



1 **The Status and Future of Small Uncrewed Aircraft Systems (UAS) in Operational**
2 **Meteorology**

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61 **Capsule**

62 Small weather-sensing Uncrewed Aircraft Systems are becoming reliable and accurate enough to
63 be considered as a cost-effective solution for filling observational gaps that could enhance National
64 Meteorological and Hydrological Services around the world.

65 **Abstract**

66 The boundary layer plays a critical role in regulating energy and moisture exchange between the
67 surface and the free atmosphere. However, the boundary layer and lower atmosphere (including
68 shallow flow features and horizontal gradients that influence local weather) are not sampled at
69 time and space scales needed to improve mesoscale analyses that are used to drive short-term
70 model predictions of impactful weather. These data gaps are exasperated in remote and less
71 developed parts of the world where relatively cheap observational capabilities could help
72 immensely. The continued development of small, weather-sensing Uncrewed Aircraft Systems
73 (UAS), coupled with the emergence of an entirely new commercial sector focused on UAS
74 applications, has created novel opportunities for partially filling this observational gap. This article
75 provides an overview of the current level of readiness of small UAS for routinely sensing the lower
76 atmosphere in support of National Meteorological and Hydrological Services (NMHS) around the
77 world. The potential benefits of UAS observations in operational weather forecasting and
78 numerical weather prediction are discussed, as are key considerations that will need to be
79 addressed before their widespread adoption. Finally, potential pathways for implementation of
80 weather-sensing UAS into operations, which hinge on their successful demonstration within
81 collaborative, multi-agency-sponsored testbeds, are suggested.

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83 A significant *in situ* observational gap resides in the lower atmosphere which encompasses
84 the surface layer, atmospheric boundary layer and lower free troposphere (National Research
85 Council 2009; National Academies of Sciences, Engineering, and Medicine 2018; Geerts et al.
86 2018; NOAA 2020). This observational gap is most acute in remote locations and is further
87 exacerbated in less developed regions of the world (WMO 2018). A schematic representation of
88 this *in situ* observation gap is shown in Figure 1. While surface meteorological stations including
89 airport-based observing stations and mesonets provide good spatio-temporal near-surface
90 coverage over land areas in developed countries, radiosondes are launched just twice daily and are
91 generally spaced over 300 km apart. Thus, radiosondes alone greatly under-sample mesoscale and
92 diurnal variability of the atmosphere. While aircraft-based observations (which may be obtained
93 via AMDAR, TAMDAR, ADS-B, Mode S) can capture diurnal variations of the lower atmosphere
94 (Zhang et al. 2019), these observations are confined to arrival and departure ascent/descent legs at
95 major airports and have reduced temporal coverage overnight.

96 Radar networks and satellite observations help to fill these *in situ* observational gaps, but
97 these remote-sensing platforms also have limitations. Doppler weather radar networks (e.g., U.S.
98 NEXRAD) require the presence of scatterers (bugs, precipitation) for sensing velocities, provide
99 limited thermodynamic information and have significant gaps in coverage at lower altitudes,
100 particularly in mountainous areas. Moreover, advanced radar networks are not available in many
101 parts of the world because they are expensive to operate and maintain. While geostationary
102 satellites provide outstanding horizontal and temporal sampling of multi-channel radiances,
103 retrievals of thermodynamic properties of the lower atmosphere are too coarse to resolve horizontal
104 variability important for short term predictions and are often hindered by the presence of clouds
105 (e.g., Wulfmeyer et al. 2015).

106 Reducing gaps in the observation of thermodynamic and kinematic properties of the lower
107 atmosphere is critical for achieving more skillful mesoscale predictions of high-impact weather.
108 For example, the U.S. National Oceanic and Atmospheric Administration's goal of developing a
109 Warn-On-Forecast capability (Stensrud et al. 2009, 2013) hinges on improved observation of the
110 lower atmosphere at space and time scales relevant for accurately predicting hazardous severe
111 weather at the county scale. Temporal and spatial gaps in existing observing systems contribute to
112 model forecast uncertainty (e.g., Dong et al. 2011; James and Benjamin 2017; James et al. 2020).
113 In fact, the maximum skill of regional numerical weather prediction (NWP) models will not be
114 realized until spatio-temporal sampling of the lower atmosphere is comparable to the model's
115 effective resolution (Dabberdt et al. 2005).

116 Operational meteorologists have also pointed to the need for increased observation of the
117 lower atmosphere to improve the accuracy of short-term (< 24 hour) forecast guidance products
118 (e.g., Houston et al. 2020, 2021). Surveys of operational meteorologists in the U.S. have indicated
119 a need for increased sampling of remote environmental locations during periods of rapidly
120 changing conditions (e.g., evolution of temperature, moisture, wind profiles in pre-convective
121 environment) to improve their short term forecast products (Houston et al. 2020, 2021). Moreover,
122 the dearth of lower-atmospheric observations is particularly significant in less developed regions
123 of the world, making it particularly challenging for both NWP models and NHMS meteorologists
124 to produce accurate short-term forecasts of high impact weather events like severe thunderstorms
125 (e.g., Woodhams et al. 2018) and dust storms (e.g., Wang 2015).

126 In the late 1990s, small Uncrewed Aircraft Systems (UAS)¹ began to emerge as a new
127 system for obtaining *in situ* measurements within the lower atmosphere (Holland et al. 2001; Curry
128 et al. 2004). Note that the term “Uncrewed” is used to remove gender specificity following Bell et
129 al. (2020), and is our preferred terminology for describing these aircraft systems. Here, the term
130 “small UAS” refers to a class of autonomous aircraft weighing less than 25 kg (55 lbs) as defined
131 by Federal Aviation Administration’s (FAA) Part 107 regulation (and similar European laws). In
132 the last 10 years, the number of research programs focused on the development of UAS and UAS
133 weather-sensing capabilities has flourished. At the same time, commercial applications for small
134 UAS has grown dramatically (e.g., Gangwal et al. 2019; Rigby 2020) with this trend being
135 expected to continue for several years (FAA 2020).

136 Today’s small weather-sensing UAS (hereafter referred to as WxUAS following Chilson
137 et al. (2019) and Bell et al. (2020) are nearly 100% reusable (as opposed to radiosondes of which
138 only 20% are recovered and a smaller fraction reused), can rapidly sample the lower atmosphere,
139 are powered with batteries that can be recharged using locally generated solar energy, and are
140 extremely adaptable; capable of flying targeted missions or performing routine systematic
141 profiling (Elston et al. 2015). The term WxUAS is used here to distinguish between UAS that are
142 dedicated to observing the atmosphere and those that may collect atmospheric data coincidentally
143 while performing some other primary service (e.g., commercial delivery).

144 Recent development efforts have resulted in the production of fully autonomous systems
145 that can automatically progresses through all stages of flight including take-off and landing,
146 profiling, system checks and recharging (Leuenberger et al. 2020). Special permissions have been

¹ Also known as drones, remotely piloted aircraft or unmanned aircraft systems.

