1	Sampling strategies to optimize coincident remote sensing and in situ		
2	cloud and precipitation observations from multiple aircraft		
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27			
28	ABSTRACT		
29	The combination of simultaneous, collocated aircraft in situ measurements and remote		
30	sensing data at multiple wavelengths is of tremendous value in physical process studies but is hard		

to obtain in practice. Appropriate multi-aircraft and multi-sensor resources for a given project 31 32 must be coupled with agile mission support (people and tools) and close coordination with the 33 Federal Aviation Administration to implement successfully. Obtaining closely coordinated in situ and remote sensing measurements was key to meeting the science objectives for the NASA 34 Investigation of Microphysics and Precipitation for Atlantic Coast-Threatening Snowstorms 35 36 (IMPACTS) and it required a team effort. IMPACTS flew a complementary suite of remote-37 sensing and in situ instruments in three 6-week deployments on the NASA ER-2 and P-3 aircraft to provide observations critical to understanding the mechanisms of snowband formation, 38 39 organization, and evolution. The collocated IMPACTS data subset encompassed 106 flight legs 40 during 22 storms, and it included over 21 hours where the NASA ER-2 and NASA P-3 were only up to 5 min and 4 km apart. This unique dataset on winter storm conditions in the Northeast and 41 42 Midwest US provides a wealth of information which will have lasting value for the community. This paper explains how the science team, engineers, air crews, and NASA mission support 43 44 accomplished the measurement goals and key aspects of the IMPACTS coordinated data set. 45 Future field campaigns with similar science applications can maximize their flight hours by 46 leveraging the lessons learned from IMPACTS coordination.

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CAPSULE

49 A field campaign coordinated two aircraft, one with remote sensors and another with in 50 situ instruments, to collect a comprehensive collocated airborne dataset for clouds and 51 precipitation.

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BODY of ARTICLE

54 Introduction

Winter snowstorms disrupt transportation, commerce and public safety, while their mesoscale precipitation variability presents significant challenges for operational weather forecasting. Substantial precipitation forecast errors result from relatively small spatial errors in rain-snow boundaries and snowband locations (Zhang et al. 2002; Ganetis and Colle 2015; Greybush et al. 2017), while remote sensing retrievals often assume uniform particle types despite ground and airborne measurements revealing complex mixtures of ice particle habits and rime fraction (Stark et al. 2013, Finlon et al. 2016). Major knowledge gaps exist regarding snowband

initiation, organization, vertical structure, microphysical properties, and their representation in 62 63 numerical models. Understanding the complex interactions between flow structure, 64 thermodynamics, and microphysical processes across convective, mesoscale, and synoptic scales remains critical for predicting the spatial and temporal variability of precipitation within 65 extratropical cyclones (Ralph et al. 2005). However, past observations and simulations of these 66 67 interactions have not achieved adequate temporal and spatial resolution to diagnose particle growth processes within these storms (Hashino 2007). High-resolution collocated remote sensing and in 68 69 situ observations of the vertical structure of snowbands and retrieved microphysical properties 70 (Plummer et al. 2014, 2015; Finlon et al. 2016; Grecu et al. 2016, 2018), in conjunction with 71 numerical models, are needed to assess the relative performance of different microphysical 72 parameterizations (Han et al. 2010, 2013, 2018; Putnam et al. 2017) and improve these microphysical schemes. 73

74 Several field campaigns over the past 30 years have collected high-resolution remote 75 sensing and in situ observations of precipitation structure and cloud microphysical properties, with 76 varying success at collocating the datasets. Early field campaigns were able to collocate two 77 aircraft that fly at similar speeds using pre-planned flight patterns. For example, the Convection and Moisture Experiment (CAMEX-3) in 1998 collocated the NASA DC-8 and ER-2 within 3 km 78 79 and 5 minutes for over 19 hours based on aircraft navigation data (Kakar et al. 2006). In the Central 80 Equatorial Pacific Experiment (CEPEX) in 1993 (Central Equatorial Pacific Experiment Design 81 Document 1993), collocation between the Aeromet Learjet and NASA ER-2, as well as with the NOAA P-3, was attempted but complicated by different speeds of the aircraft. However, in both 82 83 experiments there were no capabilities to monitor instrument health or changes in targeted weather phenomena in real time, making it difficult to ensure the collocated data of the targeted 84 85 phenomenon were captured.

Starting in the late 2000s, real-time downlinking of instrument and weather data became possible through tools like NASA's Real Time Mission Monitor (RTMM; Blakeslee et al. 2007) and later the Mission Tools Suite (MTS; Airborne Science Program 2025). However, budget constraints often limited field campaigns to a single aircraft or required different agencies or programs to fund participation by additional aircraft. For example, the Olympic Mountains Experiment (OLYMPEX), flown concurrently with the Radar Definition Experiment (RADEX) in late 2015, included the NASA DC-8, the ER-2, and University of North Dakota (UND) Citation

93 but was funded by multiple NASA programs, limiting the coordination (5 km and 5 min) between 94 the aircraft to a total of roughly 31 minutes (Houze et al. 2017). Even when multiple aircraft were 95 funded by the same program, such as the NASA ER-2 and P-3 during the 2016 Observations of 96 Aerosols above Clouds and their Interactions (ORACLES) campaign, close coordination (e.g., 97 Redemann et al. 2021) was defined to mean sampling the same cloud at different altitudes at the 98 same time, rather than exact coordinated flying. Recent NASA projects, with more robust budgets, 99 have been able to collocate multiple aircraft with more success, such as the Aerosol Cloud 100 meTeorology Interactions oVer the western ATlantic Experiment (ACTIVATE) that collected 101 278.5 hours of collocated data to within 6 km and 5 minutes (Schlosser et al. 2024), but had science 102 objectives focused on small-scale cloud and aerosol interactions that do not meet the needs of the 103 winter precipitation community.

