# Impact of Climate-Indicators on Continental-Scale Potential Groundwater

# **Recharge in Africa**

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## Abstract

In the last decades, human activity has been contributing to climate change that is closely associated with an increase in temperatures, increase in evaporation, intensification of extreme dry and wet rainfall events, and widespread melting of snow and ice. Understanding the intricate linkage between climate warming and the hydrological cycle is crucial for sustainable management of groundwater resources, especially in a vulnerable continent like Africa. This study investigates the relationship between climate-change drivers and potential groundwater recharge (PGR) patterns across Africa for a long-term record (1960-2010). Water-balance components were simulated by using the PCR-GLOBWB model and were reproduced in both gridded maps and latitudinal trends that vary in space with minima on the Tropics and maxima around the Equator. Statistical correlations between temperature, storm occurrences, drought, and PGR were examined in six climatic regions of Africa. Surprisingly, different effects of climate-change controls on PGR were detected as a function of latitude in the last three decades (1980-2010). Temporal trends observed in the Northern Hemisphere of Africa reveal that the increase in temperature is significantly correlated to the decline of PGR, especially in the Northern Equatorial Africa. The climate indicators considered in this study were unable to explain the alarming negative trend of PGR observed in the Sahelian region, even though the Standardized Precipitation-Evapotranspiration Index (SPEI) values report a 15% drought stress. On the other hand, increases in temperature have not been detected in the Southern Hemisphere of Africa, where increasing frequency of storm occurrences determine a rise of PGR, particularly in southern Africa. Time analysis highlights a strong seasonality effect while PGR is in-phase with rainfall patterns in the summer (Northern Hemisphere) and winter (Southern Hemisphere) and out-ofphase during the fall season. This study helps to elucidate the mechanism of the processes influencing groundwater resources in six climatic zones of Africa, even though modeling results need to be validated more extensively with direct measurements in future studies.

Keywords: Potential groundwater recharge, rainfall, drought, seasonality, storm frequency, Africa

### **1. Introduction**

Groundwater is an essential natural resource for urban, industrial, and agricultural water supply and socioeconomic development (de Vries and Simmers, 2002). It has been projected that rapid population growth will force a significant increase in freshwater demands in many regions of the world (Wada *et al.*, 2011). The degree to which groundwater resources can sustain increased abstraction is highly uncertain, given the current scarcity of groundwater recharge estimates (Döll *et al.*, 2014; McLaughlin and KinzelBach, 2015). Climate-change impacts on renewable freshwater resources add another layer of uncertainty with regard to future pressures on water availability, accessibility, and demand.

Understanding how climate drives the spatiotemporal distribution of groundwater recharge rates triggered below the vadose zone has been explored in existing studies in different areas of the world (Eckhardt and Ulbrich, 2003; Brouyere *et al.*, 2004; Keese *et al.*, 2005; Ng *et al.*, 2010; Green *et al.*, 2011; Kim and Jackson, 2011; Crosbie *et al.*, 2013; Panwar and Chakrapani, 2013; Faramarzi *et al.*, 2013). Rainfall can be considered as the major climatic driver in affecting recharge rates, with particular emphasis on rainfall intensity and frequency distribution. Studies related to climate change suggest that a greater increase in extreme precipitation will exacerbate the frequency of heavy rainfall events in some regions and increase the risk of more prolonged drought spells in other regions (Milly *et al.*, 2005). Regional-scale studies in North America provide evidence of a decreasing trend in groundwater recharge under (historical or projected) global warming (Rosenberg *et al.*, 1999; Ferguson and George, 2003). Other important climatic factors include solar radiation and vapor deficit, which significantly impact groundwater recharge rates in tropical zones (Vivoni *et al.*, 2009; Barron *et al.*, 2012).

Groundwater is a crucial resource in Africa, where the population at risk of increased water stress is projected to be over 500 million people by 2050 (Burke *et al.*, 2009; Carter and Parker, 2009; MacDonald *et al.*, 2012). A sustainable and efficient water management carefully regulates withdrawals such that water is not overused when rainfall is not able to replenish the groundwater stocks. Thus, there is an urgent necessity to examine the sensitivity of groundwater recharge to

climate-change indicators in Africa (MacDonald et al., 2009). The challenge is to assist regional policy makers in Africa in acquiring more knowledge on vulnerable groundwater resources by capturing recharge response to extreme climate impacts (Shongwe et al., 2009; Shongwe et al., 2011). Field studies of groundwater recharge in Africa have had uneven spatial coverage (Edmunds, 2009; Bonsor and MacDonald, 2010). As a result, there is currently a pressing need for large-scale recharge estimates to underpin sustainable water-resource management strategies (Döll and Fiedler, 2008; Nyenje et al., 2009). In order to fill the information gap, this study addresses spatiotemporal variability in continental-scale groundwater recharge rates using a global hydrologic model (PCR-GLOBWB) with daily boundary forcings and a high spatial resolution. We note that the model has not been calibrated for Africa; hence, this investigation de facto represents a large-scale sensitivity analysis. The term "potential groundwater recharge" is therefore used to refer only to a synthetic model approximation of direct measurements in Africa. The results have been validated by few available direct-recharge measurements retrieved in the scientific literature. The main objectives of this study are twofold: (1) assessing the impact of climate-change drivers on spatial distribution and spatiotemporal trends of average annual PGR in Africa; and (2) analyzing inter-annual and seasonal variability of PGR. This novel contribution provides valuable insights in detecting the historical relationships between climate-change indicators and groundwater recharge in Africa. The groundwater recharge rates simulated with a continental-scale bucket model represent, to date, the primary data used to assess the influence of climatic controls on unconfined aquifers in Africa. Therefore the observed trends in six climatic regions in Africa can be used to develop analogs for groundwater recharge responses that can be exploited in scenario-based approaches under future climate-change projections.