147 obtained to allow WxUAS to fly up to 6 km AGL enabling sampling of rapidly varying weather
148 features with high vertical resolution over a deep layer of the atmosphere (Figure 2). In this
149 example, a profiling Meteodrone WxUAS (using Meteomatics Meteobase for automatic
150 recharging) captured the evolution of temperature, humidity and winds during a recent fog
151 evolution study. Note the deepening layer of relative humidity exceeding 90% just above the
152 surface associated with a shallow layer of northeasterly winds. In addition to reliability and
153 efficiency, the accuracy of wind, temperature, and humidity measurements obtained with WxUAS
154 is now comparable to that of calibrated tower and radiosonde measurements (e.g., Leuenberger et
155 al. 2020; Bell et al. 2020) with the consistency of observational errors also improving (Barbieri et
156 al. 2019). These attributes, coupled with decreasing costs of UAS production, operation, and
157 maintenance, are making WxUAS an economically viable option for use by NMHS to fill
158 observational data voids (e.g., McFarquhar et al. 2020).

159 While the utility of WxUAS for collecting research quality datasets within the lower
160 atmosphere is now well established (e.g., Houston et al. 2012; Elston et al. 2015; Bärfuss et al.
161 2018; de Boer et al. 2018; Vömel et al. 2018; Kral et al. 2020), their use in operational meteorology
162 has been limited due, in part, to limitations on the accessibility of airspace for land-based flights
163 (Houston et al. 2012), measurement accuracy, and the accessibility of data to forecasters (Koch et
164 al. 2018). In recent years, UAS and WxUAS flights over land areas have become common-place
165 (e.g., Chilson et al. 2019; Lee et al. 2019; Lee and Buban 2020; de Boer et al. 2020; Bailey et al.
166 2020; Frew et al. 2020) but linkages to operational meteorology have only just begun (Koch et al.
167 2018). Koch et al. (2018) found that forecasters didn't use WxUAS observations to full effect
168 because the data were not integrated into their operational display tools. In a separate, short-
169 duration testbed, Cione et al. (2020) attempted to demonstrate the utility of WxUAS observations

170 in short-term hurricane forecasting by providing WxUAS observations within the boundary layer
171 of hurricanes (including their eyewalls) to NOAA's National Hurricane Center (NHC) in real time.
172 These initial exercises by Cione et al. (2020) were critical for assessing the readiness of WxUAS
173 and other system technologies (e.g., communications, flight systems, airframe design) required to
174 permit the collection and transmission of unprecedented targeted measurements to hurricane
175 forecasters, and pointed to the need for additional research and development efforts.

176 WxUAS observations can also influence operational meteorology and NHMS through their
177 assimilation into operational numerical weather prediction models. As will be discussed briefly
178 below, several studies have demonstrated the benefit of assimilating WxUAS observations into
179 regional NWP research models; however, additional research is needed to fully assess their
180 potential. Work is also needed to establish direct lines of communication between WxUAS and
181 operational modeling centers such that the values of these new observations can be assessed in an
182 operational environment.

183 The goal of this paper is to discuss the main factors influencing the adoption of WxUAS
184 by NHMS. The role that recent, current, and planned testbed demonstrations will play in
185 facilitating the potential adoption of WxUAS by NHMS is discussed. Finally, potential pathways
186 from WxUAS testbed demonstrations to operational meteorology are proposed and
187 recommendations for getting involved in the research-to-operations process are given.

188 **FACTORS INFLUENCING THE PATH TO OPERATIONS FOR WxUAS.** The adoption of
189 WxUAS by NHMS will only occur if the value of improved forecasts and resulting support
190 services significantly exceeds the cost of implementation. Because of massive investments in
191 developing UAS technologies over the past 10 years by private industry, governments, and

192 university research groups, the cost of WxUAS has decreased dramatically (Belton 2015; Nath
193 2020). The current cost of operating a WxUAS, particularly due to the requirement of one pilot
194 per UAS and, in many cases, the need for human observers to meet sense-and-avoid (SAA)
195 requirements, drives a relatively high cost to operate. However, progress toward widespread
196 autonomous WxUAS flight without direct human management is being made via allowances for
197 beyond visual line of sight (BVLOS) flight (Jacob et al. 2020) through both SAA technologies
198 which requires intercommunication between UAS and detect-and-avoid (DAA) systems (Mitchell
199 et al. 2020) which use sensors to detect obstacles (e.g., power lines, cell towers, other UAS, and
200 piloted aircraft) and do not rely communications with other UAS to maintain air space separation.

201 While the cost associated with implementing WxUAS continues to decline, an increasing
202 number of studies have demonstrated that WxUAS data assimilation (DA) can improve the skill
203 of mesoscale weather predictions. An Observing System Experiment (OSE) study by Leuenberger
204 et al. (2020) showed that the assimilation of WxUAS observations improved the short-term
205 prediction of radiation fog events. Jensen et al. (2021a,b) used OSEs to demonstrate that WxUAS
206 DA dramatically reduced biases in the analyses of low- and mid-level moisture and winds that
207 were critical for more accurately predicting the timing and location of thunderstorms and
208 subsequent outflows. An example of the impact of WxUAS DA on improving the representation
209 of the pre-convective environment and subsequent storm prediction is shown in Figure 3. Here the
210 assimilation of observations collected with several distributed profiling WxUAS (see Jensen et al.
211 2021b for details) reduced biases in both temperature and moisture profiles that made the
212 atmosphere more conducive for the development of convective storms. These findings are
213 consistent with the results of Moore (2018) and Chilson et al. (2019) that demonstrated the value
214 of WxUAS DA in predicting thunderstorm evolution. Both Jensen et al. (2021b) and Moore (2018)

215 demonstrated the value of assimilating targeted profile of winds, temperature and humidity using
216 WxUAS to improve prediction of storm evolution compared to that obtained when assimilating
217 conventional observations alone.

218 While these results demonstrate the potential for UAS DA in NWP and hint at some of the
219 requirements for data accuracy and sampling strategies, a great deal of work remains to assess the
220 effectiveness of UAS DA across a range of challenging high impact weather prediction scenarios.
221 In addition, strategies for implementation of WxUAS within an operational environment, with or
222 without subsequent DA, need to be developed via close coordination with NHMS and operational
223 forecast offices (Houston et al. 2020, 2021). Much of these efforts are in need of end-to-end testbed
224 demonstrations that can help facilitate the establishment of linkages between WxUAS
225 development efforts, operational meteorologists, and modeling centers.

226 **DEMONSTRATION TESTBEDS.** While there have been a few short-duration testbed
227 demonstrations that looked at the use of WxUAS in operational environments (e.g., Koch et al.
228 2018; Cione et al. 2020), these testbeds have been limited in scope and duration. Longer duration
229 testbeds of increasingly broader scope are needed to more completely assess the utility of WxUAS
230 in operational environments. Such testbeds can also be used to develop data protocols,
231 requirements, and standards (e.g., de Boer et al. 2020). Several testbeds described below are
232 already underway or planned that will further facilitate research-to-operations of WxUAS. Key
233 goals of these testbeds are to establish connections between the UAS operators and the weather
234 enterprise and to facilitate interactions between the WxUAS developers, commercial UAS
235 operators, operational meteorologists, and other stakeholders.