104 The Investigation of Microphysics and Precipitation for Atlantic Coast-Threatening Snowstorms (IMPACTS) flew a complementary suite of remote-sensing and in situ instruments in 105 106 three 6-week deployments on the NASA ER-2 and P-3 aircraft to provide observations critical to 107 understanding the mechanisms of snowband formation, organization, and evolution (McMurdie et 108 al. 2022). Deployments were conducted in 2020, 2022, and 2023 during the months of January and 109 February (a planned 2021 deployment was delayed due to COVID). The NASA P-3 aircraft 110 deployed from the Wallops Flight Facility in Wallops Island, VA, its headquarters, minimizing 111 costs and logistical issues. To reduce the impact of adverse winter weather on operations, the 112 NASA ER-2 aircraft was based out of the Southeast US but changed each year due to hangar 113 availability: Hunter Army Airfield (Savannah, GA) in 2020, Pope Army Airfield (Fayetteville, 114 NC) in 2022, and Dobbins Air Reserve Base (Marietta, GA) in 2023. During these three 115 deployments, IMPACTS conducted a total of 35 science flights (total flights, coordinated and 116 uncoordinated), 26 ER-2 flights for 218 hours and 33 P-3 flights for 267 hours, during a variety of 117 winter storms.

118 IMPACTS scientists employed a three-level sampling strategy to observe winter storms 119 and achieve its science objectives. The NASA ER-2 aircraft served as an advanced cloud and 120 precipitation remote sensing platform capable of simulating satellite sensors from above the clouds 121 and precipitation, but with advanced measurement capabilities (multi-frequency, nadir viewing 122 radars) and much higher spatial and temporal resolution. The P-3 served as the IMPACTS in situ 123 platform at storm level for identifying microphysical particle characteristics, the local environment of the particles, and vertical thermodynamic and kinematic profiles from dropsondes. On the ground, mobile radar systems and radiosondes provided additional large-scale perspectives of the thermodynamic environments and storm system structures. We present in this paper the strategies utilized by the IMPACTS team to coordinate the two aircraft, including the 5-point flight legs that enabled the collection of a robust collocated remote sensing and in situ (microphysical and thermodynamic) dataset that is critical to improving snowfall retrieval algorithms and numerical weather prediction microphysics schemes.

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132 IMPACTS Aircraft and Sensors

133 Aircraft. The NASA ER-2's range, altitude, and real-time data downlinking capabilities 134 made it ideally suited to provide the remote sensing measurements required for IMPACTS. The 135 nominal altitude, speed, range, endurance, and IMPACTS base locations for both aircraft are provided in Table 1. For IMPACTS, the ER-2 flew above cloud systems at ~65,000 feet (20 km), 136 137 carrying radars, a lidar, radiometers, and electric field meters that have a long history of flying on 138 the aircraft. The NASA P-3 is designed for low-altitude heavy-payload applications, making it 139 ideal for the IMPACTS suite of in situ instrumentation. The P-3 nominal altitude, speed, range, endurance, and IMPACTS base location are reported in Table 1. The vertical range of the P-3 with 140 141 the IMPACTS payload configuration varied from 300 ft (90 m) over water (conditions permitting) 142 up to 22,000 ft (6.7 km), ensuring the full vertical sampling of cloud and precipitation structures. 143 During IMPACTS, the P-3 rarely flew below the freezing altitude on science legs, or flew legs at 144 temperatures above freezing last, to mitigate in situ probe icing.

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Table 1. Aircraft sp	ecifications of re	levance for IMPA	CTS collocation.

Parameter	ER-2	Р-3
Cruise Altitude	20 km	0.5 to 6.7 km
Cruise Speed	210 m s ⁻¹	144-175 m s ⁻¹
Range	>5,000 km	7,000 km
Endurance	7-8 hr	10 hr
IMPACTS Location	Southeast US	Wallops Island, VA

148 **Remote Sensing.** A total of eight remote sensing instruments flew on the ER-2 during 149 IMPACTS, providing the vertical and horizontal structure of storms. Key specifications of each 150 instrument are provided in Table 2. Three radars flew on the ER-2, the Cloud Radar System (CRS), High Altitude Wind and Airborne Profiler (HIWRAP), and ER-2 Doppler Radar (EXRAD). All 151 152 three radars measure the reflectivity and radial velocity of precipitation and clouds with nadir-153 looking beams (Li et al. 2015; Walker McLinden et al. 2021; Heymsfield et al. 2023). The Cloud 154 Physics Lidar (CPL) is a multi-wavelength elastic backscatter lidar that measures vertical profiles of cloud and aerosol properties (McGill et al. 2002). Depolarization ratio estimates, which provide 155 156 information about particle sphericity (Yorks et al. 2011a), are provided at 1064 nm. IMPACTS 157 flew three different microwave radiometers, two at a given time, during the three deployments to provide brightness temperatures. The Advanced Microwave Precipitation Radiometer (AMPR; 158 Amiot et al. 2021, Richter and Lang 2024) is a four-frequency, dual-polarized, cross-track-159 160 scanning microwave radiometer. The Conical Scanning Millimeter-wave Imaging Radiometer 161 (CoSMIR) and the Configurable Scanning Submillimeter-wave Instrument/Radiometer (CoSSIR) 162 are a pair of microwave radiometers that share a common configurable scanning architecture 163 (Kroodsma et al. 2019; Liu and Adams 2025). During the first deployment of IMPACTS (2020), CoSMIR was flown in two new configurations — forward/aft conical and conical/along-track. The 164 165 latter scan strategy was adopted for the following two deployments, in 2022 for CoSMIR and in 166 2023 for CoSSIR. The Lightning Instrument Package (LIP) flew on the ER-2 during all three 167 IMPACTS campaigns (Schultz et al. 2021). Three-dimensional (3D) electric field vectors are 168 retrieved as well as electric field changes due to lightning.