## 2. Methodology

## 2.1 PCR-GLOBWB Model

The PCR-GLOBWB (PCRaster-GLOBal Water Balance) is a grid-based bucket model of global terrestrial hydrology (van Beek *et al.*, 2011; Wada *et al.*, 2014; http://pcraster.geo.uu.nl/). For each

grid cell (spatial resolution of  $0.1^{\circ} \times 0.1^{\circ}$ ), this model simulates the water exchange between the atmosphere and two vertically-stacked soil layers connected to a third underlying groundwater layer. The PCR-GLOBWB model refers to the units of meter (m) and day (d) for the dimensions of length (L) and time (T), respectively. Rainfall (L T<sup>-1</sup>) and snowmelt (L T<sup>-1</sup>) represent the inflow to the hydrologic system. Runoff (L T<sup>-1</sup>) and evaporation (L T<sup>-1</sup>) contribute to water loss from the soil surface. Vegetation affects the water balance through transpiration (L T<sup>-1</sup>) and canopy interception (L T<sup>-1</sup>). The soil hydraulic parameters are assigned according to eleven soil-texture classes corresponding to different soil types based on the FAO Digital Soil Map of the World (FAO, 2003).

Groundwater recharge (downward flux) and capillary rise (upward flux) represent the vertical water exchanges between the second and third subsurface soil layers (Wada *et al.*, 2010). Downward water fluxes (percolation and groundwater recharge) among the three soil layers are calculated using the unit-gradient form of the Darcy-Buckingham equation. Groundwater recharge interacts with groundwater storage through capillary rise if the top of the groundwater level is within ~5 m of the topographical surface (Wada *et al.*, 2012; Wada *et al.*, 2014). Groundwater storage is fed by groundwater recharge and drained as baseflow by a reservoir coefficient that includes information on lithology and topography (*e.g.*, hydraulic conductivity of the subsoil; maps can be visualized in http://www.bgs.ac.uk/research/groundwater/international/africangroundwater/maps.html).

The ensuing capillary rise is calculated as the upward moisture flux that can be sustained when an upward gradient exists and the moisture content of the soil is below field capacity. Additionally, the capillary rise, cannot exceed the available storage in the underlying groundwater reservoir.

The model was forced with daily fields of precipitation, reference (potential) evapotranspiration, and temperature for the 1960-2010 period. For the 1960-1978 period, boundary forcings were prescribed by the ERA40 re-analysis data (Uppala *et al.*, 2005). The associated climate data are publicly available at http://www.ecmwf.int/en/forecasts/datasets/era-40-dataset-sep-1957-aug-2002. In order to extend our analysis to the year 2010, we forced the model by comparable daily climate fields taken from the ERA-Interim re-analysis data, from which we obtained daily fields of temperature, reference potential evapotranspiration, and GPCP-corrected precipitation (GPCP: Global Precipitation Climatology Project; http://www.gewex.org/gpcp.html). For compatibility with our overall analysis,

the ERA40 data analysis was bias-corrected on a grid-by-grid basis by scaling the long-term monthly means of the daily climate fields (precipitation, evapotranspiration, and temperature) to those of the ERA-Interim data for the overlapping reference climate for 1979-2001. We then further bias-corrected the modified climate dataset by scaling the long-term monthly means of the daily climate fields to those from the Climate Research Units (CRU) TS 2.1 time-series (Mitchell and Jones, 2005), wherever station coverage by the CRU is adequate for the overlapping period. The associated climate data are publicly available at http://www.cru.uea.ac.uk/data. For the ERA40 re-analysis data, we separately bias-corrected for the 1960s and 1970s period to correct the overestimation of precipitation present over the tropics during the 1970s (Hempel *et al.*, 2013; Wada *et al.*, 2014; Wada and Bierkens, 2014).

#### **2.2 Model Limitations**

We caution that the continental-scale lumped-parameter approach provides only a crude approximation of the water budget and is possible at the cost of several simplifications: (a) ignoring preferential flow into the soil, fractured rock, saprolite (Neumann, 2005; Beven and Germann, 2013), and karst regions (Hartmann *et al.*, 2014); (b) not considering the role of macroporosity (Owor *et al.*, 2009; Taylor *et al.*, 2013); (c) missing proper soil-hydraulic properties pertaining to datasets related to specific Tropical and Equatorial zones in Africa (Minasny and Hartemink, 2011); (d) missing information on land-use changes (urbanization, deforestation) in Africa (Githui *et al.*, 2009; Baker and Miller, 2013); (e) lacking records of water pumping and the role of human consumption of water; and (f) ignoring focused recharge via leakage from surface waters (ephemeral streams, wetlands, or lakes). Therefore, we caution that the model might not be able to simulate actual recharge rates in some regions and, for this reason, we will refer only to "potential groundwater recharge" (PGR) in the remainder of the manuscript (Rushton, 1997).

