Abstract: The regional climate model (RegCM4) was used to project the potential evapotranspiration (PET) of Egypt under two future scenarios: RCP45 and RCP85. Spatially, the RegCM4 has a higher PET under the RCP85 than the RCP45. Among all locations, the RegCM4 was able to capture the monthly variability of PET with respect to the Climate Research Unit (CRU). In addition, the simulated PET was notably improved when the Linear Regression Model (LRM) was used. Also, the future PET projects a strong increased trend under the RCP85; meanwhile the future PET projects a weak increased trend under the RCP45.

Keywords: Climate Research Unit; Egypt; Linear Regression Model; Regional climate model; Potential evapotranspiration

1. Introduction

North Africa is vulnerable to climate change impacts as General Circulation Models (GCMs) participated in the fifth phase of the coupled model Intercomparison project (CMIP5) show a gradual increase of annual temperatures in Northern Africa, higher than the average [1]. Also, the potential evapotranspiration (PET) demand will increase with a severe stress on water resources in the region. PET is an important component in the global terrestrial hydrology cycle and it is used for calculating the water needs of different crops and assessing important hydrological impacts such as meteorological droughts, water balance analysis, and designing and operating irrigation projects.

Penman-Monteith method (PM) is a standard procedure for calculating the PET and to validate other PET methods [2]. PET (or $\lambda E$ as a source of latent heat) is computed as [3]:

$$\lambda E = \frac{\Delta s \left( R_n - G \right) + \rho c_p VPD \left( T_e - T_a \right) \gamma}{\Delta s + \gamma \left( 1 + \frac{1}{T_e - T_a} \right)}$$

where $\lambda E$, $R_n$, and $G$ expressed in W.m$^{-2}$, VPD is the vapor pressure deficit (kPa), $\Delta s$ is the slope of saturation vapor pressure curve (kPa °C$^{-1}$) at air temperature, $\rho$ is the density of air (Kg m$^{-3}$), $c_p$ in J Kg$^{-1}$ °C$^{-1}$, $\gamma$ in kPa °C$^{-1}$, $r_s$ is the aerodynamic resistant (s m$^{-1}$), $r_h$ is the surface resistance to vapor transport (s m$^{-1}$). However, the PM method shows an important weak point because it requires specific thresholds and unlimited supply of moisture. Such condition doesn’t exist under extreme dry conditions [4].

[5, 6] reported that the Hargreaves–Samani method (HS) can be recommended after the PM approach to compute the PET. In fact, the HS method has been widely used in many studies [7]. In data-scarce regions, the HS can be used to compute the PET with a reasonable accuracy [8, 9]. According to the HS method, PET is calculated as:

$$PET_{HS} = 0.0135 \times R_s \times (T_{mean} + 17.8)$$

$T_{mean}$ is the 2-m mean air temperature (in °C). $R_s$ is expressed in units of mm day$^{-1}$ to show how much energy is used to evaporate water [2], [5] showed that inclusion of wind speed and relative humidity didn’t add much to the calculated PET. The HS method is initially proposed to work under different climate conditions. For instance, [10] found that the standardized-HS and the Variable Infiltration Capacity (VIC-3L) models show a reasonable accuracy for estimating the regional and grid-scale variability of ET in data-scarce regions when the field scale measured data is not available.

Correcting the RCMs in both historical and future periods (with respect to the CRU product using the LRM approach) wasn’t conducted over Egypt till present day. To address such a topic, the current study aims to:
1. Examine the spatial pattern of the simulated PET in the historical period (1986-2005) as well as the PET anomaly in the time segments 2021-2040, 2041-2060, 2061-2080 and 2081-2100 under the two future scenarios: RCP45 and RCP85.
2. Bias-correct the simulated PET with respect to the CRU product in the period (1981-2005) for twelve locations (indicated in table 1).
3. Correct the projected PET of the two future scenarios using the LRM approach for the twelve locations (between the RegCM4 and CRU of the historical period).

