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Journal Name	Environmental Monit	foring and Assessment
Corresponding Author	FamilyName	Almeida
	Particle	de
	Given Name	Ceres Duarte Guedes Cabral
	Suffix	
	Division	Dom Agostinho Ikas Agricultural School
	Organization	Federal Rural University of Pernambuco
	Address	São Lourenço da Mata, Pernambuco, Brazil
	Phone	
	Fax	
	Email	ceres.cabral@ufrpe.br
	URL	
	ORCID	http://orcid.org/0000-0001-6073-3853
Author	FamilyName	Araújo
	Particle	de
	Given Name	Caio Sérgio Pereira
	Suffix	
	Division	Department of Agricultural Engineering
	Organization	Federal Rural University of Pernambuco
	Address	Recife, Brazil
	Phone	
	Fax	
	Email	caiosergio.ufersa@gmail.com
	URL	
	ORCID	http://orcid.org/0000-0002-6747-7028
Author	FamilyName	Silva
	Particle	e
	Given Name	Ivis Andrei Campos
	Suffix	
	Division	Department of Agricultural Engineering
	Organization	Federal Rural University of Pernambuco
	Address	Recife, Brazil
	Phone	
	Fax	
	Email	ivisandrei@gmail.com
	URL	
	ORCID	http://orcid.org/0000-0002-9555-4238
Author	FamilyName	Ippolito
	Particle	
	Given Name	Matteo
	Suffix	
	Division	Department Agricultural, Food and Forest Sciences (SAAF)

	Organization Address Phone	University of Palermo Palermo, Italy
	Fax Email URL ORCID	matteo.ippolito@unipa.it http://orcid.org/0000-0001-6919-5000
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Abstract	This study aimed to analy 10 years (01/01/2011 to 1 different cities in the state air temperature data were validated using statistical ERA5-Land reanalysis ag highest accuracy of the av Ibimirim ($R2 = 0.98$), whi mean-square error (RMSI Pernambuco. The highest from the city of Caruaru (Recife (0.41 °C). Accordi temperature data compare highest underestimation (overestimated the tempera Therefore, based on the re air temperature for the state	ze the average air temperature data estimated by ERA5-Land reanalysis over 2/31/2020), based on data from 12 automatic weather stations located in c of Pernambuco, northeast of Brazil. For more careful evaluation, the average stratified by mesoregions of the state. ERA5-Land reanalysis data were indices that evaluated the accuracy. The average air temperature estimated by ree well with weather stations in almost the entire state of Pernambuco. The verage air temperature estimated by ERA5-Land occurred in the city of le the lowest accuracy was measured in the city of Caruaru ($R2 = 0.57$). Root-E) generated by the ERA5-Land reanalysis was lower than 0.60 °C in most RMSE of ERA5-Land for average air temperature was calculated using data 1.11 °C), whereas the lowest RMSE was obtained with data from the city of ng to mean bias errors, (MBE) the ERA5-Land underestimated the average air to automatic weather stations data, especially in Ouricuri, which had the -0.80 °C). On the other hand, among the municipalities where ERA5-Land ature values, the highest overestimation was identified in Garanhuns (0.35 °C). esults of this study, ERA5-Land reanalysis successfully estimated the average te of Pernambuco.
Keywords (separated by '-')	Climatology - ECMWF -	Automatic weather stations - Semiarid
Footnote Information		



¹ Evaluation of air temperature estimated by ERA5-Land ² reanalysis using surface data in Pernambuco, Brazil

- ³ Caio Sérgio Pereira de Araújo[®].
- ⁴ Ivis Andrei Campos e Silva^D · Matteo Ippolito^D ·
- ⁵ Ceres Duarte Guedes Cabral de Almeida¹⁰

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Abstract This study aimed to analyze the average 8 air temperature data estimated by ERA5-Land reanal-9 ysis over 10 years (01/01/2011 to 12/31/2020), based 10 on data from 12 automatic weather stations located in 11 different cities in the state of Pernambuco, northeast 12 of Brazil. For more careful evaluation, the average air 13 temperature data were stratified by mesoregions of 14 the state. ERA5-Land reanalysis data were validated 15 using statistical indices that evaluated the accuracy. 16 The average air temperature estimated by ERA5-Land 17 reanalysis agree well with weather stations in almost 18 the entire state of Pernambuco. The highest accuracy 19 of the average air temperature estimated by ERA5-20 Land occurred in the city of Ibimirim $(R^2 = 0.98)$, 21 while the lowest accuracy was measured in the city of 22

