

Regional co-variability of spatial and temporal soil moisture - precipitation coupling in North Africa: an observational perspective

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Abstract. The magnitude and sign of soil moisture–precipitation coupling (SMPC) is investigated using a probability-based approach and 10 years of daily microwave satellite data across North Africa at 1° horizontal scale. Specifically, the co-existence and co-variability of spatial (i.e. using soil moisture gradients) and temporal (i.e. using soil moisture anomaly) soil moisture effects on afternoon rainfall is explored. The analysis shows that in the semi-arid environment of the Sahel, the negative spatial and the negative temporal coupling relationships do not only co-exist, but are also dependent of one another. Hence, if afternoon rain falls over temporally drier soils, it is likely to be surrounded by a wetter environment. Two regions are identified as SMPC "hot-spots". These are the south-western part of the domain (7–15° N, 10° W–7° E) with the most robust negative SMPC signal, and the South Sudanian region (5–13° N, 24–34° E). The sign and significance of the coupling in the latter region is found to be largely modulated by the presence of wetlands and is susceptible to the amount of long-lived propagating convective systems. The presence of wetlands and an irrigated land area is found to account for about 30 % of strong and significant spatial SMPC in North African domain. This study provides the first insight into regional variability of SMPC in North Africa, and supports potential relevance of mechanisms associated with enhanced sensible heat flux and mesoscale variability in surface soil moisture for deep convection development.

1 Introduction

Soil moisture can affect the state of the lower atmosphere through its impact on evapotranspiration and surface energy flux partitioning (e.g., Eltahir, 1998; Klüpfel et al., 2011). Especially in the "hot-spots" of soil moisture–precipitation coupling (SMPC), like the semi-arid Sahel (Koster et al., 2004; Taylor et al., 2012; Miralles et al., 2012), soil moisture exerts strong control on evapotranspiration (e.g., Timouk et al., 2009; Dirmeyer, 2011; Lohou et al., 2014), influencing the development of the daytime planetary boundary layer (BL), and hence convective initiation and precipitation variability. Most of the physical understanding on how soil moisture could alter BL properties and affect development of convection comes from 1-D to 3-D model analyses (Seneviratne et al., 2010; Nicholson, 2015). Observational evidence of the SMPC largely relies on the measurements of recent field campaigns (like African HAPEX and AMMA: Goutorbe et al., 1994; Redelsperger et al., 2006) and hence, is often limited to a short spatio-temporal scale. Such observational analyses present a unique evidence of environmental

conditions preceding convection development (e.g., Lothon et al., 2011) and can be further used as a test-ground to evaluate and improve the physical parametrizations of models (e.g., Couvreur et al., 2013). Both, observational and modelling studies agree reasonably well on the effect of soil moisture availability and heterogeneity on the lower atmospheric stability (e.g., Kohler et al., 2010) and convective initiation at mesoscales (e.g., Taylor, 2010; Birch et al., 2013). However, the impact of soil moisture on convective precipitation remains more uncertain. At mesoscale, there is a disagreement in the sign of the SMPC between models which use parametrizations of deep convection and observations (Hohenegger et al., 2009; Taylor et al., 2012, 2013). Recent satellite-based analysis has demonstrated that the choice of soil moisture parameter itself (temporal anomaly or spatial gradient) and related differences in physical mechanisms have a direct effect on the resulting sign of the coupling. Guillod et al. (2015) found a positive temporal (i.e. using soil moisture anomaly) but negative spatial (i.e. using spatial soil moisture gradient) SMPC over most of the globe at 5° horizontal scale. Our study addresses the question of co-existence of spatial and temporal SMPC at a finer $1^\circ \times 1^\circ$ horizontal grid. Specifically, we use 10-year satellite records of daily soil moisture from the AMSR-E and 3-hourly TMPA precipitation product to investigate spatio-temporal co-variability of observed coupling relationships in the region of North Africa.