236 The NOAA Air Resources Laboratory (ARL) Atmospheric Turbulence and Diffusion
237 Division (ATDD) in Oak Ridge, Tennessee has established a long-term WxUAS testbed to
238 demonstrate the benefit of routine WxUAS observations to operational meteorology, to perform
239 calibration of WxUAS observations, and to support boundary layer research. As part of this
240 testbed, WxUAS are being used to obtain up to 8 profiles per day that are transmitted to the
241 Morristown, TN Weather Forecast Office (WFO) in near real time to support their short-term
242 forecast desk (Figure 4). As of this writing, over 350 flights have been performed for Morristown
243 forecasters which is located 80 km away from the nearest radiosonde site. Forecasters have
244 reported that the WxUAS observations have improved their understanding of local processes that
245 contribute to boundary layer evolution and indicated that, once fully implemented into the forecast
246 process, the high rate, local observations afforded by WxUAS would lead to greater skill and
247 specificity of short term forecasts of winds, fog, and thunderstorms initiation needed to produce
248 the Terminal Aerodrome Forecasts (TAFs).

249 In Finland, researchers at FMI (Finnish Meteorological Institute) have been operating a
250 WxUAS testbed since mid-2020, collecting 1-2 profiles every hour during the day. These WxUAS
251 observations are being validated against radiosonde observations. Work is also underway to
252 establish a real-time feed of WxUAS data to Météo-France and eventually to the European Center
253 for Medium-range Weather Forecasting (ECMWF) to facilitate data assimilation studies. More
254 broadly, the World Meteorological Organization (WMO) is organizing a year-long global
255 demonstration period (starting as early as the Fall of 2022) geared toward increasing the visibility
256 of WxUAS for use in operational meteorology and will work toward establishing international
257 standards for data protocols and data quality criteria that will facilitate usage of WxUAS
258 observations by major modeling centers (WMO 2021).

259 Several WxUAS demonstration testbeds are currently being planned throughout the world
260 over the next few years. A 6-month testbed focused on assessing the performance and reliability
261 of WxUAS in strong winds and icing conditions will commence in Switzerland in November 2021.
262 The WxUAS observations will be compared with conventional radiosonde and remotely sensed
263 observations. Moreover, the WxUAS observations will be transmitted to MeteoSwiss in near real
264 time for parallel DA studies designed to evaluate the impact of WxUAS observations on NWP
265 skill. The Oklahoma State University is leading a team of universities and private partners to
266 develop a weather-aware UAS Traffic Management (UTM) system similar to the concept
267 described in the sidebar. As discussed in the sidebar, WxUAS, possibly along with commercial
268 UAS, will collect and transmit observations of the lower atmosphere that will inform other UAS
269 operating nearby of winds and other potential UAS hazards (Jacob et al. 2021). The WxUAS
270 observations will be made available for assimilation into experimental NWP models to evaluate
271 the impact of these observations on the accuracy of predicted low-level winds and other UAS
272 weather hazards. Similar studies have recently been initiated in the U.K. and involve the U.K. Met
273 Office (Stonor 2021).

274 The utility of data collected with WxUAS will be fostered by the establishment of common
275 data formats and reporting standards. Interactions within testbeds are needed to help define
276 metadata requirements, which might include details of how the data were collected (type of
277 aircraft, commercial UAS vs dedicated WxUAS) and provide additional information on the data
278 quality and level of post-processing. It will be important to coordinate these activities with those
279 ongoing within the ASTM (not an acronym) International F38 UAS committee which is focused
280 on developing standards to support routine WxUAS operations². Similarly in Europe, the UAS

² <https://www.astm.org/COMMIT/SUBCOMMIT/F3802.htm>

281 Task team within **PRO**filin^g the atmospheric **B**oundary Layer at **E**uropean scale (**PROBE**)
282 initiative is working to develop standards and minimum data quality requirements for application
283 to operational meteorology through industry engagement and testbed demonstrations (Cimini et
284 al. 2020).

285 Testbeds can also be used to develop cost-sharing approaches that can be evaluated in a
286 real world environment (see the sidebar). For example, the cost of collecting and transmitting
287 weather data can be weighed against the added value of improved situational awareness among
288 UAS operators, as well as the impact of these data on short-term weather forecasts needed to plan
289 and execute safe and efficient commercial UAS operations. Likewise, the cost of maintaining and
290 operating a small fleet of WxUAS by a NHMS can be weighed against the value of increased lead
291 time in the prediction of severe or high-impact weather or improved air quality forecasts.

292 Within these testbeds, experiments can be designed to tackle hurdles with moving WxUAS
293 into operational use by NHMS. Some of the most pressing hurdles to widespread adoption of
294 WxUAS by NHMS are discussed below.

295 **HURDLES TO ROUTINE OPERATIONS.** Key challenges to the adoption of WxUAS by
296 NHMS around the world include the cost of implementing new technologies, limitations on the
297 range of operating conditions under which WxUAS can operate, system reliability and
298 measurement accuracy, regulatory limitations (which vary from country to country) and societal
299 acceptance.

300 *Cost.* With the continued expansion of commercial UAS operations, the cost of acquiring
301 components to build WxUAS has declined markedly in the past decade. Nonetheless, the cost of
302 purchasing a fully autonomous WxUAS including an automated recharging system is roughly

303 \$100K with operations/maintenance adding roughly \$20K/year. These costs compare favorably to
304 the cost of radiosonde launches (including materials and labor) which has been estimated at around
305 \$300/launch or roughly \$200K/year/site (depending on the number of special launches requested),
306 though automated radiosonde systems are beginning to reduce some of the operational costs
307 (Madonna et al. 2020). WxUAS offer the advantage of providing near-continuous profiling of the
308 lower atmosphere at a fraction of the cost of radiosondes without the extra burden of polluting the
309 environment with circuit boards and batteries that are seldom recovered. Moreover, as the
310 reliability of WxUAS increases and automation continues to become more sophisticated, the cost
311 of operations should continue to decline whereby, ultimately, a single operator will be able to
312 monitor an entire fleet of autonomously profiling WxUAS.

313 *Range of operating conditions.* Assuming that permission can be obtained to operate BVLOS
314 within and above cloud layers (as is already being demonstrated at some testbeds), the conditions
315 most impactful to unhindered autonomous profiling of the lower atmosphere include in-flight icing
316 (from snow, supercooled cloud droplets, and freezing precipitation) and excessive winds. Icing
317 can cause light-weight, low-powered UAS to quickly loose lift causing the platform to drop from
318 the sky (e.g., Roseman and Argrow 2020). Strong winds that exceed UAS airspeed and heavy rain
319 exceeding aircraft lift capacity can also ground operations. Algorithms that can automatically
320 detect these conditions (either directly by the UAS or indirectly using external observations like
321 those from weather radar), warn operators, and automatically commence abort sequences are
322 needed. Recent efforts to mitigate icing have been pioneered by Meteomatics who demonstrated
323 an active icing mitigation system that heats the blades to prevent icing (Figure 5). While this new
324 anti-icing capability enables safe operations within some icing conditions, such capabilities need

325 to be further developed and fielded in testbeds to determine the true range of safe operating
326 conditions.

327 *Reliability and accuracy.* Related to that discussed above, work remains to demonstrate the
328 reliability of fully autonomous WxUAS under a range of environments (Petritoli et al. 2018).
329 Estimates for a required level of reliability must be developed based on perceived risk. The light-
330 weight nature of WxUAS makes them very unlikely to cause harm to human life or property in the
331 event of loss of control (Barr et al. 2017); however, a low incident rate is still critical for reducing
332 the cost of operations and, as will be described further below, to support positive public perception.
333 The accuracy of WxUAS measurements still varies significantly as a function of UAS type (fixed-
334 wing versus multi-rotor), mounting, shielding and aspiration of the sensors, and wind retrieval
335 techniques (Barbeiri et al. 2019). Standards for calibration and metadata requirements for WxUAS
336 measurements must be established (e.g., Jacobs et al. 2018) and methods for on-the-fly evaluation
337 of the quality of commercial UAS observations and possibly automated calibration methods are
338 needed.