Section 2 describes the study area and experiment design; section 3 shows the results of the study. Section 4 provides the discussion and conclusion.

2. Materials and Methods

2.1. Study Area

Egypt lies in the north-eastern corner of the African continent and has a total area of about one million km$^2$. Climatologically, Egypt is characterized by hot dry summers and mild winters. Summer temperatures are extremely high, reaching 38°C to 43°C with extremes of 49°C in the southern and western deserts. The northern areas on the Mediterranean coast are much cooler, with 32°C as a maximum. Annual rainfall ranges between a maximum of about 200 mm in the northern coastal region to a minimum of nearly zero in the south, with an annual average of 51 mm.

2.2. Model Description and Experiment design

In this study, the Abdus Salam International Centre for Theoretical Physics (ICTP) regional climate model version 4.7 (RegCM-4.7.0, [11]; hereafter RegCM) was used. The RegCM was downscaled by Earth System Model of the Max Planck Institute (MPI-ESM; [12]) of medium resolution (MR; 96 × 192 grid points i.e., 1.875 × 1.875 horizontal degree resolution). This study comprises two domains: (1) the coarse domain which covers the Middle East/North Africa (MENA) region with 50 km horizontal grid spacing, 235 grid points in the zonal direction and 121 grid points in the meridional direction, centered at latitude 19.5° and longitude 24.5°; (2) the nested domain which covers Egypt and surrounding regions with 20 km horizontal grid spacing, 121 grid points in both zonal and meridional directions, centered at latitude 25.5° and longitude 30.5°. The domain dimension and topography height for the MENA and Egypt domains are presented in Figure 1. The Model simulations were integrated over the historical period 1981-2005 and 2006-2100 for both the moderate future scenario RCP45 and extreme future scenario RCP85.

2.3. Validation Data

Measurement of PET in Egypt is important for monitoring agricultural activity, assessing the water needs and drought monitoring over interval of time scales (ranging from daily, seasonal and annual). However, availability of the in-situ PET is not feasible for a long time (using the PM method). Instead, the Climate Research Unit (CRU; version 4.05; [13]) dataset was used as the ground truth of observation. CRU product is integrated over the period 1901-2020 and it includes various variables: cloud cover, diurnal temperature range, frost day frequency, PET, precipitation, 2-m mean temperature, maximum and minimum air temperature, as well as vapor pressure. CRU product is available in 0.5 × 0.5 horizontal degree resolution. Also, CRU is considered as the best available reference PET data and it is used as the ground truth of observation for global assessment of PET [14]. The current study used the CRU gridded product to correct the simulated PET in both the historical period and the two future scenarios (RCP45 and RCP85) for twelve locations (indicated in table 1).

3. Results

3.1. Spatial Pattern of 2-m mean air temperature and PET under two future scenarios
Figure 2 shows the future projected changes of the 2-m mean air temperature (hereafter TMP) over Egypt during the period 2021-2100 of the future scenario RCP45 compared to the reference period 1986-2005. From Figure 2, it can be observed that Western, Eastern Coast and Delta region experiences an increase in TMP by 0.5 – 1°C; meanwhile Upper Egypt experiences a decrease in TMP by 0.5 – 1°C relative to the reference period. Such increase/decrease is noted the time segments 2061-2080 and 2081-2100 (Figure 2d, e). Figure 3 shows the projected TMP under the RCP85 future scenario. From Figure 3, it can be observed that the TMP shows a similar increase to the one observed in the time segments 2021-2040 and 2041-2060 (Figures 3b, c). However during the time segments 2061-2080 and 2081-2100; the TMP shows a higher increase (ranging from 3 - 5°C) overall Egypt especially during the time segment 2081-2100 (Figures 3d, e).