169 In situ sensors. IMPACTS employed several different cloud probes to provide redundancy across a wide particle size range, as shown in Figure 1. The Cloud Droplet Probe (CDP) and Fast 170 171 Cloud Droplet Probe (FCDP) use forward-scattering principles to measure the size distributions of 172 cloud water droplets. The Particle Habit Imaging and Polar Scattering (PHIPS) combines a high-173 resolution stereo-microscopic imager and a single particle polar nephelometer to determine cloud 174 particle shape, size, and habit using a 3 x 2 mm field-of-view (Abdelmonem et al. 2016; Schnaiter 175 et al. 2018; Waitz et al. 2021). For cloud droplets and larger ice particles, IMPACTS relied on the 176 Two-Dimensional Stereo (2D-S) and High-Volume Precipitation Spectrometer (HVPS-3) imaging probes. The Hawkeye probe, consisting of a FCDP, 2D-S, and Cloud Particle Imager (CPI), was 177 178 impacted by shattering during IMPACTS, but provided redundancy for comparative purposes.

179 Cloud liquid content was measured with a King Probe and Science Engineering Associates Model 180 WCM-3000. IMPACTS also includes a Rosemount Ice Detector (RICE) to detect the occurrence 181 and amount of supercooled water. The Water Isotope System for Precipitation and Entrainment Research (WISPER) provided high-accuracy measurements of total water content that is the sum 182 183 of liquid and ice content (Twohy et al. 1997). The Turbulent Air Motion Measurement System (TAMMS) instrument measures 3D winds, humidity, and temperature at the P-3 flight level 184 185 (Brown et al. 1983). Derived measurements of the 3D wind components, temperature, and moisture are computed from the raw 100 Hz data and archived at 20-Hz resolution. The Diode 186 Laser Hygrometer (DLH; Podolske et al. 2003), flown during the 2023 deployment, is a laser-187 188 based hygrometer that measures water vapor via differential absorption techniques at isolated 189 spectral lines near 1.4 µm. While the in situ probes all have varying raw collection rates, data 190 products are reported at 1 Hz for all sensors.

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 Table 2. Key specifications of the IMPACTS remote sensing instruments.

Instrument	Frequency or Wavelength	Resolutions/Swath	Other Specifications
CRS	W-band (94 GHz)	Horizontal: 200 m Vertical: 115 m	Nadir pointing; 50 m horizontal sampling; Vertical sampling: 14 m
HIWRAP	Ku-band (14 GHz) Ka-band (35 GHz)	Horizontal (Ku): 800 m Horizontal (Ka): 350m Vertical: 130 m	Nadir pointing; 100 m horizontal sampling; Vertical sampling: 26 m
EXRAD	X-band (9.6 GHz)	Horizontal:1 km Vertical: 150 m	Nadir pointing; 100 m horizontal sampling; Vertical sampling: 13 m; Additional conical scanning beam with ~30° tilt angle (20 km swath width)
CPL	355, 532, 1064 nm	Horizontal: 200 m Vertical: 30 m	Depolarization ratio at 1064 nm; Nadir pointing
AMPR	10.7, 19.35, 37.1, 85.5 GHz	Swath width: 38 km IFOV: 0.6-2.8 km	Dual-polarized; Cross-track-scanning; Four scene sweeps and calibration sequence (10-12 sec)
CoSMIR	50.3, 52.8, 89.0, 165.5, 183.31 GHz	Swath width: 50-60 km Res: 1.4-3.9 km	Configurable scanning architecture; New scan configurations in 2020; Conical/along-track scan in 2022
CoSSIR	170.5, 183.31, 325.15, 684.0 GHz	Swath width: 50-60 km Res: 1.4-3.9 km	Used in 2023 deployment
LIP	N/A	50 Hz sampling rate	Dynamic range: ~10 ⁰ -10 ⁶ V m ⁻¹ ; ~10% measurement error

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Figure 1. Hydrometeor size ranges as measured by the IMPACTS cloud probes.

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198 Collocation Challenges

199 One of the biggest challenges of collocating the NASA ER-2 and P-3 was the differences 200 in the cruise speeds of the two aircraft, which was further exacerbated by the variations in the P-3 201 cruise speed based on altitude and winds. Figure 2 shows the variation of the ground-speed ratio 202 (ER-2 to P-3) with the P-3 altitude. When the P-3 flew at higher altitudes (4 to 8 km), the ground 203 speed ratio was typically 1.0 to 1.2, meaning the two aircraft were flying nearly the same cruise speeds. However, the ground-speed ratio was greater than 1.4 when the P-3 flew below 4 km in 204 205 altitude. The differences in cruise speeds also necessitated longer legs and longer turns between 206 legs for the ER-2 compared to the P-3, which reduced the amount of coincidental sampling time. 207 Furthermore, the lower altitude of the P-3 required timely and accurate communication with local 208 air traffic control (ATC) centers. If the P-3 flight plans were altered in real time during flight or 209 the aircraft was flying near a busy airport (New York, Chicago, etc.) or ATC was busy with other 210 aircraft, delays to the P-3 led to further delays to the ER-2.



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Figure 2. The ground speed ratio (ER-2/P-3) versus P-3 altitude for the deployments in 2020 (a), 2022 (b), 2023 (c), and all deployments (d). Histograms of the P-3 altitudes are provided on the left side of the plots, while histograms of the ground speed ratio are provided on the top. A best fit line in 2020 shows the origin of the ground speed ratios for the coordinated legs.

