



EMADDC: high-volume, high-quality, and timely wind and temperature observations from aircraft surveillance data (Mode-S EHS)

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Abstract. Wind and temperature observations from aircraft are of major importance for aviation meteorology and numerical weather prediction (NWP). The European Meteorological Aircraft Derived Data Centre (EMADDC) system processes aircraft surveillance data received from air traffic control (ATC) and other partners and converts them into upper-air observations of wind and temperature. Only so-called Mode-S Enhanced Surveillance (Mode-S EHS) data can be used because these data contain the air vector and ground vector of the aircraft from which a wind vector can be inferred. Temperature is derived from true airspeed and Mach number measurements. To produce high-quality observations, the data are processed in three steps: pre-processing, processing, and post-processing. The pre-processing is needed to obtain high-quality information and to calculate several correction values for correcting temperature observations and heading values. Processing converts the aircraft data into meteorological information, and finally post-processing guarantees that only high-quality information is made available.

The EMADDC system processes around 75×10^6 surveillance observations per day and produces over 55×10^6 observations of quality-controlled wind observations and 32×10^6 temperature observations in the European airspace per day. The average age of the observation is around 5 to 10 min, depending on the method of data delivery (files via ftp or streaming constantly).

The quality of the observations produced is verified by comparing these observation to other upper-air wind and

temperature observations from radiosondes and Aircraft Meteorological Data Relay (AMDAR) and comparing them with NWP data. The quality of wind observations is almost identical to AMDAR, and the quality of the temperature of EMADDC observations is lower but with a bias of around zero, while AMDAR exhibits a positive bias of 0.5 K.

This paper presents the EMADDC (R2.2) system, operational since 2019.

1 Introduction

For normal, and safe, operation, aircraft are equipped with sensors to measure height and velocity with respect to the surrounding air. These sensors can be exploited to observe wind and temperature at the aircraft's location (WMO, 2023). For many years, aircraft observations have been an essential component of the global observing system, which is used as input for numerical weather prediction models during assimilation (Cardinali et al., 2003; James et al., 2020; Li, 2021; Strajnar et al., 2015; de Haan, 2013). For almost 30 years, aircraft measurements have been collected using the Aircraft Meteorological Data Relay (AMDAR), where meteorological information is automatically sent to national weather services using either satellites or ground stations (Ingleby et al., 2020; Barwell and Lorenc, 1985; Cardinali et al., 2003; James et al., 2020; Lange and Janjić, 2016; Li, 2021; Petersen, 2016; Zhu et al., 2015; Benjamin et al., 2010). Some commercial aircraft are equipped with software to collect

the relevant information from the onboard computer systems. Observations are collected with specified observation strategies to optimize coverage with respect to data transmission costs. Over the last decade, a different manner of collecting meteorological information has been developed, utilizing the operational infrastructure for aircraft safety in Europe, starting in the area of Germany, Belgium, and the Netherlands. The infrastructure used by the European air traffic control (ATC) is based on mode-selective (Mode-S) radars which (selectively) interrogate all aircraft in view of the radar on information on the heading, true airspeed, etc., to guide aircraft through its airspace (de Haan, 2011). Although in the whole European airspace Mode-S radars are used for ATC, not all received information can be used to refer to meteorological information; only Mode-S Enhanced Surveillance (Mode-S EHS) radars can interrogate the necessary Broadcast Dependent Surveillance (BDS) 5.0 and 6.0 registers. Fortunately, most Mode-S radars in Europe have EHS capabilities. The observation frequency is determined by the interrogation frequency of the Mode-S radar. Since the information is aimed at ATC, unfortunately no direct meteorological parameters are present in the received information. To extract meteorological information from the received BDS5.0 and BDS6.0 registers, processing and corrections are needed (de Haan, 2011; Stone and Pearce, 2016). Obviously, receiving meteorological parameters directly is preferred to deriving and correcting non-meteorological parameters, since corrections are inherently imperfect, which may lead to complicated errors. The unique character of these Mode-S EHS observations is that (almost) all aircraft are interrogated every 4 to 20 s, which results in (locally) very dense datasets.

In light of the COVID-19 pandemic, there was a significant reduction in the number of flights, which in turn impacted the availability of temperature and wind observations collected through the Aircraft Meteorological Data Relay (AMDAR). Nevertheless, as certain airlines continued to operate (e.g. as cargo flights) the European Meteorological Aircraft Derived Data Centre (EMADDC) was still producing valuable observations, exploiting the ATC information received for surveillance of all flying aircraft. These observations were used by ECMWF (Ingleby et al., 2021) to address the gap resulting from the lack of AMDAR observations.

This paper describes the current state-of-the-art processing and correction methodology (R2.2) as implemented at EMADDC.

2 EMADDC data collection

Secondary surveillance radar (SSR) is a two-way system where an ATC radar interrogates an aircraft requesting specific parameters. In Europe, all large aircraft (with minimum take-off weight larger than 5700 kg) are required to broadcast Mode-S Elementary Surveillance (ELS) and Enhanced Surveillance (EHS) (European Commission, 2011).

EMADDC exploits these to derive wind speed, wind direction, and temperature observations from surveillance data requested from aircraft for ATC purposes. Where elementary surveillance uses only aircraft broadcasts of altitude and identity, the enhanced surveillance interrogation complements these basic parameters with data of the aircraft state, such as roll angle, airspeed, and Mach number. These additional parameters are requested in groups as a BDS request. To derive wind and temperature, EMADDC requires both BDS5.0 and BDS6.0 to be interrogated and the aircraft to respond.

Additional to these mandatory BDS registers (BDS5.0 and BDS6.0), the BDS4.4 register, known as the Meteorological Routine Air Report or MRAR, can be also interrogated. This register contains observed temperature, wind, static pressure, and humidity (where available). However, this register is not mandatory, and only fewer than 5 % of aircraft respond to such interrogation requests (Strajnar, 2012), and few countries actively interrogate this register.

2.1 Mode-S EHS interrogation

ATC radar initiates a request to an aircraft for specific BDS data registers. If the aircraft is appropriately equipped, it will respond by broadcasting the requested register information. Each radar employs a distinct interrogation scheme that outlines the frequency of requests and the specific registers to be queried. Different countries or an Air Traffic Service Air Navigation Service Provider (ATS ANSP) may utilize varying interrogation protocols, including different frequencies and rates. The response sent by the aircraft is received by ATC radar but can also be received through a commercially available local receiver, as data are not encrypted. Unfortunately, it is more difficult to decode data received by these receivers as the type of register is not contained in the transmission, and hence fuzzy logic or other techniques are applied to decode the type transmitted properly (de Haan, 2011; Stone and Pearce, 2016). C-code software, developed in-house, performs this task (similar to the Python library developed by Sun, 2021).

2.2 Automatic Dependent Surveillance–Broadcast

Automatic Dependent Surveillance–Broadcast, or ADS-B, as its name suggests, allows an aircraft to broadcast aircraft state data using the transponder. Data are autonomously broadcast about every 0.5 s and contain the aircraft's onboard sensed position (through GPS and inertial systems), which is often more accurate than radar-derived position from ATC. These data are available to ATC but can also be displayed on the Navigation Display of newer aircraft for situational awareness. ADS-B does not broadcast wind and temperature, nor does it broadcast all required parameters to derive wind and temperature, although the difference between GNSS height and pressure altitude is transmitted frequently

and could be used in data assimilation (Stone and Kitchen, 2015).

2.3 Data handling

As outlined in the previous sections, EMADDC receives data in two ways: (1) aircraft data are collected by ATC, or (2) aircraft data are collected using a local ADS-B/Mode-S receiver. In both cases data are then forwarded to EMADDC through an FTP file transfer. New methods are currently being developed to enhance real-time data transfer, including NewPENS (the pan-European Network Service for real-time exchange of air traffic control data).

Data collected by ATC are of high quality as the contents are properly decoded since the content of each transmission is known to the interrogator. However for local receivers, the content of a received register is unknown, and logic is required to verify that a register is correctly decoded. All data have quality control and filtering applied.

ATC data can be of ASTERIX CAT48 format, which is mono-radar data, or CAT62 data from a radar tracker combining multiple radars. This latter data format uses filtering to sample all radar plots to a typical 4 s interval. The content of the formats CAT48 and CAT62 is similar, but the typical resolution of the Mach number in CAT62 is 0.008 (0.004 for CAT48), and as a consequence the derived temperatures are of lower quality (see Sect. 6). EMADDC is working with EUROCONTROL Maastricht Upper Area Control Centre (MUAC) to develop a solution that provides the Mach number with a resolution of 0.004 and share this solution with other ATC providers.

An advantage of ATC radar data is that the individual BDS register messages corresponding to a single revolution of the radar are combined into a single “observation” message. For ATC radar, the position is determined by ATC radar. In contrast, for data received from local receivers, the position is decoded from the Compact Position Report (CPR) format, which is included in the ADS-B message (and hence not present in radar/tracker data as these data are not available in ASTERIX CAT48 or CAT62 data). The timestamp is not generated by the aircraft; instead, it is created by the radar or receiver at the time of reception. Local receivers utilize either a GPS antenna or a time server for time synchronization. Furthermore, the EMADDC system must combine the various BDS registers to derive observations, as previously outlined.

The techniques utilized by EMADDC are generating significant volumes of high-quality data from Mode-S EHS data. It is essential to implement various quality control that checks identify and address any observation discrepancies and assure that generated observations are of high quality.