In order to test the above limitations of the model, we compiled 13 published groundwater recharge rates (Table 1) obtained primarily by field techniques, namely chloride mass balance (CMB), water table fluctuations (WTF), groundwater residence time (GRT) with the use of tracers, and water balance (WB) (BredenKamp et al., 1995; Scanlon et al., 2002; Wang et al., 2010). The spatial

resolution (SP), mean ( $\mu$ ) values, and ranges of average annual observed groundwater recharge rates are reported with the corresponding simulated values in PCR-GLOBWB (Table 1).

The Nash-Sutcliffe Efficiency (NSE%) index has been calculated for the analysis of agreement between observed ( $GR_{obs}$ ) and simulated ( $GR_{sim}$ ) groundwater recharge (GR):

$$NSE\% = \left[1 - \frac{\sum_{i=1}^{N} \left(GR_{obs,i} - GR_{sim,i}\right)^{2}}{\sum_{i=1}^{N} \left(GR_{obs,i} - \overline{GR_{obs}}\right)^{2}}\right] \times 100$$
(1)

where *N* is the number of recharge observations (N = 13), and  $\overline{GR_{obs}}$  is the arithmetic mean of observed groundwater recharge. High values of NSE% (near 100%) indicate best agreements, whereas a value of 50% (*i.e.*, NSE% > 50%) can be assumed as a threshold for efficient and acceptable agreements (Moriasi *et al.*, 2007). Near-zero or negative values of NSE% ( $0 > NSE\% > -\infty$ ) indicate unacceptable performances provided by the model simulations (Bennet *et al.*, 2013).

Despite the fact that the NSE% = 30.6% is below the threshold of satisfactory efficiency (NSE% < 50%), the performance can be considered sufficiently acceptable, considering that the model was not calibrated. The relative differences (RD), expressed in percentages of the  $log_{10}$ -values listed in the right-hand column of Table 1, indicate that most of the modeled recharge rates tend to overestimate the measurements. The best model performances are observed in countries in southern Africa (Botswana, Zambia, and Zimbabwe), albeit preferential flow plays an important role in controlling recharge rates measured with chloride mass balance (de Vries *et al.*, 2000). Higher discrepancies are reported in countries of Sahelian Africa (Niger, Sudan).

### 2.3 Data Processing

The set of input and output data in Africa for the 51-year time period (1960-2010) was taken from the existing database of worldwide simulations executed by the PCR-GLOBWB model (Wada *et al.*, 2014). Gridded parameter values, boundary forcings, and simulation results were mapped using

ArcMap 10.1. Statistical analyses were completed in a parallel computing environment that adapted MATLAB into a super-computing cluster (http://hcc.unl.edu/tusker/index.php). The scripts can be made available upon request.

The analysis of extreme-rainfall variability considered two statistical indicators describing storm occurrences ( $f_{max}$ ) and mean dry consecutive days ( $\tau_{dry}$ ) (Small, 2005). The former expresses the number of storms (all those values above the annual mean value) over the total number of rainfall events, while the latter is identified through the mean dry interval time between two separate rainfall events.

Average annual values of boundary forcings, namely rainfall (P) and potential evapotranspiration  $(ET_p)$ , are displayed in Fig. 1, whereas simulated water-balance components, namely potential groundwater recharge (PGR), actual evapotranspiration  $(ET_a)$ , and runoff ( $Q_r$ ), are shown in Fig. 2.

The continental-scale spatial distribution of average annual PGR and  $Q_r$  broadly follows the precipitation distribution (Goddard and Graham, 1999; Sultan and Janicot, 2000).

The African continent has a heterogeneous distribution of renewable freshwater resources. Water availability is mostly concentrated in the humid zones around the Equator. The remaining areas experience permanent arid conditions with modest or insignificant recharge generation (Schuol *et al.*, 2008). Nevertheless, the humid areas are characterized by a high degree of spatial and temporal variability of groundwater recharge rates. On a latitudinal-averaged basis, the mean annual PGR reaches maxima near the Equator ( $0.42 \text{ m yr}^{-1}$  at latitude  $-0.45^{\circ}$ N) and minima near the tropics ( $0.00028 \text{ m yr}^{-1}$  at 20.05°N and 0.012 m yr<sup>-1</sup> at -25.35°N, respectively) (Fig.2).