Both modelling and observational studies reported the possibility of negative as well as positive SMPC (Nicholson, 2015). Spatial gradients in soil moisture can affect BL state and convection initiation through thermally-induced mesoscale circulations on 10 to 100 km scales (Taylor and Ellis, 2006; Taylor et al., 2007; Rochetin et al., 2017). In association with this mechanism and under favourable thermodynamic conditions, convection is likely to be initiated over spatially drier soils, indicating a negative SMPC (Taylor et al., 2011; Garcia-Carreras et al., 2011). However, whether the further development and propagation of moist convection will occur over drier or wetter soils remains less clear. The modelling study of Froidevaux et al. (2014) suggested that negative SMPC is possible under very weak surface wind conditions, and is associated to stationarity of convective systems once initiated. The opposite sign is expected under a stronger horizontal advection, which will support propagation of the developing moist convection downwind, i.e. from dry to wet soils, and its further amplification over wetter areas. Another important factor is related to the life-cycle of mesoscale convective systems (MCS) and thus their organization in space and time (Mathon et al., 2002). Small-scale convective systems are expected to be particularly sensitive to surface moisture variability and will propagate preferentially towards spatially drier soil, bounded by a wetter surrounding (Taylor and Ellis, 2006). Alternatively, larger organized systems have been found to evolve towards wetter soils - areas of increased latent heat flux, convective available potential energy (CAPE) and moist static energy (MSE) (Taylor and Lebel, 1998; Clark et al., 2003). Hence, these systems are expected to be more sensitive to soil moisture availability.

The impact of temporal anomalies of soil moisture on the atmospheric BL and moist convection is largely governed by thermodynamical processes, and may likewise result in a coupling of both signs. Wet soils are expected to lead to an increase of boundary layer MSE or similarly equivalent potential temperature, through a decreased boundary layer height (BLH) and subsequent less vigorous entrainment (e.g. Eltahir, 1998; Alonge et al., 2007). The enhanced MSE over wet soils is favourable to convective rainfall formation. Dry soils, on the contrary, are associated with a reduced MSE and thus provide lower potential for convection development and may even suppress existing MCS (Gantner and Kalthoff, 2010) or deviate its propagation direction (Wolters, 2010). However, modelling and observational evidence indicate that both dry and wet soils, can favour

moist convection, depending on the morning stratification of the lower atmosphere (Findell and Eltahir, 2003a, b) into which the daytime convective BL is growing (Ek and Mahrt, 1994; van Heerwaarden and Guerau de Arellano, 2008; Couvreux et al., 2012).

The relevance of mesoscale spatial heterogeneity of soil moisture in favouring moist convection over wet and dry temporal soil moisture anomalies was demonstrated by e.g. Clark et al. (2003) and Taylor and Ellis (2006) respectively. However, until recently no attempts were made to directly compare the temporal and spatial aspects. The first comparison of the spatial and temporal effects of soil moisture on precipitation was presented by the study of Guillod et al. (2015) (hereafter, G15) at 5° horizontal scale. Applying the probability-based approach of Taylor et al. (2012) (hereafter, T12) to 10-years of global satellite-based soil moisture and precipitation data, they demonstrated that a negative spatial (rain over spatially drier soils) and a positive temporal (rain over temporally wetter soils) SMPC dominate over most of the globe and do not exclude one another. G15 suggested that the two effects might be interconnected through spatial coupling mechanisms, in which adjacent precipitation would provide required moisture to enhance convection development over spatially but not temporally drier soil. Using multiple data sets, G15 showed that the signal is robust across different input data sets. However, in a few regions, including the Sahel in Africa, an opposite temporal relationship was revealed: spatially and temporally negative coupling was found to co-exist in opposition to the global relationship.

