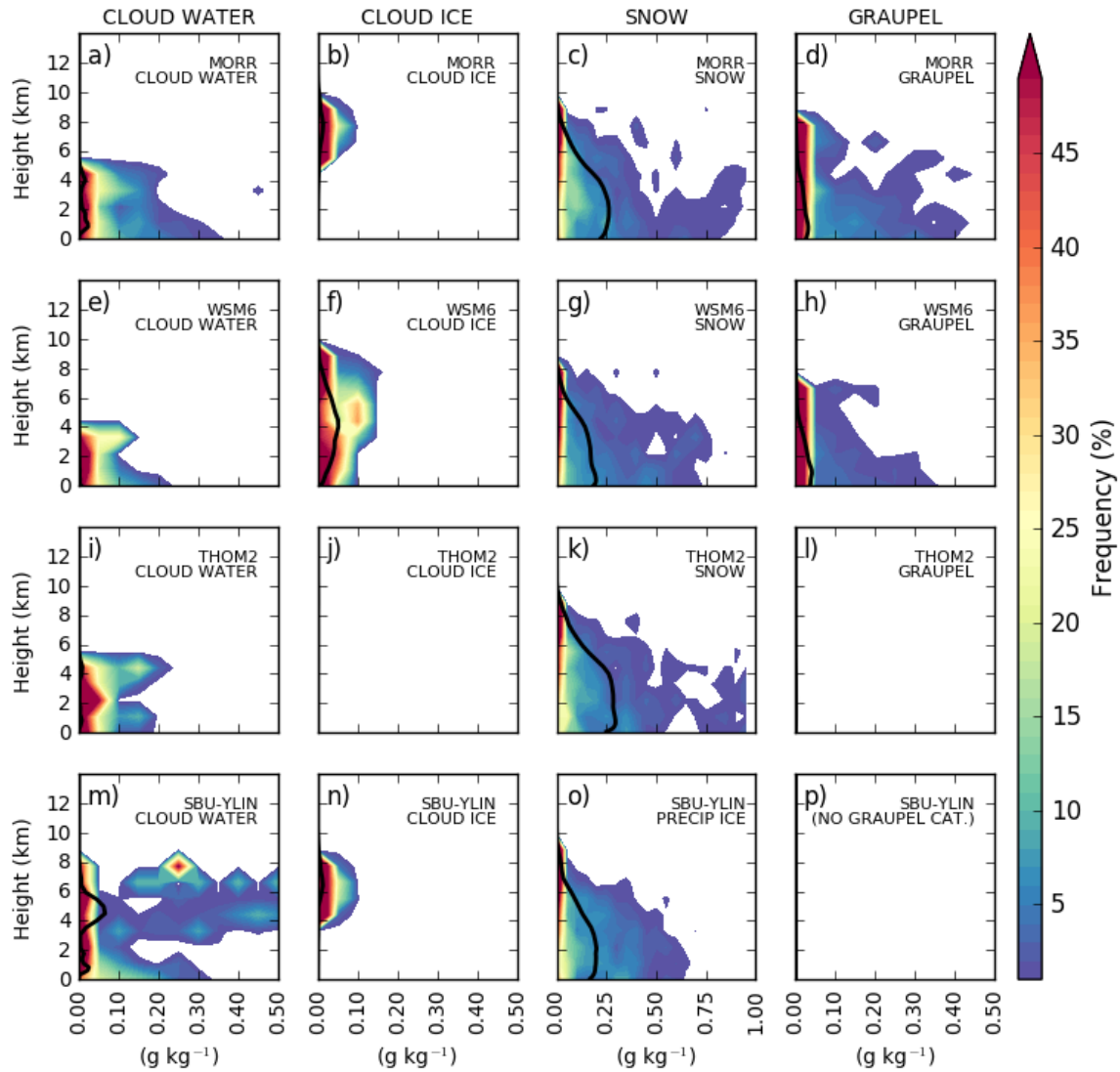
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 918 when model profiles were extracted for performing comparisons.