The Mann-Kendall nonparametric test (Mann, 1945; Kendall, 1975) has been calculated in order to detect eventual temporal trends in long-time series for annual temperature (T), rainfall indicators ( $f_{max}$ ,  $\tau_{dry}$ ), and PGR. The Mann-Kendall test is expressed by:

$$S = \sum_{i=1}^{N-1} \sum_{j=i+1}^{N} sgn(x_j - x_i)$$
(2)

where  $x_i$  and  $x_j$  are the sequential data values, N (the total number of years is 51 between 1960 and

2010) is the length of data series, and *sgn* is the sign function. For a large number of years, S is assumed to be normally distributed with an expected value E(S) = 0 and

$$\operatorname{var}(S) = \frac{N(N-1)(2N+5)}{18}$$
(3)

The standardized test statistic  $Z_c$  is calculated as:

$$Z_{c} = \begin{cases} (S-1)/\operatorname{var}(S)^{1/2} & S > 0\\ 0 & S = 0\\ (S+1)/\operatorname{var}(S)^{1/2} & S < 0 \end{cases}$$
(4)

The null hypothesis indicating no trend is accepted when  $|Z_c| \le Z_{1-\alpha/2}$ , where  $\alpha$  is the significance level, and  $Z_{1-\alpha/2}$  is the standard normal quantile at probability level of  $1-\alpha/2$ . Therefore, the trend test is based on the following two conditions:

- (1)  $Z_c < -Z_{1-\alpha/2}$  indicates decreasing trend; and
- (2)  $Z_c > Z_{1-\alpha/2}$  indicates increasing trend.

In this study, we fixed  $\alpha$  at the 0.05 significance level, and  $Z_{1-\alpha/2} = 1.96$ .

### 3. Results and Discussions

### **3.1 Spatial Distribution of PGR**

The objective of this subsection is to characterize PGR spatial patterns in Africa and to identify statistical relationships between average annual rainfall and PGR on a latitudinal-averaged basis. Trambauer *et al.* (2014) recently defined six climatic regions, namely Mediterranean (MED), Sahel (SAH), North Equatorial Central Africa (NEQ), Horn of Africa (HORN), South Equatorial Central Africa (SEQ), and Southern Africa (SAFR). These six regions were based on the existing map of five climate classes (hyper-arid, arid, semi-arid, dry sub-humid, and humid) according to the aridity index, defined as the ratio between mean-annual rainfall and mean-annual evaporation (Fig.3a).

PGR occurs significantly only in the North (21% area) and South (12% area) Equatorial Central Africa (NEQ+HORN and SEQ) regions with 0.27 m yr<sup>-1</sup> and 0.26 m yr<sup>-1</sup>, respectively (Fig. 3b). PGR

is lower than 0.05 m yr<sup>-1</sup> in the remaining four regions (MED, SAH, HORN, and SAFR) that cover the major part of Africa. In addition to climate drivers, biophysical controls, such as geomorphology, soil stratigraphy, lithology, and land use, play a paramount role in determining the spatial variability of the water-balance components (Beauvais, 1999; Farmer *et al.*, 2003; Kim and Jackson, 2011). Longitudinal variabilities of PGR,  $ET_a$ , and  $Q_r$  increase with rainfall along the Equatorial latitudes, as designated by the larger gray-shaded areas depicted in Fig. 2d-f.

The highest rates of PGR are observed in the NEQ and SEQ regions (0.29 and 0.26 m yr<sup>-1</sup>, respectively) (Table 2). The coefficients of variation (CV) values are relatively low (12% and 15%, respectively). The recharge rate in the Mediterranean region is one order of magnitude lower than that of central Africa (0.0215 m yr<sup>-1</sup>), whereas the remaining three regions have recharge rates of two orders of magnitude lower than those in Equatorial Africa with the highest degree of spatial variability (CV > 100 %). Average recharge rates of 0.05 m yr<sup>-1</sup> as simulated in the PCR-GLOBWB model in the SAH region support previous measurements in chloride profiles that range between 0.03 and 0.15 m yr<sup>-1</sup> (Scanlon *et al.*, 2006)

The latitudinal relationships between average annual rainfall and PGR in each climatic region (HORN is combined with NEQ) demonstrate that all correlation coefficients ( $R^2$ ) are significantly high (0.75 <  $R^2 < 0.88$ ), implying that PGR is highly dependent on average annual rainfall patterns (Fig. 4). We observe that the lowest correlation value ( $R^2 = 0.75$ ) in the MED region is significantly affected by a strong drying latitudinal (north-south) gradient. The northern semi-arid half portion ( $37^\circ < lat < 32^\circ$ ) of the MED region is characterized by high spatial variability and very low correlation between P and PGR. Conversely, in the southern hyper-arid half portion outside of the mountainous headwaters ( $32^\circ < lat < 27^\circ$ ) of the MED region, P and PGR are reduced by three times but exhibited high correlation. The regression line grade (indicated by the slope-values) gets steeper in Central Africa (NEQ + HORN and SEQ) and three times lower in the other three climatic zones (MED, SAH, and SAFR), where the average annual rate of PGR is only between 14% and 24% of P on a latitudinal-averaged basis.

The spatial distribution of climate-indicators, namely storm occurrences and drought, reflects different patterns in the two Hemispheres (Fig.5)

The extreme rainfall parameter,  $f_{max}$ , increases between the Equator and the Tropic of Capricorn, with the peak at -12.95°N ( $f_{max} = 0.42$ ), meaning that about 40% of rainfall events have intensities higher than their mean values. The values of  $\tau_{dry}$  emphasize the importance of episodic rainfall events in the hyper-arid areas of the Northern Hemisphere and follow a similar trend along the latitude with a cusped shape in proximity to the Tropic of Cancer characterized by a high degree of longitudinal variability. A modest increase of  $\tau_{dry}$  is detected above the Tropic of Capricorn.