3 Aircraft measurement methodology

A modern aircraft is equipped with sensors that can measure static pressure, Mach number, temperature, position and heading, and geometric altitude. This section contains a brief description of measurements of pressure, Mach number, and temperature. The information flow is depicted in Fig. 1.

3.1 Mach number and static and total pressure

A crucial measurement in any aircraft is the measurement of the airspeed, which can be obtained from the combination of a pitot-tube measurement and a temperature measurement. The pitot tube measures the static pressure p_s and the total pressure p_t (Ruijgrok, 1990). Both pressure observations suffer from inaccuracies related to, for example, a (small) angle between the flow and the probe (Rodi and Leon, 2012). The Mach number is the true airspeed of the aircraft relative to the speed of sound. Let $q_t = p_t - p_s$ be the dynamic pressure, which is more accurately measured because the first-order error of p_t and p_s is cancelled. The Mach number M is defined as (Ruijgrok, 1990)

$$M = \sqrt{\frac{2}{\gamma - 1} \left(1 + \frac{q_t}{p_s} \right)^{\frac{\gamma - 1}{\gamma}} - 1}, \quad (1)$$

where $\gamma = c_p/c_v$ is the ratio of specific heats. Note that the dependence of M on the (inaccurate) p_s remains.

3.2 Temperature and true airspeed

The temperature is measured with a temperature probe (Ruijgrok, 1990). The measured total temperature T_t needs to be corrected to obtain the temperature T ,

$$T = T_t \left(1 + \lambda \frac{(\gamma - 1)}{2} M^2 \right)^{-1}. \quad (2)$$

The true airspeed A can now be determined neglecting the effect of humidity,

$$A = M \sqrt{\gamma R_d T}, \quad (3)$$

where R_d is the universal gas constant of dry air.

4 EMADDC measurement methodology

Temperature and wind information is not directly available in Mode-S EHS downlinked information. The wind vector needs to be computed, and the temperature needs to be derived; see Fig. 2.

4.1 Downlinked parameters

The (most relevant) parameters obtained through interrogation of Mode-S EHS radars are shown in Table 1. The timestamp is created at the moment of reception of the information. All parameters that originate from interrogation have

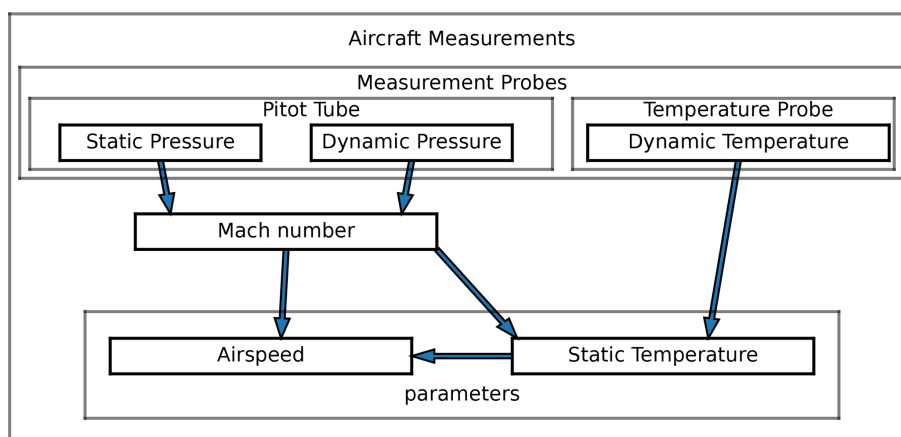


Figure 1. Information flow of aircraft measurements.

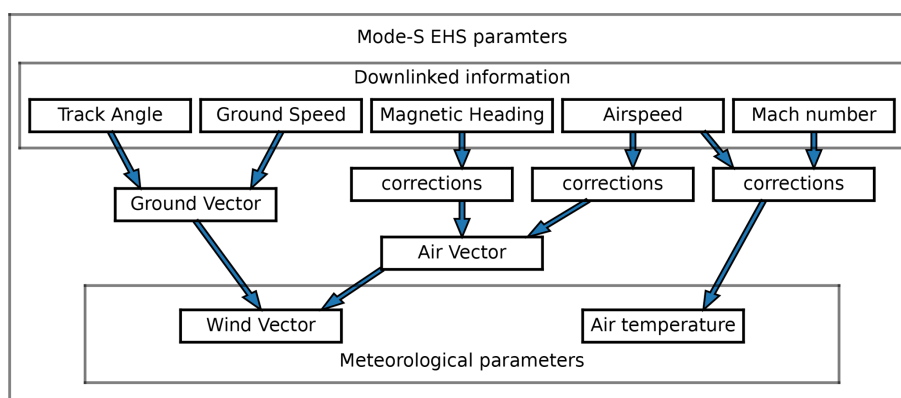


Figure 2. Downlinked data flow of Mode-S EHS parameters to acquire meteorological information.

an observation frequency depending on the radar; ADS-B information can have an observation frequency of twice per second. Table 1 provides information on the downlinked parameters. The information flow of the downlinked parameters is depicted in Fig. 2. Also shown in this figure are the corrections applied to the magnetic heading, true airspeed, and Mach number, discussed later.

4.2 Raw data input control

The EMADDC quality procedure has been systematically developed and enhanced over the past decade. The first step in defining the quality is to check the input for obvious errors or measurements in conditions where calculation is not possible, as listed in Table 2. Measurements failing one of these checks are discarded from further processing.

4.3 Output control and safe list

Output control is necessary to obtain good-quality observations. The parameters for output quality control are related to the correction methods applied to the temperature and wind measurement.

Additionally, a safe list is created for wind speed and temperature independently, based on the difference between observations and forecast statistics. Aircraft with a 14 d wind standard deviation exceeding 4 kn are designated as ineligible, while for temperature, the standard deviation threshold is set at 1.23 K. EMADDC currently uses the operational ECMWF model with a minimal forecast lead time of 9 h for this comparison by collocating observations with numerical weather prediction (NWP).

5 Derived wind measurement

The wind vector is the difference between ground vector and air vector, where all vectors are with respect to true north.

$$V \begin{pmatrix} \cos(d) \\ \sin(d) \end{pmatrix} = G \begin{pmatrix} \cos(t) \\ \sin(t) \end{pmatrix} - A \begin{pmatrix} \cos(h) \\ \sin(h) \end{pmatrix}, \quad (4)$$

where V denotes the wind speed with wind direction d , G and t are the ground speed and track angle, and A and h the airspeed and heading. Note that this equation is valid under the assumption that the vertical wind speed is negligible, the

Table 1. The reported precision and observation frequency of downlinked parameters, all parameters are rounded.

Parameter	Abbreviation	Symbol	Reported precision	Frequency	BDS
Position (latitude–longitude)	lat, long	λ, ϕ	$1 \times 10^{-5}^\circ$	0.5–2 s	ADS-B
Position (latitude–longitude)	lat, long	λ, ϕ	$2\text{--}5 \times 10^{-5}^\circ$	5–20 s	radar echo
Flight level	1 flight level = 100 ft (30.48 m)		25 ft (7.62 m)	5–20 s	ADS-B
Roll angle	ra		0.175°	5–20 s	5.0
True track angle	tta	t	0.175°	5–20 s	5.0
Ground speed	gspd	G	2 kn (1.02 m s^{-1})	5–20 s	5.0
Track angle rate	tar		$0.03125^\circ \text{ s}^{-1}$	5–20 s	5.0
True airspeed	tas	A	2 kn (1.02 m s^{-1})	5–20 s	5.0
Magnetic heading	mhdg	h_m	0.352°	5–20 s	6.0
Indicated airspeed	ias	A_I	1 kn (0.51 m s^{-1})	5–20 s	6.0
Mach number	mach	M	0.004	5–20 s	6.0

Table 2. Current input quality checks used in operation.

Input data quality checks	Occurrence
1 absolute value of roll angle larger than 2.5°	16 %
2 absolute difference between track angle and magnetic heading larger than 25°	1 %
4 true airspeed larger than 570 kn or smaller than 100 kn	1 %
5 ground speed larger than 850 kn or smaller than 50 kn or when below flight level 50 smaller than 100 kn	2 %
6 Mach number equal to 0	2 %
7 constant flight level and decreasing ground speed and indicated airspeed when flight level is lower than 50	2 %
8 position consistency	< 1 %

sideslip is zero, and the roll angle is small. The heading is reported with respect to the magnetic north pole and needs to be converted into a heading with respect to true north. For this purpose, geomagnetic declination tables from Maus and Macmillan (2005) and Chulliat (2015) are applied; thus

$$h = h_m + \Delta(y, \lambda, \phi), \quad (5)$$

where y is the datum of the (static) heading correction table on board the aircraft, (λ, ϕ) is the location of the aircraft, and Δ is the heading correction from the declination table. As it turns out, the heading correction is aircraft-dependent; that is y is aircraft-dependent and even may change in time after an aircraft is being serviced, for example when the computer software is updated (Mirza et al., 2016).