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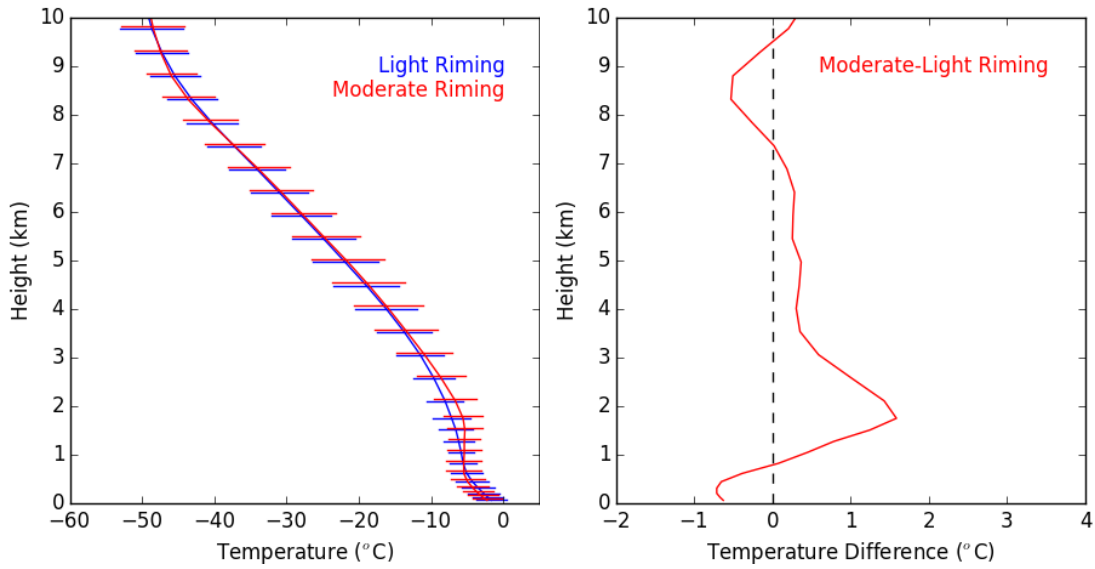
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 921 simulations sampled during observed periods of light riming shown in Figure 2.
 922 Combined mixing ratios of cloud ice and snow are shown as a black dashed line.



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924 Figure 4. Contoured frequency with altitude diagrams (CFADs) for hydrometeor species
 925 obtained from surveyed microphysics schemes during the light riming periods shown in
 926 Figure 2. Mean profiles of each hydrometeor type are provided as a solid black line.
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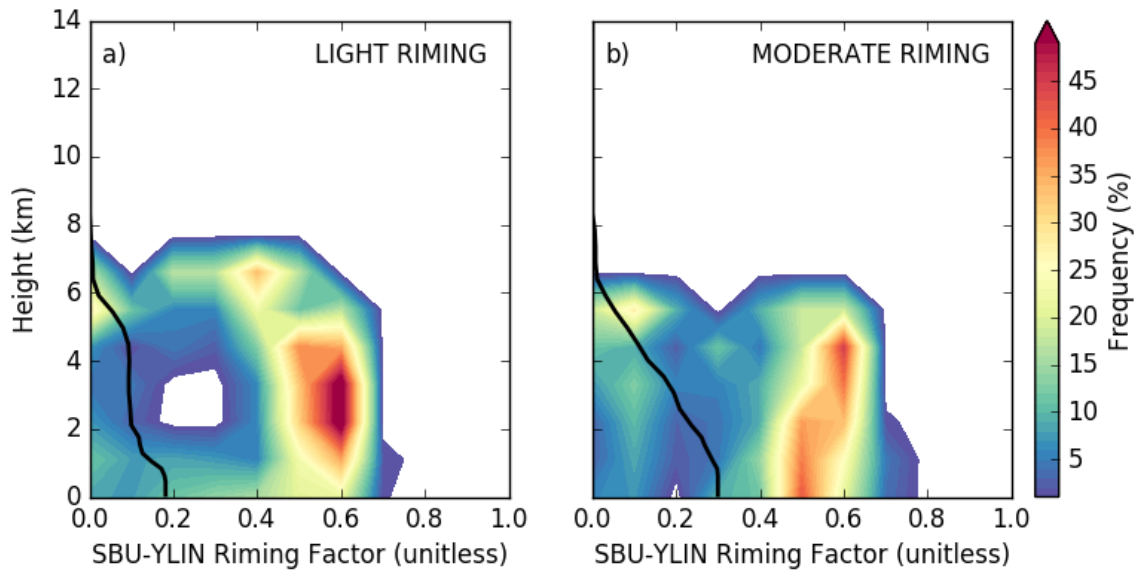
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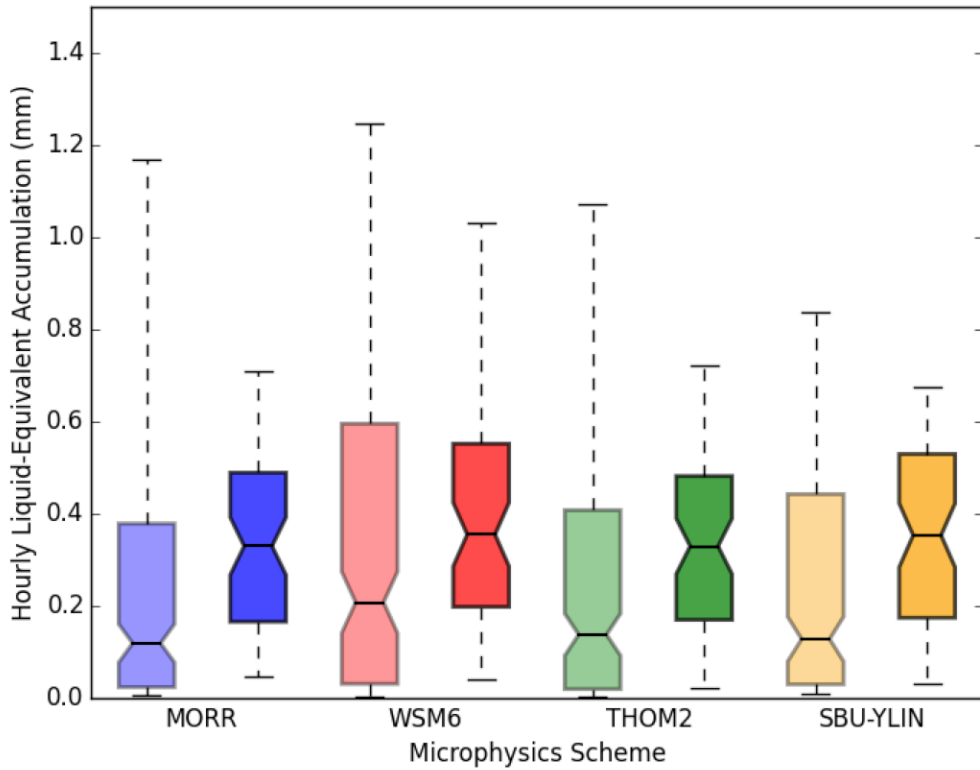
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 933 periods of light and moderate riming shown in Figure 2, and (right) difference of the
 934 mean temperature profile between moderate and light riming periods.