#### **3.2 Temporal Variability of PGR**

#### 3.2.1 Inter-annual variability of PGR

This section addresses the empirical understanding of how climate-change drivers have been related to PGR in the last three decades (1980-2010).

The Spearman's correlation coefficients between PGR and climate-change indicators (temperature and the two extreme rainfall distribution indicators; see Table 3)highlight a general negative correlation between temperature (T) and PGR, with zero effect on the MED, SAH, and HORN regions, with a modest impact on SEQ (-0.20), and with a significant influence on NEQ (-0.48) and SAFR (-0.76) (Tashie *et al.*, 2015). While the relationship between rainfall and temperature is difficult to characterize, the correlation between potential evapotranspiration ( $ET_p$ ) and temperature is often positive (Kingston and Taylor, 2010).

The parameter,  $f_{max}$ , has a very low impact in the arid regions of Africa, with the exception of SAFR, whereas significant correlation is observed in Equatorial Africa. The increase in storm occurrences favors both PGR and  $Q_r$  generation in humid Equatorial regions of Africa. Indeed, in the model setup, the former is driven by gravity and is greatest under saturated conditions, while the latter occurs under

the same circumstances but when rainfall exceeds soil-infiltration capacity. The parameter  $\tau_{dry}$  significantly affects the two extreme regions (MED and SAFR) with correlations values of -0.45 and - 0.38, respectively, while the impacts in the humid regions are relatively low.

The Mann-Kendall test results pertaining to the last three decades of model simulations (1980-2010)elucidate the temporal evolution of climate-indicators and PGR (Table 4). The temperature exhibits increasing trends in the MED, SAH, and NEQ zones, but not in the Southern Hemisphere of Africa. The extreme rainfall indicators also signal an increasing trend of  $f_{max}$  in SAFR (Min *et al.*, 2011). Nevertheless, PGR shows decreasing trends in SAH and NEQ and an increasing trend in SAFR. In view of the significant trends reported in Table 4 and considering the correlation coefficients listed in Table 3, we can infer that climate change is significantly impacting PGR only in the SAH, NEQ, and SAFR climatic zones. The two zones above the Equator indicate negative correlation between temperature and PGR, whereas the latter shows a high positive correlation value between  $f_{max}$  and PGR.

It is interesting to note that temperature is increasing only in the Northern Hemisphere of Africa. This implies a general decrease in PGR in the zones above the Equator, in contrast to a general temporal stability of climatic-change drivers and hydrological balance in the Southern Hemisphere. The climate-temporal variation in each climatic zone has been objectively analyzed through the Standardized Precipitation-Evapotranspiration Index (SPEI) (Vicente-Serrano *et al.*, 2010) downloaded from the website http://sac.csic.es/spei/database.html (Beguería *et al.*, 2014). This multi-scalar drought index has the advantage of combining the effect of precipitation and potential evapotranspiration demand (affected by temperature) and is particularly valuable for the application of future projections (Lorenzo-Lacruz *et al.*, 2010). For the purpose of showing an illustrative example, a transect along longitude 15.25°E has been identified by considering latitudes 28.25°N, 15.25°N, 5.25°N, -5.25°N, and -22.25°N for the MED, SAH, NEQ, SEQ, and SAFR zones, respectively. Very dry conditions characterized by negative SPEI values are observed in the 1930s, 1940s, and 1970s in the MED zone, the 1970s, 1980s, and 1990s in the SAFR zone (Fig.6).

This qualitative visual inspection confirms the high-climatic variability among the different zones of Africa. However, we considered the temporal SPEI values in all cells pertaining to each climatic zone. Thus, it is possible to compare the frequency distribution of the SPEI values in the time range (1901-2010) with the distribution belonging to the last three decades of model simulations (1980-2010) in order to identify the causes of the trends observed in Table 4. The frequency distributions of SPEI values (blue bars) in the entire range with an expected SPEI mean around zero approximately partition 50% of wet and 50% of dry conditions (Fig.7). The frequency distributions of SPEI values (red bars) pertaining to the last three decades in the Northern Hemisphere of Africa are affected by a shift to the left by lowering the probabilities of observing wet conditions by 4.4%, 14.8%, and 6% for MED, SAH, and NEQ, respectively. Decreasing trend of PGR is observed in the SAH and NEQ zones, but not in the MED region (Table 4). The devastating drought spells in the SAH zone in the last three decades probably represent the most prevalent climatic changes in Africa that have significantly decreased PGR. On the other hand, the two zones (SEQ and SAFR) in the Southern Hemisphere report a slight increase in wet conditions in the last three decades, increasing by 1.5% and 1% in the SEQ and SAFR zones, respectively. The climatic stability observed in the SEQ zone impedes any trend for PGR, although the climate and PGR statistics of NEQ and SEQ (observed in Tables 1, 2, and 3; Fig. 4) are very similar. Based on the historic data, our findings support the future projections described by Faramarzi et al. (2013) that predict an increase up to 400% of blue water in Southern Africa (albeit small) and a decrease of renewable water resources by 25-100% in the semi- and hyperarid regions of Africa that will become more susceptible to severe drought conditions. Döll and Flörke (2005) evaluated the impact of climate change on long-term, average annual diffuse groundwater recharge, as simulated by the global hydrological WaterGAP Global Hydrology Model. By applying four different climate-change scenarios between the present day and the 2050s, they have predicted a significant increase in groundwater recharge by up to 30% in the SAH and HORN regions, while decreasing trends are expected in the MED and southwestern SAFR regions.