Figure 4 shows the projected PET under the RCP45 scenario. In Figure 4, it can be noted that the projected PET experiences an increase (by 0.1 – 0.3 mm day⁻¹) over the Western Coast and Western desert (Figure 4b) during the time segment 2021-2040. In the time segment 2041-2060, the RegCM shows an increase in the simulated PET over the Western Egypt, Western desert and Delta regions (Figure 4c). Under the RCP85 scenario, the situation is quite different, as in the time segment 2021-2040; the PET noted changes are ±0.1 mm day⁻¹ (Figure 5b). While in the time segment 2041-2060, the RegCM shows an increase of the simulated PET (by 0.1 – 0.3 mm day⁻¹) overall Egypt (Figure 5c). In accordance with the noted changes in TMP, the RegCM shows a higher increase of the simulated PET (by 0.3 – 0.6 mm day⁻¹) than the one observed in the time segments 2021-2040 and 2041-2060 (Figure 5d). Lastly in the time segment 2081-2100, the RegCM exhibits the highest increase of the simulated PET over majority of Egypt (by 0.5 – 0.9 mm day⁻¹; Figure 5e).

3.2. Correcting the PET in the historical period and future scenarios

Performance of the RegCM model was evaluated with respect to the CRU gridded product for twelve locations (indicated in table 1) to represent different climate zones of Egypt: 1 - Alexandria, Marsa-Matruh and Siwa representing the north zone; 2 - Ismailia, Port-Said and Arish representing the eastern zone; 3 - Giza and Asyout representing Lower Egypt zone; Dakhla and Kharga representing the Middle Egypt zone and 4 - Luxor and Asswan representing the Upper Egypt zone (see table 1). The LRM approach ([15]) was applied by plotting a scatter plot between the RegCM model output and CRU observational-based product in the historical period 1981-2005. The RegCM performance was quantified (before and after applying the LRM) in terms of the statistical metrics: mean bias (MB; calculated as CRU minus RegCM) and standard deviation ratio (SD; which is defined as the ratio between standard deviation of the RegCM to standard deviation of the CRU). Figure 6 shows the monthly time-series of the PET in comparison with the CRU in the historical period before and after applying the LRM method. In general, the RegCM is able to capture the monthly variability with respect to the CRU product; however, the model underestimates/overestimates the monthly PET depending on the study location. To project the relative future PET changes for the twelve locations, the following steps were conducted:

1. The mean PET of the corrected RegCM output (in the historical period) is calculated.
2. The LRM is used to correct the projected PET of the two scenarios: RCP45 and RCP85.
3. The relative PET changes are calculated following [16]:

   \[
   \text{PET}_{\text{relative changes}} = \frac{\text{PET}_{\text{future annual}} - \text{PET}_{\text{present annual}}}{\text{PET}_{\text{present annual}}}
   \]  

   (3)

Figure 7 shows the corrected relative future projection of PET under the RCP45 and RCP85 scenarios of the twelve locations. From Figure 7, it can be noticed that, the relative changes of PET ranges from -10% to +12%. Also, the two scenarios are in agreement in the period of 2006-2054; after that the divergence between the two scenarios becomes evident. In addition, the future PET projects a strong increased trend under the RCP85; meanwhile the future PET projects a weak increased trend under the RCP45.

3.3. Figures, Tables
Figure 1 shows the MENA domain (left panel) and Egypt domain (right panel).

Figure 2 (left side) Average 2-m air temperature (hereafter TMP; in degrees Celsius) over Egypt during 1986-2005 (RF) (a) and the potential change during the period 2021-2040 (b), the period 2041-2060 (c), the period 2061-2080 (d), the period 2081-2100 (e) according to the RCP45 scenario. Figure 3 (right side) Average 2-m air temperature (hereafter TMP; in degrees Celsius) over Egypt during 1986-2005 (RF) (a) and the potential change during the period 2021-2040 (b), the period 2041-2060 (c), the period 2061-2080 (d), the period 2081-2100 (e) according to the RCP85 scenario.

