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COVID-19 Pandemic Imperils Weather Forecast

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5

6 **Key points:**

7 1. The COVID-19 pandemic eliminated 50-75% of aircraft meteorological
8 observations.

9 2. Accuracy of weather forecast reduced significantly especially over southeast
10 China and US, and error develops as forecast goes longer.

11 3. This could handicap early warning of extreme weather, establishing more
12 meteorology stations can buffer the impact of global emergencies.

13

14 **Keywords:** COVID-19 pandemic, weather forecast, aircraft, assimilation, accuracy

15

16 **Abstract:**

17 Weather forecasts play essential parts in economic activity. Assimilation of
18 meteorological observations from aircraft improves forecasts greatly. However, global
19 lockdown during the COVID-19 pandemic (March-May 2020) has eliminated 50-75%
20 aircraft observations and imperils weather forecasting. Here, we verify global
21 forecasts against reanalysis to quantify the impact of the pandemic. We find a large
22 deterioration in forecasts of surface meteorology over regions with busy air flights,
23 such as North America, southeast China and Australia. Forecasts over remote regions
24 are also substantially worse during March-May 2020 than 2017-2019, and the
25 deterioration increases for longer-term forecasts. This could handicap early warning
26 of extreme weather and cause additional economic damage on the top of that from the
27 pandemic. The impact over Western Europe is buffered by the high density of
28 conventional observations, suggesting that introduction of new observations in
29 data-sparse regions would be needed to minimize the impact of global emergencies on
30 weather forecasts.

31

32 **Plain Language Summary:**

33 Weather forecasts play essential parts in daily life, agriculture and industrial
34 activities, and have great economic value. Meteorological observations on
35 commercial aircraft help improve the forecast. However, the global lockdown during
36 the COVID-19 pandemic (March-May 2020) chops off 50-75% of aircraft
37 observations. Here, we verify global weather forecasts (1-8 days ahead) against the

38 best estimates of atmospheric state, and quantify the impact of the pandemic on
39 forecast accuracy. We find large impacts over remote (e.g., Greenland, Siberia,
40 Antarctica and the Sahara Desert) and busy air-flight regions (e.g., North America,
41 southeast China and Australia). We see deterioration in the forecasts of surface
42 meteorology and atmospheric stratification, and larger deterioration in longer-term
43 forecasts. This could handicap early warning of extreme weather and cause additional
44 economic damage on the top of that from the pandemic itself. The impacts over
45 Western Europe are small due to the high density of conventional observations,
46 suggesting that introduction of new observations would be needed to minimize the
47 impact of global emergencies on weather forecasts in future.

48

49 **1. Introduction**

50 Weather forecasts play an essential part in daily life [*Böcker et al.*, 2013],
51 agriculture [*Calanca et al.*, 2010] and industrial activities [*Teisberg et al.*, 2005], and
52 have great economic value [*Zhu et al.*, 2002]. The accuracy of forecasts is largely
53 dependent on the quality of initial conditions used in numerical weather prediction
54 models. The number of meteorological observations has increased steadily over the
55 past decades globally, and their assimilation has greatly improved model initial
56 conditions and forecasts [*Kanamitsu*, 1989]. Aircraft observations from commercial
57 airlines around the world are a critical component of global meteorological
58 observations. Assimilation of aircraft observations exerts the largest improvements in
59 global weather forecasts compared with each individual category of conventional

60 observations (exclude satellite), both for long-term average and for individual events
61 [*Ota et al.*, 2013; *Petersen*, 2016].

62 However, availability of these critical aircraft observations has reduced
63 remarkably since March 2020, resulting from the global lockdown in response to the
64 COVID-19 pandemic. According to the International Civil Aviation Organization, by
65 the end of March 2020, more than 20 commercial airlines have stopped flights
66 entirely and about 12 airlines stopped all international flights. This eliminates about
67 50-75% of aircraft observations globally during March-May 2020, according to the
68 World Meteorological Organization (WMO, [*WMO*, 2020]), the European Centre for
69 Medium-Range Weather Forecasts (ECMWF, [*ECMWF*, 2020]) and the Aircraft
70 Meteorological DATA Relay programme (<https://amdar.noaa.gov>). Lack of critical
71 aircraft observations could imperil the weather forecast. WMO, ECMWF and
72 scientists expressed concerns over the impacts to the public regarding the possible of
73 unreliable weather forecasts [*ECMWF*, 2020; *Viglione*, 2020; *WMO*, 2020]. The lack
74 of aircraft data may become worse as the COVID-19 pandemic develops further and
75 the associated lockdown extends, and this will lead to larger impacts on weather
76 forecasting and impose an additional economic cost on the top of that from the
77 pandemic itself. Therefore, a quantifying understanding of the potential impacts of the
78 pandemic on weather forecasting and development of mitigation approaches are
79 critical for protecting current living standards and economic activity.