The heterogeneity of winter storms also led to challenges in coordinating the IMPACTS aircraft. Numerical weather prediction models can have errors in the forecast of the rain-snow lines or locations of snowbands on the order of tens to hundreds of kilometers (Zhang et al. 2002, Ganetis and Colle 2015, Greybush et al. 2017), which made planning the exact location of flight lines difficult 24-48 hours before the storm. Even when models properly predict the locations of 222 the snowbands, these bands typically evolve and move through the region of interest quickly 223 during an 8-hr flight, necessitating adjustments to the planned flight patterns. The evolution of 224 these snowbands occurs rapidly (on the order of minutes in some cases) further challenging multi-225 aircraft coordination. Finally, vertical variations in horizontal wind speeds as the P-3 flies through 226 frontal boundaries further introduce variability in the true ground speed of the P-3 compared to 227 planned flight patterns, especially at lower altitudes. Another weather-related challenge to 228 planning coordinated flights was the local weather conditions for takeoff and landing, which 229 sometimes caused takeoff delays or early to return to base if landing conditions were forecast to deteriorate. 230

231 Instrument sampling rates and measurements volumes of the remote sensors also impact the ability to collocate the remote sensing and in situ data, even when the aircraft themselves are 232 233 well coordinated. For example, the CPL points nadir and has a 100 microradian field of view, providing a narrow 1-2 m diameter footprint at the altitude where many IMPACTS cloud tops 234 were sampled (Yorks et al. 2011b). Additionally, cloud vertical profiles are limited to optical 235 236 depths less than 3.0, causing CPL to only penetrate roughly 1-3 km deep into the cloud systems 237 observed during IMPACTS. Thus, to collocate the in situ sensors with the lidar data without assumptions of particle homogeneity across large spatial scales, the P-3 must be flying near cloud 238 top and within meters horizontally of this small lidar sampling "volume". The collocation 239 240 constraints for the high-altitude radars are less stringent, as the HIWRAP and EXRAD sensors 241 have a footprint diameter of ~1 km. While these radars are sensitive to hydrometeors through most 242 of the cloud depth, they are sometimes insensitive to small particles at the cloud tops that the lidars 243 are sensitive to, which must be considered when flying the P-3 near cloud top. Attenuation in 244 moderate to heavy rain can also occur at lower altitudes, particularly for higher-frequency radars 245 like HIWRAP and CRS. Collocation between the microwave radiometers is simple in the 246 horizontal direction, given the wide swaths of these sensors, but can be a challenge vertically if 247 the instrument does not have a frequency sensitive to the cloud vertical structure at the P-3 altitude. 248 Given most of the remote sensors and in situ instruments report their data at 1 Hz, there are minimal 249 sampling issues related to data rates.

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251 Sampling strategies employed during IMPACTS

252 Separation requirements. Ice processes occurring at higher altitudes, such as riming, 253 influence a significant portion of global precipitation patterns (Heymsfield et al. 2020). Although 254 these mechanisms can enhance ice water content (IWC) within mid-latitude storm systems (Waitz 255 et al. 2021), researchers have not yet fully determined their quantitative impact on surface snowfall 256 accumulation. Deng et al. (2024) suggests that small-scale ice clusters, areas of several kilometers 257 exhibiting elevated ice particle concentrations or IWC, are primarily responsible for the non-258 uniform distribution of ice within cloud formations. Horizontal winds within the storm, often > 20 259 m/s or more at altitudes greater than 4 km (20 m/s yields 6 km horizontal motion in 5 minutes), 260 transport hydrometeors sideways an order of magnitude faster than they fall (Finlon et al. 2022, 261 Tomkins et al. 2025). To observe these variations in cloud microphysical properties and transport 262 over small spatial and temporal scales, the IMPACTS project defined a minimum collocation requirement for the two aircraft of 4 km and 5 min. Several recent publications have used 3-5 mins 263 264 as a collocation threshold when using the combined remote sensing and in situ IMPACTS dataset 265 (e.g., Finlon et al. 2022; Maherndl et al. 2024, Allen et al. 2025). Field campaigns that target other 266 atmospheric applications may not require as stringent of collocation temporal and spatial scales.

267 The 5-point flight legs (Figure 3) became the standard IMPACTS flight pattern later in 2020 operations due to the ease of the design, the effectiveness for coordinating aircraft timing, 268 269 and the reduction in confusion when retasking one or both aircraft was required mid-flight. It uses 270 points defining the flight leg beginning and end to coordinate the P-3 (P1 and P2 in Fig. 3) and the 271 ER-2 (E1 and E2) that are based on the typical ground speed ratios of the two aircraft. The center 272 point (C1), where the aircraft are intended to overfly the same ground point at the same time, acts 273 as a reference point for both aircraft to communicate estimated overflight times and coordinate 274 changes in speed or turning locations to maintain close timing. The lines were initially planned 275 based on the forecasted conditions for the first pass. However, as the P-3 would repeat passes at 276 lower altitudes, the slower P-3 airspeeds at these altitudes and the vertical variations of the 277 horizontal winds limited the collocation success and instrument collection so delay maneuvers 278 were performed by one or both aircraft to accommodate the 5-minute collocation goal. Initially, a 279 P-3 to ER-3 leg length ratio of 1:1.2 led to coordination issues when the P-3 was flying slower 280 than planned at lower altitudes. The ER-2 would overtake the P-3 before the center point, forcing it to ad lib leg extensions to maintain coordination, wasting sampling time. Flying shorter flight 281 282 legs improved the temporal coordination between the airplanes. However, shorter lines required

283 relatively more time turning around at the end of each line, reducing data collection time for many 284 of the remote sensors, which require straight and level flight. The line lengths were sometimes 285 adjusted during flights based on the width of the snowbands or features being sampled. By the 286 2023 deployment, the sampling strategy changed to using two 5-point flight lines during planning: 287 a high-altitude line that used a leg length ratio of 1.20 and a low-altitude line that used a ratio of 288 1.45 (Figure 3). Model initialization winds at the P-3 flight level were also used to estimate the 289 influence of crosswind/headwind for the P-3 and gauge additional reductions in aircraft true airspeed, both during planning as well as during flights. 290



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Figure 3. The general 5-point line concept for ground speed ratios of 1.2 (P-3 high altitudes) and
1.45 (P-3 low altitudes), with the aircraft start/end point labels.