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937 Figure 6. CFADs (shaded) and mean profiles (black line) for the unitless riming factor
 938 used to parameterize the mass-diameter and diameter-fall speed relationships for the
 939 precipitating ice category within the SBU-YLIN scheme, partitioned into model
 940 simulations of observed a) light and b) moderate riming periods shown in Figure 2.



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942 Figure 7. Box and whisker plots of liquid equivalent precipitation from various
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 945 with median values inset. Extended, dashed lines represent the 10th and 90th percentiles.

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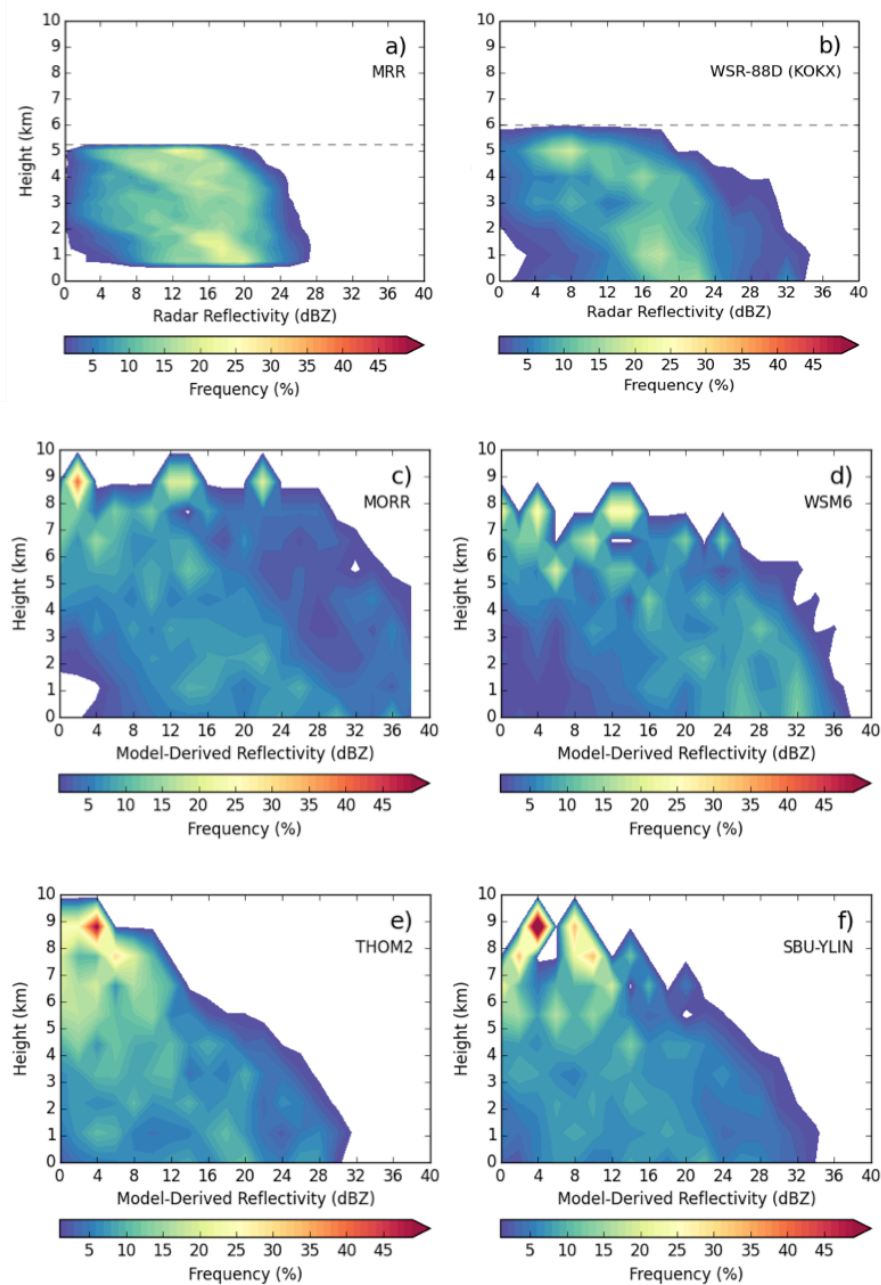
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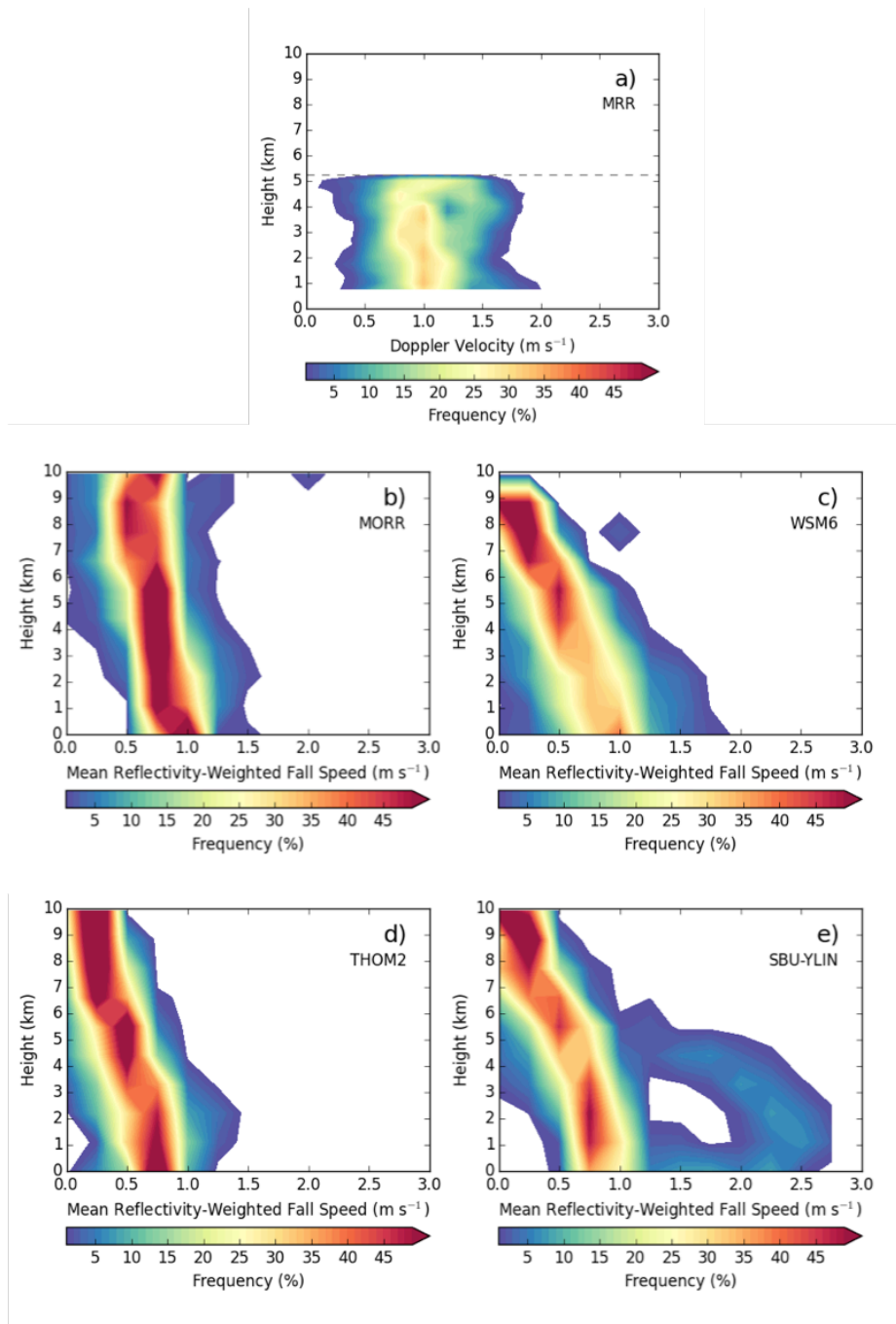
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954 Figure 8. CFADs of observed reflectivity (dBZ) for light riming periods shown in Figure
 955 2. (a) from the MRR and (b) from the WSR-88D at OKX. (c) Simulated reflectivity
 956 (dBZ) from the MORR scheme, (d) from the WSM6 scheme, (e), from the THOM2
 957 scheme, and (f) from the SBU-YLIN scheme. The dashed line at 5.25 km in panel a) and
 958 6 km in panel b) indicates the altitude at which point the observed CFADs were truncated
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 960 Section 2b.

