



1 **Brief communication: Significant cold bias in ERA5 output for** 2 **McMurdo region, Antarctica**

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8
9 **Abstract.** The ERA5 climate reanalysis dataset plays an important role in applications such as monitoring
10 and modelling climate system changes in polar regions, so the calibration of the reanalysis to ground
11 observations is of great relevance. Here, we compare the 2-meter air temperature time series of the ERA5
12 reanalysis to the near-ground air temperature measured in 17 Automatic Weather Stations in the
13 McMurdo Sound, Antarctica. We find that the reanalysis data has a systematic cold bias of $\sim 5^{\circ}\text{C}$. Our
14 results show that future work should rely on secondary observations to calibrate when using the ERA5
15 reanalysis in polar regions.

16
17 **Short Summary.** By analyzing temperature time series over more than 20 years, we have found a
18 discrepancy between the 2-meter temperature values reported by the ERA5 reanalysis and the Automatic
19 Weather Stations in the McMurdo Sound, Antarctica. The ERA5 reanalysis temperatures are
20 systematically colder by $\sim 5^{\circ}\text{C}$.

21 **1. Introduction**

22
23 ERA5 dataset represents the fifth iteration of ECMWF (European Center for Medium-Range Weather
24 Forecasts) global climate hindcasting derived by combination of climate data assimilation and climate
25 simulations (Hersbach et al, 2020). With its global coverage, high temporal resolution, and relatively high
26 spatial resolution of 31 km this dataset may prove particularly useful for research in polar regions such as



27 Antarctica, where long-term climate observations are geographically sparse and often temporally
28 discontinuous (Lazzara et al, 2012). A recent study found encouraging agreement between ERA5 output
29 and AWS (Automatic Weather Station) data from 13 stations located in the southern section of Antarctic
30 Peninsula (Tetzner et al., 2019).

31

32 Here, we report the results of a comparison between monthly 2-meter air and ground temperatures in the
33 McMurdo region, Antarctica, reported in the ERA5 dataset and corresponding observations from 17 AWS
34 locations across this region. We focus our analysis on this region because of the relatively high spatial
35 and temporal coverage of AWS observations and due to the high multidisciplinary research interest in
36 this region which contains the main USA and New Zealand research stations and is proximal to Italian
37 and Korean research stations. Despite the encouraging results found by Tetzner et al. (2019) for the
38 southern Antarctic Peninsula, we find a significant cold bias in ground temperatures and air temperatures
39 in our study area.

40 **2. Data and methods**

41 We analyze the daily surface temperature (2-meter temperature) recorded at 17 AWS (Figure 1) managed
42 by the McMurdo Dry Valleys Long Term Ecological Research Project (LTER) since 1992, although some
43 of the stations have been reporting data only since 1986 (Doran et al., 2002;). We compare the AWS data
44 to the monthly ECMWF ERA5 climate reanalysis surface temperature data (Muñoz Sabater, 2019) and
45 we also we tested against the near-surface bias-corrected reanalysis dataset (Cucchi et al., 2022). For each



46 LTER AWS, where daily 2-meter air temperature data was available, we ran a 30-day moving average
47 filter with 0% overlap to obtain monthly time series. The ERA5 grid node used in comparisons to each
48 individual AWS was selected by minimizing the haversine distance between each AWS and all the nodes
49 in the reanalysis grid. Finally, we interpolated both time series to a regular monthly sequence. The time
50 series for the ERA5 node data were truncated to match the periods where data was available at their
51 corresponding AWS. We report the mean temperature for the span of each time series and the standard
52 error of the mean for each sample.

53 Furthermore, we compare the two data sets by analyzing the correlograms and performing a linear
54 regression. Figure 2.b shows an example of this comparison. We report the squared correlation
55 coefficients (R^2) as a metric of the goodness of fit and the p-values from the F-statistic to assess the level
56 of statistical significance.