339 *Regulatory challenges.* For maximum effectiveness in thunderstorm prediction and other weather
340 prediction scenarios, WxUAS will need to operate up to at least 1000 m AGL (Chilson et al. 2019)
341 which is considered BVLOS. In the U.S. and Europe, UAS are generally allowed to fly up to 120
342 m AGL under Visual Line of Sight (VLOS) conditions by the FAA and the European Aviation
343 Safety Agency (EASA) without requiring a waiver. Exceptions to VLOS and the 120 m AGL rules
344 can be obtained in both the U.S. and Europe. In the U.S., this is done by obtaining a Certificate of
345 Authorization (COA) from the FAA. In Europe, starting in 2021, a waiver that will permit BVLOS
346 called a PDRA-01 (Pre-Defined Risk Assessment) may be obtained from the EASA. On a case-

347 by-case basis, WxUAS have already been granted permission to fly up to 2 km AGL in the U.S.
348 and up to 6 km AGL in Europe. However, in order to maximize the potential of WxUAS for
349 operational meteorology, the process for obtaining these waivers needs to be streamlined and
350 standardized. Testbeds can be used to further demonstrate WxUAS capabilities in a safe
351 environment, while working with regulators to streamline procedures for obtaining permissions
352 for WxUAS operations.

353 Sense-and-avoid technologies that minimize the risk of collision with other low-flying aircraft will
354 further help alleviate regulatory restrictions. The development and implementation of sense-and-
355 avoid and remote identification systems are already underway within the FAA UAS Traffic
356 Management System Pilot Program (UPP)³. The outcomes of this work will lead to greater access
357 to airspace above 120 m and further enable BVLOS operations in the U.S., paving the way for
358 more routine WxUAS sensing of the lower atmosphere including the entire depth of the boundary
359 layer. Once again, working with regulatory agencies (e.g., FAA) within a testbed framework will
360 provide a safe environment for developing and testing protocols needed to integrate WxUAS
361 operations with low-altitude crewed air traffic.

362 *Societal acceptance.* Another barrier that must be overcome in order to routinely operate WxUAS
363 and to expand commercial uses for UAS is societal acceptance (Walther et al. 2019). UAS flights
364 near homes and over people raise legitimate privacy and safety concerns. To address safety
365 concerns, UAS operators must demonstrate that the risk of a UAS failure leading to injury or
366 property damage is negligible. This can be done through proven mitigation engineering strategies
367 and procedures such as equipping the UAS with a parachute and/or making them able to

³https://www.faa.gov/uas/research_development/traffic_management/utm_pilot_program/

368 disintegrate on impact. Privacy issues also need to be considered, particularly since forecasters
369 have noted a desire to equip WxUAS with cameras to target and monitor rapidly evolving, high-
370 impact weather conditions such as assessing whether convective cap is breaking or getting a view
371 of supercell storm structure to evaluate severity (Houston et al. 2021). Significant outreach to
372 educate the public will be required and can be achieved via testbeds. Additional steps such using
373 consistent non-threatening colors that would make WxUAS readily identifiable by the general
374 public would also help to alleviate public privacy concerns and mistrust.

375 Another notable aspect of UAS acceptance involves the development of regulations that
376 promote national security. This aspect of UAS operations has been considered since the inception
377 of UTM concepts (e.g., Kopardekar 2014). Remote identification technology will enable UTM
378 system operators, as well as public safety or government agencies, to interrogate any UAS to
379 determine its intent, operating parameters, and pilot information. Demonstrations of remote
380 identification systems are planned for 2021 in the U.S. (Garret-Glasser 2020) and, where possible,
381 should be coordinated with upcoming testbeds. Ultimately, this new remote identification system
382 will be required by all UAS operating within the U.S. Resolution of security concerns will allow
383 commercial UAS operations to expand while at the same time improving local, state, and national
384 security, as well as the safety of the general public.

385 **PATHWAYS TO OPERATIONAL IMPLEMENTATION.** There are several potential
386 pathways to implementation of WxUAS in support of operational meteorology. These pathways
387 are being funded via research efforts at universities and government programs. For example, the
388 U.S. NOAA has several ongoing programs designed to utilize UAS in support of their missions
389 (uas.noaa.gov). Another approach could involve augmentation of existing surface observing

390 networks with profiling WxUAS to create a three-dimensional mesonet (e.g., Chilson et al. 2019).
391 This implementation approach takes advantage of existing infrastructure while offering a notable
392 expansion of observational capabilities. Such augmentations could be implemented by NHMS,
393 local governments, and/or the private sector. Depending on resources and funding mechanisms,
394 there may be opportunities for cost sharing that can help build out WxUAS observing capabilities.

395 Another pathway for implementation may be through the introduction of targeted
396 observations which take full advantage of the flexible nature of UAS. Under this approach
397 WxUAS could be tasked to deploy to areas that drive uncertainty in the prediction of high impact
398 weather event or to augment existing radiosonde launches with more frequent sampling of the
399 lower-atmosphere in highly evolving weather scenarios. For example, multiple WxUAS could be
400 deployed in the vicinity of a dry line to more accurately assess gradients, stability profiles and
401 surface boundaries which can improve prediction of the location and intensity of severe
402 thunderstorms. A recent survey of meteorologists revealed that targeted surveillance may be the
403 preferred operational modality for forecasters in the U.S. (Houston et al. 2021).

404 A separate, yet potentially parallel pathway to operations for obtaining weather
405 observations via UAS follows the Aircraft Meteorological Data Relay (AMDAR) model whereby
406 commercial aircraft downlink weather observations for use by NWP modeling centers (Benjamin
407 et al. 2010; Peterson 2016). Many commercial UAS that fly BVLOS already carry basic
408 meteorological sensors that measure temperature and humidity in support of their operations (e.g.,
409 package or medical supply delivery). In fact, Robinson et al. (2020) posit that if even a small
410 fraction of commercial UAS reported this weather information in the future this could have a
411 profound influence on the safety and efficiency of their operations through improved situational

412 awareness and improved weather guidance. As described in the sidebar, it will require additional
413 infrastructure to get this weather information onto data servers where it can be made available to
414 weather service providers, weather forecast offices, and modeling centers. Yet the benefits of these
415 low costs are likely to pay for themselves many times over.

416 Finally, there will be opportunities for data sharing and developing new cost models to
417 determine the value of UAS-based weather observations in the private sector. Commercial UAS
418 may find that the weather data they collect can provide an opportunity for additional revenue.
419 Agreements will need to be developed once the value of weather data collected by commercial
420 UAS is better quantified. At the same time, data sharing may be an equitable approach whereby
421 commercial UAS observations are provided to modeling centers and in turn weather prediction
422 needed by UAS operators is improved, resulting in a win-win solution for all stakeholders involved
423 (see Sidebar).