80 In this study, we quantify the impact of the COVID-19 pandemic on weather
81 forecasts by verifying global weather forecast against reanalysis data, which is the

82 best available estimate of the atmospheric state. We also present the difference in
83 impacts over different regions. Based on scientific evidences, we provide suggestions
84 to minimize the impact of global emergencies, such as the COVID-19 pandemic, on
85 weather forecasting in future.

86

87 **2. Materials and Methods**

88 **2.1 COVID-19 pandemic impacts on aircraft meteorological observations**

89 Coronavirus disease 2019 (COVID-19) broke out globally during February 2020
90 and became a global pandemic in March 2020 [*WHO*, 2020]. Since March, lockdown
91 has been enforced by countries across the world to control the spread and save lives.
92 For example, Italy announced lockdown on March 9th, followed by Spain on 14th,
93 France on 17th, Germany on 22nd, UK on March 23rd, US banned travel from Europe
94 since March 14th and Australia banned all international visitors since March 19th.
95 These restricting measures have produced a remarkably reduction in meteorological
96 observations from commercial airlines since March 2020 [*WMO*, 2020].

97 Aircraft Meteorological DATA Relay programme (AMDAR), initiated by WMO,
98 includes more than 3500 aircraft from ~40 commercial airlines globally, and provides
99 more than 680,000 temperature and wind reports per day [*Petersen*, 2016]. About 100
100 aircraft, mainly over the United States, can also provide moisture observations
101 [*Petersen*, 2016]. These observations are reported every few minutes when aircraft are
102 at cruise levels and every few seconds for profiles during take-off or landing. As part

103 of WMO protocols [*Moninger et al.*, 2003], these AMDAR observations are quality
104 controlled by National Centers for Environmental Prediction (NCEP). The AMDAR
105 dataset has the highest density over North America and Europe, where the airspace is
106 busiest, but these regions are also the centres of the COVID-19 pandemic in
107 March-May 2020. Due to lockdown during the pandemic, the total number of
108 meteorological profile aircraft-reports reduced by more than 50% from about 102,000
109 per week in February 2020 to about 52,000 in the last week of March 2020, and
110 further reduced to about 26,000 in the last week of April and May 2020 (data source:
111 <https://amdar.noaa.gov>).

112 **2.2 Weather forecast dataset and observation-based datasets**

113 In order to investigate the impact of the reduction in aircraft observations on
114 global weather forecasts in March-May 2020, the NCEP Global Forecast System
115 (GFS) dataset (ds084.1, [*NCEP*, 2015a]) is verified against the high resolution NCEP
116 Global Data Assimilation System (GDAS) reanalysis dataset (ds083.3, [*NCEP*,
117 2015b]) and Global Precipitation Climatology Centre monthly precipitation dataset
118 (GPCP, [*Ziese et al.*, 2011]). The GFS model couples atmosphere, ocean, land/soil
119 and sea ice modules to produce the weather forecast. There are 64 hybrid
120 sigma-pressure layers in the atmospheric module, from the ground surface to about
121 0.27 hPa [*Sela*, 2009]. More details of GFS model are given in the website of
122 NCEP-GFS [*NCEP*, 2019]. The GDAS data is our best estimate of the atmospheric
123 state; it assimilates the greatest number of meteorological observations from global
124 sources, including aircraft, radiosonde, satellite-based and ground-based observations,

125 details given in *NCEP* [2018]. GPCC data provides monthly total precipitation over
126 land-surface on a global scale at a resolution of $1.0^\circ \times 1.0^\circ$ (latitude \times longitude),
127 based on the surface synoptic observations from WMO [*Ziese et al.*, 2011]. GDAS
128 and GFS datasets with a horizontal resolution of 0.25 degree are adopted, and forecast
129 results up to 192 hours are analysed in this study.

130 We demonstrate the reduction of forecast accuracy in temperature, relative
131 humidity (RH), wind speed and pressure with a special focus on temperature, because
132 temperature is widely observed by commercial airlines with high quality and
133 assimilated in the GDAS [*Petersen*, 2016]. In this study, we mainly focus on the
134 surface layer and at 00:00 UTC. These GDAS reanalyses are believed to be the
135 highest quality ones, since most surface meteorological observations are still working
136 properly during the COVID-19 pandemic, and the largest availability of radiosonde
137 observations is at 00:00 UTC around the world [*Ingleby et al.*, 2016]. We also discuss
138 the impacts of elimination in aircraft observations during the pandemic on
139 precipitation forecasts by validating GFS forecasts against the observation-based
140 GPCC dataset.

141 To investigate the impact of the COVID-19 pandemic on the weather forecast,
142 we compare the forecast accuracy for March-May 2020 (during global lockdown)
143 against the average of March-May 2017-2019. In addition, we conduct the same
144 analysis for February 2020 before global lockdown, in order to demonstrate that this
145 impact on accuracy is associated with the pandemic in March-May 2020 rather than
146 the meteorological characteristics of 2020.