57 **3. Results**

58 Table 1 summarizes the results of our comparison. Even though some of the largest differences in the
59 mean between the AWS station and the closest ERA5 grid node are observed for stations at high altitudes
60 (e.g., Beacon Valley and Mt. Fleming) there are other stations at relatively high altitudes that report a
61 smaller bias (e.g., Friis Hills) and conversely, the Taylor Valley AWS station is not at a high altitude, but
62 it does report a significant bias. The only station where the ERA5 average temperature was warmer than
63 the corresponding AWS was located at Lake Vida, in Wright Valley.



Table 1. List of available AWS in the McMurdo Dry Valleys region and comparison to ERA5 closest node.

AWS Location name	AWS ID	Latitude	Longitude	Elevation (m.a.s.l.)	Distance to ERA5 node (km)	AWS data date range	Avg 2m air temperature @ AWS	Avg 2m air temperature @ ERA5 node	AWS _{mean_temp} - ERA5 _{mean_temp}
Beacon Valley	BENM	-77.828	160.6569	1176.0	3.27	2000-12-11 - 2012-11-19	-21.48 ± 0.7	-33.39 ± 0.69	11.91 ± 1.39
Lake Bonney	BOYM	-77.7147	162.4646	64.0	1.84	1993-12-08 - 2018-10-09	-17.26 ± 0.61	-23.85 ± 0.43	6.59 ± 1.04
Lake Brownworth	BRHM	-77.4344	162.7036	279.0	3.83	1995-01-23 - 2018-11-10	-19.94 ± 0.66	-25.28 ± 0.52	5.34 ± 1.18
Canada Glacier	CAAM	-77.6133	162.9644	264.0	1.71	1994-12-18 - 2011-01-05	-16.36 ± 0.72	-22.93 ± 0.61	6.57 ± 1.33
Commonwealth Glacier	COHM	-77.5646	163.2823	290.0	3.96	1993-12-06 - 2018-10-30	-17.69 ± 0.47	-21.94 ± 0.51	4.25 ± 0.98
Explorer's Cove	EXEM	-77.5887	163.4175	25.0	1.32	1997-12-05 - 2018-11-23	-18.97 ± 0.7	-21.51 ± 0.55	2.54 ± 1.25
Mt. Fleming	FLMM	-77.5327	160.2714	1870.0	3.7	2011-01-22 - 2018-11-11	-24.2 ± 0.58	-33.84 ± 0.76	9.65 ± 1.34
Lake Fryxell	FRLM	-77.6113	163.1701	19.0	1.45	1994-12-12 - 2018-11-19	-19.78 ± 0.7	-22.22 ± 0.51	2.44 ± 1.21
Friis Hills	FRSM	-77.7474	161.5162	1591.0	5.28	2011-01-04 - 2018-11-06	-22.56 ± 0.63	-26.69 ± 0.75	4.13 ± 1.38
Garwood Ice Cliff	GAFM	-78.0259	164.1315	51.0	2.97	2012-01-24 - 2012-12-19	-16.66 ± 2.79	-23.49 ± 2.28	6.84 ± 5.07
Howard Glacier	HODM	-77.6712	163.0773	472.0	3.25	1993-12-04 - 2018-10-31	-17.18 ± 0.44	-20.6 ± 0.47	3.42 ± 0.91
Lake Hoare	HOEM	-77.6254	162.9005	77.0	2.82	1987-11-25 - 2018-11-29	-17.61 ± 0.51	-23.53 ± 0.42	5.92 ± 0.93
Miers Valley	MISM	-78.1011	163.7877	51.0	0.31	2012-02-11 - 2018-11-06	-16.69 ± 0.97	-23.1 ± 0.91	6.41 ± 1.88
Taylor Glacier	TARM	-77.74	162.1314	334.0	4.51	1994-12-05 - 2018-11-05	-16.9 ± 0.5	-25.23 ± 0.43	8.34 ± 0.93
Upper Howard	UHDM	-77.686	163.145	N/A	1.89	2001-11-28 - 2003-12-24	-16.56 ± 1.49	-20.15 ± 1.73	3.59 ± 3.22



Lake Vanda	VAAM	-77.5257	161.6913	296.0	2.87	1994-12-08 - 2018-12-07	-19.58 ± 0.75	-24.96 ± 0.44	5.38 ± 1.19
Lake Vida	VIAM	-77.3778	161.8007	351.0	2.47	1995-12-08 - 2018-11-14	-26.68 ± 0.96	-23.93 ± 0.48	-2.74 ± 1.44