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 965 computed mean reflectivity-weighted fall speed simulated from the MORR scheme, (c)
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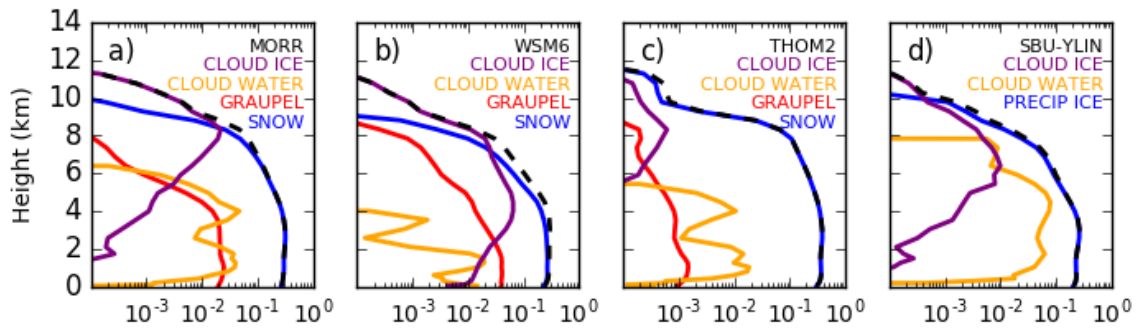
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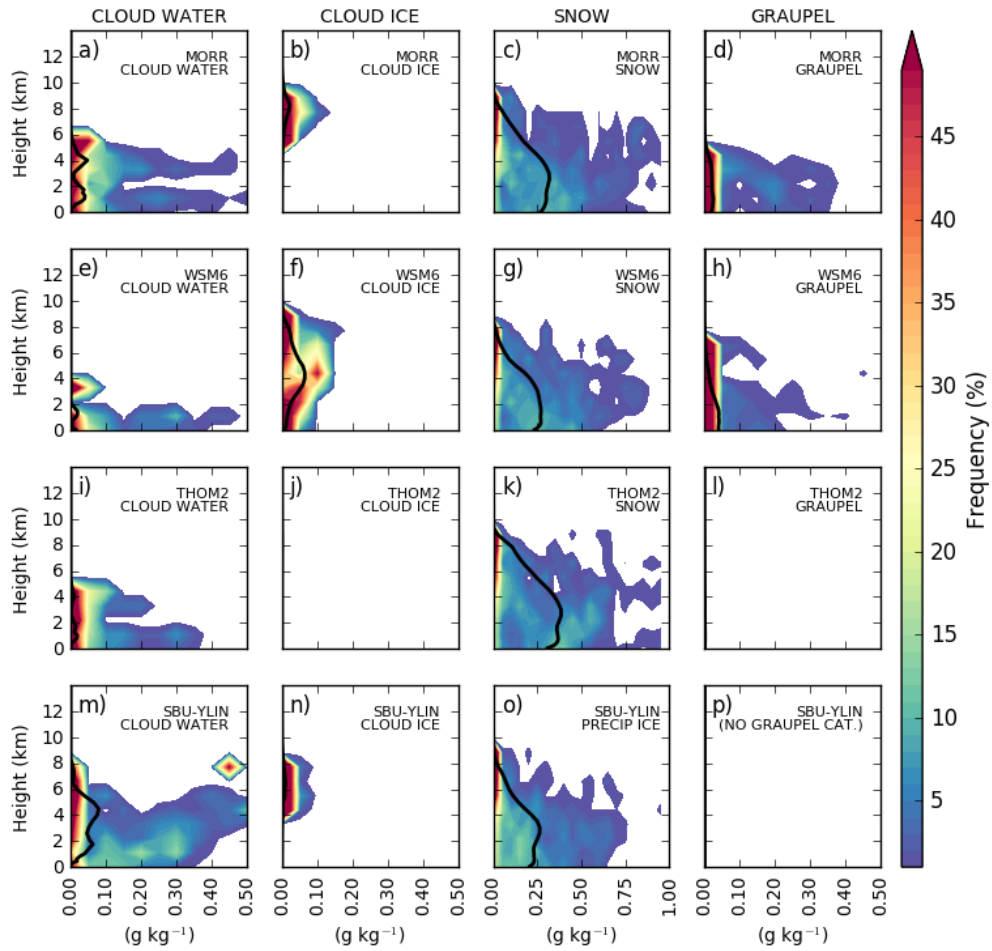


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976 Figure 10. As in Figure 3, but for moderate riming periods shown in Figure 2.

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980 Figure 11. As in Figure 4 but for moderate riming periods shown in Figure 2. Note that
 981 the x-axis scale for the snow or precipitating ice column differs from the remaining
 982 panels.