A1 C. S. P. de Araújo · I. A. C. e Silva

- A2 Department of Agricultural Engineering, Federal Rural
- A3 University of Pernambuco, Recife, Brazil
- A4 e-mail: caiosergio.ufersa@gmail.com

A5 I. A. C. e Silva

- A6 e-mail: ivisandrei@gmail.com
- A7 M. Ippolito
- A8 Department Agricultural, Food and Forest Sciences
- A9 (SAAF), University of Palermo, Palermo, Italy
- A10 e-mail: matteo.ippolito@unipa.it
- A11 C. D. G. C. de Almeida (\boxtimes)
- A12 Dom Agostinho Ikas Agricultural School,
- A13 Federal Rural University of Pernambuco,
- A14 São Lourenço da Mata, Pernambuco, Brazil
- A15 e-mail: ceres.cabral@ufrpe.br

Caruaru ($R^2 = 0.57$). Root-mean-square error (RMSE) 23 generated by the ERA5-Land reanalysis was lower 24 than 0.60 °C in most Pernambuco. The highest RMSE 25 of ERA5-Land for average air temperature was calcu-26 lated using data from the city of Caruaru (1.11 °C), 27 whereas the lowest RMSE was obtained with data 28 from the city of Recife (0.41 °C). According to mean 29 bias errors, (MBE) the ERA5-Land underestimated 30 the average air temperature data compared to auto-31 matic weather stations data, especially in Ouricuri, 32 which had the highest underestimation (-0.80 °C). 33 On the other hand, among the municipalities where 34 ERA5-Land overestimated the temperature values, 35 the highest overestimation was identified in Gara-36 nhuns (0.35 °C). Therefore, based on the results of 37 this study, ERA5-Land reanalysis successfully esti-38 mated the average air temperature for the state of 39 Pernambuco. 40

KeywordsClimatology · ECMWF · Automatic41weather stations · Semiarid42

Introduction

The use of meteorological data is characterized as an important information database for society in general with applications in different sectors (de Oliveira Castro AQ2 et al., 2019). An example of this is the study carried out by de Gois et al. (2019) in which the annual and monthly AQ3 variability of rainfall and air temperature were evaluated, 49

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as well as quantified the components of the climatic 50 water balance in the municipality of Resende, state of 51 Rio de Janeiro, in 22 years of meteorological data. In 52 addition, Ventura et al. (2016) with computational sup-53 port applied four methodologies of genetic algorithms 54 derived from artificial intelligence and neural networks to 55 be used to fill gaps in climatological databases. Another 56 example is the study carried out by dos Santos et al. AQ4 (2017) in which used the weather stations databases from 58 the National Institute of Meteorology (INMET) and the 59 National Institute of Space Research (INPE), both in Bra-60 zil, to investigate the relationship between the variables 61 land use and occupation, rainfall, maximum temperature, 62 and relative air humidity in southwestern of the state of 63 Piaui in the 1984–2015 period, supported by images of 64 Landsat 5 and Landsat 8 satellites. Franco et al. (2019) 65 constructed a rainfall zoning for cacao cultivation in the 66 state of Bahia, Brazil, using the fuzzy logic based on his-67 torical series of 519 meteorological stations. 68

According to da Silva et al. (2017), the atmosphere AQ5 has been suffering from high temperatures in recent 70 decades, with a significant increase in this variable 71 from the 1980s due to global warming. Automatic 72 weather stations are considered one of the main tools 73 for monitoring and assessing environmental condi-74 tions of a given locality, consisting mainly of pro-75 viding measurements of parameters that indicate the 76 instantaneous, or average, state of the atmosphere 77 (Lopes et al., 2021). However, the standardization of 78 observation schedule, the weather station location, 79 preventive maintenance operations, calibration pro-80 cedures and measuring instruments testing, and the 81 cost for maintaining these weather stations are some 82 of the reasons that restrict their expansion in Brazil 83 (de Oliveira Castro et al., 2019). The quality of this 84 dataset provides assessing the consequences of natu-85 ral resource management, such as the soil-crop-water 86 relationship in an irrigated area. 87

