Evaluation of ERA5 Reanalysis over the Deserts in Northern China¹

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Abstract: Deserts cover a vast area of the world's land surface, but the study of desert climate has been impeded by a lack of ground meteorological observations. In recent years, the climate reanalysis products provide an important data source to investigate climate change in observation-limited areas, however, their accuracy in desert regions has been poorly investigated. Here, we first comprehensively evaluated the performance of the latest ERA5 reanalysis datasets on the climatic conditions over the deserts of northern China (DNC), including temperature and precipitation climatology, climate extremes and detection skills of daily precipitation, in comparison with ground observations (OBS) as well as other datasets. The results show that ERA5 well captures the observed pattern of annual and seasonal temperature, as well as the warming trend during the past 40 years in DNC, compared to the OBS. However, both annual and seasonal precipitation have been greatly overestimated over DNC and large bias of ERA5 is found in estimating precipitation trend. Compared to the OBS, ERA5 has lower variability in interannual precipitation, smaller precipitation intensity and maximum 1-day precipitation, shorter continuous dry days and more wet days. The bias in extreme precipitation may be traced to the overestimation of rainfall

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occurrence in ERA5, because the false alarm ratio (FAR) is 66% over the whole DNC and can be as high as 90% in extremely arid western deserts. Overall, ERA5 outperforms other reanalysis such as MERRA2, NCEP1, NCEP2 and NOAA-20C for precipitation and extreme precipitation over DNC, although the performance is not as good as gauge-based precipitation datasets. This study provides reference for use of ERA5 in climate change studies over DNC and other observation-limited desert regions.

Key words: ERA5, Deserts in Northern China, desert climate, climate extremes, rainfall occurrence



Fig.1. Spatial distribution of DNC and meteorological stations. The shade in the map refers to the deserts and dune fields while the dots indicate the meteorological stations. The box labels each sub-region. 1: Taklamakan Desert, 2: Gurban Tunggut Desert, 3: Qaidam Desert, 4: Kumtag Desert, 5: Badain Jaran Desert, 6: Tengger Desert, 7: Ulan Buhe Desert, 8: Hobq Desert, 9: Mu Us Dune Field, 10: Otindag Dune Field, 11: Horqin Dune Field, 12: Hulunbuir Dune Field.



Fig. 2. Spatial distribution of statistical indices of annual and seasonal temperature at each station. (a)-(e): Correlation coefficient (CC) between OBS and ERA5. (f)-(j): Mean bias (MB) between OBS and ERA5. (k)-(o): Root mean square error (RMSE) between OBS and ERA5. The dots represent the geographical location of each station.



Fig. 3. Boxplot of statistical indices of annual and seasonal temperature. (a): Correlation coefficient (CC) between OBS and ERA5. (b): Mean bias (MB) between OBS and ERA5. (c): Root mean square error (RMSE) between OBS and ERA5.



Fig. 4. Spatial distribution of statistical indices of annual and seasonal precipitation at each station. (a)-(e): Correlation coefficient (CC) between OBS and ERA5. (f)-(j): Mean bias (MB) between OBS and ERA5. (k)-(o): Root mean square error (RMSE) between OBS and ERA5. The dots represent the geographical location of each station.



Fig. 5. Boxplot of statistical indices of annual and seasonal precipitation. (a): Correlation coefficient (CC) between OBS and ERA5. (b): Mean bias (MB) between OBS and ERA5. (c): Root mean square error (RMSE) between OBS and ERA5.



Fig. 6. Time series and linear trend of annual mean temperature based on the OBS and ERA5 during 1979-2019 over six sub regions of DNC. 'trend_E' and 'trend_O' are the linear trend in ERA5 and OBS, respectively. The significant level of the trend was determined by Mann-Kendall test. **, * indicate 99%, 95% significant levels respectively.