424 **VISION FOR THE FUTURE.** In addition to WxUAS observations improving weather
425 prediction and supporting operational forecasters, UAS have demonstrated utility in a number of
426 other NHMS services (see Manfreda et al. 2018 for a review of environmental applications).
427 Specific examples of demonstrated capabilities include the use of UAS to perform detailed surveys
428 of severe thunderstorm damage (e.g., Wagner et al. 2019), to assess the impact of tropical systems
429 and synoptic storms on coastal erosion (Kaamin et al. 2016), and to monitor inland water body
430 flooding (Imam et al. 2020; Dyer et al. 2020). UAS have also been used to collect measurements
431 within volcanic plumes (Galle et al. 2020; Schellenberg et al. 2020) to assess the potential for
432 hazardous air quality or volcanic ash impacts to passenger aircraft. These applications should

433 continue to be explored and augmented through testbed demonstrations that facilitate partnerships
434 between researchers and NHMS.

435 With the cost of UAS platforms, operations, and maintenance continuing to decline and as
436 UAS move toward greener technologies (e.g., solar-powered battery recharging stations, more
437 efficient engines, longer-lived batteries), the economics of using WxUAS to observe the lower
438 atmosphere has become quite compelling. Efforts over the next five years should focus on
439 establishing larger-scale testbeds that strengthen partnerships between WxUAS developers and
440 potential stakeholders while at the same time facilitating the collection of observations over larger
441 areas for a more complete assessment of potential benefits. In particular, the potential for serving
442 as a means of filling observation gaps in less developed regions of the world should be explore in
443 future demonstration testbeds. Commercial UAS operators can use testbeds to develop business
444 models to determine the market value for weather observations through the demonstration of the
445 impact of these observations on forecast skill (e.g., Zhang et al. 2016). The approach used here
446 could be similar to that which unfolded with commercial transport aircraft observations via the
447 Tropospheric Airborne Meteorological Data Reporting (TAMDAR) program (Daniels et al. 2006).

448 While a number of smaller scale WxUAS testbeds have been undertaken, these endeavors
449 need to be expanded in scope and duration in order to fully demonstrate the value of WxUAS
450 observations in improving NHMS services. Key to furthering the use of WxUAS observations will
451 be making the data available for modeling centers for use in side-by-side data assimilation
452 experiments which is most easily done in a real-time environment. Having NHMS (both modeling
453 centers and operational meteorologists) get involved with current and future testbeds will be
454 critical for increasing the acceptance of this emerging source of weather observations. In addition,

455 societal acceptance is also critical and should continue to be nurtured through outreach activities
456 such as issuing press releases, talking to local news outlets, increasing presence on social media,
457 holding public open houses during demonstration projects, and K-12 education opportunities (de
458 Boer et al. 2020).

459 Given the rapid progress made over the last few years, there is little doubt that in the near
460 future, WxUAS and commercial UAS will begin to fill the observational data gap in the lower
461 atmosphere which will lead to significant advances in weather forecasting and the skill of regional
462 NWP models.

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476 **Sidebar SB1. WEATHER-AWARE UAS TRAFFIC MANAGEMENT SYSTEM**

477 As commercial UAS operations continue to expand throughout the world, government regulatory
478 agencies are working with aviation stakeholders and research partners to develop UAS traffic
479 management systems (UTM) to organize and monitor this airspace. With small UAS being more
480 susceptible to weather conditions than larger aircraft, a new suite of much higher resolution
481 weather guidance products, such as that demonstrated by Pinto et al. (2021), will be needed to
482 support UTM. The inclusion of tailored weather information in UAS flight path planning tools
483 will aid in route optimization by helping operators find favorable winds that optimize power
484 consumption along a user-defined flight path (yellow curve). Thus, it will be critical to improve
485 the accuracy of low-level wind speed and direction analyses and forecasts at scales less than 1 km.
486 In addition, more accurate prediction of weather conditions that are hazardous to commercial UAS
487 operations (e.g., low ceilings or fog and areas of enhanced turbulence) will be vital for high mission
488 success rates (Thibbotuwawa et al. 2020).

489 Weather prediction at sub-kilometer scales requires mesoscale-to-microscale coupling (e.g., Haupt
490 et al. 2019). Operational mesoscale model predictions can be improved by filling the data gap in
491 the lower atmosphere with observations obtained with WxUAS and commercial UAS. These new
492 UAS-borne observations will be downlinked and transmitted to modeling centers to improve initial
493 conditions used in their regional models (e.g., James and Benjamin 2017). Observations from
494 dedicated profiling WxUAS may be used to complement existing observing networks like that of
495 New York State Mesonet. Additional observations from commercial UAS would further increase
496 data coverage and enhance forecast skill. In this way, commercial UAS can contribute to
497 improving their own safety and efficiency while at the same time improving short-term weather
498 prediction for the benefit of society.

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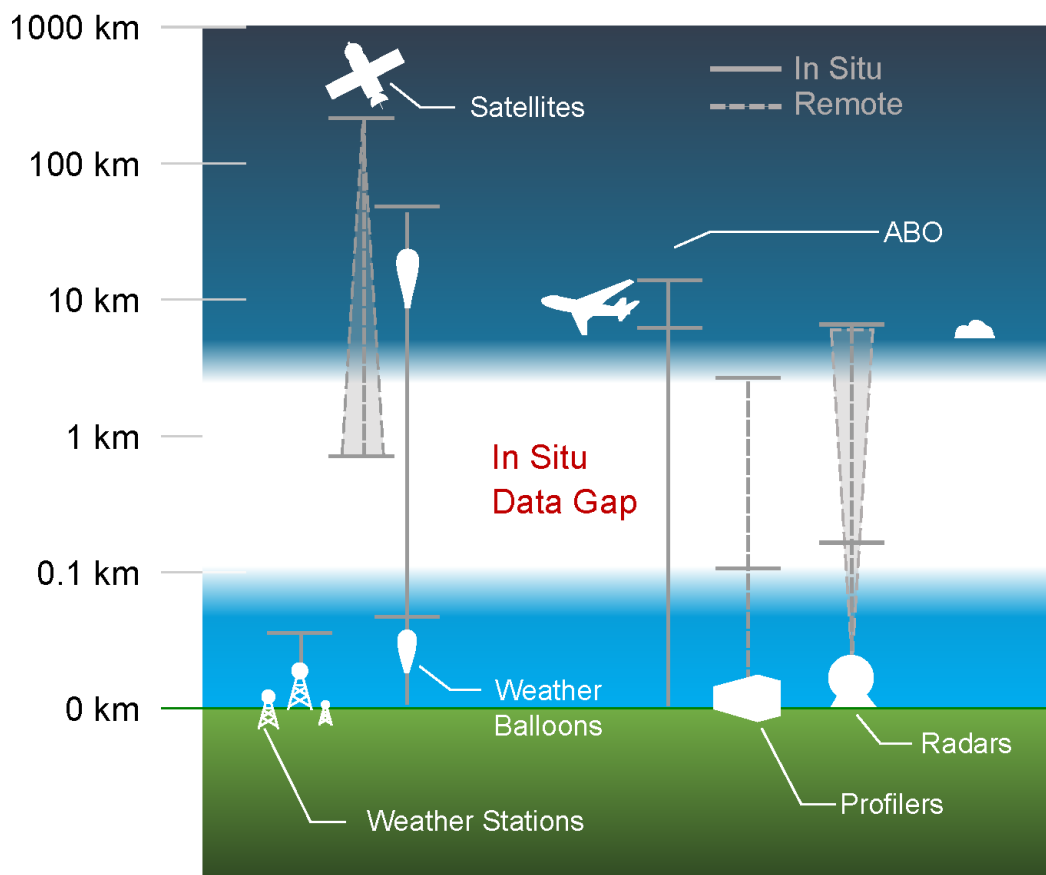


Figure 1. Schematic illustrating the *in situ* observation gap in the lower atmosphere. Horizontal lines indicate nominal regions of primary data collection. Diagonal lines indicate changing size of foot print with distance from remote sensor. ABO refers to commercial Aircraft Based Observations.