To date, there are about 786 automatic and con-88 ventional weather stations in Brazil belonging to the 89 INMET network, and 219 of which are in the north-90 east region (146 automatic and 73 conventional). The 91 state of Pernambuco has 13 automatic weather stations 92 and seven conventional weather stations that provide 93 hourly and monthly data of numerous climatic vari-94 ables (INMET, 2021). Weather stations have fast data AQ6 processing of various climate variables that are gen-96 erated by electronic sensors and analog measurement 97

equipment. However, these stations have some limi-98 tations in data acquisition, including reading errors, 99 caused by equipment failures in automatic weather 100 stations and human observation errors in conven-101 tional weather stations (Bier & Ferraz, 2017; Lopes 102 et al., 2021). Consequently, the user should consider 103 the impact of these occurrences on historical series, 104 which are more evident in daily data and may mask 105 the variability of climatic data, resulting in unreliable 106 data. Therefore, the identification and remotion of 107 the discrepancies of the historical series are essential 108 before running the analysis, through homogenization 109 procedures, which are extremely important for studies 110 of climatic variability (Azorin-Molina et al., 2016). 111

In recent years, the use of climatological data from 112 global atmospheric models has become an important 113 alternative for assessment and environmental monitor-114 ing conditions. This database provides measurements 115 of several variables that can help minimize the scar-116 city of observations and historical data series in sur-117 face weather stations. Thus, the reanalysis technique of 118 climatological data is an effective alternative for many 119 scientific and operational studies in recent decades (de 120 Oliveira Castro et al., 2019; Moraes et al., 2012). 121

Reanalysis is obtained through a process that 122 involves data assimilation, mainly based on data 123 observed in global weather stations and on numerical 124 models of forecasting aiming to estimate the weather 125 conditions with greater consistency (Parker, 2016). 126 Consequently, reanalysis is among the most used 127 data sets in climatological and environmental studies 128 at different scales for the most distinct purposes (de 129 Oliveira Castro et al., 2019). However, this data set 130 suffers interference from climate and local topogra-131 phy due to the spatial resolution, which is considered 132 a disadvantage for reanalysis (Chen et al., 2021). 133

The Tiled European Centre for Medium-Range 134 Weather Forecasts (ECMWF) Scheme for Surface 135 Exchanges over Land (TESSEL) is used operation-136 ally in the Integrated Forecast System (IFS) for 137 explaining the development of soil, vegetation, and 138 snow at various spatial resolutions over Earth. A 139 revised land surface hydrology (HTESSEL) is intro-140 duced in the ECMWF operational model, to report 141 failings of the land surface scheme, specifically the 142 lack of surface runoff and the choice of a global uni-143 form soil texture (Hersbach et al., 2020). ERA5 pro-144 duced by the Copernicus Climate Change Service 145

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146 (C3S) is the fifth generation ECMWF atmospheric 147 reanalysis of the global climate covering the period 148 from January 1950 to the present, with spatial res-149 olution $0.25^{\circ} \times 0.25^{\circ}$ and the high temporal scan 150 hourly available also aggregate at monthly scale.

ERA5-Land released in July 2019 is an improved 151 version of ERA5 that was created by forcing the ter-152 restrial component of ERA5 reanalysis, covering the 153 same period. The ERA5-Land product is delivered 154 at the same time resolution as ERA5 but with a spa-155 tial resolution of 0.1°×0.1°against 0.25°×0.25° of 156 ERA5 (C3S, 2021) and without data available for the 157 oceans. Thus, the reanalysis technique has become a 158 monitoring system designed to estimate many atmos-159 pheric and land climate variables. 160

The air temperature estimated by ERA5-Land 161 reanalysis has been widely studied. Liu et al. (2021) 162 analyzed the spatial-temporal consistency of Earth 163 surface temperature trends derived from satellites 164 (AIRS and MODIS) and ERA5-Land reanalysis. Cao 165 et al. (2020) used the ERA5-Land reanalysis database 166 to assess the accuracy of soil temperature in perma-167 frost regions. Graham et al. (2019) evaluated five dif-168 ferent types of atmospheric reanalysis in an Arctic 169 portal and concluded that all products well estimated 170 air temperature, but ERA5 reanalysis provided best 171 performance. 172