295 Moving Lines software. The time and effort scientists spend planning flight paths 296 according to science objectives, aircraft performance, airspace availability, meteorological and 297 sampling conditions, airborne instruments needs, and coordination with other airborne research platforms can distract from accomplishing the science objectives. To mitigate this issue, a research 298 299 flight planning tool called Moving Lines (LeBlanc 2018) was used and subsequently modified 300 during IMPACTS. This flight planning tool focuses on building airborne sampling strategies for 301 better resolving the environment surrounding aerosol, clouds, radiation, and atmospheric 302 dynamics. To date, Moving Lines has been used during at least 10 NASA field campaigns 303 including IMPACTS. It was built as an open-source Python library with a graphical user interface 304 portraying mapping (cartopy) and interfaces through simple spreadsheets (Excel). The 305 fundamental interface is one spreadsheet tab per desired flight path, often used as different aircraft,

306 like the NASA ER-2 and the P-3 used during IMPACTS, for planning coordinated science observations. This interface includes multiple waypoints for identifying latitudes and longitudes 307 308 of sampling in concert with the vertical aircraft location. Moving Lines incorporates a 309 parameterized set of aircraft characteristics, like typical cruise speed, altitude, turn bank angles, 310 flight speed as a function of altitude, and climb rate for the different research aircraft in addition 311 to solar geometry calculations, satellite overpass predictions, common flight modules, 312 model/satellite imagery overplotting, and multiple aircraft plans. It also expands with predetermined flight sampling modules, for which many were custom designed for IMPACTS, like 313 314 the 5-point line for coordinated ER-2 and P-3 sampling (Figure 4). As part of the design and 315 building of the flight plans, Moving Lines calculates the flight time, from the parameterized 316 aircraft specifications and sampling design, and can output pilot-friendly files for easier 317 dissemination, as well as a multitude of figures and summary presentations for scientific feedback.

318 The primary benefit of the 5-point-line scheme is the ease of use and planning. An 319 IMPACTS flight leg was typically on the order of 150 to 250 km or 15-25 minutes of flight time, 320 with the ER-2 having longer legs. Given the typical spatial and temporal evolution scales of winter 321 storms, it was necessary to treat each leg as an independent measure. Complex flight schemes such 322 as lawnmower and bowtie patterns, which aim to collect horizontally oriented aerial or volumetric 323 sampling of the storms, generally had limited use during IMPACTS since the storms evolve more 324 rapidly than a multi-leg pattern can be flown. As such, it is inappropriate to make spatial linkages 325 between different flight legs. Lagrangian schemes, which attempt to follow storm features as they advect, were taxing to plan and execute. They required waypoints to be calculated, communicated, 326 327 manually entered into flight systems, and cleared with ATC in real time based on the storm 328 advection. Waypoint changes can take 15 to 30 minutes from calculation to clearance, leaving little 329 margin for error in executing Lagrangian flight patterns. Thus, simple flight patterns have higher 330 probability of successfully meeting their design goals, and the 5-point-leg flight scheme's 331 simplicity provided a benefit in terms of flight planning and execution since different flight leg 332 altitudes or temperatures were flown using a common set of waypoints. The 5-point-line module 333 in the Moving Lines flight planning software allowed mission planners to quickly define all the 334 necessary flight points for both the P-3 and ER-2 by choosing a starting location, a bearing, and a distance (Fig. 3). The software generates all the applicable waypoints for translation into MTS and 335 336 the Flight Management System of the aircraft. This made transitions between legs easier and

quicker to clear with ATC since altitude is the only changing factor. The advection of the storm
through the flight path curtain allowed for diversity of sampling with respect to distance and
bearing relative to the low-pressure center. Since the number of waypoints was small, changing
the flight legs during the flight became a simple translation task, as explained in Sidebar 1.



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Figure 4. The 5-point line module in the Moving Lines software enabled the IMPACTS mission
scientist to enter the leg length and angle (orientation of the line) to create flight plans ahead of
each flight (red lines).

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346 Assessment of IMPACTS collocation success

The IMPACTS dataset is an exceptional collection of winter storm observations from thenortheastern and midwestern United States, providing invaluable data that researchers will

349 continue to benefit from for years to come. During 22 storm events, the IMPACTS project 350 collected a specialized dataset comprising more than 21 hours of measurements taken within 5 351 minutes and less than 4 kilometers horizontal separation between NASA's ER-2 and P-3 aircraft, 352 spanning 106 flight legs. These IMPACTS collocation estimates are determined using the sensor 353 data, ensuring functional instruments sampling of features of interest (i.e., cloud and precipitation). 354 Collocation statistics cited for CAMEX and ACTIVATE earlier in the paper used the aircraft 355 navigation data, which do not consider sensor functionality or the presence of features. Figure 5 shows a map of the coordinated flight legs. Not all 106 appear on the map, as many of these 356 357 coordinated legs were flown over the exact same line but at different P-3 altitudes. While most of 358 the coordinated lines are over the Northeast US, there are several off the Mid-Atlantic US coast 359 and over the Midwest. As many as 6 different storm types were sampled, including Miller Type A 360 and B, Gulf Coast cyclones, Alberta Clippers, cold fronts, and Great Plains cyclones (Zaremba et 361 al. 2024, Lundstrom et al. 2025a,b).