Site-specific studies in Africa support our findings. Ruelland *et al.* (2012) predict a decrease of rainfall by 15-17% and a rise in potential evapotranspiration by 16-18% in the Bani River catchment located in Mali (SAH region), resulting in a critical reduction of water availability. The effects of

climate change on groundwater recharge in the upper Ssezibwa catchment, Uganda (the SEQ region) have been investigated by Nyenje and Batelaan (2009), who implemented a hydrological model for simulating future climate-change scenarios. They predict a combined increase of rainfall and groundwater recharge by 100% in 2080.

#### 3.2.2 Seasonality of PGR

The rainfall regimes during the four seasons are complex in both the Northern and Southern Hemispheres (Nicholson, 2000). Seasonal variation of P and PGR is considered both in gridded maps (Fig.8) and in latitudinal averaged basis (Fig.9). Seasonal rainfall maxima are observed in the SAH zone in summer and in the SAFR zone in winter. Nevertheless the Equatorial zones (NEQ and SEQ) are characterized by two rainy seasons (spring and fall). However, large longitudinal variability is observed in the MED and SAFR zones in winter.

In general, PGR is in-phase with rainfall patterns in the summer and winter, while the two regimes are out-of-phase in the Southern Hemisphere during the spring and in the Northern Hemisphere in the fall.

In the MED region (above the Tropic of Cancer), rainfall is mostly concentrated in the spring, fall, and winter. Episodic and intense rainfall events without PGR response occur in spring, especially over the Atlas Mountain range in northwestern Africa (Jasechko *et al.*, 2014). The summer in this region is very dry, whereas fall and winter are characterized by rainfall events with recharge occurring only in winter (peak of 0.1 m yr<sup>-1</sup> in the northern extreme). The winter is affected by a significant frequency of intense storms (high  $f_{max}$ -values; see Fig. 5h) that trigger PGR occurrences in the vadose zone.

The region delimited by the two Tropics (which includes the southern portion of Sahel, Equatorial central Africa, Horn of Africa, and the northern portion of southern Africa) is affected by marked seasonal trends of rainfall and PGR. It is interesting to note that P and PGR oscillate with a moving bell-shaped curve, with minima corresponding to the two tropical lines and the maximum point over the Equator in spring and fall.

In the spring, P is almost symmetrically partitioned by the Equatorial line which delimits the maximum point (0.44 m yr<sup>-1</sup>); conversely, PGR is 70% in the Southern Hemisphere and 30% in the

Northern Hemisphere, with maxima (about 0.44 m yr<sup>-1</sup>) spanning from the Equator down to the border between SEQ and SAFR. The rainfall over the Tropic of Capricorn is 0.08 m yr<sup>-1</sup>. In the summer, the rainfall line shifts towards the north (79% rainfall in the north), with its peak at 11.05°N (about 0.60 m yr<sup>-1</sup>), followed by PGR (91% recharge in the north), with its peak (0.13 m yr<sup>-1</sup>) at 7.35°N. In the fall, the rainfall line moves back and reaches its peak (0.51 m yr<sup>-1</sup>) immediately below the Equator (Table 5). Surprisingly, we observed a rainfall value of 0.14 m  $yr^{-1}$  in the Tropic of Capricorn, indicating a rainfall concentration over the area that includes Zimbabwe, Mozambique, and Botswana. Indeed, rainfall in this season is not symmetrically subdivided by the Equator, but with a disproportion of 66% in the Southern Hemisphere (mostly in the South Equatorial central Africa; Fig. 9c) and 34% in the Northern Hemisphere. On the contrary, PGR remains concentrated in the Northern Hemisphere (76%) with its peak (0.14 m yr<sup>-1</sup>) above the Equator and a gradual decrease northward with zero values registered in central Sahel. Finally, in the winter, the rainfall line moves southward with its maximum point (0.66 m yr<sup>-1</sup>) at -11.25°N and 86% of rainfall in the Southern Hemisphere. The severe rainfall events prompt PGR, even in zones that remain arid for the rest of the year. In this period, the PGR follows the rainfall with a very similar proportion (82% in the south) and its peak of 0.13 m yr<sup>-1</sup> at -11.75°N.

### 4. Conclusions

Groundwater levels of shallow aquifers worldwide have been experiencing a decreasing trend over the last few decades. This is generally due to groundwater pumping surpassing groundwater recharge rates (Wada *et al.*, 2014). Nonetheless, climate-related changes in groundwater recharges merit further analysis because their impact might alleviate or exacerbate the observed general decreasing trend (Loaiciga *et al.*, 2000).