The most common climatic variable measured is 173 air temperature due to being easy to obtain and the 174 importance for hydrology and agricultural applications, 175 besides for research in meteorology and climatology 176 and the society in general (Pohlmann & Lazzari, 2018). 177 Air temperature data from the atmospheric reanaly-178 sis model, such as the ECMWF, can be used to over-179 come the data absence of historical series from surface 180 weather stations. Thus, the objective of this study was 181 to analyze the accuracy of the monthly air temperature 182 data estimated by ERA5-Land reanalysis over 10 years 183 (01/01/2011 to 12/31/2020) in the state of Pernambuco, 184 northeastern Brazil, based on the data set observed in 185 automatic weather stations. The state of Pernambuco 186 is an interesting case study region for the evaluation of 187 the accuracy of the air temperature dataset estimated 188 by ERA5-Land reanalysis since there are three mesore-189 gions with different climate units; as a result, this study 190 could provide valorous information to ERA5-Land 191 application in other regions. 192

Material and methods

The study was conducted in the state of Pernambuco 194 (8°20'24" S, 37°48'36" W), located in northeastern 195 Brazil (Fig. 1a) in a tropical zone, which occupies an 196 area of approximately 98,146 km² and has three dif-197 ferent climatic classes (Tropical monsoon; Tropical 198 savannah and Arid), according to Köppen's classifi-199 cation (Alvares et al., 2013). This region stands out 200 in agricultural production, especially in the sugar-201 cane cultivation and, in recent decades, with the con-202 tinuous increase in the irrigated area cultivated with 203 tropical fruit, for example, grape, mango, guava, and 204 banana. The data set used as a reference to validation 205 of ERA5-Land was provided from automatic weather 206 stations (AWS) distributed throughout the state of 207 Pernambuco, in different mesoregions. These sta-208 tions are situated at the municipalities of Arcoverde, 209 Cabrobó, Caruaru, Floresta, Garanhuns, Ibimirim, 210 Ouricuri, Petrolina, Serra Talhada, Palmares, Recife, 211 and Surubim, totaling 12 municipalities (Fig. 1b). 212 The data set from the weather station located in the 213 municipality of Salgueiro was not used due to the 214 absence of data in the period defined for this study. 215 In fact, the absence of climatic data was identified at 216 various periods and in several municipalities in the 217 state, with different intensities. Thus, the use of rea-218 nalysis data can be an alternative to address this lack 219 of public information. Hourly air temperature data 220 (°C) were acquired through automatic weather sta-221 tions belonging to the network of the National Insti-222 tute of Meteorology (INMET), covering 10 years 223 (01/01/2011 to 12/31/2020). 224

The second database of air temperature adopted in 225 this study was provided by ERA5-Land atmospheric 226 reanalysis produced by the European Center for 227 Medium-Range Weather Forecast (ECMWF) through 228 virtual stations corresponding to the same cities and 229 the same period as automatic weather stations, on a 230 monthly scale with a spatial resolution of 0.10° and 231 measured in Kelvin (Fig. 1c). The monthly data of air 232 temperature were converted from Kelvin to degrees 233 Celsius to unify both databases and enable the global 234 atmospheric model validation. The geographic infor-235 mation of the weather station is presented in Table 1, 236 in addition, to the percentage of missing data in each 237 AWS. AQ7



Fig. 1 Location of the Pernambuco state in northeastern Brazil (a); Spatial distribution of automatic weather stations belonging to National Institute of Meteorology (INMET) net-

The absence of data observed in some periods and 239 locations in the historical series analyzed required 240 the application of a gap-filling methodology known 241 as the simple average method. For this purpose, 242 hourly measurements of automatic weather stations 243 from other geographically close locations, which had 244 similar climatic characteristics, were used to obtain 245 the arithmetic mean and estimate the missing value 246 according to respective hour (Kashani & Dinpashoh, 247 2012). The monthly average air temperature was cal-248 culated from the hourly data, using the arithmetic 249 mean for each month of the year. 250