5.1 Aircraft-dependent heading correction

Although the correction should be in the order of the (actual) declination, research showed that a simple correction is not enough (de Haan, 2011; Pourret et al., 2021). Each aircraft may use their own version of a declination lookup table, which implies that each aircraft corrects the true north to magnetic north in a different way. The correction method uses the assumption that the correction is determined by a geomagnetic reference table for a certain datum (or epoch) and is static until it is updated through aircraft maintenance. The optimal datum is found by minimizing a cost function,

depending on datum, by comparing corrected winds from observations with NWP model forecast winds. The cost function is constructed by the vector length difference between the unit heading vector from the aircraft and the unit heading vector formed by the ground vector and NWP wind vector, that is

$$\delta^i(y) = \left(\frac{\cos(h_N^i)}{\sin(h_N^i)} \right) - \left(\frac{\cos(h_m^i + h_c(y, \lambda^i, \phi^i))}{\sin(h_m^i + h_c(y, \lambda^i, \phi^i))} \right), \quad (6)$$

where y is the date of the inclination table, i the index of an observation, h_m^i is the observed magnetic heading, and $h_c(y, \lambda^i, \phi^i)$ is the value of the declination table with datum y at location (λ^i, ϕ^i) . NWP heading angle h_N^i is defined as

$$h_N^i = \text{atan} \left(\frac{G^i \sin(t^i) - V^i \sin(d^i)}{G^i \cos(t^i) - V^i \cos(d^i)} \right), \quad (7)$$

with G^i the ground speed, t^i the track angle, V^i the wind speed, and d^i the wind direction. The first vector in Eq. (6) is the NWP heading direction vector; the second is the corrected heading direction vector.

The cost function is defined as the sum of all vector length differences over all observations, that is

$$C(y) = \frac{1}{2} \sum_i \|\delta^i(y)\|^2 \quad (8)$$

$$= \frac{1}{2} \sum_i \left(\sin(h_N^i) - \sin(h_m^i + h_c^i(y)) \right)^2 + \left(\cos(h_N^i) - \cos(h_m^i + h_c^i(y)) \right)^2 \quad (9)$$

$$= \sum_i 1 - \cos(-h_N^i + h_m^i + h_c^i(y)). \quad (10)$$

Next, magnetic declination is linearized with datum, in order to find a minimum, that is

$$h_c^i = H_0^i + (y - y_{\text{ref}}) \Delta H^i, \quad (11)$$

where H_0^i is the value of magnetic declination for given location (λ^i, ϕ^i) on datum y_{ref} , and ΔH^i is the value of the change in magnetic declination per year (this approximation is valid, as is discussed in Appendix A). The cost function is approximated by a quadratic function in the datum offset, which yields

$$C(y) \approx 1 - \cos(H_0^i - h_N^i + h_m^i) + \delta_y \sum_i \Delta H^i \sin(H_0^i - h_N^i + h_m^i) + \frac{1}{2} \delta_y^2 \sum_i (\Delta H^i)^2 \cos(H_0^i - h_N^i + h_m^i), \quad (12)$$

where

$$\delta_y = y - y_{\text{ref}}. \quad (13)$$

The (offset) datum value for which the cost function attains a minimum is found by setting the derivative to δ_y of the cost function to zero. Let x_1^i and x_2^i for observation i be defined by

$$x_1^i = -\Delta H^i \sin(H_0^i - h_N^i + h_m^i) \quad (14)$$

$$x_2^i = (\Delta H^i)^2 \cos(H_0^i - h_N^i + h_m^i), \quad (15)$$

then the datum minimizing the cost function is given by

$$y = y_{\text{ref}} + \frac{\sum_i x_1^i}{\sum_i x_2^i}. \quad (16)$$

Using the value for an aircraft EMADDC calculates the corresponding magnetic declination tables at the location of an observation to find the declination and converts the reported magnetic heading to true heading and calculates the wind according to the equation above.

The above method uses NWP wind forecasts extracted from the operational ECMWF forecast (IFS model). The forecast lead time is at least 9 h, such that observations from EMADDC are not used as input for assimilation and correction simultaneously.

5.2 Heading correction results

Figure 3 shows the results of the heading correction for all 19 006 aircraft operational in 2021–2023. Each aircraft is represented by a vertical line in the top panel. The white colour indicates that no correction was found and aircraft with a lot of white pixels are not flying regular in the EMADDC domain. The top panel shows the offset with the maximum reported heading correction in 2021–2023. In general, the change in heading correction over 2023 is small, except for aircraft that have large corrections. The most constant correction values (offsets smaller than 2 years, red- to brown-coloured) are found for values close to 2023, while higher offsets are only found with high maximum values, which gives us reason to believe that for these aircraft, the datum correction algorithm may not perform optimally. Note that aircraft near the right of Fig. 3 have declination tables decades into the future and are obviously not correctly estimated, which is also reflected in the noisy pattern for these aircraft. The corrections for these aircraft are invalidated, and the aircraft are marked as bad and will not be present in dissemination.

5.3 Dependence on NWP wind vector information

The magnetic heading is calibrated using NWP wind vector information. Consequently, the obtained correction depends on the quality of the NWP information. The magnitude of this dependence is rather small and of the order of 1 over the magnitude of ground speed, which is explained by the following.

Suppose we have a biased NWP wind direction, that is the true wind direction d is biased by β , then

$$\tilde{h}_N = \text{atan} \left(\frac{G \sin(t) - V \sin(d + \beta)}{G \cos(t) - V \cos(d + \beta)} \right) \approx h_N + \beta \frac{V(-G \cos(d - t) + V)}{G^2 - 2GV \cos(d - t) + V^2}, \quad (17)$$

which implies that

$$|\tilde{h}_N - h_N| \lesssim \left| \beta \frac{V(V + G)}{(G - V)^2} \right| \lesssim \frac{|\beta|}{G} \ll |\beta|. \quad (18)$$

The offset from the true heading using wind-direction-biased information is substantially smaller than the actual bias.

The difference in heading from an unbiased and a biased NWP model is smaller than the bias divided by the ground speed (G). The ground speed is in general around 100 to 200 m s⁻¹, and thus the bias in heading is around 100 times smaller than the bias in the wind vector.

Similarly, the offset from the true heading based on wind speed biased (by α) is given by

$$|\tilde{h}_N - h_N| \lesssim \left| \alpha \frac{G \sin(d - t)}{G^2 - 2GV \cos(d - t) + V^2} \right| \lesssim \frac{|\alpha|}{G} \ll |\alpha|. \quad (19)$$

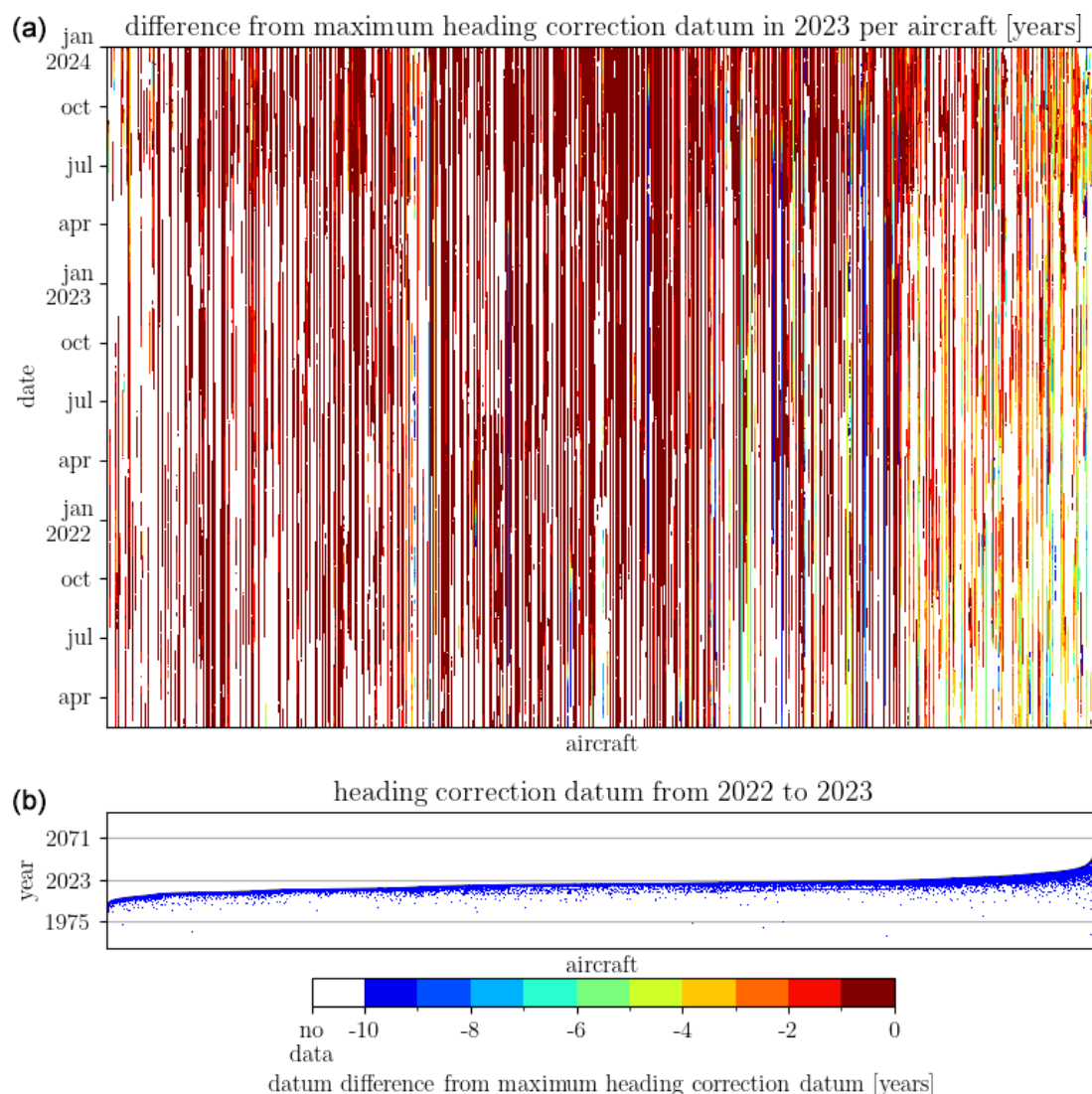


Figure 3. Panel (a) shows the difference of heading correction in 2021–2023 with the reported maximum heading correction datum in this period. The white colour indicates that no (or bad) heading correction value was reported. Panel (b) shows the maximum heading correction found in 2023 (top black line), with the deviations from its maximum as small blue dots. The 19 006 aircraft are sorted by maximum heading correction.