Fig. 7. Time series and linear trend of annual precipitation based on the OBS and ERA5 during 1979-2019 over six sub regions of DNC. 'trend_E' and 'trend_O' are the linear trend in ERA5 and OBS, respectively. The significant level of the trend was determined by Mann-Kendall test. **, *

indicate 99%, 95% significant levels respectively.



Fig. 8. Spatial distributions of temperature extreme indices based on OBS and ERA5, respectively, and their probability density distribution (the rightmost panel). (a): TX calculated from OBS. (b): TX calculated from ERA5. (c): Probability density distribution of TX based on OBS and ERA5. (d): TN calculated from OBS. (e): TN calculated from ERA5. (f): Probability density distribution of TX based on OBS and ERA5. (g): DTR calculated from OBS. (h): DTR calculated from ERA5. (i): Probability density distribution of DTR based on OBS and ERA5.



Fig. 9. Spatial distributions of five different precipitation extreme indices based on OBS and ERA5, respectively, and their probability density distribution (the rightmost panel). (a): CV

calculated from OBS. (b): CV calculated from ERA5. (c): Probability density distribution of CV based on OBS and ERA5. (d): SDII calculated from OBS. (e): SDII calculated from ERA5. (f): Probability density distribution of SDII based on OBS and ERA5. (g): RX1d calculated from OBS. (h): RX1d calculated from ERA5. (i): Probability density distribution of RX1d based on OBS and ERA5. (j): CDD calculated from OBS. (k): CDD calculated from ERA5. (l): Probability density distribution of CDD based on OBS and ERA5. (m): RD calculated from OBS. (n): RD calculated from OBS. (o): Probability density distribution of RD based on OBS and ERA5.



Fig. 10. Spatial distribution of daily precipitation indices that used to evaluate the capacity of ERA5 in detecting daily precipitation events. (a): Probability of detection (POD). (b): False alarm ratio (FAR). (c): Critical success index (CSI).

Fig.	11. Boxplot	of mean bias	s calculated for	or daily	precipitation	of different	grades.
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Short Full Name		Deserts/ Dune fields	Geographical Location					
Name								
TakD	Taklamakan	Taklamakan Desert (1)	Tarim Basin south of the Tianshan					
	Desert Region		Mountains					
GTD	Gurban	Gurbantunggut Desert	Junggar Basin north of the Tianshan					
	Tunggut	(2)	Mountains and west of the Altai Mountain					
	Desert Region							
QaiD	Qaidam Desert	Qaidam Desert and	Qaidam Basin south of Qilian Mountain					
	Region	Kumtagh Desert (3, 4)						
AlxaD	Alxa Desert	Badain Jaran, Tengger	Alxa Plateau west of Helan Mountain and					
	Region	and Ulan Buh Deserts	north of Qilian Mountain					
0V		(5, 6, 7)						
OrdosD	Ordos Desert	Hobq Desert and Mu Us	Ordos Plateau east of Helan Mountain and					
Region		dune field (8, 9)	west of Taihang Mountain					
ED	Eastern Desert	Otindag, Horqin and	Mongolia Plateau in northeastern China					
	Region	Hulunbuir dune fields						
		(10, 11, 12)						

 Table 1 Information of sub-regions of DNC.

Table 2 Information of the statistical indices used in this study. In the formula, n is the length of time series, E_i is the climate variables in ERA5 during i period, and O_i is the climate variables based on ground meteorological observations, \overline{E} and \overline{O} are the average value in corresponding

periods.