The performance of ERA5-Land reanalysis in estimating air temperature data was evaluated using statistical indices of accuracy, adopting the data from automatic weather stations as standard series. For more careful evaluation, the data set of average air work in the state of Pernambuco, identified by the city and the respective World Meteorological Organization, WMO code (b); Spatial distribution of ECMWF virtual stations (c)

temperature were compared by city and grouped by 256 climatic mesoregions. Thus, the root-mean-square 257 error (RMSE) (Eq. (1)) and the mean bias error 258 (MBE) (Eq. (2)) were calculated. In addition, the 259 graphical analysis of the data dispersion around the 260 1:1 line was performed, and the corresponding coef-261 ficient of determination (R^2) was obtained (Eq. (3)), 262 which indicates the data set accuracy degree, ranging 263 from 0 to 1; the closer to 1, the better the fit of the 264 evaluated model. The temperature observed is that 265 measured by the electronic sensor at the automatic 266 weather station and the estimated temperature is that 267 derived from the ERA5-Land reanalysis. 268

$$RMSE = \sqrt{\frac{\sum_{i=1}^{N} \left(T_{obsi} - T_{esti}\right)^2}{N}}$$
(1)

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 Table 1 Geographical characteristics of the automatic weather stations used in this study

Station code ^a	City	Elevation (m)	Mesoregion	Climatic classification ^b	Latitude (S)	Longitude (W)	Missing observations (%)
A307	Petrolina	372.5	Semiarid	BSh	9.39°	40.52°	2.74
A309	Arcoverde	683.9	Semiarid	Aw	8.43°	37.05°	3.66
A329	Cabrobó	342.8	Semiarid	BSh	8.50°	39.31°	4.20
A349	Ibimirim	434.2	Semiarid	Aw	8.51°	37.71°	2.52
A350	Serra Talhada	499.0	Semiarid	BSh	7.95°	38.29°	3.18
A351	Floresta	327.4	Semiarid	BSh	8.60°	38.60°	4.20
A366	Ouricuri	462.0	Semiarid	BSh	7.88°	40.10°	11.66
A322	Garanhuns	827.8	Agreste	Aw	8.91°	36.49°	5.17
A328	Surubim	421.4	Agreste	Aw	8.84°	35.80°	6.67
A341	Caruaru	852.0	Agreste	Aw	8.36°	36.03°	7.52
A301	Recife	11.3	Forest/Coastal	Am	8.06°	34.96°	2.04
A357	Palmares	164.0	Forest/Coastal	Aw	8.67°	35.57°	9.30

Am tropical, monsoon, Aw, tropical, savannah, BSh, arid, desert, hot

^aWorld Meteorological Organization (WMO)

^bAlvares et al. (2013)

²⁷¹
$$MBE = \frac{1}{N} \sum_{i=1}^{N} (T_{esti} - T_{obsi})$$
 (2)

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$$R^{2} = \left[1 - \frac{(T_{obsi} - T_{esti})^{2}}{(T_{obsi} - \overline{T})^{2}}\right]$$
274

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where T_{obsi} represents the monthly observed temperature (°C); T_{esti} is the monthly estimate temperature (°C); and N is the number of the observations available for analysis (*i*=1, 2, ..., *N*); \overline{T} is the observed average values (°C).

RMSE expresses error data the magnitude of the 280 error between estimated values by ERA5-Land and 281 measured/observed, ranging from zero to ∞ , and 282 the closer to zero, the better the model (Janssen 283 & Heuberger, 1995). The mean bias error (MBE) 284 index expresses the mean deviation of the estimated 285 values relative to the observed values, indicating the 286 model tends to overestimate or underestimate; thus, 287 the closer to zero, the lower the magnitude of the 288 model systematic error (Leite & de Andrade, 2002). AQ8 Statistical analyses were performed by the R 290 291 software (R CoreTeam, 2020), which uses a multiparadigm programming language to provide, manip-292

²⁹³ ulate, analyze, and visualize data.