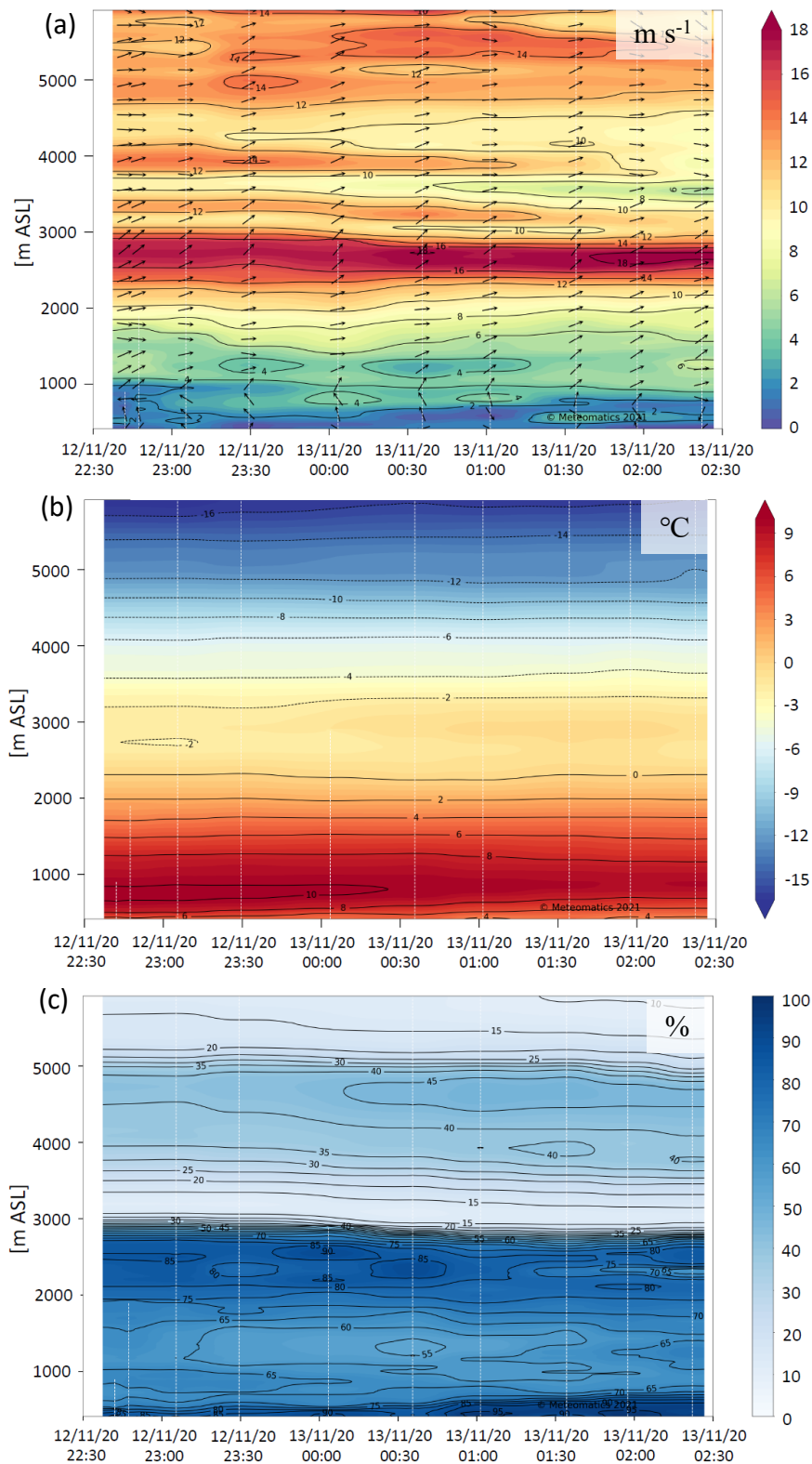


Figure 2. Time-height profiles of (a) wind speed (color fill) and direction (arrows pointing with the wind), (b) temperature and (c) relative humidity obtained with a Meteomatics Meteodrone collected prior to and during a fog event at Amlikon, Switzerland. Profiles were obtained every 30 min up to 6 km AMSL.