147

148 **3. Results and discussion**

149 **3.1 COVID-19 pandemic reduces accuracy of weather forecast**

150 As shown in Fig. 1, the accuracy (absolute error) of surface meteorology forecast
151 in March-May 2020 decreases remarkably (with respect to March-May 2017-2019;
152 red colours indicate worse forecasts, blue colours indicate better forecasts) over north
153 and south polar regions (latitude > 70 degree), throughout the 1-8 day forecasts.
154 Temperature forecast in March-May 2020 shows an extra 0.5-1.0 °C bias compared
155 with that in March-May 2017-2019 over south polar regions. The deterioration in
156 temperature forecasts over north polar regions is less than south polar regions, by an
157 extra 0-0.5 °C bias; however, before the global lockdown in February, the temperature
158 forecast over north polar regions is generally improved by 0.5-1.5 °C in 2020 against
159 2017-2019, with small exceptions in the 24-48 hour forecast (Fig. 2a). The surface
160 RH, pressure and wind speed forecasts in March-May 2020 are also remarkably worse
161 than the forecast in February 2020 (Fig.1b-1d and Fig. 2b-2d). The deterioration of
162 the temperature forecast in March-May 2020 develops in the upper layers as the
163 forecast is extended, with large deterioration (~ 1.0 °C) over polar regions from ground
164 up to ~ 300 hPa in the 168-hour forecast (details in Fig. S1). This could lead to larger
165 uncertainties in longer forecasts in the descriptions of atmospheric stratification and
166 synoptic scale weather systems, with impacts on medium-range (3-10 days ahead) to
167 long-range (15-30 days ahead) forecasts. And the deterioration in global model

168 accuracy could worsen predictions for mesoscale and microscale systems using
169 high-resolution models, whose boundary conditions are constrained by global
170 produces.

171 No notable deterioration in the surface pressure and wind speed forecasts in
172 March-May 2020 is observed in 24-96 hours forecasts (Fig. 1c and 1d), but there is a
173 slight improvement in February 2020 (Fig. 2c and 2d). However, the errors develop as
174 the forecasts are extended. In northern polar regions, the 96-192 hour forecasts of
175 surface pressure are worsened by 1-3 hPa in March-May 2020, even though an
176 improvement of 1-4 hPa is seen in the February results (Fig. 2c). Similar for wind
177 speed forecast, error in March-May forecasts develops as the forecasts are extended
178 and the accuracy is worsened by up to 0.8 m/s in north polar regions when forecast is
179 more than 100 hours ahead. However, wind forecast of February 2020 shows an
180 improvement by 0.2-0.5 m/s against February 2017-2019, throughout 24-192 hours
181 forecasting period (Fig. 2d). Very limited diurnal variation in the deteriorations is
182 observed (Fig. S2), indicating these deteriorations in the forecasts of surface
183 meteorology are consistent throughout a day.

184 The total precipitation forecasts during March-May 2020 are validated against
185 the observation-based GPCP dataset, and compared the accuracy in March-May 2020
186 with March-May in 2017-2019 (Fig. S3). No significant deterioration in precipitation
187 forecasts during March-May 2020 is observed, compared with 2017-2019. Although
188 there is some deterioration in a small area of southeast China, the deterioration in
189 precipitation forecasts does not consistently present over a large scale of the regions

190 with busy air flights as the deterioration in temperature forecasts does (Fig. 3, detail
191 discussion in the next section). This is not surprising, since previous studies show that
192 aircraft observations play a critical role in the forecasts of temperature, humidity and
193 wind from troposphere to lower-stratosphere [James and Benjamin, 2017; Ota et al.,
194 2013; Petersen, 2016]; while, cloud properties from satellites are important for rainfall
195 forecasts [James and Benjamin, 2017], and are not eliminated during the global
196 lockdown.

197 In summary, better forecasts of surface meteorology are expected in 2020 as
198 indicated by February results, but significant worse forecasts are shown in
199 March-May 2020. This discrepancy strongly suggests that the COVID-19 pandemic
200 imperils weather forecasting of surface temperature, RH, pressure and wind speed due
201 to the lack of aircraft observations during the global lockdown. However,
202 precipitation forecasts are not remarkably affected.