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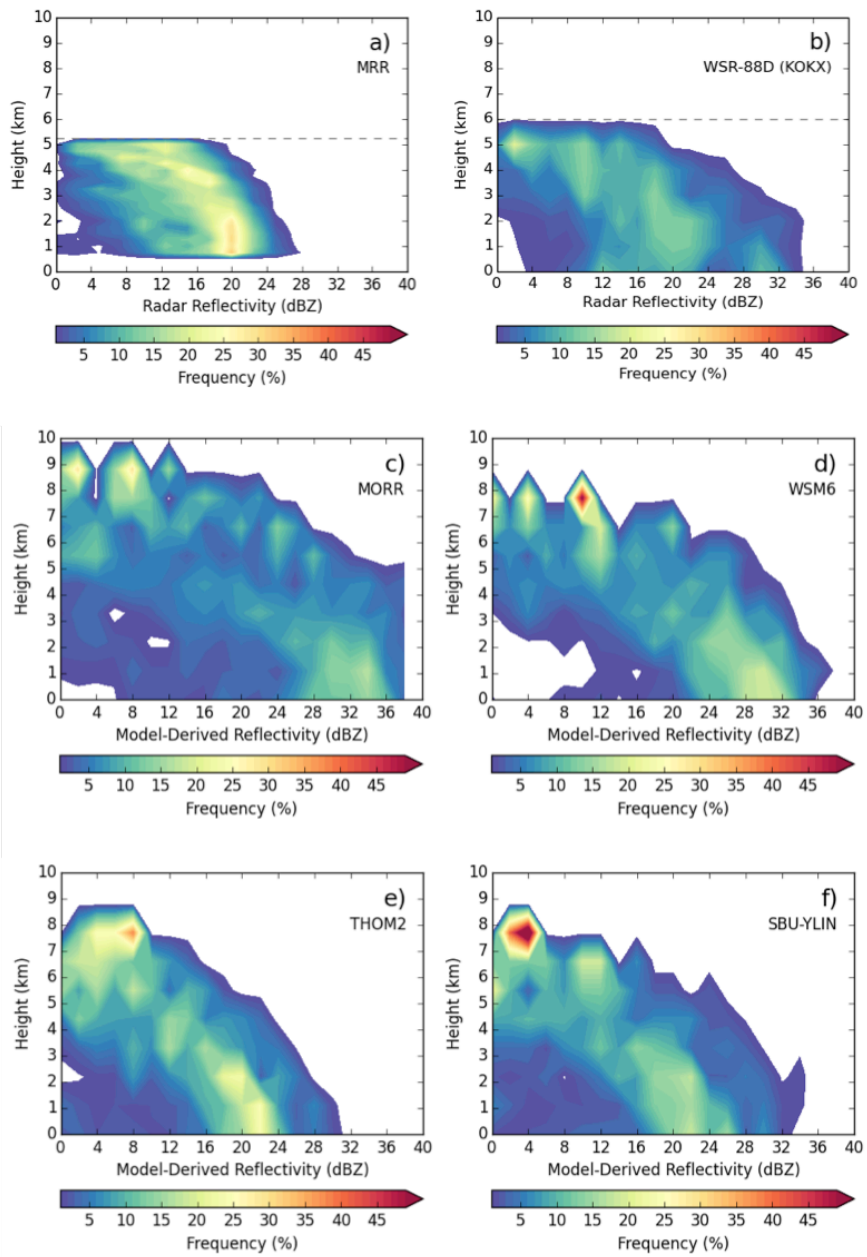
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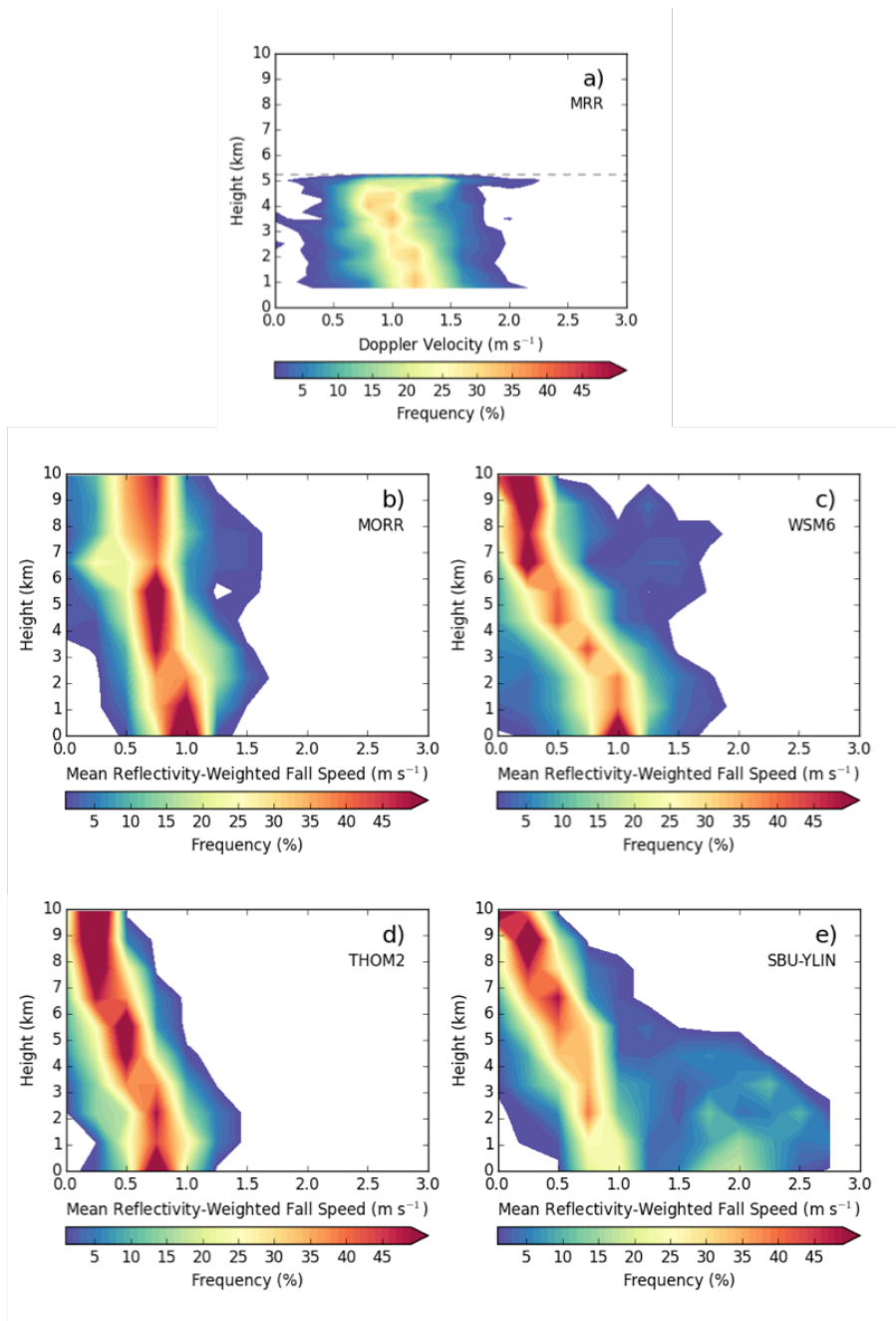
990 Figure 12. As in Figure 8, but for moderate riming periods shown in Figure 2.