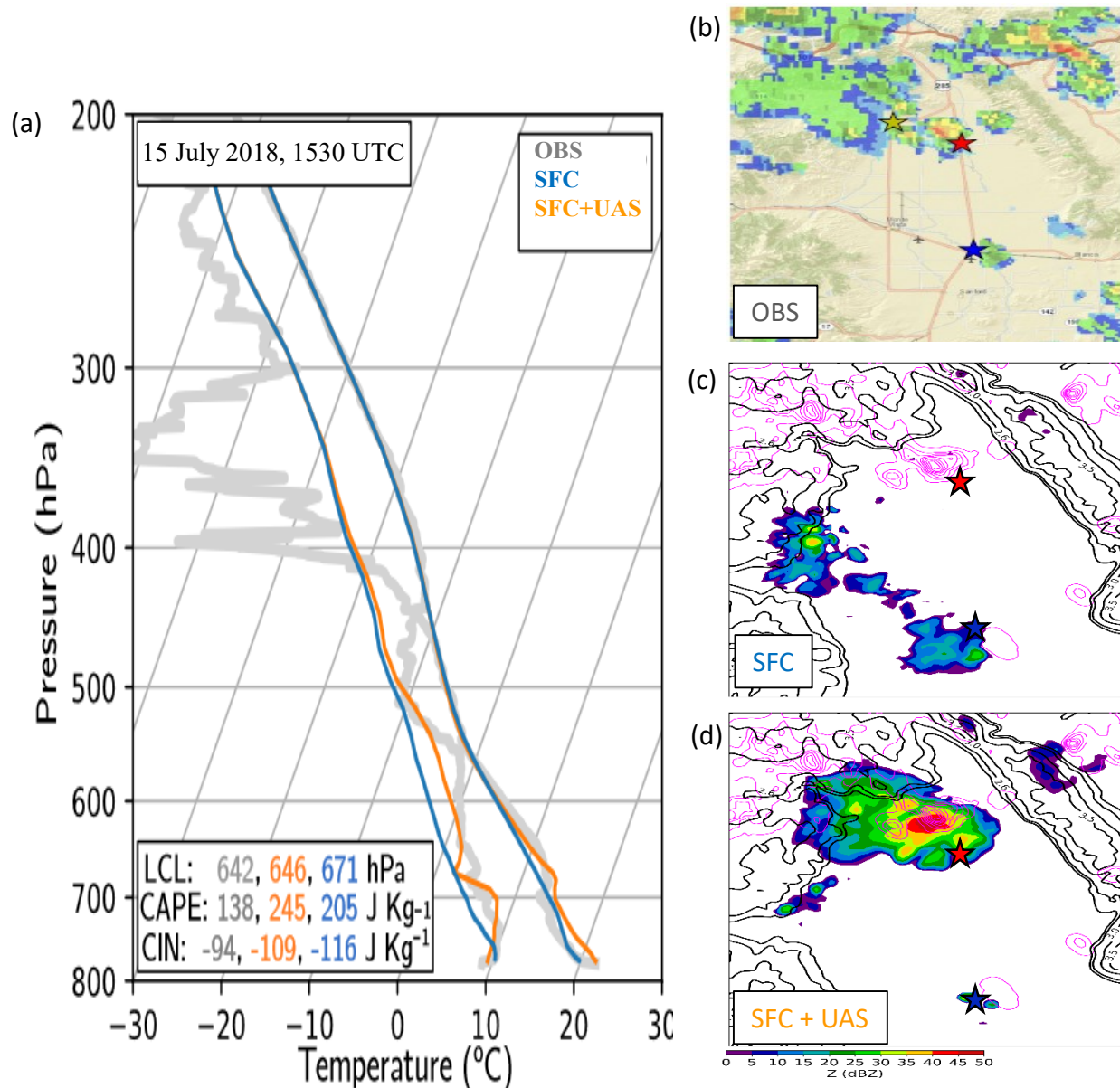


Figure 3. Analyses of temperature and dew point profiles obtained in the center of the San Luis Valley of Colorado using EnKF data assimilation with 15-minute cycling of surface observations (SFC) and surface plus WxUAS observations (SFC+UAS) along with independent observed values obtained with a radiosonde (OBS). Column on right shows (b) observed composite reflectivity and that obtained with a 30 min forecast with (c) SFC DA and (d) SFC+UAS DA valid at 2000 UTC. Magenta contours in (c) and (d) denote 3 hour accumulations estimated from the observed composite reflectivity. Red and blue stars denote locations of Moffat and Alamosa, respectively.

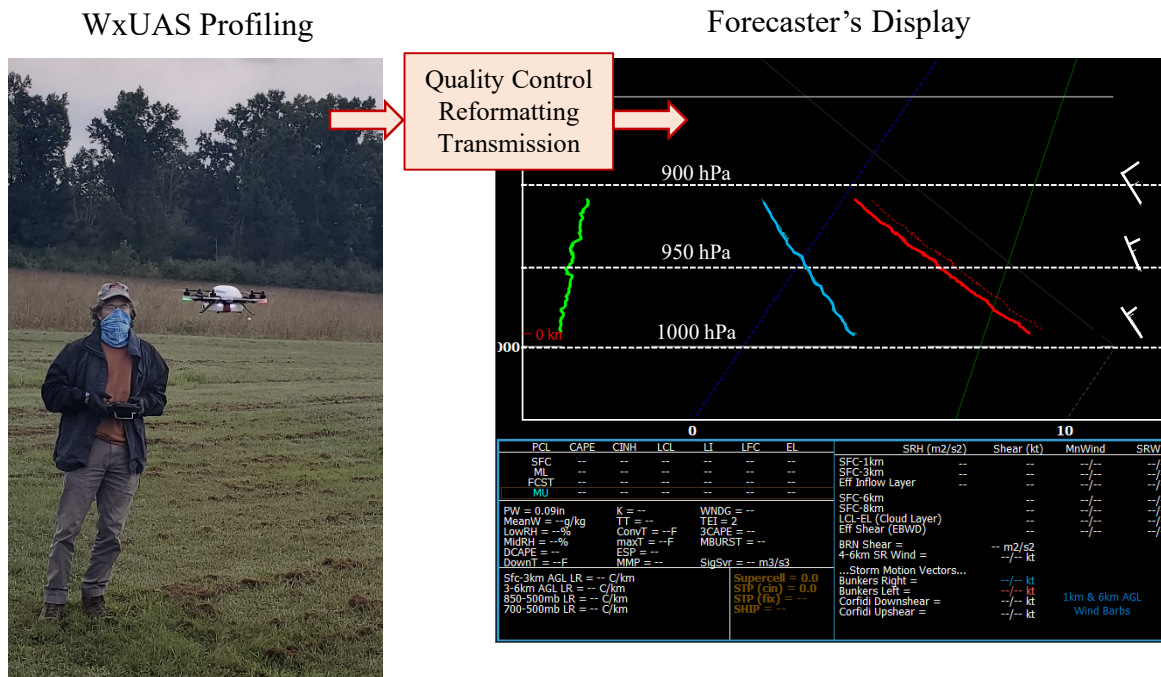


Figure 4. On left, Meteomatics WxUAS being flown during demonstration testbed to assess their use in operational meteorology. Data are relayed to a ground station, post-processed with quality control software and reformatted before transmission to the WFO in Morristown, TN in real-time. On right, WxUAS data can be displayed at the WFO using SHARPPy display tool. Observations shown are dew point temperature (green), wet bulb temperature (blue), air temperature (red), and wind speed and direction (barbs on right) are used to support the development of short-term forecasts.