Results

(3)

The average monthly air temperature was calculated 295 from the hourly data observed in each automatic 296 weather station, during the period from 2011 to 297 2020 (Fig. 2). The average monthly air temperature 298 amplitude in the Semiarid mesoregion ranged from 290 20.6 to 29.2 °C in the municipalities of Arcoverde in 300 July and Cabrobó in November, respectively. While 301 in the Agreste mesoregion, the average air tempera-302 ture values ranged from 19.0 to 25.6 °C in Garan-303 huns in August and Surubim in March, respectively. 304 Finally, in the Forest/Coastal mesoregion, the aver-305 age monthly air temperature ranged from 22.6 to 306 26.9 °C, in Palmares in July and Recife in December, 307 respectively. 308

The average monthly air temperature in the state 309 of Pernambuco varied according to the geographical 310 location of the municipalities (longitude), distance 311 from water bodies (rivers and lakes), and elevation. 312 These data show that there has been a slight decrease 313 in the monthly air temperatures in the municipalities 314 located at higher elevations, whereas municipali-315 ties in the Semiarid mesoregion have high average 316 monthly air temperatures. 317

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Fig. 2 Monthly averages of air temperature (°C) measured by automatic weather stations in the state of Pernambuco, from 2011 to 2020



The average monthly air temperature values estimated by ERA5-Land, in general, were overestimated, whatever of the mesoregion in the state of Pernambuco. Figure 3 shows the error (E) caused 321 by the difference between the average monthly air 322 temperature estimated in the ERA5-Land reanalysis 323



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Table 2 Performance statistical indices of	Statistical	Municipalities					
ERA5-Land reanalysis	indices	Ouricuri	Floresta	Palmares	Petrolina	Cabrobó	Ibimirim
temperature compared	R^2	0.92	0.96	0.96	0.96	0.96	0.98
to the automatic weather	RMSE (°C)	0.96	0.80	0.60	0.53	0.54	0.43
stations in the state of Pernambuco, for the period	MBE (°C)	-0.80	-0.67	-0.53	-0.40	-0.37	-0.32
from 2011 to 2020		Arcoverde	Caruaru	Garanhuns	Serra Talhada	Recife	Surubim
	R^2	0.94	0.57	0.93	0.93	0.93	0.94
	RMSE (°C)	0.58	1.11	0.57	0.57	0.41	0.47
<i>RMSE</i> root-mean-square	MBE (°C)	-0.23	0.13	0.35	0.22	-0.27	-0.22

and the average monthly air temperature calculated 324 using the data recorded by automatic weather sta-325 tions over 10 years. The largest underestimation and 326 327 overestimation generated by the ERA5-Land reanalysis in the state of Pernambuco occurred in the 328 Semiarid mesoregion, in the municipalities of Serra 329 Talhada (-0.68 °C) in April and Floresta (1.22 °C) 330 in September, respectively (Fig. 3a). In the Agreste 331 mesoregion (Fig. 3b), the largest underestimation 332 of ERA5-Land was -0.60 °C, in August, in Gara-333 nhuns, while the highest overestimation occurred 334 in the municipality of Surubim, in April (0.42 °C). 335 In the Forest/Coastal mesoregion, ERA5-Land 336 data overestimated monthly air temperature by up 337 to 0.64 °C in the cities of Recife and Palmares in 338 339 November (Fig. 3c).

ERA5-Land successfully simulates the average 340 air temperature throughout the state of Pernambuco, 341 according to performance statistical indices (Table 2). 342 In general, the highest values of R^2 correspond to the 343 lowest values of RMSE, and the opposite is equally 344 true. The estimate of average air temperature by 345 ERA5-Land reanalysis obtained accuracy higher than 346 90% in almost the entire state of Pernambuco, espe-347 cially for the city of Ibimirim ($R^2 = 0.98$) located in 348 the Semiarid mesoregion (Table 2) while the lowest 349 accuracy of ERA5-Land was identified in Caruaru 350 351 $(R^2 = 0.57)$ in the Agreste mesoregion.

According to RMSE values, lower than 0.60 °C in 352 most of the state of Pernambuco and with a minimum 353 value of 0.41 °C (Table 2), the ERA5-Land reanalysis 354 presents suitability to estimated monthly air temperature. 355 Again, the city of Caruaru (Agreste mesoregion) resulted 356 in the largest error (RMSE=1.11 °C) followed by Floresta 357 and Ouricuri (0.80 °C and 0.96 °C, respectively), both 358 municipalities located in the Semiarid mesoregion. 359

The mean bias error (MBE) indicates that ERA5-360 Land reanalysis generated underestimated data of 361 average air temperature compared to the data observed 362 in the AWS in the Semiarid and Forest/Coastal mes-363 oregions of Pernambuco (Table 2). This result is con-364 firmed in the data graphical analysis (Fig. 4), which 365 shows that of the 12 automatic weather stations used 366 in this study, ERA5-Land underestimated the average 367 air temperature values in nine of them, correspond-368 ing to 75% of the stations. The highest underestima-369 tion occurred in Ouricuri (-0.80 °C), in the Semiarid 370 region of Pernambuco, while the highest overestima-371 tion of ERA5-Land occurred in Garanhuns (0.35 °C), 372 in the Agreste mesoregion. 373