362 IMPACTS was very successful at collocating the ER-2 and P-3 aircraft, incorporating 363 lessons learned after each deployment to accumulate 21.3 hours of collocated data based on 364 HIWRAP. For this paper, a matching routine was developed using a k-d tree search algorithm with the Minkowski p-norm (Euclidean distance, p = 2) to identify the nearest 30 HIWRAP radar 365 366 profiles to the P-3 for every 5 s of flight (Finlon et al. 2022). This radar matching algorithm was 367 modified from the ones described in Chase et al. (2018) and Ding et al. (2020). From there, a 368 Barnes (1964) interpolation procedure was applied to the 30 gates to obtain a spatially-weighted 369 reflectivity value for each 5 s collocated point. Guided by spatial autocorrelation analysis of the in 370 situ microphysics and remotely sensed measurements among all the coordinated flight legs, 4 km 371 was determined as the optimal distance threshold based on the global Moran's I autocorrelation 372 index (Moran 1950). Table 3 shows the number of hours IMPACTS collected collocated (2 mins 373 and 1 km or 5 mins and 4 km) data for each deployment year. During the 2020 deployment, only 374 3.7 hours of collocation within the IMPACTS goal of 5 minutes and 4 km were achieved, due to 375 the limited number of joint flights and flight plans that did not optimize coordination. The 376 implementation of the 5-point line for the 2022 deployment helped the team improve overall 377 collocation to 5.1 hours. However, opportunities for collocation were often limited in 2022 due to strong crosswinds at the Pope airfield that were out of limits for a safe ER-2 takeoff or landing. 378 379 For the 2023 deployment, the implementation of the 5-point line module using 2 ground speed

ratio options, the Nystrom line tool in MTS (Figure SB2), better communication between the
aircraft pilots in real time, and less weather-related ER-2 delays all led to 12.6 hours of collocation,
with 5.6 hours coordinated to within 2 mins and 1 km.







Figure 5: A map of the coordinated flight legs over all three IMPACTS deployments. The black lines represent the ER-2 flight track, while the red represents the P-3 flight track. Many of these tracks are oriented NNW to SSE to be perpendicular to snowbands and frontal zones. This had the added benefit of reducing cross track separation error between the two aircraft caused by the amount of great circle correction inherent in flying lines of different lengths. A true north/south oriented line has no great circle correction while east/west lines have small persistent corrections dependent on length.

Collocation	2020	2022	2023	TOTAL
2 min/1 km	1.5 h	2.0 h	5.6 h	9.1 h
5 min/4 km	3.7 h	5.1 h	12.6 h	21.3 h

Table 3. The IMPACTS collocation sampling hours for each deployment year.

Given that the lidar and radars flown during IMPACTS are sensitive to different portions 395 396 of the vertical extent of the clouds, collocation with respect to cloud depth is an important factor 397 when combining the in situ and remote sensing datasets. Figure 6 shows the normalized frequency 398 of collocated observations (5 min and 4 km) versus the P-3 altitude with respect to depth below 399 cloud top as observed by the CPL. The 2020 deployment had few (if any) collocations within 2 km of cloud top, but many observations deep (8-10 km) into the clouds, mostly due to concerns 400 401 about flying near cloud tops that year. In the following deployment years, there was a strong desire 402 to sample a variety of temperature ranges and to sample near cloud top, especially in 2023, in order to sample the full range of microphysical growth regions, as defined by Bailey and Hallett (2009), 403 404 i.e., the polycrystalline growth layer in temperatures less than -18°C, dendritic growth layer from 405 -18 to -12°C, the plate growth layer from -12°C to -8°C, and the needle growth layer from -8°C to -3°C. However, these layers were not always present in every storm or could not be sampled for 406 407 variety of reasons such as being too close to the ground. This resulted in an uneven distribution of 408 sampling by depth below cloud top shown in Fig. 6, especially for the 2022 deployment.

409 IMPACTS collocations between the in situ sensors and the radar data were more robust 410 than the collocations between the in situ sensors and lidar, given the frequent sampling 2-6 km deep into the clouds. Figure 7 displays the 2D histogram and cumulative distribution functions of 411 412 all the 5 s collocated observations as a function of aircraft distance and time offset for HIWRAP 413 (Fig. 7a) and CPL (Fig. 7b). There were 6.28 hours of data where the HIWRAP radar and in situ 414 sensors were collocated to within 1 min and 17.15 hours when they were collocated to a distance 415 offset of less than 1 km, which is within the footprint of the HIWRAP (Ku band) and EXRAD 416 radars. During many flights, the P-3 could be observed as a "skin paint" echo in real-time images 417 of the EXRAD radar reflectivity, as shown in Figure 8 at 21:42 UTC and 4.8 km altitude on 25 418 January 2023. The wider swath microwave radiometers, especially AMPR and CoSMIR, being 419 sensitive to the lower vertical regions of the clouds, have similar collocation statistics to the radars. 420 However, lidar collocation is not as good as the radars and radiometers due to (1) attenuation of the laser beam ~1-3 km into the cloud, (2) limited observations near cloud top in 2020 and 2022,
and (3) the narrow footprint diameter (1 m) of CPL. Despite these limitations, IMPACTS still
collected 2.83 hours of collocated data within a distance offset of less than 1 km and 3.47 hours
with a time offset of less than 5 minutes.



427 Figure 6: The normalized frequency of collocations versus the P-3 depth below cloud top for 2020
428 (left panel), 2022 (middle panel), and 2023 (right panel).



Figure 7. 2D histogram of the number of 10 s collocated observations as a function of distance
and time offset between the HIWRAP data (a), the CPL data (b), and P-3 aircraft's in situ
observation. The cumulative distribution functions are provided along each axis side panel.