203 **3.2 Impact in different regions**

204 We notice that the degradation of the weather forecast is more substantially in the
205 northern hemisphere than the southern hemisphere. This is because there is a much
206 larger number of aircraft observations in this region to constrain the initial conditions
207 of the forecast model. We notice a much larger degradation in the March-May
208 forecast over some regions than others, as shown by the 168-hour forecast as an
209 example in Fig. 3. Remote regions (magenta boxes), such as the Greenland, Siberia,
210 Antarctica and the Sahara Desert, are impacted greatly. This is because assimilation of
211 aircraft observations provides a much larger improvement in forecasts over regions

212 where very limited conventional observations are available [*Ota et al.*, 2013]. Regions
213 with busy air flights are also affected greatly, such as North America, southeast China
214 and Australia (green boxes in Fig. 3). The accuracy of the surface temperature
215 forecasts over these regions is reduced or occasionally slightly improved (0-0.5 °C) in
216 March-May 2020 (Fig. 3a), but we could reasonably expect a larger improvement of
217 0.5-1.5 °C over these regions as seen in the February result (Fig. 3b). Therefore, this
218 gap of 0.5-1.5 °C improvement between forecasts in March-May and February
219 (calculated as “Fig. 3a – Fig. 3b”, shown in Fig. 3c) could be attributed to the lack of
220 aircraft observations during the COVID-19 pandemic. This is supported by the
221 reduced availability of aircraft observations as discussed in the Methods section,
222 where only about 25-50% of aircraft observations were available globally during
223 March-May 2020 compared with February 2020.

224 As reported in previous studies [*Ota et al.*, 2013; *Petersen*, 2016] (see also Fig.
225 S4), North America, southeast China and Australia are regions with a large number of
226 aircraft observations under normal conditions. Western Europe (blue box in Fig. 3)
227 also has a large amount of aircraft observations, which reduced greatly during the
228 COVID-19 pandemic with strict lockdown over most European countries. However,
229 nearly no impact on the surface temperature forecasts is observed. This is because
230 there is a dense network of meteorological stations over western Europe compared
231 with other regions, 1519 stations in the small blue box of Fig. 3 (sourced from WMO:
232 <https://oscar.wmo.int/>), providing a good constraint on the initial conditions of
233 forecast model and hence a reliable weather forecast. Additional aircraft observations

234 make limited improvement over regions where observation information is almost
235 “saturated” [Ota *et al.*, 2013], such as western Europe. Therefore, the high density of
236 conventional meteorological observations buffers the impact of the COVID-19
237 pandemic on weather forecasts over western Europe.

238

239 **4. Summary**

240 Weather forecasts play an essential part in daily life, agriculture and industrial
241 activities, and their accuracy is largely dependent on the amount of meteorological
242 observations assimilated in forecast models. The COVID-19 pandemic has led to a
243 global lockdown and greatly reduce the number of flights and the associated aircraft
244 observations during March-May 2020. In this study, we verify global weather
245 forecasts in March-May 2020 against high resolution global reanalysis dataset and an
246 observation-based global precipitation dataset, which are the best estimate of the
247 atmospheric state. To investigate the forecast deterioration during the pandemic, the
248 forecast accuracy during March-May 2020 is further compared with the average
249 accuracy during March-May 2017-2019. We report a significant deterioration in the
250 forecasts of surface temperature, RH, wind speed and pressure, but no significant
251 deterioration in precipitation forecast is observed. A similar analysis for February
252 2020 suggests that the forecast accuracy of surface meteorology could have been
253 expected to improve in 2020 compared with 2017-2019, if aircraft observations were
254 carried out as usual.

255 Forecasts over remote and busy air-flight regions are more vulnerable due to the
256 lack of aircraft observations. Over the Greenland and Siberia, the accuracy of surface
257 temperature forecasts could be reduced by up to 2 °C, and the deterioration in the
258 forecasts of surface wind speed and pressure develops as the forecasts are extended.
259 Forecasts over North America, southeast China and Australia are also greatly affected
260 by the COVID-19 pandemic, but the impact over western Europe is compensated to
261 some extent by the high density of meteorological observations stations available.

262 The lack of aircraft observations may become more severe as the COVID-19
263 pandemic develops and the associated lockdown extends. This study warns that
264 further worsening of weather forecasts may be expected and that the error could
265 become larger for longer-term forecasts. This could handicap early warning of
266 extreme weather and cause additional hardship for daily life in the near future. The
267 results also highlight that establishing more meteorological stations in
268 observation-sparse regions and report data to WMO can improve the weather forecast
269 and effectively buffer the impact of global emergencies, such as the COVID-19
270 pandemic, in future.