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67 Figure 2.a illustrates the comparison of AWS and ERA5 monthly temperature time series for one of 17
68 locations used in this study (Lake Hoare) over the time span of two decades. The cold bias is clearly
69 visible and persistent throughout the period covered by the AWS data. The monthly temperature mismatch
70 is particularly large during the summer months, when observations indicate actual temperatures were up
71 to +10°C higher than ERA5 temperatures (e.g., Figure 2b). Over the rest of the year the mismatch shrank
72 to the range of 2-6°C. Figure 2.b suggests that there is a strong seasonality in the relationship between the
73 data sets. During the austral Winter and Summer seasons the temperatures are generally closely clustered
74 together, systematically being closer correlated during the Winter and more dispersed during the Summer.
75 The Spring and Fall seasons show a hysteresis that is repeated over all the comparisons. As the
76 environment warms up during the Spring months the ERA5 temperatures are above the best-fit line and
77 drop below it during the Fall. These seasonal biases may ultimately be helpful in revealing what climate
78 processes must be better represented in the ERA5 reanalysis to eliminate the strong observed temperature
79 bias.

80 **4. Discussion**

81 Our results differ significantly from the findings reported by Tetzner et al. (2019) for the Southern
82 Antarctic Peninsula - Ellsworth Land region. For that region there is a slight cold bias of the ERA5 surface
83 temperatures close to the coast ($-0.51^{\circ}\text{C} \pm 0.74$) and a slight warm bias in the mountain range escarpment



84 (+0.14°C ± 0.72) which has encouraging implications for using the reanalysis data where there is no AWS
85 coverage, which represents most of Antarctica. In contrast, we find no clear topographic dependence on
86 the temperature differences between AWS and ERA5 data, even though the largest differences are indeed
87 in two high altitude areas, Beacon Valley and Mount Fleming. The magnitude of the overall cold bias
88 (average of all differences) is more than an order of magnitude larger (5.33 ± 0.76 °C) as compared to the
89 study of Tetzner et al. (2019) and seems to be systematic. The ERA5 temperatures show a large overshoot
90 during the summer, with an average difference of 6.7 ± 0.8 °C (e.g., Figure 2). This may be a particularly
91 significant problem given the fact that warm summer temperatures determine the annual melt rate of
92 snow, glaciers, and permafrost in Antarctica. The bias presented here is also present when using the near-
93 surface bias-corrected reanalysis dataset (Cucchi et al., 2022). Modelling of snow or ice melting driven
94 by ERA5 temperatures (e.g., Costi et al., 2018) with a strong cold bias, as observed in our study region,
95 will result in a significant underestimate of summer melt production. Although the ERA5 reanalysis is
96 an outstanding source of global climate variables, the discrepancy between our results and those obtained
97 by Tetzner et al. (2019) suggests that secondary observations should be used to test the reliability of the
98 ERA5 dataset in polar regions.

99 **5. Conclusions**

100 We have compared the surface temperature (2-meter temperature) recorded at 17 AWS in the McMurdo
101 Dry Valleys, Antarctica with temperatures from the ERA5 reanalysis dataset. We found that the
102 temperatures reported by the global climate reanalysis are, on average, 5.34 ± 0.76 °C colder than the
103 temperatures recorded at the permanent weather stations. The cold temperature bias appears to be the



104 largest during the warm summer months (6.7 ± 0.8 °C), when loss of snow and ice to melting is the largest.
105 We advise using secondary observations to assess the accuracy of parameters included in ERA5 reanalysis
106 for polar regions.