In light of future climate-change predictions, we are particularly interested in understanding how PGR reacts to temperature increases and changing storm characteristics, particularly because their impact on groundwater recharge remains poorly understood in Africa. The conditions that favor or limit PGR over evapotranspiration or runoff generation depend largely on the nonlinear unsaturated zone

response to rainfall. These results highlight the three main following potential consequences in climate-change conditions caused by continental warming: (1) increasing temperatures occurring in the Northern Hemisphere can potentially trigger a decrease of PGR, especially in the NEQ region; (2) the decreasing trend of PGR observed in the last three decades in the SAH region is not correlated to the temperature or  $\tau_{dry}$ , even though the SPEI values point out a notable decrease of 15% of probability for wet conditions; and (3) storm occurrences affect the humid regions of Equatorial Africa (NEQ and SEQ), and the increasing trends of  $f_{max}$  and PGR have been observed in SAFR for the last 30 years. Therefore, the rise in temperature and storm occurrences, which can be considered warning signs of climate change, impact PGR in contrasting modes in the Northern and Southern Hemispheres of Africa. These outcomes align with previous continental-scale predictions carried out by Faramarzi *et al.* (2013), and the observed historical trends partially contrast future projections provided by Döll and Flörke (2005).

Temporal analysis reveals the role of seasonality with evident latitudinal patterns and disproportional distribution of rain and PGR between the Northern and Southern Hemispheres. The rainfall and PGR regimes are out of phase in the spring and fall, and synchronous in the summer and winter.

In view of the empirical relations observed in this study, future scenarios can concentrate on specific statistical analyses to link storm characteristics and temperature variations with PGR in simplistic models used for water-resources management. The observed model uncertainty is high all across Africa, with the exception of SAFR. Therefore, the outcomes observed in our analysis might be potentially realistic only for southern Africa. Nonetheless, we emphasize the importance of offering coarse guidelines on the relations between climate change and PGR as representing preliminary outcomes for future applications, that will be hopefully tested by direct groundwater-recharge measurements.

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**Figure 1.** Gridded maps of average annual (a) rainfall (P) and (b) potential evapotranspiration  $(ET_p)$ ; latitudinal patterns of average annual (c) P and (d)  $ET_p$ . The ranges of values between the 25<sup>th</sup> and 75<sup>th</sup> percentiles are represented by the gray bands. The horizontal dotted lines represent the Tropic of Cancer and the Tropic of Capricorn, while the horizontal dashed line depicts the Equator.

Accept



**Figure 2.** Gridded maps of average annual (a) potential groundwater recharge (PGR), (b) actual evapotranspiration ( $ET_a$ ), and (c) runoff ( $Q_r$ ); latitudinal patterns of average annual (d) PGR, (e)  $ET_a$ , and (f)  $Q_r$ . The ranges of values between the 25<sup>th</sup> and 75<sup>th</sup> percentiles are represented by the gray bands. The horizontal dotted lines represent the Tropic of Cancer and the Tropic of Capricorn, while the horizontal dashed line depicts the Equator.

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**Figure 3.** (a) Gridded patterns of the five climate classes and six climatic regions in Africa; (b) average annual PGR (black circles) delimited by the  $25^{\text{th}}$  and  $75^{\text{th}}$  percentiles in each climatic region (vertical black lines). The gray bar charts depict the relative contributing areas, AC (%).

Accepted



**Figure 4.** Relationship between average annual rainfall (P) and potential groundwater recharge (PGR) values (gray squares) along with trend lines (dotted lines) and correlation coefficients (R<sup>2</sup>) in (a) MED, (b) SAH, (c) NEQ+HORN, (d) SEQ, and (e) SAFR.

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**Figure 5.** Gridded maps of (a) storm occurrences ( $f_{max}$ ) and (b) mean dry consecutive days ( $\tau_{dry}$ ); latitudinal patterns of (c)  $f_{max}$  and (d)  $\tau_{dry}$ . The ranges of values between the 25<sup>th</sup> and 75<sup>th</sup> percentiles are represented by the gray bands. The horizontal dotted lines represent the Tropic of Cancer and the Tropic of Capricorn, while the horizontal dashed line depicts the Equator.

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**Figure 6.** Temporal evolution of 12-month SPEI values from 1901-2013 along the transect at longitude 15.25°E.

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**Figure 7.** Frequency distribution of the 12-month SPEI values from 1901-2010 (blue bars) and from 1980-2010 (red bars) in the six climatic regions of Africa (MED, SAH, NEQ+HORN, SEQ, and SAFR).



Figure 8. Gridded maps of P (top panel) and PGR (bottom panel) in spring, summer, fall, and winter.



**Figure 9.** Latitudinal trends of P (black lines) and PGR (red lines) in spring, summer, fall, and winter. The horizontal dotted lines represent the Tropic of Cancer and the Tropic of Capricorn, while the horizontal dashed line depicts the Equator.

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Table 1 - List of bibliographic references in 13 countries of Africa reporting spatial resolution (SP), average ( $\mu$ ), range (min and max) and method of measured recharge. PCR-GLOBWB simulated recharge values are reported as well and compared to measured recharge rates through the relative differences (RD).