When the model contains a wind speed bias, the bias in the heading is smaller than the wind speed bias divided by the ground speed and thus will be a factor of 10 to 100 smaller than the wind speed bias in the model.

Since the heading correction is based on many observations, over a large period (at least 15 d) it can be regarded as independent of the NWP information. Moreover, the EMADDC system is fitting a declination table over a large domain with a single parameter; the effect of locally biased or erroneous NWP estimates will be small. Aircraft visiting this region will have deviating quotients from which the datum is calculated (see Eq. 16). This deviation is detected by checking the standard deviations of the quotients separately.

5.4 True airspeed correction

The measurement of true airspeed (TAS) depends on the temperature and Mach number; see Eq. (3). Since the observed Mach number is corrected (Rodi and Leon, 2012), the true airspeed measurement might be improved likewise.

The current algorithm uses an aircraft-dependent constant correction value. However, in the next update of the EMADDC processing (EMADDC 3.0), a true airspeed-dependent correction will be investigated. This improvement is reasoned by comparing the observed true airspeed with a NWP-model-equivalent true airspeed, thus creating a correction that depends on the observed value of true airspeed. The model true airspeed is obtained by the vector difference be-

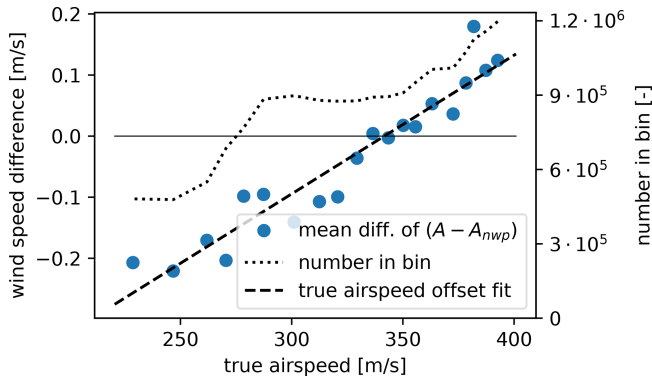


Figure 4. The mean difference of true airspeed as derived from observed ground vector and model wind versus the observed true airspeed. Data from 3 full days (5 June 2021, 2 December 2021 and 1 August 2023) and two 6 h intervals (1 August 2022 06:00–12:00 UTC, and 1 January 2023 00:00–06:00 UTC) are shown.

tween the observed ground vector and wind vector from the NWP model. Figure 4 displays the average difference of the observed true airspeed versus the model-based true airspeed. The differences are averaged over true airspeed bins for all data points observed in the EMADDC domain in 3 d. Note that the difference in wind speed is of the order of a few tenths of a metre per second. Clearly there is a relation between mean airspeed difference and true airspeed itself.

As a first-order approach, the EMADDC system currently applies a true airspeed bias correction depending on aircraft and phase of flight. Future research will study physical methods of true airspeed correction.

6 Derived temperature

Although the temperature is measured by the sensors on board the aircraft, the information is not transmitted in the Mode-S EHS request BDS5.0 and BDS6.0. However, the Mach number M and the true airspeed A are available and from these parameters the temperature can be deduced using the relation between the speed of sound and temperature and the ideal gas law,

$$M = \frac{A}{C}, \quad (20)$$

where $C = \sqrt{\gamma R_d T}$, and R_d is the universal gas constant for dry air. Note that the dependence of the speed of sound on humidity is neglected. So, given M and A , the temperature T can be calculated by

$$T = \frac{1}{\gamma R_d} \left(\frac{A}{M} \right)^2, \quad (21)$$

where A is in $[\text{m s}^{-1}]$.

6.1 Onboard aircraft temperature correction

The aircraft measurements are improved by algorithms on board the aircraft. The applied corrections are not available and may be aircraft-dependent, or aircraft-type-dependent, or both. It is known that the measurement of the static pressure p_s suffers from airflow instabilities and/or angle of attack (Rodi and Leon, 2012). The static pressure is corrected, which consequently results in a correction of the Mach number M and temperature T .

6.2 Temperature measurement improvements

The temperature is calculated from the Mach number and the true airspeed (TAS) on board the aircraft. Actually, there are two types available of the Mach number, one being the downlinked data and one determined using indicated airspeed (IAS) information. The downlinked Mach number is of worse accuracy than the Mach number determined from the indicated airspeed, as will be discussed next. An estimate of the formal error of the temperature T calculated using Eq. (21) is constructed following (Taylor, 1997)

$$\sigma_T^2 \approx \left(\frac{\partial T}{\partial A} \right)^2 \sigma_A^2 + \left(\frac{\partial T}{\partial M} \right)^2 \sigma_M^2 = \frac{4T^2}{A^2} \sigma_A^2 + \frac{4T^2}{M^2} \sigma_M^2. \quad (22)$$

Rounding leads to an additional error of $r/\sqrt{12}$, where r is the rounding (see Appendix B). One of the downlinked parameters is indicated airspeed, which is defined as the airspeed measured as if the aircraft were flying at mean sea level ($p_0 = 1013.25 \text{ hPa}$, $T_0 = 288.15$, and $\rho_0 = 1.225 \text{ kg m}^{-3}$, the standard pressure, temperature, and density at mean sea level)

$$A_I = \sqrt{\frac{2}{\gamma - 1} \frac{p_0}{\rho_0} \left(\left(\frac{q_t}{p_0} + 1 \right)^{\frac{\gamma - 1}{\gamma}} - 1 \right)} \\ \Rightarrow q_t = p_0 \left(\left(\frac{\gamma - 1}{2} \frac{\rho_0}{p_0} A_I^2 + 1 \right)^{\frac{\gamma}{\gamma - 1}} - 1 \right) \quad (23)$$

The dynamic pressure q_t can be calculated from A_I , which in turn can be used to recalculate M , according to Eq. (1), so an (first-order) estimate of the error in Mach number becomes

$$\sigma_M^2 \approx \left(\frac{\partial M}{\partial q_t} \frac{\partial q_t}{\partial A_I} \right)^2 \sigma_{A_I}^2 \quad (24)$$

$$= \frac{A_I^2}{M^2} \frac{\rho_0^2}{p^2} \left(1 + \frac{q_t}{p} \right)^{-\frac{2}{\gamma}} \left(\frac{A_I^2 \rho_0 (\gamma - 1)}{2 p_0} + 1 \right)^{\frac{2}{\gamma - 1}} \sigma_{A_I}^2. \quad (25)$$

Figure 5 shows the formal temperature errors as calculated with the downlinked Mach number ($\sigma_T^2(M)$, red line) and the temperature error based on the indicated airspeed ($\sigma_T^2(A_I)$, blue line) and the formal error term related to the A . The effect on introducing the Mach indicated airspeed is a reduction in formal error of a factor of 4. The largest part in the

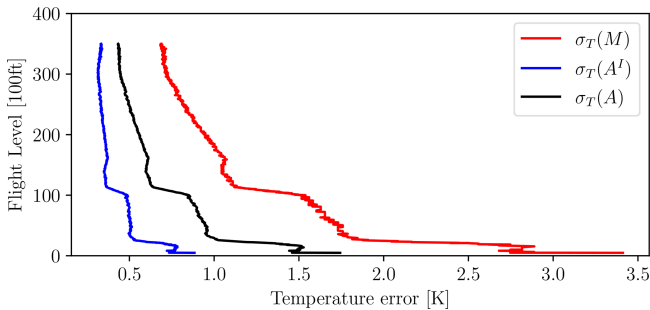


Figure 5. Different terms of the formal error for temperature. Black line represents the formal error due to uncertainty in A ; red line represents the term related to the temperature derived using the (coarse) Mach number; blue line depicts the formal error when the Mach number is recalculated from A^I . Aircraft data from ICAO24 A2BD72 (Airbus A320-214), valid from 1 August 2023 05:17:07 to 1 August 2023 05:26:40.

formal error is (with the A_1) now related to the true airspeed error. This implies that to further improve the temperature observation, reduction of true airspeed error needs to be accomplished.

The temperature correction is performed by first averaging, followed by correction. An average over 20 s is determined to reduce the noise in the temperature observation, that is

$$\bar{T} = \frac{1}{N} \sum_i^N T_i^I, \quad (26)$$

and average observations are marked as bad when the standard deviation of the observations T_i^I exceeds 5 K.