Figure 4 (left side) shows the average evapotranspiration (mm/day) over Egypt during 1986-2005 (RF) (a) and potential change during the period 2021-2040 (b), the period 2041-2060 (c), the period 2061-2080 (d), the period 2081-2100 (e) according to the RCP45 scenario. Figure 5 (right side) shows the average evapotranspiration (mm/day) over Egypt during 1986-2005 (RF) (a) and potential change during the period 2021-2040 (b), the period 2041-2060 (c), the period 2061-2080 (d), the period 2081-2100 (e) according to the RCP85 scenario.
Figure 6 shows the monthly times series of Potential Evapotranspiration (PET; in mm/day) of by the RegCM in the historical period (1981-2005) of the twelve locations in comparison with the CRU product (in red), before applying the linear regression model (RegCM; in blue) and after applying the linear regression model (RegCMnew; in green).

Figure 7 shows the future corrected PET changes (in %) under the two future scenarios (RCP45; in blue) and (RCP85; in red) for the twelve locations.

Table 1 shows the linear regression model (LRM) for each station to correct the PET calculated by the RegCM with respect to the CRU in the historical period 1981-2005 and statistical metrics (MB and SD ratio) before and after applying the linear regression model.

4. Discussion and Conclusion

PET is essential variable for assessing both hydrological and agricultural activities as well as monitoring of droughts on a regional scale. The FAO organization recommends the PM method to
calculate the PET; however, calculating PET using the PM is not recommended for the following reasons: 1) the PM method works under specific thresholds and unlimited supply of moisture [4], 2) the uncertainty of various meteorological inputs can amplify the error of the calculated PET because the PM equation is not linear and 3) calculating the PET using the PM method using version 4.7 of the RegCM (and afterwards) is not fully tested yet over a hierarchy of time scales. To overcome such problems, an empirical equation (with a minimum number of meteorological inputs) is needed to track the PET changes either on daily or longer time scales using the TMP as a proxy. Therefore, the HS method was chosen as an alternative option to calculate the simulated PET particularly over arid/hyper arid regions (as in this study).

In the present study, the RegCM was downscaled by the MPI-ESM-MR over the MENA and then nested over Egypt for the reference period 1980-2005 and under the two future scenarios: RCP45 and RCP85. After that, the RegCM model performance was evaluated with respect to the CRU. The results showed that the projected PET showed a gradual increase from the North zone to Upper Egypt zone; such finding is consistent with the results reported in [17]. Moreover, the PET was subjected to a moderate increase under the RCP45 scenario and a high increase under the RCP85 scenario. In addition, the RegCM model was able to capture the monthly variability with respect to the CRU observational-based product; however, the RegCM over/underestimated the PET depending on location under study.

Furthermore, the RegCM showed an improved performance when the LRM was applied at the location of interest. The highest fluctuations of the future projected PET changes were observed at Asyout, Dakhlia, Kharga and Luxor particularly under the RCP85 scenario. It is important to highlight that, this work provides a first insight for projecting PET of Egypt using a high-resolution regional climate model and bias-correcting the PET in both the historical and two future scenarios using the LRM approach. The current study relies on downscaling of one GCM from the pool of the Fifth phase of the Coupled Model Intercomparison Project (CMIP5; [18]). Therefore, a future work will consider the following points:

1. Using multi-GCMs (CMIP5/CMIP6; [18, 19 and 20]) and their ensemble to further examine the sensitivity of the RegCM to atmospheric forcing and therefore the simulated \( T_{\text{mean}}, R_u \) and hence PET.

2. Examining the potential role of aerosols on the simulated \( T_{\text{mean}}, R_u \) and eventually PET.

3. Evaluating the RegCM4 output with high-resolution gridded PET product (e.g., [21, 22]) along with the CRU (with 0.5°) to account for the uncertainty of the observational-based dataset.

**Supplementary Materials:** Not applicable

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**Conflicts of Interest:** The authors declare no conflict of interest.
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