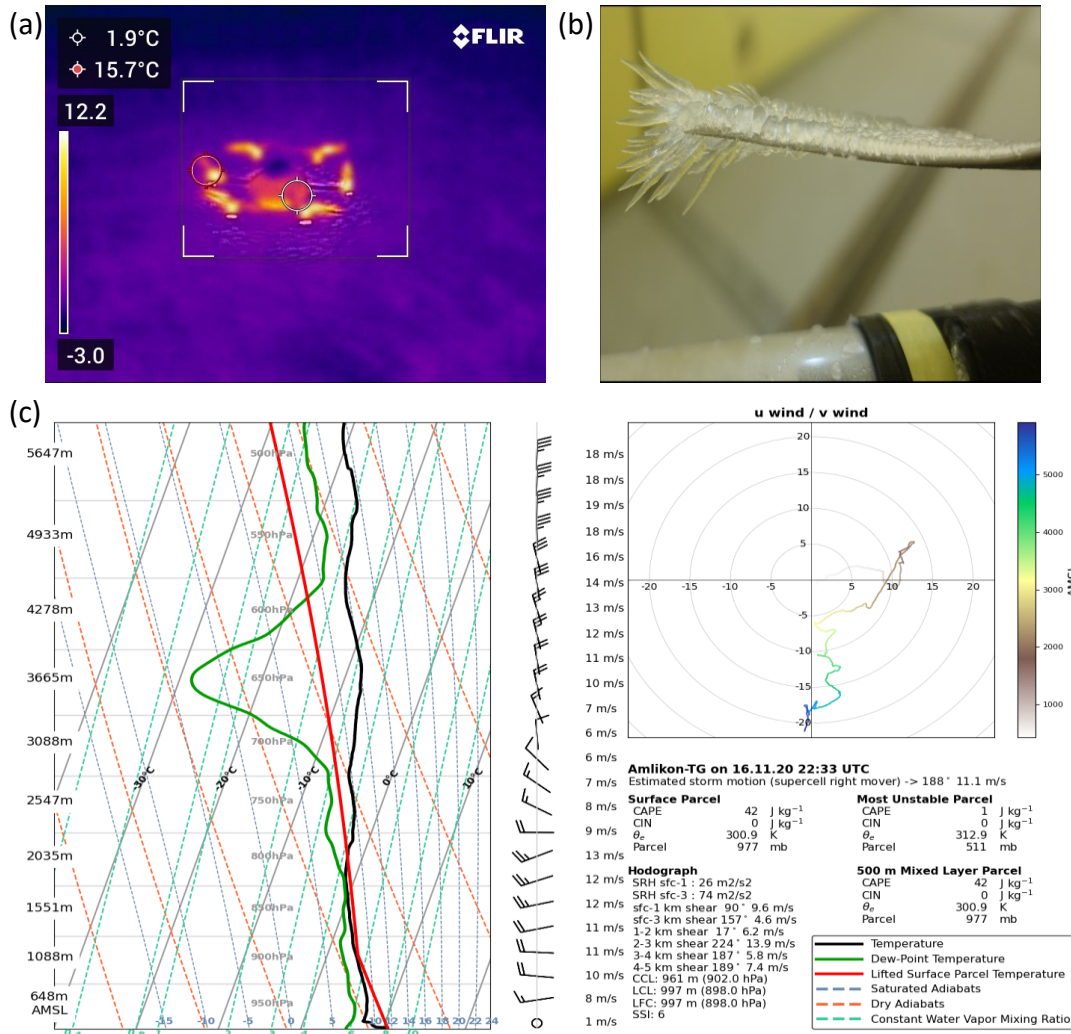
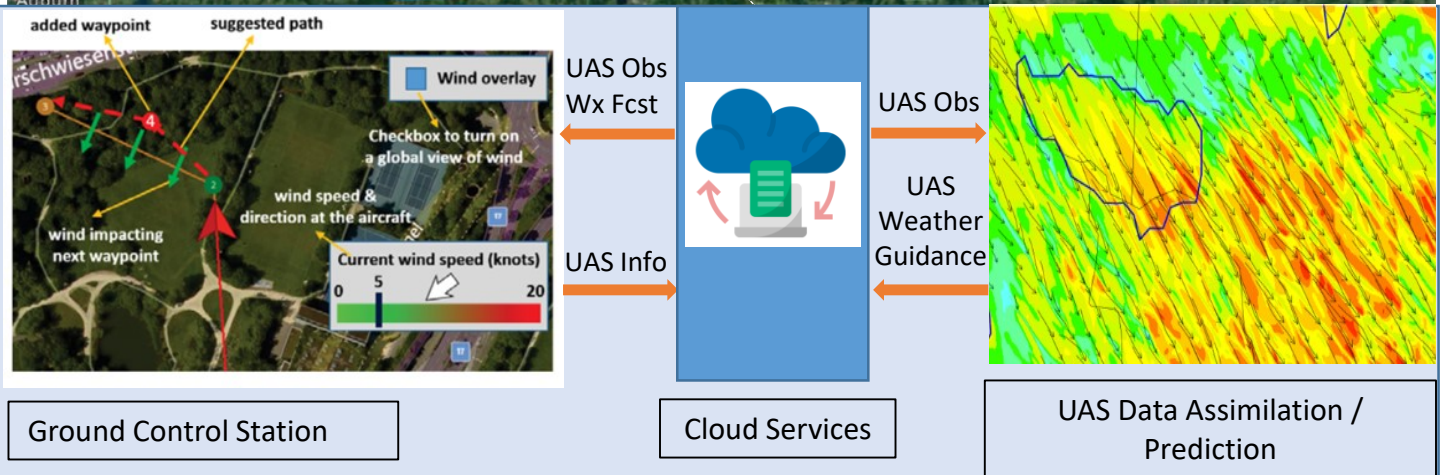
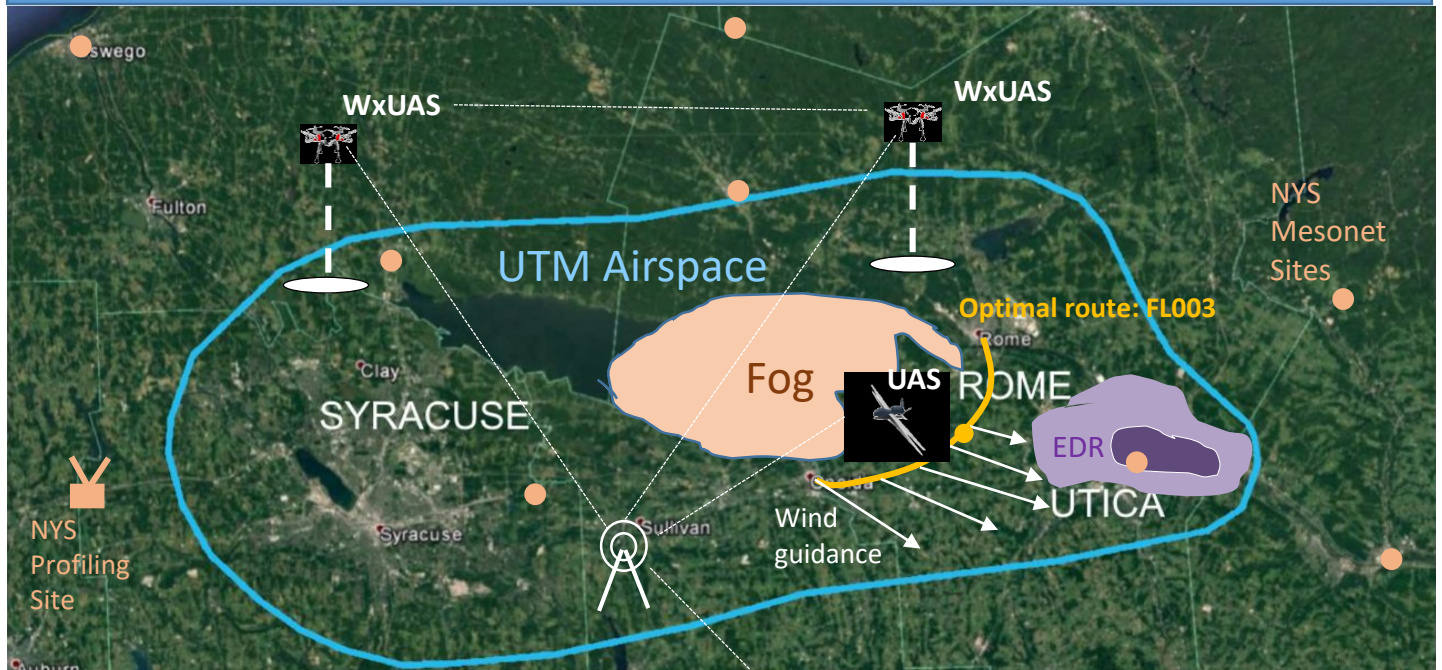


Figure 5. (a) Thermal imagery of heated Metedrone flying within icing conditions. (b) Picture showing horning icing on the rotary blade of a Metedrone. (c) Observations of temperature, dew point and winds collected with Metedrone WxUAS during flight into icing conditions including (left) Skew-T diagram and (right) hodograph and convective parameters. The freezing level is at 1600 m MSL.

WEATHER-AWARE UAS TRAFFIC MANAGEMENT SYSTEM



Sidebar SB1. Example of a testbed where WxUAS and commercial UAS combine to collect weather observations along with an existing sensor network centered over Upstate New York. The UAS communicate observations with each other when within line of sight. Weather information can be transmitted to cell towers and processed / stored in the cloud for operators and other subscribers including modeling centers to access. Modeling centers process data to improve initial conditions needed to drive weather prediction models whose outputs are used as guidance for UAS flight planning. Other applications process model data to determine likelihood of conditions that are hazardous to commercial UAS operations such as low ceilings and reduced visibility caused by fog, icing layers and areas intense turbulence or wind shift boundaries. Wind information is used directly to optimize route planning and to estimate departure and arrival times, optimize commercial UAS fleet mix, set up metering/spacing between commercial UAS, etc.