The temperature correction applied is based upon correcting the static pressure measurements and recalculating subsequently the Mach number, the total temperature, and finally the temperature. The correction used is described in detail in de Haan et al. (2022). The corrected static pressure correction p_c depends on the static pressure itself and the static pressure divided by the square of true airspeed, that is

$$p_c = a + p_s \left(b + \frac{c}{A^2} \right), \quad (27)$$

where a , b , and c are determined by fitting NWP temperatures. These coefficients are aircraft-dependent. The corrected temperature is determined as follows: first the corresponding total temperature is calculated,

$$\bar{T}_t = \left(1 + \lambda \frac{(\gamma - 1)}{2} M^2 \right) \bar{T}, \quad (28)$$

and then the corrected temperature calculated using the corrected pressure p_c becomes

$$\bar{T}_c = \left(1 + \lambda \frac{(\gamma - 1)}{2} M_c^2 \right)^{-1} \bar{T}_t, \quad (29)$$

where

$$M_c^2 = \frac{2}{\gamma - 1} \left(1 + \frac{q_t}{p_c} \right)^{\frac{\gamma}{\gamma - 1}} - 1. \quad (30)$$

Figure 6 displays the effect of the three steps (green “raw” temperature, improving M by the indicated airspeed (red), followed by the averaging (blue) and pressure correction (black)). Applying a simple smoothing of adjacent points, the standard deviation is reduced to values between 1 K in the mid-troposphere and up to 1.5 K at higher altitudes and near the surface (Fig. 6 compares red and black dashed lines); The bias (and standard deviation to a lesser extent) is then further reduced by applying the raw pressure correction method (Fig. 6 black lines).

Figure 7 displays the necessary steps needed in the processing to turn Mode-S EHS observations into meteorological information.

7 Data processing methodology

EMADDC suppliers generally deliver data in batches every 5 to 15 min, allowing the system to retrieve new files for ingestion into the EMADDC database. Receiver data are processed by the decoder, which combines the BDS5.0 and BDS6.0 registers to generate observations that are subsequently entered into the EMADDC database. In contrast, for ATC radar/tracker data, the observations are decoded (from ASTERIX to an EMADDC internal format) and inserted into the database. For this type of data, no intermediate combining step is necessary, as the registers have already been consolidated by the respective tracker or radar system.

Once the data have been ingested into the hourly database table, the processing scheduling system triggers three specific processing jobs. The first job handles observations within a 15 min time window, operating with an approximate delay of 30 min. This job functions as the primary processing task, indicating that the designated time window has been successfully processed. Figure 8 shows the coverage of EMADDC on 21 April 2024.

In 2022, two additional processing jobs were introduced to process observations in 5 min batches at 13 and 23 min past the initial observation in the designated time window. These “fast” files account for approximately 70 % and 90 % of the total data available in the standard 15 min interval files, respectively. The implementation of these new fast files has notably enhanced the timeliness of EMADDC by reducing the delivery delay to approximately 10 min. These fast files should be utilized when 15 min interval data are not yet available/processed.

A processing job begins by collecting all available data within the specified time window. The input data undergo quality control, as previously outlined. To ensure continuity in flight profiles and phase determination, the data from the last 5 min of the preceding time window are included for

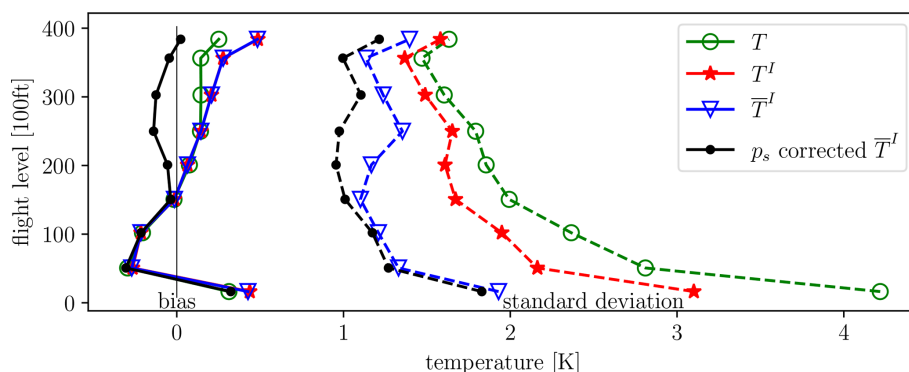


Figure 6. Results of temperature correction based on data from 1 August 2023 06:00–12:00 UTC. The mean difference between model and observation is shown by a solid line and the standard deviation of the difference (dashed line). The green line represents the raw temperature statistics, the red line temperatures calculated using the Mach–IAS relation, and the blue line the additional time averaging statistics, while the black line is the final result after the raw pressure correction.

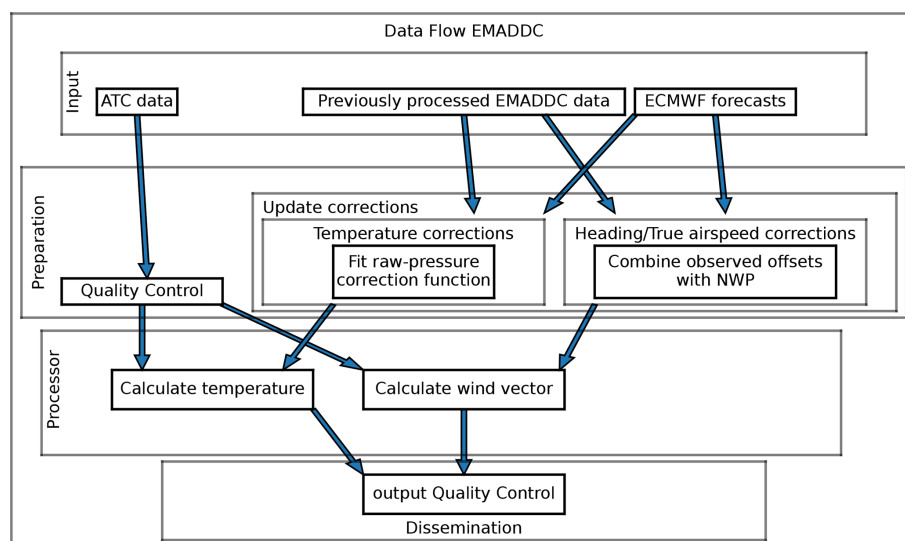


Figure 7. Functional data flow in EMADDC, needed to derive wind vector and temperature observations.

continuation of flight profiles and phase determination. The flight profile and flight phase are used in the application of the corrections and quality control. Wind and temperature are derived using equations for wind speed and wind direction (see above), and the detected magnetic table datum is used in the World Magnetic Model (Chulliat, 2015) to determine the magnetic declination at the location of the observation and obtain the true heading. Subsequent corrections and post-processing are applied, after which outliers are identified using linear regression over individual aircraft's 30 s time windows.

EMADDC currently receives data from multiple overlapping sources. For instance, in the EUROCONTROL MUAC area, EMADDC acquires radar data from MUAC, in addition to data from approximately four receivers managed by Air Support, which has now been rebranded as ADSB Support. KNMI operates ADSB Support receivers located in De

Bilt and Cabauw at 180 m, which captures data extending to Paris. Since these receivers collect similar data to those of the radars used in ATC, the EMADDC database contains duplicate entries. To address this issue, a duplicate detection algorithm has been implemented, whereby data from the same aircraft recorded within 1 s of another observation are designated as a duplicate of the primary observation. Observations identified as duplicates are not included in the EMADDC output. It is important to note that observations obtained directly from air traffic control – such as radar or tracker data – are not marked as duplicates, as the ATC system applies its own filtering and quality control measures to eliminate duplicates. Consequently, it is possible for an observation to appear in a fast file but be absent in subsequent 23 min fast files or full time-window files if it was identified as a duplicate and replaced by another observation. The current approach to identifying duplicates has demonstrated effective-

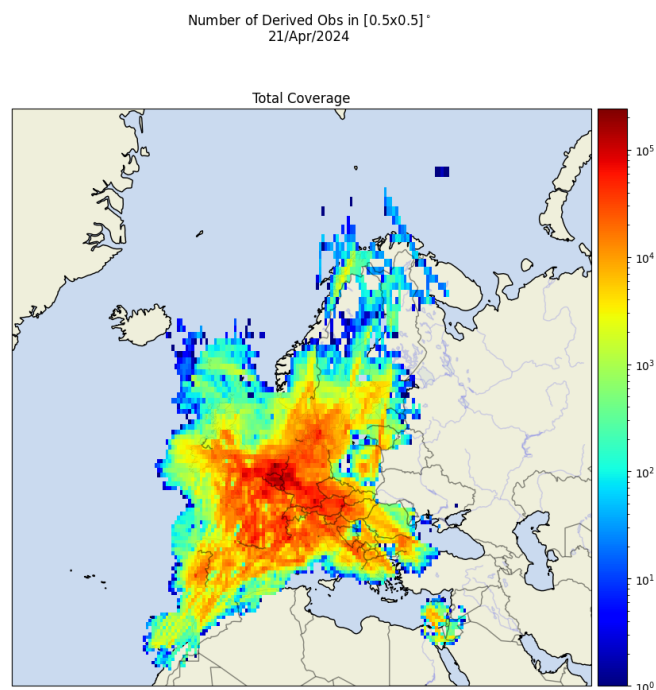


Figure 8. Data coverage of EMADDC on 21 April 2024. The colour indicates the number of observations per 0.5° squared box.

ness; however, it is not without imperfections. The intent is to enhance this process in future developments of EMADDC R3.0. In addition to this step of identifying duplicate observations, the decoding and combining process effectively addresses the potential for duplicate registers that may be received from multiple overlapping receivers within a receiver file. Furthermore, ATC data does not contain duplicate observations, as radars exclusively process received interrogations that pertain to their own interrogations, discarding those produced by other radars.