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275 **Author contributions**

276 YC conceived the study, performed the analysis, interpreted the results and wrote the
277 manuscript.

278 **Notes**

279 The author declare no competing financial interest.

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281 The GFS (ds084.1) and GDAS (ds083.3) global datasets are available from (National
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290 **References:**

- 291 Böcker, L., M. Dijst, and J. Prillwitz (2013), *Impact of Everyday Weather on Individual Daily Travel*
292 *Behaviours in Perspective: A Literature Review*, *Transport Reviews*, 33(1), 71-91.
- 293 Calanca, P., D. Bolius, A. P. Weigel, and M. A. Liniger (2010), *Application of long-range weather*
294 *forecasts to agricultural decision problems in Europe*, *The Journal of Agricultural Science*,
295 149(1), 15-22.
- 296 ECMWF (2020), *Drop in aircraft observations could have impact on weather forecasts*,
297 [https://www.ecmwf.int/en/about/media-centre/news/2020/drop-aircraft-observations-could-ha](https://www.ecmwf.int/en/about/media-centre/news/2020/drop-aircraft-observations-could-have-impact-weather-forecasts)
298 [ve-impact-weather-forecasts](https://www.ecmwf.int/en/about/media-centre/news/2020/drop-aircraft-observations-could-have-impact-weather-forecasts), (last access: 20 April 2020).
- 299 James, E. P., and S. G. Benjamin (2017), *Observation System Experiments with the Hourly Updating*
300 *Rapid Refresh Model Using GSI Hybrid Ensemble–Variational Data Assimilation*, *Monthly*
301 *Weather Review*, 145(8), 2897-2918.
- 302 Ingleby, B., M. Rodwell, and L. Isaksen (2016), *Global radiosonde network under pressure*,
303 [https://www.ecmwf.int/en/newsletter/149/meteorology/global-radiosonde-network-under-pres](https://www.ecmwf.int/en/newsletter/149/meteorology/global-radiosonde-network-under-pressure)
304 [sure](https://www.ecmwf.int/en/newsletter/149/meteorology/global-radiosonde-network-under-pressure), *Number 149 - Autumn 2016 (ECMWF Newsletter)*.
- 305 Kanamitsu, M. (1989), *Description of the NMC Global Data Assimilation and Forecast System*,
306 *Weather and Forecasting*, 4(3), 335-342.
- 307 Moninger, W. R., R. D. Mamrosh, and P. M. Pauley (2003), *Automated Meteorological Reports from*
308 *Commercial Aircraft*, *Bulletin of the American Meteorological Society*, 84(2), 203-216.
- 309 NCEP (2015a), *NCEP GFS 0.25 Degree Global Forecast Grids Historical Archive*, edited, *Research*
310 *Data Archive at the National Center for Atmospheric Research, Computational and Information*
311 *Systems Laboratory, Boulder, CO*.
- 312 NCEP (2015b), *NCEP GDAS/FNL 0.25 Degree Global Tropospheric Analyses and Forecast Grids*,
313 *edited, Research Data Archive at the National Center for Atmospheric Research, Computational*
314 *and Information Systems Laboratory, Boulder, CO*.
- 315 NCEP (2018), *The Development and Success of NCEP's Global Forecast System*.
- 316 NCEP (2019), *The Global Forecast System (GFS) - Global Spectral Model (GSM)*, (last access: 20
317 April 2020).
- 318 Ota, Y., J. C. Derber, E. Kalnay, and T. Miyoshi (2013), *Ensemble-based observation impact*
319 *estimates using the NCEP GFS*, *Tellus A: Dynamic Meteorology and Oceanography*, 65(1),
320 20038.
- 321 Petersen, R. A. (2016), *On the Impact and Benefits of AMDAR Observations in Operational*
322 *Forecasting—Part I: A Review of the Impact of Automated Aircraft Wind and Temperature*
323 *Reports*, *Bulletin of the American Meteorological Society*, 97(4), 585-602.
- 324 Sela, J. (2009), *The implementation of the sigma-pressure hybrid coordinate into the GFS.*, *NCEP*
325 *Office Note #461*, pp25.
- 326 Teisberg, T. J., R. F. Weiher, and A. Khotanzad (2005), *The Economic Value of Temperature*
327 *Forecasts in Electricity Generation*, *Bulletin of the American Meteorological Society*, 86(12),
328 1765-1772.
- 329 Viglione, G. (2020), *How COVID-19 could ruin weather forecasts and climate records*, *Nature*.
- 330 WHO (2020), *Novel Coronavirus-2019*, (last access: 20 April 2020).
- 331 WMO (2020), *COVID-19 Impacts on Global Observing System*,
332 <https://public.wmo.int/en/resources/meteoworld/covid-19-impacts-global-observing-system>, (last
333 [access: 20 April 2020](https://public.wmo.int/en/resources/meteoworld/covid-19-impacts-global-observing-system)).