In the graphical analysis of the municipality of Caru-374 aru (Fig. 4), it can be noticed that there was a change 375 in the relationship between data observed in the AWS 376 and those estimated by the ERA5-Land, overestimat-377 ing in a certain period of monitored time and underes-378 timating in another period. This finding encouraged the 379 monthly analysis of each year, over the 10 years, and 380 thus, it was identified that this change in trend occurred 381 from July/2018 (Fig. 5, dashed line). According to 382 information obtained from INMET, this AWS was ini-383 tially installed in the central part of the city, with 570 m 384 (-08.23 S, -35.98 W); however, at 07/18/2018, this 385 AWS was transferred to an area with preserved veg-386 etation with 852 m (-08.35S, -36.02 W) a little far 387 from the city. This place transfer absolutely justifies 388 the change mentioned above, since lower temperature 389 data started to be recorded, distancing a little further 390 from those estimated by ERA-5-Land, underestimating 391 around 2 °C from July 2018. 392

The means of R^2 , RMSE, and MBE for all stations 393 corresponded to 0.92, 0.63 °C, and -0.26 °C, respectively, indicating the quality of fit and agreement 395



Estimated Air Temperature (°C)

Fig. 4 Monthly averages of air temperature measured by automatic weather stations (observed) and estimated by ERA5-Land in municipalities in the state of Pernambuco, in the period from 2011 to 2020

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between ERA5-Land reanalysis and most automatic 396 weather stations. The association between eleva-397 tion and temperature estimation errors is identified 398 (Fig. 6). Specifically, Fig. 6 shows the digital terrain 399 model (DTM) of Pernambuco, as well as the RMSE 400 (Fig. 6a) and MBE (Fig. 6b) values associated to air 401 temperature, computed in each weather station. In 402 terms of RMSE and MBE, the best performance was 403 obtained in weathers station located in proximity of 404 the coastal region, with RMSE minimum and MBE 405 close to zero. On the other hand, both statistical indi-406 cators are worse for inland weather stations, probably 407 due to the orographic characteristics of the terrain. 408

409 Discussion

This study analyzed the air temperature estimated by 410 ERA5-Land based on data from 12 automatic weather 411 stations, belonging to the National Institute of Mete-412 orology (INMET) network, in the state of Pernam-413 buco (98,146 km²), northeast of Brazil. However, 414 the density of this network is relatively low compar-415 ing to earlier studies, such as Valeriano et al. (2019) 416 analyzing data from 10 stations in southeastern Brazil 417 (10,014 km²), Moraes et al. (2012) studying simulated 418

decennial data of precipitation, maximum tempera-419 ture, and minimum air in São Paulo (248,209 km²) 420 with 33 weather stations, Pelosi et al. (2020) with data 421 from 18 stations in the Campania region (14,000 km²), 422 in Italy, and Paredes et al. (2018) with data from 24 423 stations in Portugal (92,200 km²) and 42 agrometeoro-424 logical stations distributed over Sicily (25,711 km²), 425 Italy (Negm et al., 2017). Therefore, one of the limita-426 tions is the low availability of specialized meteorologi-427 cal data, measured on the surface, which becomes one 428 of the main motivations for increasing the application 429 of reanalysis data and this study seeks to obtain data 430 that will help to address these research gaps. 431

The ERA5-Land showed, in general, good perfor-432 mance in estimating the average air temperature through-433 out the state of Pernambuco. According to Pelosi et al. 434 (2020), this performance can be explained by the fact that 435 air temperature is one of the variables predicted by math-436 ematical modeling since it can be directly estimated by 437 the numerical integration of the laws describing atmos-438 phere dynamics. Therefore, it can be better predicted than 439 diagnostic variables, which are determined from observa-440 tions in surface weather stations. 441

The RMSE values ranged from 0.41 to 1.11 °C, 442 which was considered slightly lower than those found 443 in the previous studies. Valeriano et al. (2019), studying 444