435

Figure 8: The EXRAD nadir reflectivity from the IMPACTS flight on 25 January 2023. The P-3
aircraft (indicated with red arrow) caused an observed reflectivity of about 40 dBZ (4800 m
altitude around 21:42 UTC).

439

440 Concluding Remarks, Significance and Summary

441 IMPACTS, a three-year NASA field campaign, deployed complementary aircraft - the high-altitude ER-2 with remote sensing instruments flying above storms and the P-3 equipped with 442 443 in-situ probes operating within clouds at various altitudes - to comprehensively study snowband 444 formation, organization, and evolution. By flying the two aircraft in a vertically stacked pattern 445 with a 5-point flight leg, IMPACTS was able to collect over 21 hours (106 flight legs) of collocated 446 remote sensing and in situ microphysics data for 22 winter storms. This robust collocated dataset, 447 with temporal (5 mins or less) and spatial (4 km or less) scales to diagnose particle growth 448 processes within storms, enables scientists to accurately interpret the remote sensing 449 measurements with respect to microphysical processes and environmental conditions, advancing our understanding of clouds and precipitation. 450

451 Multiple studies have resulted from the analysis of IMPACTS collocated data. These 452 include in-depth case studies, such as Varcie et al. (2023) documenting the differences in 453 microphysical processes in the stratiform and convective portions of a deepening cyclone, 454 DeLaFrance et al. (2024) illustrating the effects of riming on radar moments and precipitation 455 fallout, Zhang et al. (2025) examining elevated convection and banded precipitation in a broad 456 frontal band, and Han et al. (2025) highlighting the role of supercooled liquid water at cloud top 457 in an intense east coast storm. Studies focusing on microphysical and ice growth processes leveraging the full IMPACTS collocated remote sensing and in situ dataset include Allen et al. 458 459 (2025) and Heymsfield et al. (2023). Tomkins et al. (2025) utilizes multi-year IMPACTS datasets 460 as well as NOAA NWS operational observations to quantify the impact of mesoscale snowbands on surface snowfall rates. Finlon et al. (2022), Zaremba et al. (2024), and Nicholls et al. (2025) 461 demonstrated improvements to remote sensing retrievals of snowfall using the collocated 462 463 IMPACTS dataset. Furthermore, combining the IMPACTS dataset with numerical weather 464 prediction models enables assessments and improvements to model microphysical schemes (e.g., 465 Colle et al. 2023) as well as the evaluation of the impact of assimilating various types of 466 observations on model analyses and forecasts. While this paper focuses on the direct applications 467 for winter storm and precipitation processes, there are needs across many different atmospheric 468 composition communities (clouds, aerosols, trace gases, etc.) to understand the connections 469 between spatial structure, microphysical properties, and thermodynamic processes. IMPACTS 470 provides a model for future atmospheric science field campaigns to coordinate remote sensing and 471 in situ aircraft and address these science community needs. Continued investment in collocated 472 airborne platforms capable of collecting radar, lidar, and in situ microphysics remains essential for 473 disentangling the complex vertical and horizontal variability of cloud processes that shape 474 precipitation formation.

475

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489

490 Data Availability

491 Mission scientist reports, weather discussions, and quick-look images from all three
492 IMPACTS deployments are provided in the IMPACTS field catalogs below:

- 493 2020: <u>http://catalog.eol.ucar.edu/impacts_2020</u>
- 494 2022: <u>https://catalog.eol.ucar.edu/impacts_2022</u>
- 495 2023: <u>https://catalog.eol.ucar.edu/impacts_2023</u>

All the IMPACTS data can be obtained from the Global Hydrology Resource Center Distributed
Active Archive Center at https://ghrc.nsstc.nasa.gov/uso/ds_details/collections/impactsC.html
and McMurdie et al. (2019). Python code that matches nadir-pointing data from the CPL, CRS
and HIWRAP to the P-3 location is publicly available (Finlon et al. 2025).

500

501 Sidebar 1 - Dynamic modifications during flight

502 Real-time coordination of the aircraft in flight was facilitated by NASA MTS, which 503 enabled real-time tracking of the aircraft, comparison of actually flown tracks to the original flight 504 plans, and simultaneous visualization of geostationary satellite imagery and radar products for in-505 flight guidance of the aircraft. All ER-2 instruments used Inmarsat to downlink instrument data 506 for "quicklook" plots in real time during IMPACTS, which were made available in MTS or other 507 websites accessible to IMPACTS scientists. Dynamic modifications of the flight plan, based on 508 the real-time monitoring using MTS, were coordinated between the lead mission scientist, ER-2 509 mission scientist, and aircraft coordinator on the ground, as well as the P-3 mission scientist 510 onboard who had direct interaction with P-3 pilots. The lead mission scientist had intimate knowledge of a specific flight plan and authority to change the flight plan after take-off. 511

512 Communication of changes to the flight plan and the experience of the aircraft coordinator,

513 former ER-2 pilot Jan Nystrom (Fig. SB1), was key to successful implementation. The aircraft 514 coordinator used the Nystrom tool (named in honor of him) in MTS (Fig. SB2) to compute the 515 coordinates of the start, mid, and end points for the aircraft. The tool provides latitude/longitude 516 of the waypoints in decimal degrees, radial Distance Measuring Equipment arc, and degree 517 minutes simply by dragging the line to a new location with a computer mouse. The aircraft 518 coordinator then passed new way points using Internet Relay Chat communicated directly to the 519 ER-2 pilot as well as to the P-3 mission scientist, who then relayed the points to the P-3 pilots. The flight management system in both aircraft estimated the time at which they would overfly the 520 521 center point of the next leg in the series considering headwind and planned altitude changes. When 522 a 5-point line is repeated at multiple altitudes, the pilots of the aircraft communicated their estimated time to the next center point to each other via VHF/UHF radios (if within a range of 523 524 \sim 500 km) or Internet Relay Chat messages to plan whether to cut short or extend their current leg 525 to set up for well-coordinated segment on the next leg. Typically, the ER-2 performed any 526 necessary deviations for the flight legs because of its high altitude, well above conflicting air 527 traffic, maintaining an area clearance from ATC to fly freely within a certain altitude band without 528 prior permission. The P-3 pilots communicated new waypoints to ATC when necessary, which sometimes resulted in a holding pattern until approved. One advantage of the 5-point line is that it 529 530 required fewer waypoints (3 per aircraft) to be updated, reducing the task load on the flight crews 531 and accelerating communication with ATC.