334 *Zhu, Y., Z. Toth, R. Wobus, D. Richardson, and K. Mylne (2002), THE ECONOMIC VALUE OF*
335 *ENSEMBLE-BASED WEATHER FORECASTS, Bulletin of the American Meteorological Society,*
336 *83(1), 73-84.*

337 *Ziese, M., A. Becker, P. Finger, A. Meyer-Christoffer, B. Rudolf, and U. Schneider (2011), GPCP*
338 *First Guess Product at 1.0°: Near Real-Time First Guess monthly Land-Surface Precipitation*
339 *from Rain-Gauges based on SYNOP Data, doi: 10.5676/DWD_GPCP/FG_M_100.*

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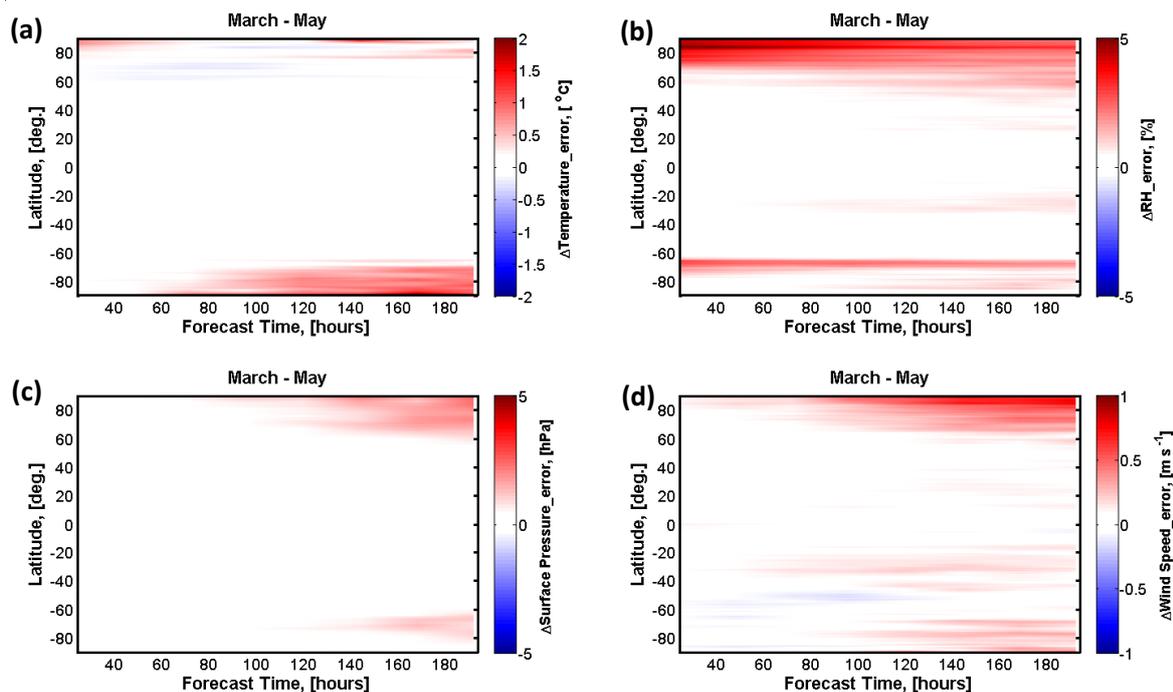


Figure 1. Deviation in absolute error of weather forecasts between 2020 and the average of 2017-2019. Forecasts of 24-192 hour (1-8 day) ahead in the period of March to May, all variables are at 00:00 UTC and in surface layer: (a) temperature; (b) RH; (c) pressure; (d) wind speed. Only deviations with significance higher than 95% confidence level according to t-test are shown. Red colours indicate worse forecasts in 2020, blue colours indicate better forecasts in 2020.

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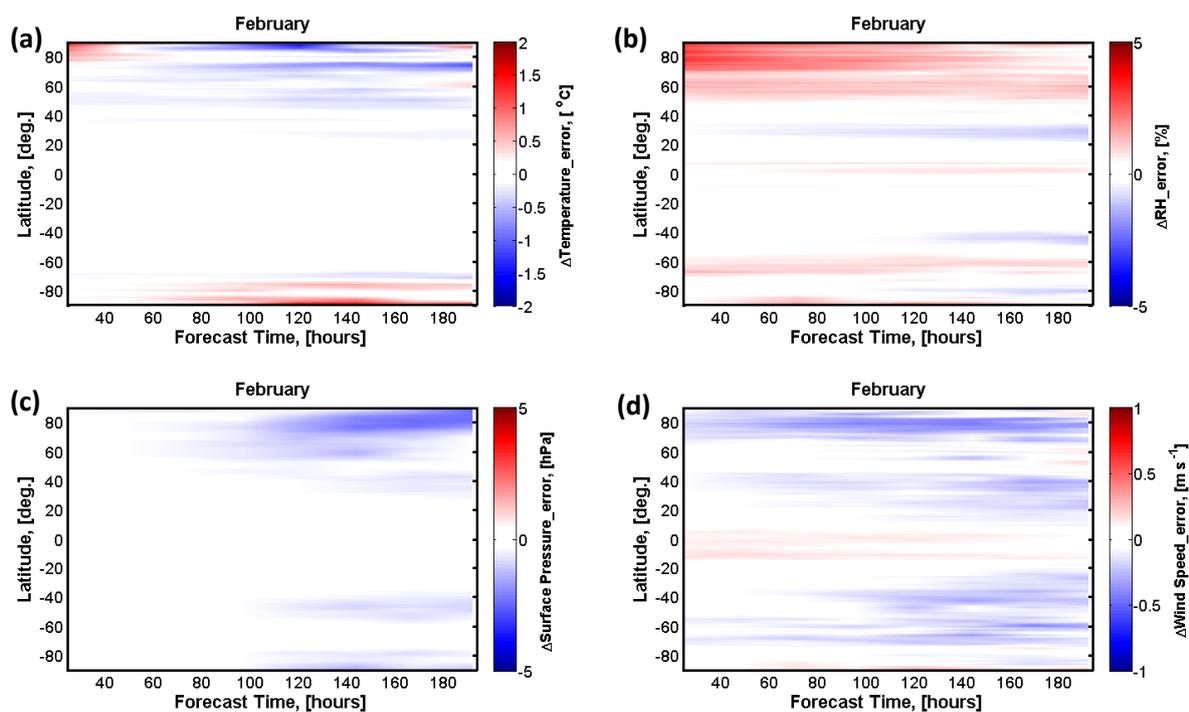