The final processing step involves verifying whether the observations are from aircraft that have been approved for use through a safe list, as explained in Sect. 4.3.

Ultimately, all validated, non-duplicate, and quality-checked data are compiled into CSV and BUFR file formats. These files have been made accessible via KNMI's FTP servers and the KNMI Data Platform (KDP). Please visit <https://datapatform.knmi.nl/dataset/access/emaddc-hist-repro-data-1-0> (last access: 4 June 2025). Access to these historical files or datasets is open, whereas operational data are currently restricted to authorized users.

8 Quality assessment

Observations derived with the above described processing system are validated with upper-air in situ measurements and numerical model equivalents. This section shows the com-

parison of Mode-S EHS wind and temperature observations against the ECMWF NWP model, radiosonde wind and temperature observations, and AMDAR wind and temperature observations. For the NWP model, the forecast lead time is a minimum of 9 h, in order to avoid comparing observations that are used in assimilation. The comparison with NWP equivalents is based on 3 months of data (January–March 2024). Section 8.2 discusses EMADDC versus radiosondes and NWP (January–March 2023). And finally, in Sect. 8.3, EMADDC, AMDAR, and NWP are compared over an 8-month period.

8.1 Numerical weather prediction model comparison

Wind and temperature observations are corrected using NWP information, as was discussed above. The heading correction applied may introduce some (NWP) correlation but not directly to wind components and temperature and when correlations exist it will be small (see Sect. 5.3). Furthermore, corrections, as well as parameters for the safe list, are based on past NWP data, while the comparison is made using current (forecast) data.

From 1 January to 1 April 2024, a total of 4.5 billion observations were derived by the EMADDC system. From these observations 2.8 billion unique and allowed wind observations are made available to the users, and in total nearly 1.8 billion temperature observations are disseminated in these 3 months. The quality of wind observations compared to ECMWF-IFS is around 2.5 m s^{-1} in standard deviation, with a small bias of 0.3 m s^{-1} ; see Table 3. Table 3 also shows the statistics of wind observations for different height levels: the error in wind speed with respect to the model increases with height from 2.2 m s^{-1} near the surface to 2.8 m s^{-1} at a height of 11 km. The wind direction statistics show a different signal; near the surface the wind direction error is nearly 15° , with a minimum at around flight level 350, increasing again to an error of 10° at around 11 km. Note that only observations with wind speed larger than 4 m s^{-1} are used for wind direction. Wind is in general more variable near the surface. These values are all within the acceptable range for use in for example data assimilation. According to the WMO Rolling Review of Requirements, the requirements for temperature used in high-resolution NWP are 0.5 K (goal), 1 K (threshold), and 3 K (breakthrough) and for wind speed 1, 2, and 5 m s^{-1} , respectively (WMO, 2025).

The temperature statistics are shown in Table 4. The temperature error in total is slightly smaller than 1 K, with a minimum error of 0.8 K around flight level 250. The maximum error is found at cruising level (11 km). Note that the bias with the model is around zero.

8.2 Comparison with radiosonde observations

Radiosondes are regarded as anchor observations for meteorology and are generally launched 45 min prior to the main

Table 3. Statistics of EMADDC wind observations against the operational ECMWF model.

1 January to 1 April 2024								
		Wind speed, EMADDC-NWP				Wind direction, EMADDC-NWP		
		Raw volume	Number	Bias [m s ⁻¹]	SD [m s ⁻¹]	Number	Bias [°]	SD [°]
All data		4 546 047 080	4 384 070 442	0.34	4.76	4 281 120 981	0.14	9.62
Allowed and unique		–	2 868 355 459	0.30	2.52	2 800 011 753	0.17	8.67
Flight level	Pressure (hPa)	Raw volume	Number	Bias [m s ⁻¹]	SD [m s ⁻¹]	Number	Bias [°]	SD [°]
0–100	1013–696	235 608 797	151 947 715	0.20	2.20	135 707 126	–0.16	14.24
100–200	696–465	361 851 780	252 516 206	0.22	2.28	243 224 271	0.45	10.77
200–300	465–300	594 829 968	386 369 304	0.27	2.53	378 271 790	0.33	9.32
300–400	300–187	3 133 630 884	2 016 820 254	0.32	2.60	1 983 182 045	0.12	7.94
> 400	< 187	220 023 701	131 427 728	0.36	2.81	128 500 903	0.02	10.03

SD: standard deviation.

Table 4. Statistics of EMADDC temperature observations against the operational ECMWF model.

January 2024–March 2024					
		Temperature, EMADDC-NWP			
		Raw input	Number	Bias [K]	SD [K]
All data		4 546 047 080	3 138 758 482	0.02	1.05
Allowed and unique		–	1 763 880 586	–0.00	0.95
Flight level	Pressure (hPa)	Raw volume	Number	Bias [K]	SD [K]
0–100	1013–696	235 608 797	51 837 791	0.13	1.08
100–200	696–465	361 851 780	140 307 524	0.05	0.83
200–300	465–300	594 829 968	241 388 676	0.06	0.77
300–400	300–187	3 133 630 884	1 309 098 407	–0.01	1.04
> 400	< 187	220 023 701	83 204 202	0.05	1.24

SD: standard deviation.

synoptic hours 00:00 and 12:00 UTC to achieve a profile centred around 00:00 UTC (12:00 UTC). Note that a few sites also launch at 06:00 and 18:00 UTC. Due to budget restrictions some radiosondes are only launched once a day. Table 5 shows collocated observations statistics of wind (speed and wind components). Table 6 contains temperature statistics. Aircraft and radiosonde observations will never be exactly collocated in both space and time. Nevertheless, collocations are made by having the maximum distance between aircraft and radiosondes of at most 50 km, a maximum height difference of 100 m, and a time difference of 1800 s.

For all wind parameters, the comparison between radiosonde and EMADDC mostly has a standard deviation lower than that of the comparison model and EMADDC or radiosonde. Furthermore, the difference between model and radiosonde or model and EMADDC is similar and of the order of 0.3 to 0.5 m s⁻¹, while the mean difference between

aircraft and balloon is small. The reason for this is the model representation error due to the model grid size. The temperature statistics show that all three systems have the same average temperature (all biases are small and near zero). Not surprisingly, the temperature observations of EMADDC are clearly of less quality than radio soundings, although above 850 hPa the quality is reasonable.

8.3 Comparison with AMDAR observations

Finally, a comparison is made between EMADDC observation and AMDAR observations. AMDAR observations are extracted from the onboard computer and should have a large resemblance to EMADDC observation, and thus the mutual statistics are expected to be small. Figure 9 shows the statistics of EMADDC, AMDAR, and NWP for temperature and vector RMSE. A lookup table is used to connect AMDAR aircraft to an ICAO aircraft identification. This lookup ta-

Table 5. Statistics of wind observations comparison against radiosondes and the operational ECMWF model.

January–March 2023								
East–west wind component [m s^{-1}]								
			EMADDC-RS		EMADDC-NWP		RS-NWP	
Flight level	Pressure (hPa)	Number	Bias [m s^{-1}]	SD [m s^{-1}]	Bias [m s^{-1}]	SD [m s^{-1}]	Bias [m s^{-1}]	SD [m s^{-1}]
< 0	> 1023	166	0.58	1.16	0.40	1.41	−0.09	1.08
0–100	1013–696	541 153	0.07	2.42	−0.20	2.59	−0.19	2.46
100–200	696–465	790 741	0.11	2.08	−0.11	2.32	−0.17	2.17
200–300	465–300	976 001	0.10	2.25	−0.09	2.66	−0.14	2.49
300–400	300–187	1 548 851	0.05	2.36	−0.06	2.69	−0.11	2.55
> 400	< 187	54 193	0.09	2.78	0.10	2.87	0.04	2.74
North–south wind component [m s^{-1}]								
< 0	> 1023	160 746	0.02	2.38	−0.06	2.47	−0.25	2.34
0–100	1013–696	434 029	0.02	2.12	0.00	2.31	−0.09	2.16
100–200	696–465	565 086	0.02	2.27	−0.17	2.44	−0.22	2.40
200–300	465–300	858 477	0.01	2.33	−0.25	2.71	−0.23	2.56
300–400	300–187	590 015	−0.13	2.51	−0.32	2.78	−0.18	2.68
> 400	< 187	236	0.03	1.92	−0.19	2.02	−0.17	1.76
Vector RMSE								
			EMADDC-RS		EMADDC-NWP		RS-NWP	
		Number	VRMSE [m s^{-1}]		VRMSE [m s^{-1}]		VRMSE [m s^{-1}]	
< 0		166	2.18		2.06		2.05	
0–100		541 153	3.31		3.53		3.35	
100–200		790 741	2.98		3.32		3.10	
200–300		976 001	3.30		3.81		3.66	
300–400		1 548 851	3.40		3.89		3.70	
> 400		54 193	3.84		4.19		3.88	

SD: standard deviation.

Table 6. Comparison against radiosondes.