Short name	Full name	Formula
CC	Correlation coefficient	$CC = \frac{\sum_{i=1}^{n} (E_i - \overline{E})(O_i - \overline{O})}{\Gamma}$
MB	Mean bias	$MB = \frac{1}{n} \sum_{i=1}^{n} (E_i - O_i)$
RB	Relative bias	$RB = \frac{MB}{\bar{o}}$
RMSE	Root mean square error	$RMSE = \left \frac{1}{n}\sum_{i=1}^{n}(E_{i}-O_{i})^{2}\right $
RRMSE	Relative root mean square error	$RRMSE = \frac{RMSE}{\overline{O}}$

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-Table 3 POD, FAR, and CSI are used to evaluate the capacity of ERA5 in detecting daily precipitation events. H is the number of hits, which means the frequency of precipitation events recorded in both ERA5 and meteorological stations. M is the number of misses, which means that the frequency of precipitation events recorded in meteorological stations but missed in ERA5. F is the number of false alarms which means the frequency of precipitation event occurring in ERA5 but not appearing in the meteorological stations.

Short name	Full name	Formula
POD	Probability of detection	$POD = \frac{H}{H+M}$
FAR	False alarm ratio	$FAR = \frac{F}{H + F}$
CSI	Critical success index	$CSI = \frac{H}{H + F + M}$

29	Indices	Unit	Definition							
	ТХ	°C	Averaged daily maximum temperature							
	TN	°C	Averaged daily minimum temperature							
	DTR	°C	Averaged daily temperature range							
	CV	\	Coefficient of variation of annual precipitation							
	SDII D=14	mm/d	Averaged precipitation intensity of all wet days (>1mm)							
	KXIQ	mm	Maximum 1-day precipitation							

CDD	d	Maximum number of consecutive days with precipitation < 1mm
RD	d	Number of days with precipitation >= 1mm

 Table 5 Evaluation results of several reanalysis and gauge-based gridded precipitation datasets for annual precipitation and daily precipitation detection skills.

			Daily precipitation							
Datasets			Error analys	15	dete	detection capacity				
	-	CC	RB(%)	RRMSE	POD	FAR	CSI			
	ERA5	0.65	61.4	0.76	0.74	0.66	0.30			
D 1 '	MERRA2	0.53	66.3	0.95		١				
Reanalysis	NCEP1	0.43	186.2	2.27	0.32	0.84	0.11			
datasets	NCEP2	0.43	85.6	1.37	0.24	0.81	0.11			
	NOAA-20C	0.46	159.7	1.73	0.54	0.84	0.14			
Gridded	PREC/L	0.70	17.6	0.42		\				
precipitation	GPCP	0.55	60.4	0.80		١				
dataset	CPC	<mark>0.71</mark>	<mark>13.7</mark>	0.38	0.36	0.80	0.15			
			$\langle \rangle$							

Table 6 Same as Table 5, but for extreme precipitation results.

		CV			SDII		RX1d		CDD			RD				
		CC	RB(%)	RRMSE	CC	RB(%)	RRMSE	CC	RB(%)	RRMSE	CC	RB(%)	RRMSE	CC	RB(%)	RRMSE
	ERA5	0.82	-21.4	0.31	0.43	-18.6	0.34	0.34	<mark>-7.1</mark>	0.59	0.36	-25.2	0.42	0.56	80.8	0.91
Reanalysis	NCEP1	0.22	-9.6	0.42	0.21	-7.90	0.37	0.14	-13.8	0.70	0.20	-14.7	0.50	0.42	169.5	1.98
datasets	NCEP2	0.45	4.9	0.36	0.18	<mark>-3.60</mark>	0.38	0.14	-13.8	0.69	0.28	-3.70	0.47	0.45	71.7	1.12
	NOAA-20C	0.61	-46.7	0.56	0.27	-43.7	0.52	0.20	-46.2	0.71	0.18	-25.0	0.56	0.44	153.8	1.73
Gridded	<u>d</u> v	*														
precipitation	СРС	<mark>0.92</mark>	<mark>2.1</mark>	<mark>0.18</mark>	<mark>0.62</mark>	-18.2	<mark>0.32</mark>	<mark>0.54</mark>	-16.1	<mark>0.54</mark>	<mark>0.57</mark>	<mark>-4.0</mark>	<mark>0.31</mark>	<mark>0.58</mark>	<mark>29.8</mark>	<mark>0.47</mark>
dataset																