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108 *Data availability.* The AWS data were provided by the NSF-supported McMurdo Dry Valleys Long Term
109 Ecological Research program (OPP-1637708) and can be accessed at:
110 <https://mcm.lternet.edu/meteorological-stations-location-map>. The “ERA5-Land hourly data from 1950
111 to present” (DOI: [10.24381/cds.e2161bac](https://doi.org/10.24381/cds.e2161bac)) and the “Near surface meteorological variables from 1979 to
112 2019 derived from bias-corrected reanalysis” (DOI: [10.24381/cds.20d54e34](https://doi.org/10.24381/cds.20d54e34)) were downloaded from the
113 Copernicus Climate Change Service (C3S) Climate Data Store.
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115 *Author contributions.* ST conceived the study. RGG performed the analysis. RGG and ST prepared the
116 manuscript with equal contributions.

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118 *Competing interests.* The authors declare that they have no conflict of interest.

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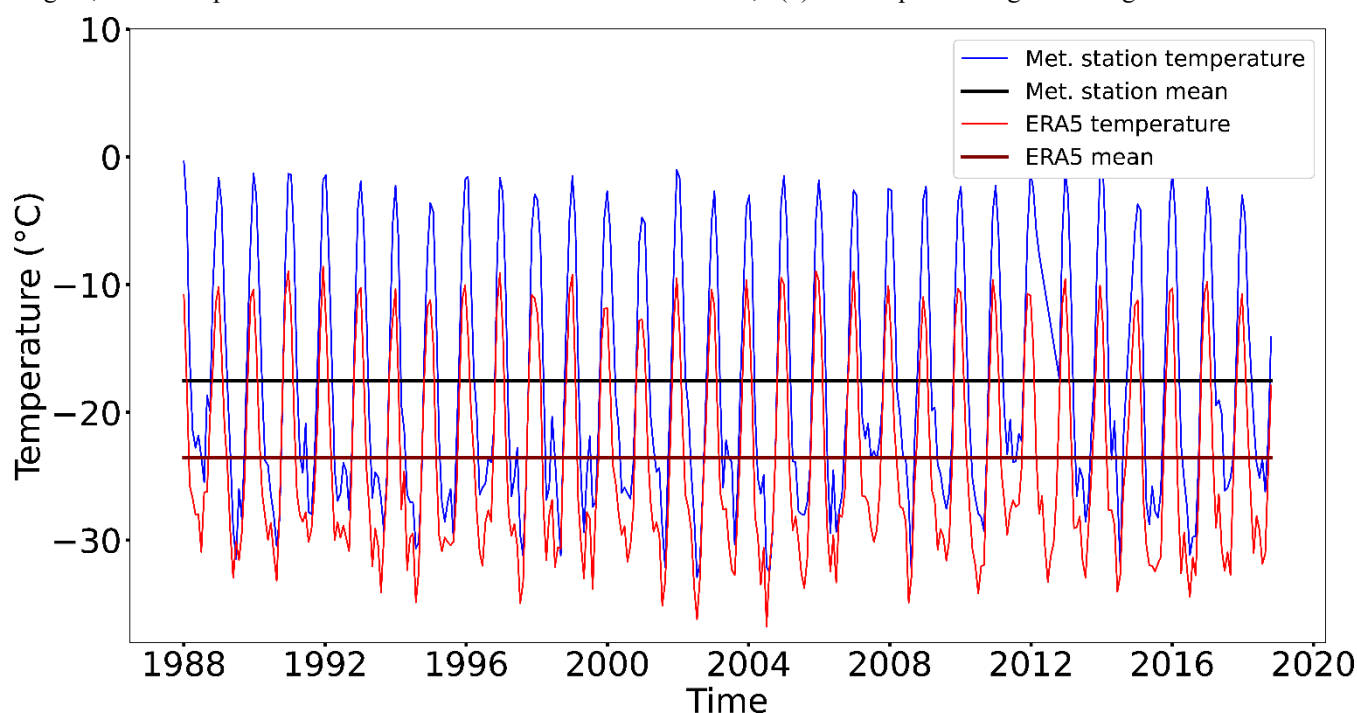
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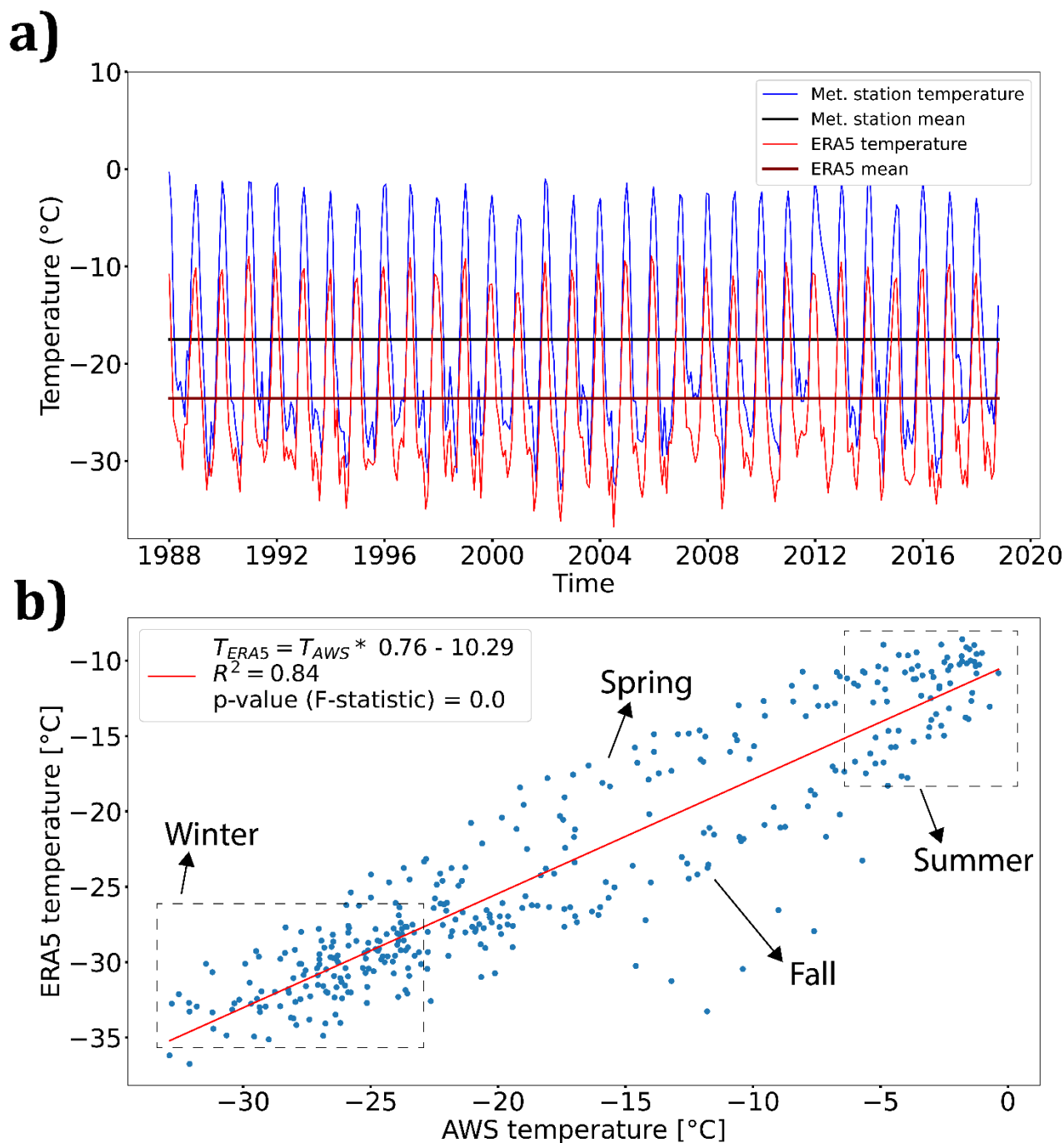
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141 **Figure 1.** Map of the McMurdo Dry Valleys region showing the location of the automatic weather
142 stations (AWS) managed by LTER and their corresponding closest ERA5 grid node.
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145 **Figure. 2** a) Comparison of the monthly averaged surface temperature time series recorded at station
146 Lake Hoare (HOEM) (blue) and the values from the closest grid node of the ERA5 reanalysis (red). b)
147 Correlogram showing the best fit line to the relationship between the AWS temperatures and the
148 reanalysis temperatures. Note the seasonal variation in the relationship.