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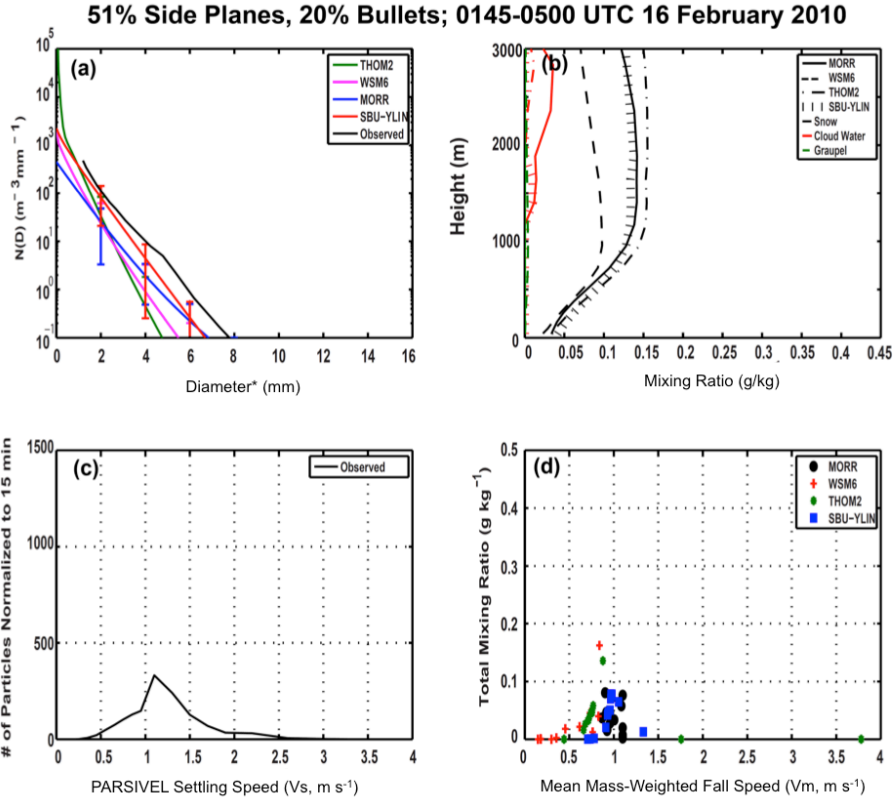


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996 Figure 13. As in Figure 9 but for moderate riming periods shown in Figure 2.