- Figure SB1: IMPACTS Aircraft Coordinator Jan Nystrom discusses real-time modifications to a
 planned IMPACTS flight with the ER-2 pilot during the 2022 deployment.



Figure SB2. The Nystrom Line tool (magenta with start, mid, and end points) in the MTS enabled
the aircraft coordinator to configure the 5-point line and drag the line as needed for dynamic
changes.

542 Sidebar 2 – Enabled Science from IMPACTS Collocations

543 Figure SB3 illustrates the value of coordinated remote sensing and in situ observations, 544 which enables interpretation of the remote sensing measurements in terms of observed microphysical processes. On 23 January 2023, the aircraft sampled a broad frontal zone over 545 546 southern Maine. The highlighted flight leg sampled across the front from northwest to southeast 547 as shown in Fig. SB3a. During this coordinated leg, the two aircraft flight tracks differed by < 220548 m in horizontal distance and the aircraft passed over the same point no more than 2 minutes apart 549 in time from one another, enabling detailed comparisons between the airborne radar and in situ 550 microphysics measurements (Fig. SB3e). The frontal zone exhibited a high degree of variability, 551 as evident in the HIWRAP Ku-band and EXRAD reflectivity fields shown in Figs. SB3b and d.

552 Using the remote sensing measurements and in situ observations together allows inference 553 of distinct microphysical processes that occurred along this flight leg. At the southern end of the 554 flight leg (44.05°N), a broad region of high reflectivity (>30 dBZ) was observed that appears to be 555 associated with fallstreaks from elevated convection above the front (Fig. SB3b). Upward vertical motions around 1 ms⁻¹ (Fig. SB3c) were observed by HIWRAP near cloud top directly above the 556 557 P-3. EXRAD conical scans of reflectivity show that this high reflectivity region had a wide horizontal extent at 4 km, near the altitude of the P-3 (Fig. SB3d). This region also coincided with 558 high IWC $(0.4 - 0.8 \text{ g m}^{-3})$, as calculated using the mass-dimension relationship of Heymsfield et 559 al. (2004), high LWC $(0.02 - 0.06 \text{ g m}^{-3})$, and high number concentration of both cloud-sized (from 560 561 2DS) and precipitation sized (from HVPS) particles (Fig. SB3f, g). Together, these fields paint the 562 picture of the high likelihood of particle growth via both vapor deposition and riming/aggregation 563 beneath cloud top.

In the vicinity of 44.3°N, the P-3 sampled the top of a narrow region of elevated reflectivity 564 565 (near 30 dBZ, Fig. SB3b). The conical scan shows a narrow region of higher reflectivity (near 30 566 dBZ) surrounded by a broader region of weak reflectivity (<15 dBZ, Fig. SB3d). The radial 567 velocities are roughly near zero (Fig. SB3c), suggesting weak upward motion. This region contains 568 very high concentrations of small particles (<0.1 mm), low IWC, and low LWC, indicating the 569 lack of supercooled liquid (SLW) droplets within the fall streak (Fig. SB3f, g, h). Substantial LWC 570 (Fig. SB3f, h) was observed outside the fallstreak on either side of it, whereas SLW was largely absent within it. In contrast to the other areas characterized by prominent fallstreaks and enhanced 571 572 ice aloft, such as the one described above, this zone appears to have avoided efficient scavenging 573 of SLW by descending ice, possibly due to the localized nature of the convective fallstreaks or 574 vertical motions that disrupted their descent. The microphysical characteristics support the idea 575 that small-scale vertical motions and gaps between fallstreaks can enable pockets of SLW to 576 persist.



- 578 Figure SB3: Data collected between 1637:30 and 1647:30 UTC on 23 January 2023 by the ER-2 and P-3 aircraft: (a) MRMS composite reflectivity (dBZ, shaded) with flight tracks overlaid for 579 580 the 23 January deployment, highlighting the track shown in subsequent panels; (b) HIWRAP Ku-581 band reflectivity (dBZ, shaded) with P-3 altitude overlaid (black line) and observed temperatures 582 at each end of the track indicated; (c) HIWRAP Ku-band Doppler radial velocity (m s⁻¹, shaded) with corrections for aircraft motions and horizontal wind applied. P-3 altitude is overlaid as in (b); 583 584 (d) Horizontal distribution of reflectivity from EXRAD conical scans spanning 10 km south and 10 km north of the ER-2 flight track at 4.0 km ASL, gridded at 0.1° latitude and 0.1° longitude 585 following Helms et al. (2020); (e) Time (black) and distance (red) offset of the P-3 and ER-2 586 aircraft during the coordinated leg; (f) Total IWC (g m⁻³) calculated using Heymsfield et al. (2004) 587 from the 2DS probe and total LWC (g m⁻³) measured by the CDP on the P-3; (g) particle number 588 concentration from the HVPS and 2DS optical array probes; and (h) merged particle size 589 590 distribution represented as the number distribution function N(D) (cm⁻⁴, shaded).
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