Fig. 6 Digital terrain model (DTM) of Pernambuco with the location of weather stations used to evaluate air temperature with the range of RMSE (a) and MBE (b) values associated with air temperature

445 different global atmospheric models in southeastern446 Brazil, observed RMSE of up to 4.59 °C. Aboelkhair

et al. (2019), in a study conducted in India, using the 447 NASA POWER reanalysis, found good performance to 448

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estimate the average air temperature, with RMSE rang-449 ing between 0.96 and 2.54 °C. Aparecido et al. (2019) 450 reported that the air temperature estimated by ERA-451 Interim proved to be accurate and feasible in the replace-452 ment of surface data in the state of Paraná, Brazil, with 453 RMSE ranging from 0.37 °C in summer to 1.11 °C in 454 spring. These results are similar to those reported by 455 Moraes et al. (2012) that found RMSE below 1 °C for 456 almost the entire state of São Paulo to the minimum 457 temperature and RMSE values lower than 4.9 °C for the 458 maximum temperature variable, corroborating with this 459 work the feasibility of using ECMWF global model data 460 for air temperature. 461

The accuracy of the average air temperature estimated 462 by ERA5-Land was higher than 90% in a great portion of 463 the Pernambuco state. This result corroborates the find-464 ings of a great deal of the previous work. This is exempli-465 fied in the work undertaken by Aparecido et al. (2019), 466 in a study in the state of Paraná, Brazil, using ERA-467 Interim reanalysis, which also found accuracy higher 468 than 90% for the average air temperature throughout the 469 state. Another example of what is meant by (Martins 470 et al., 2017), in a study in the Iberian Peninsula, observed 471 accuracy of data from reanalysis higher than 80% for the 472 maximum air temperature; however, for the minimum 473 temperature, the accuracy was lower. On the other hand, 474 Silva et al. (2020), in a study also in the state of Pernam-475 buco, observed ERA5-Land accuracy for the average air 476 temperature higher than 79% with data only of 2019. As 477 well as Monteiro et al. (2018) reported good accuracy in 478 the maximum, minimum, and average temperature data 479 from the NASA POWER reanalysis for Brazil, with R^2 of 480 0.72, 0.75, and 0.73, respectively. 481

Overall, these results indicate that ERA5-Land under-482 estimated the average air temperature values in nine of 483 the 12 stations used, which corresponds to 75%. This 484 finding was also reported by Aboelkhair et al. (2019) 485 with NASA POWER reanalysis data, which underesti-486 mated the average air temperature in 45% of the stations 487 used in India. In this context, Paredes et al. (2018) also 488 identified trends of underestimation of maximum and 489 minimum air temperatures, using ERA-Interim. 490

491 Conclusions

⁴⁹² The main goal of the current study was to determine ⁴⁹³ the accuracy of the air temperature estimated by ERA5-Land reanalysis in the state of Pernambuco, northeast of Brazil, based on the automatic weather stations data set. Overall, the results, according to the statistical indices analyzed, indicate that the ERA5-Land reanalysis successfully simulated the average air temperature of the state of Pernambuco. 499

For more careful evaluation, the average air tem-500 perature database was stratified in three climatic mes-501 oregions of the state of Pernambuco (Forest/Coastal; 502 Agreste and Semiarid). Thus, ERA5-Land reanalysis 503 obtained greater accuracy of the average air tempera-504 ture in the Semiarid, followed by the Forest/Coastal 505 and Agreste. The average air temperature data were 506 slightly underestimated in the Semiarid and Forest/ 507 Coastal mesoregions and slightly overestimated in 508 the Agreste. 509

This approach will prove useful in expanding 510 our understanding of how the use of climatological 511 data from global atmospheric models has become an 512 important alternative for understanding and monitor-513 ing environmental conditions. In general, therefore, 514 it seems that the average air temperature database 515 by ERA5-Land reanalysis can replace the respective 516 data from surface weather stations, in case of missing 517 or scarce observations. However, ERA5-Land still 518 needs improvement, especially in climatic regions 519 similar to the Agreste mesoregion of Pernambuco, 520 which showed the lowest coefficient of determination 521 and the highest RMSE. 522

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Data availabilityData will be made available on reasonable527request.528

Declarations

Conflict of interest The authors declare no competing interests. 531

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