January–March 2023								
Temperature [K]								
			EMADDC-RS		EMADDC-NWP		RS-NWP	
Flight level	Pressure (hPa)	Number	Bias [K]	SD [K]	Bias [K]	SD [K]	Bias [K]	SD [K]
0–100	1013–696	165 790	0.03	1.17	0.05	1.13	−0.01	0.97
100–200	696–465	496 154	0.07	0.84	−0.00	0.84	−0.08	0.65
200–300	465–300	704 675	−0.01	0.76	0.04	0.78	0.05	0.56
300–400	300–187	1 158 841	0.01	1.04	−0.01	0.67	−0.06	0.87
> 400	< 187	36 737	0.08	1.31	0.02	1.27	−0.08	1.21

SD: standard deviation.

ble is partially based on E-AMDAR information and is completed with results from collocation with EMADDC observations. Due to rounding of position and time, exact collocations of AMDAR and Mode-S can be tedious. Moreover, the reporting observation frequencies are different, resulting in about half of the AMDAR observations being collocated. Here an AMDAR observation is collocated with an EMADDC observation when the time difference is less than 2 min, the distance is less than 1 km, and most importantly the height difference is at most 25 ft. The height discrimination is strong because in general temperature and wind tend to change more with height than with horizontal displacements.

The temperature of AMDAR has a positive bias when compared to NWP, which is known in the literature (Zhu et al., 2015). The bias of EMADDC is around zero. The AMDAR temperature standard deviation is clearly better than EMADDC. The increase in standard deviation near the ground is due to atmospheric turbulence being more present near the surface (Schwartz and Benjamin, 1995). The EMADDC temperature error increases at lower levels because aircraft land generally with a low airspeed and Mach number, which results in a larger error due to the rounding of the airspeed observation given the relatively low resolution of these parameters.

For wind the vector RMSE shows that both AMDAR and EMADDC perform equally well; there are small differences but not significant. However, a gross error check on the zonal wind is needed to detect systematic errors in B787 aircraft (WMO, 2022).

9 Conclusions and outlook

This paper presents the current EMADDC system (R2.2) to produce wind and temperature observations derived from Mode-S EHS aircraft messages.

Mode-S EHS is a surveillance method which not only tracks an aircraft in the range of the radar but also interrogates the aircraft and requests specific information which is used by air traffic services. These downlinked data contain sufficient information to derive wind and temperature at very high spatial and temporal resolution. These data are being processed by EMADDC to produce high-quality meteorological information. First of all, to be able to generate observations of good quality, several quality checks are applied. The quality of directly derived wind is hampered by an unknown offset in heading and low resolution of the Mach number and airspeed when deriving temperature.

To obtain high-quality wind measurements, a correction from magnetic heading to true heading is necessary; this heading correction is unique for each aircraft and may change in time due to maintenance of the aircraft. The observations are compared to wind forecasts of the ECMWF model, radiosonde wind observations over a 3-month period,

and AMDAR observations over a period of 8 months. The derived wind observations compare well to the model and the radiosonde observations with similar statistics to those when comparing AMDAR to model equivalents.

The temperature is derived from the quadratic quotient of the true airspeed and the Mach number, with both values truncated. The estimate of the Mach number can be improved by exploiting the downlinked indicated airspeed. Next, a mean temperature is determined using a 20 s time window, and finally, the temperature is corrected based on the methodology developed in de Haan et al. (2022). Error analysis revealed that the quality of the true airspeed measurement limits the temperature quality. Comparison with radiosonde observations showed good quality with respect to temperature when the observation is above 700 hPa (albeit that the temperature error increases for both AMDAR and EMADDC to a maximum at 200 hPa). AMDAR comparison showed that wind observations from EMADDC are of equal quality, while temperature observations have 25 % larger error. The produced meteorological information, when thinned to avoid over-fitting, is at present widely used in regional and some global numerical weather prediction models.

The advantage of EMADDC over AMDAR is that only few aircraft are AMDAR-equipped, and Mode-S EHS is available (almost) everywhere in Europe, and therefore all aircraft are being utilized as a sensor. The costs of receiving data are limited; in most areas the data can be obtained through ATC partners or via local receivers. On the other hand, the AMDAR temperature, although biased, is better for heights below 850 hPa compared to EMADDC-derived temperatures.

Although the above-described system delivers good-quality wind and temperature observations, it is highly preferred to receive the aircraft onboard measurements of available meteorological information directly, as this will reduce and simplify the errors. Current available systems, like the changes introduced in ADS-B Version 3, may be able to provide this information in the future.

A final remark needs to be made on the use of NWP forecast for correction. It is assumed that the NWP model and forecast are (almost) bias-free. If this would not be the case then the bias might be reflected with reduced magnitude as a bias in the “corrected” observations (Eyre, 2016). This paper showed that this is not the case for the heading correction because aircraft heading is not related to forecast values.

Outlook

The EMADDC team tries to improve the quality of derived wind and temperature continuously. The team has the following items to investigate or implement:

- The current heading correction algorithm does not detect changes in the heading table datum effectively, especially when high datum correction values are de-

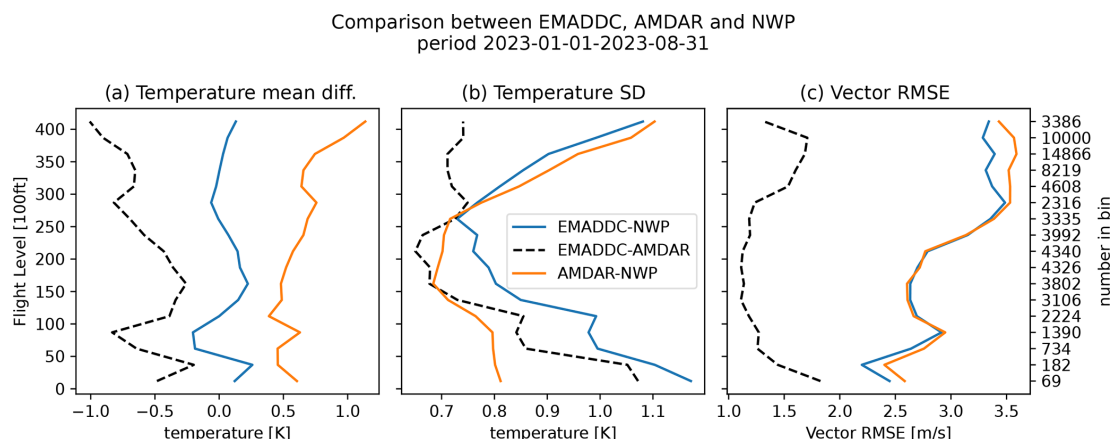


Figure 9. Statistics of an 8-month period of collocated AMDAR and EMADDC observations. Panel (a) shows the mean differences in flight bins of 25 flight levels, panel (b) shows the standard deviation of the differences in temperature, and panel (c) shows the Vector RMSE with respect to height [ft].

tected. A revisit of the heading correction algorithm is foreseen in the near future.

- Applying a general true airspeed correction for both wind and temperature.
- Similarly to temperature measurements, a formal error can be derived for wind observations.
- Information on the formal errors is of great interest for users in data assimilation, and EMADDC is looking into ways to disseminate this information.
- The vast volume of data per time period and region needs proper treatment for use in, for example, data assimilation; moreover, incorporating formal errors correctly could be accounted for. Research within EMADDC is ongoing on how to apply this most efficiently. Possible methods that will be applied are thinning and/or superobbing (i.e. averaging a number of close observations and treating them as one).
- The ADS-B information also contains geodetic height information, which might be valuable for data assimilation. Also, information about the aircraft category and positional accuracy is available. The EMADDC team intends to add this information to the observation set created.
- The current system is file-based; in the near future a (near)-real-time production is foreseen using the real-time networks of ATC in Europe (NewPENS) and real-time transmission of receiver data to EMADDC from ADSB Support and others.
- As part of an initiative of Met Office and KNMI, EMADDC Met Office Global is processing data from Flightradar24 to enhance “global coverage” (limited to

aircraft trajectories where Mode-S EHS is actively interrogated). The observations generated through this process are made available as CSV, NetCDF, and BUFR files exclusively to National Hydrometeorological Institutes. In time, these observations will be processed within the regular EMADDC system to accomplish synergy for corrections and to prevent data duplication and redundancy.

In the future ADS-B is foreseen to broadcast meteorological information (Rodriguez, 2023) creating enormous data volumes which are expected to require quality control of some kind for use in meteorological applications. The EMADDC system could fulfil the future quality assurance function.

Appendix A: Geomagnetic data

This appendix shows that the linear approximation in datum is valid for the domain of the current EMADDC processing.

The datum 1 January 2015 was set as the reference datum for the declination table in the current processing setup (R2.2 August 2023). The geomagnetic model is used in determining the declination for a given position and time (Maus and Macmillan, 2005). Figure A1 shows the value of the declination on 1 January 2015 (panel a), panel (b) shows the yearly change, and panel (c) shows the difference between declination valid for 1 January 2020 and the linear approximation. The values of magnetic declination in central Europe are small. The change in declination is strongest for high-latitude regions and close to zero for low-latitude regions (panel b). The error made by the linear approximation is small, as can be seen from panel (c).