_									
			Simulations						
	Country	Reference	SP	μ	min	max	method	PCR-GLOBWB	RD
-			km <sup>2</sup>	m yr <sup>-1</sup>	m yr <sup>-1</sup>	m yr <sup>-1</sup>		$m yr^{-1}$	%
	Botswana	de Vries et al. (2000) <sup>†</sup>	4,875	0.005	0.001	0.010	CMB	0.003	7.1
	Burundi	Bakundukize et al. (2011)	1,050	0.235	0.141	0.306	WB	0.345	-26.6
	Cameroon	Fouépé Takounjou et al. (2011)	11	0.108	0.016	0.237	WTF+CMB	0.252	-38.1
	Ethiopia	Demlie et al. (2007)	1,500	0.265	0.088	0.418	GRT	0.189	25.4
	Niger	Leduc et al. (1997)	595	0.055	0.050	0.060	WTF+CMB	0.005	84.6
	Nigeria	Edmunds et al. $(2002)^{\dagger}$	18,000	0.030	0.014	0.049	CMB	0.088	-30.8
	Senegal	Edmunds and Gaye (1994) $^{\dagger}$	1,600	0.016	0.005	0.034	CMB	0.069	-35.4
	Sudan	Abdalla (2009) <sup>†</sup>	60,000	0.006	0.004	0.008	GRT	0.036	-35.1
	Tanzania	Nkotagu (1996) <sup>†</sup>	-	0.013	0.010	0.016	CMB	0.075	-40.4
	Togo	Akouvi et al. (2008)	3,500	0.063	0.038	0.088	GRT	0.139	-28.7
	Uganda	Taylor and Howard (1996) $^{\dagger}$	840	0.200	-	-	WB	0.294	-23.9
	Zambia	Houston (1982) <sup>†</sup>	40	0.080	-	-	WB	0.118	-15.4
	Zimbabwe	Sibanda et al. (2009)	-	0.025	0.002	0.062	WTF+CMB+GRT	0.049	-18.0

<sup>†</sup> included in Bonsor and MacDonald (2010)



Table 2- Descriptive statistics of precipitation (P) and potential groundwater recharge (PGR) with mean ( $\mu$ ), standard deviation ( $\sigma$ ), coefficient of variation (CV), minimum (min) and maximum (max) in the six climatic regions of Africa (MED, SAH, NEQ+HORN, SEQ and SAFR).

	MED		SAH		NEQ+HORN		SEQ		SAFR	
	$27^\circ < \text{lat} < 37^\circ$		$10^\circ < lat < 20^\circ$		$0^{\circ} < lat < 10^{\circ}$		$-10^{\circ}$ < lat < $0^{\circ}$		$-35^{\circ} < lat < -10^{\circ}$	
	Р	PGR	Р	PGR	Р	PGR	Р	PGR	Р	PGR
$\mu$ m yr <sup>-1</sup>	0.22	0.04	0.40	0.05	1.24	0.27	1.32	0.26	0.68	0.08
$\sigma$ m yr <sup>-1</sup>	0.17	0.03	0.31	0.05	0.12	0.06	0.12	0.07	0.27	0.07
CV %	79	80	80	118	9	22	9	27	40	82
min m yr <sup>-1</sup>	0.02	0	0.03	0	0.99	0.14	1.18	0.18	0.36	0
max m yr <sup>-1</sup>	0.60	0.12	0.94	0.15	1.48	0.38	1.52	0.42	1.19	0.24

Table 3- Spearman's correlation coefficients be	etween PGR and temperature (T) a	and rainfall indicators
$(f_{max}, \tau_{dry})$ in the six climatic regions of Africa (	MED, SAH, NEQ, HORN, SEQ a	and SAFR).

	MED	SAH	NEQ	HORN	SEQ	SAFR
Т	-0.10	-0.09	-0.48	0.03	-0.20	-0.76
$\tau_{dry}$	-0.45	-0.19	-0.01	-0.12	-0.06	-0.38
$f_{max}$	0.03	-0.01	0.45	0.06	0.39	0.59

Table 4- Mann-Kendall trend test for annual temperature (T), rainfall indicators ( $f_{max}$ ,  $\tau_{dry}$ ) and potential groundwater recharge (PGR) in the six climatic regions of Africa (MED, SAH, NEQ, HORN, SEQ and SAFR). Significant positive and negative values are indicated in bold text

	MED	SAH	NEQ	HORN	SEQ	SAFR
Т	4.59	3.94	3.54	0.56	0.78	-0.68
$\tau_{dry}$	-0.08	-0.85	-1.17	-0.10	-0.22	-0.12
$f_{\rm max}$	-0.14	-1.82	1.53	0.75	-1.02	3.55
PGR	-0.03	-2.07	-2.58	1.84	-0.27	2.33

Table 5 - Seasonal partition of P and PGR (in %) between Northern and Southern Hemispheres.

	Spring		Summer		Fall		Winter	
	Р	PGR	Р	PGR	Р	PGR	Р	PGR
	%	%	%	%	%	%	%	%
Northern Hemisphere	46	29	79	91	34	76	14	18
Southern Hemisphere		71	21	9	66	24	86	82