Figure 2. Similar as Fig. 1, but forecasts for February. Red colours indicate worse forecasts in 2020, blue colours indicate better forecasts in 2020.

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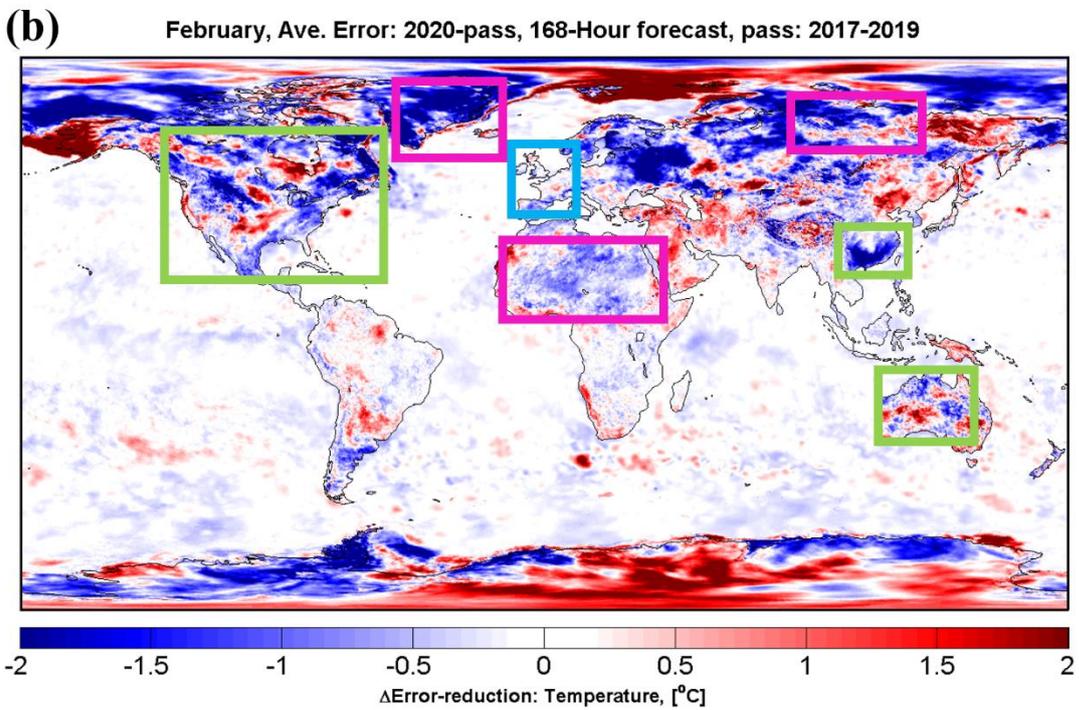
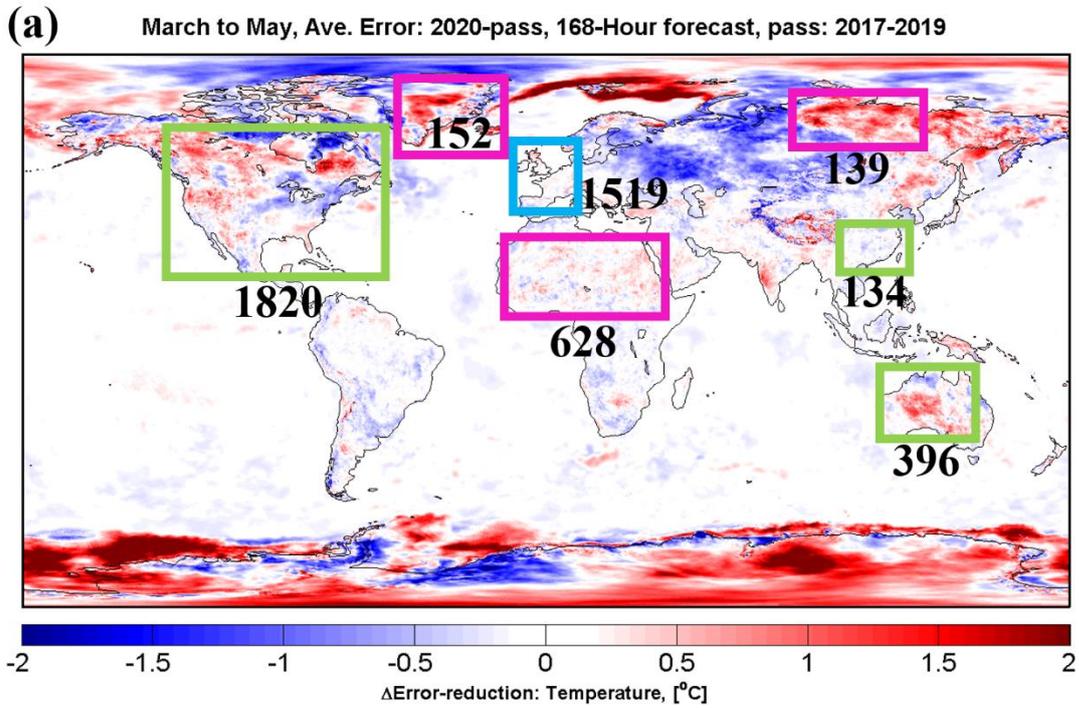