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1000 Figure 14. Period of 51% side planes and 20% bullets observed from 0145 UTC to 0500
 1001 UTC on 16 February 2010. (a) Observed and simulated surface size distribution, (b)
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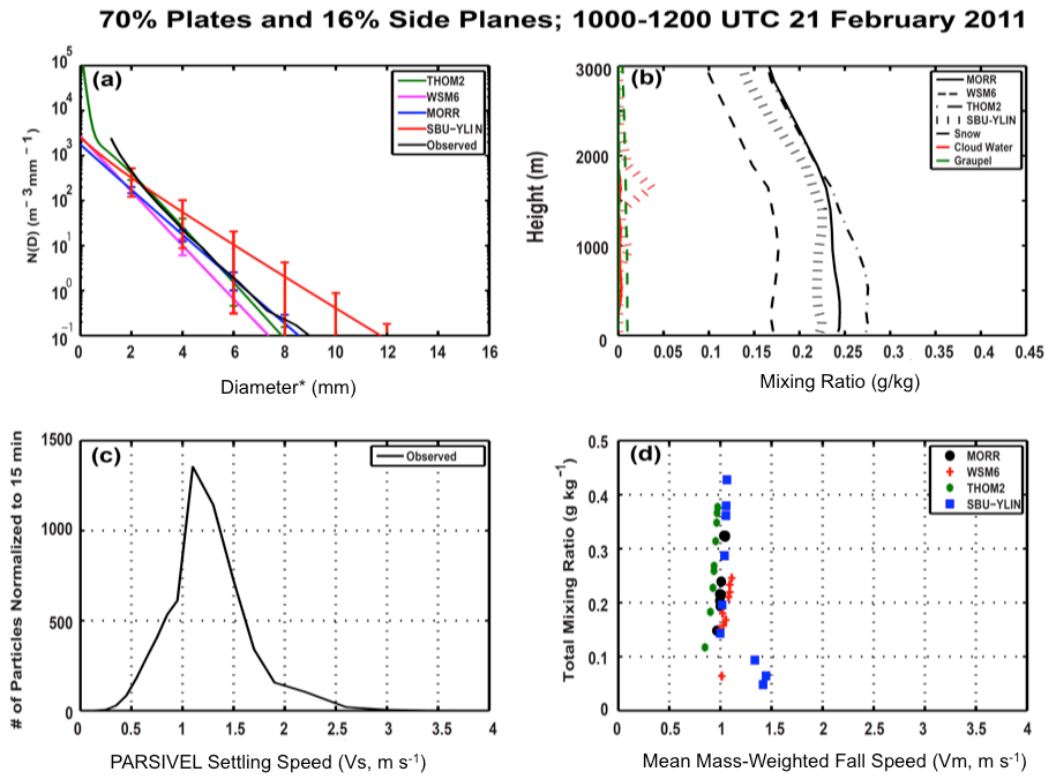
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1017 Figure 15. As in Fig. 8 except during a period of 70% plates and 16% side planes
 1018 observed from 1000 to 1200 UTC 21 February 2011.

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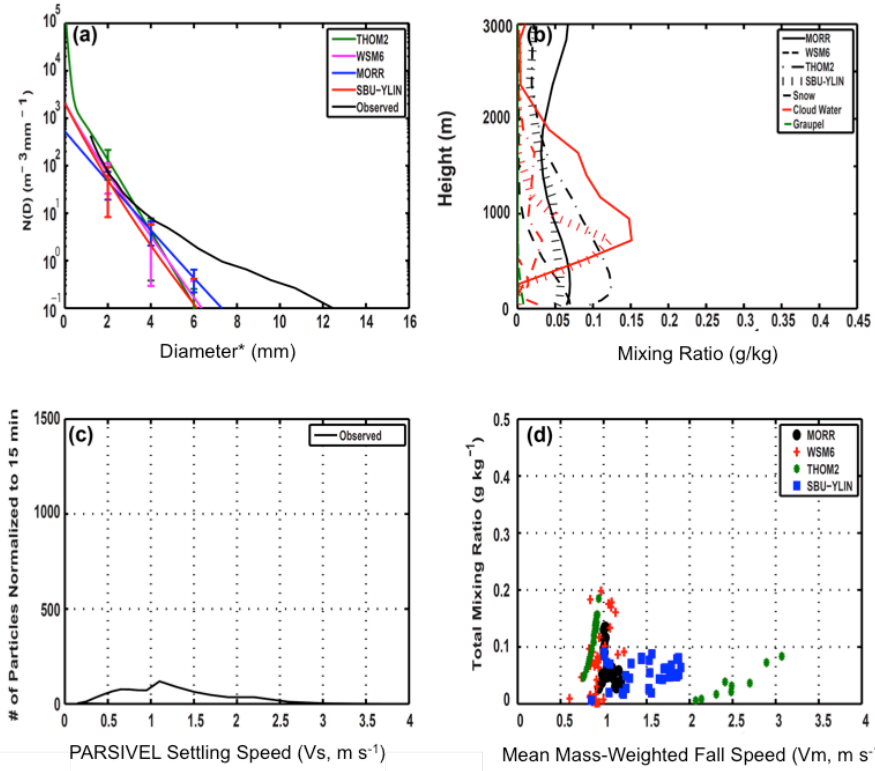
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65% Dendrites and 20% Plates; 1530 UTC 26 February 2010 to 0000 UTC 27 February 2010



1026

1027 Figure 16. As in Fig. 8 except during a period of 65% dendrites and 20% plates observed
 1028 from 1530 UTC on 26 February 2010 to 0000 UTC on 27 February 2010.

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