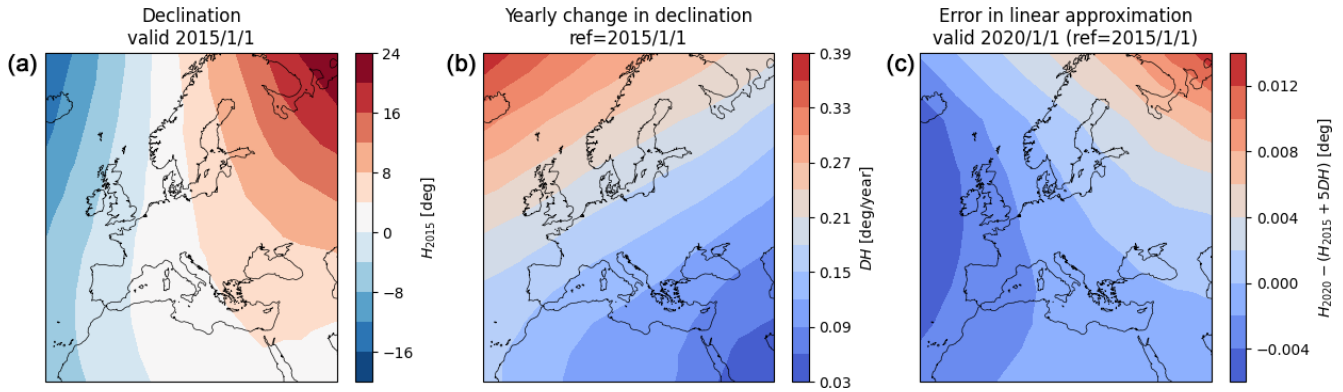


Figure A1. The effect of linearization of the declination around the datum 2015. (a) The declination on 1 January 2015, (b) the yearly change on 1 January 2015, and (c) the difference between linearization and model declination on 1 January 2020. Note that the contours differ for the three panels.

Appendix B: Rounding error

The error due to rounding is estimated as follows. Assume that the error of an measurement X is normally distributed with zero mean around the true value X_t and standard deviation σ . The mean and second moment of the error in X are given by

$$\mathcal{E}(X - X_t) = \int_{-\infty}^{\infty} x \rho_X(x) dx = 0, \quad (\text{B1})$$

$$\mathcal{E}((X - X_t)^2) = \int_{-\infty}^{\infty} x^2 \rho_X(x) dx = \sigma^2. \quad (\text{B2})$$

Let $[X]_r$ be the value of X rounded to the nearest value $r \cdot i$ with $i \in \mathbb{Z}$. The probability of a measurement $[X]_r$ is equal to probability measuring $Y = [X] + R$, with $|R| < \frac{r}{2}$, where the probability density function of R is uniform on the interval $[-\frac{r}{2}, \frac{r}{2}]$. The probability density function of $[X]_r$ is the convolution of the normal distribution of X and uniform distribution of R , that is

$$\mathcal{P}(x < [X]_r - X_t) = \int_{-\infty}^{[X]_r - X_t} \int_{-\frac{r}{2}}^{\frac{r}{2}} \rho_X(y - s) \rho_R(s) ds dy \quad (\text{B3})$$

$$\begin{aligned} \Rightarrow \rho_{[X]_r}(y) &= \int_{-\frac{r}{2}}^{\frac{r}{2}} \rho_X(y - s) \rho_R(s) ds dy \\ &= \int_{-\frac{r}{2}}^{\frac{r}{2}} \rho_X(y - s) \frac{1}{r} ds dy; \end{aligned} \quad (\text{B4})$$

thus the mean and variance of error in $[X]_r$ are

$$\mathcal{E}([X]_r - X_t) = \int_{-\infty}^{\infty} \int_{-\frac{r}{2}}^{\frac{r}{2}} y \frac{1}{r} \rho_X(x) ds dx = 0 \quad (\text{B5})$$

$$\begin{aligned} \mathcal{E}(([X]_r - X_t)^2) &= (\mathcal{E}([X]_r - X_t))^2 \\ &= \int_{-\infty}^{\infty} \int_{-\frac{r}{2}}^{\frac{r}{2}} (t + s)^2 \rho_X(t) \frac{1}{r} dt dy \\ &= \int_{-\infty}^{\infty} t^2 \rho_X(t) dt + \frac{1}{r} \int_{-\frac{r}{2}}^{\frac{r}{2}} s^2 ds = \sigma^2 + \frac{r^2}{12}. \end{aligned} \quad (\text{B6})$$

Appendix C: Data sources

Table C1 presents the different sources used in the current processing, and Fig. C1 shows the coverage of the sources processed in January 2024. Figure C2 shows the number of daily processed observations since 2016. Clearly visible is the sudden decrease in number of observations during the COVID-19 period.

Table C1. Sources of EMADDC in the processing in January 2024.

Source	Affiliation	Main coverage	ATC/local	First data provided
AS-MET	Air Support, Denmark	Europe	local receivers	15 April 2021
AU	Austro Control	Austria	ATC radar	26 September 2018
DK	DMI/NAVIAR	Denmark	ATC radar	13 November 2017
ES	AEMET	Spain	ATC radar	25 June 2019
FR	Météo France	France	local receivers	8 September 2020
IL	Israel Meteorological Service	Israel	local receivers	1 May 2023
MUAC	Maastricht Upper Area Control Centre	Benelux	ATC radar	January 2014
NO-FFI	MET Norway/FFI	Norway	local receivers	3 July 2021
RO	ROMATSA	Romania	ATC radar	1 October 2020
SE	SMHI	Sweden	local receivers	7 June 2021
SI	ARSO	Slovenia	ATC radar	8 September 2020
UK	Met Office	United Kingdom	local receivers	1 February 2020

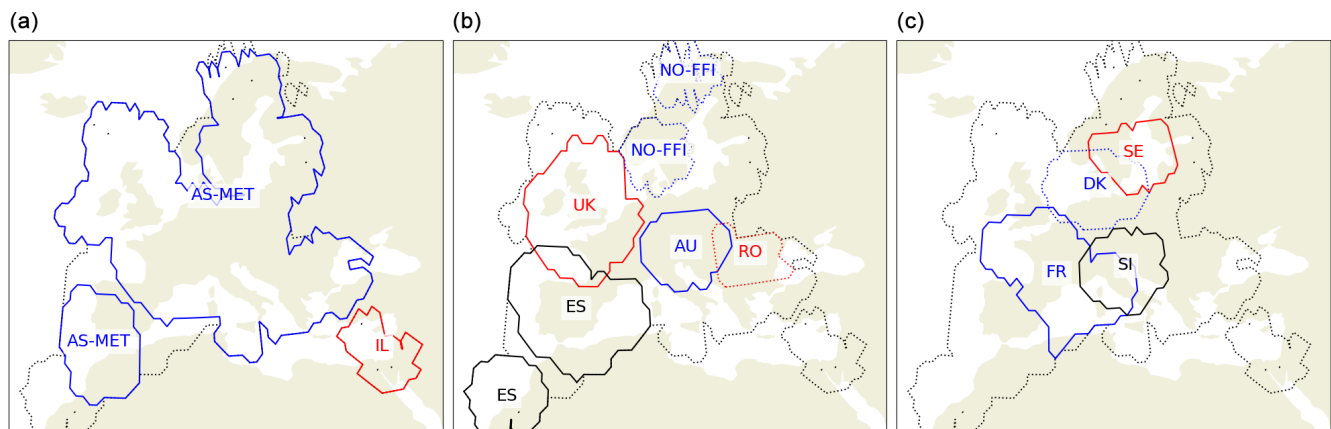


Figure C1. Coverage maps of individual sources: (a) sources AS-MET and IL; (b) ES, UK, NO-FFI, and AU; and (c) FR, DK, SE, and SI.

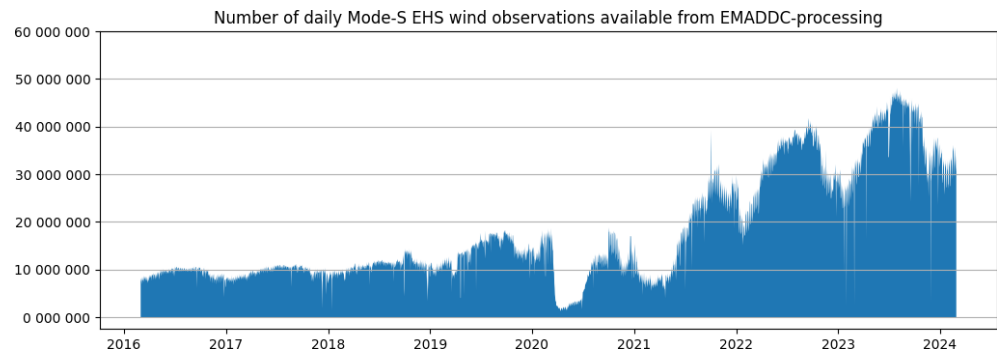


Figure C2. The number of observations per day processed by EMADDC over time.

Data availability. The processed historical data used in this article are available through the KNMI Data Platform at <https://doi.org/10.21944/7ef8-6m27> (de Haan et al., 2025).

Author contributions. SdH drafted the manuscript and built the first version of EMADDC and the heading correction algorithm and quality control. SdH and PdJ further improved the algorithms and quality control. PdJ and MK ported the (research) version to an operational processing system. JS is the EMADDC programme manager and the liaison between research and the operational aeronautical meteorological service provision and is responsible for EMADDC funding and stakeholder management. LS developed the method to improve the Mach number by recalculation using the indicated airspeed. JS, PdJ, and MK provided input for the manuscript draft.

Competing interests. The contact author has declared that none of the authors has any competing interests.

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