Fig. 3 continues in next page.

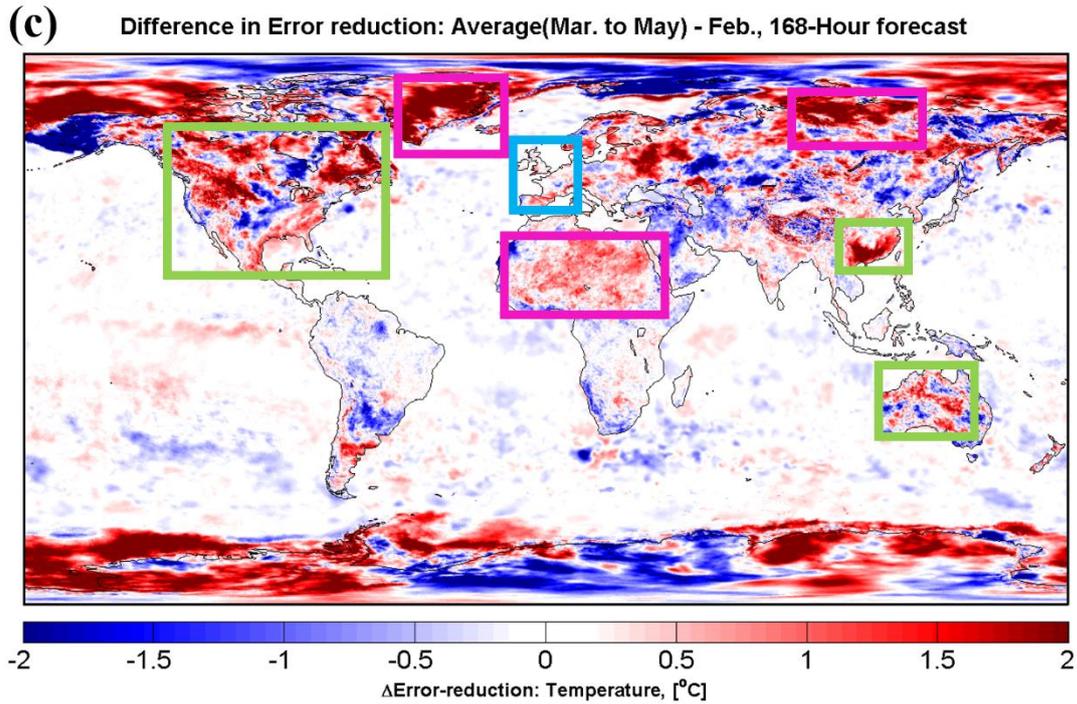


Figure 3. Global map of deviation in absolute error of surface temperature forecasts between 2020 and the average of 2017-2019. The results are for March to May (a) and February (b). The 168-hour forecasts at 00:00 UTC in surface layer are shown. Only deviations with significance higher than 95% confidence level according to t-test are shown. The number of meteorological stations in different regions (boxes) are also marked, data sourced from WMO (<https://oscar.wmo.int/>). Green boxes indicate the regions with busy air flights and large degradation in forecasts, light blue box indicates the region with busy air flights and moderate degradation in forecast, and magenta boxes indicate the remote regions with large degradation in forecasts. The difference between Fig. 3a and Fig. 3b is shown in Fig. 3c. In all panels, red colours indicate worse forecasts, blue colours indicate better forecasts.

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