In this study, we further explore spatial and temporal SMPC as well as their co-existence in North African region using the T12 method at a finer 1° × 1° horizontal grid. Furthermore, we provide insight into regional co-variability of the spatial and temporal effects on afternoon rainfall. The analysis is realized following two main steps:

1. Identification of the factors that influence the magnitude and variability of the spatial SMPC measure. By doing so we address the question: which physical processes likely underlie the observed spatial SMPC relationship if any?
2. Analysis of the link between the spatial and temporal effects of soil moisture on precipitation. This part addresses the two following questions: are spatial and temporal negative coupling relationships independent, and if not, how do they inter-relate?

We reproduce and apply the probability-based approach of T12 to 10-years of daily AMSR-E soil moisture and 3-hourly TMPA precipitation records. In contrast to the previous studies, we estimate the temporal and spatial coupling effects at a finer 1° horizontal scale, which reveals previously hidden smaller-scale effects of land cover features on the SMPC relationship.

The first part of the study is embraced in *section 4* and includes an analysis of the regional variability and robustness of the observed spatial SMPC at varying horizontal scales i.e. from the original 5° to 2.5° and 1°. Identification of the factors relevant for the observed spatial SMPC distribution includes a sensitivity analysis of the spatial coupling measure to the presence of soil moisture parameter extremes (*section 4.3*) and to the MCS life cycle (*section 4.4*). The link between the temporal and spatial SMPC is assessed using correlation analysis in *section 5.1*. Summary and discussion *section 6* is delivered in two parts. *Section 6.1* provides main conclusions on the potential physical mechanisms which likely underlie the observed coupling relationships. *Section 6.2* reviews the reasons behind the opposite sign of the temporal coupling identified in the North African region as

compared to the dominantly positive relationship identified in G15 over the globe. The paper concludes with more general discussion of the SMPC "hot-spots" and conclusions in *section 7*.

5 2 Domain and Data

2.1 Study domain

We focus our analyses on the North African region (5–20° N, 20° W–40° E) (Fig. 1, dashed rectangle) during summer period (JJAS). This region has been repeatedly pointed out as a "hot-spot" of land–atmosphere interactions (Dirmeyer, 2011; Miralles et al., 2012; Gallego-Elvira and Taylor, 2016), and one of the most vulnerable regions with respect to climate change (Dirmeyer et al., 2012; Dirmeyer and Wang, 2014). A major feature affecting the Sahelian climate is the West African Monsoon (Janicot and Thorncroft, 2008), which is associated with a high precipitation variability (Nicholson, 2013). During the monsoon, soil moisture fluctuations are strongly influenced by precipitation at a large range of spatial and temporal scales. Atmospheric and surface fields display strong meridional gradients between 10° N and 20° N (Figure 1, zonal plot) shaped by the migration of the summer time rain belt, also referred as the Inter Tropical Convergence Zone (ITCZ). Wind convergence at the surface is observed further north, around 18–20° N, along the Inter Tropical Discontinuity (ITD), where the cool and moist monsoon flow meets hot and dry Saharan air. Associated with the meridional heat gradient, the monsoon circulation and related large-scale structures like the African Easterly Jet (AEJ), as well as synoptic disturbances like the African Easterly Waves (AEWs) largely modulate convection activity over the region (Duvel, 1990; Mohr and Thorncroft, 2006). Additionally, evidence supporting a significant role of the surface state in the triggering of deep precipitating convection is steadily growing (Nicholson, 2015).

20 Middle July to August conditions may be less favourable for a strong surface influence on convection. Compared to the drier early and late monsoon months of June and September, the wetter period - from July through August - is characterized by a typically lower level of free convection (LFC) (Guichard et al., 2009; Taylor et al., 2011) and less pronounced spatial contrast between fluxes due to more dense vegetation (Kohler et al., 2010; Lohou et al., 2014). In our study, the role of the monsoon dynamics is not addressed to preserve maximum sample size for the sake of statistical significance.

25 We intentionally extend our analysis further eastwards. Despite the inherent zonal symmetry of surface and atmospheric parameters (as in precipitation in Fig. 1), considerable differences exist in rainfall and large-scale circulation regimes between East and West. Distinct orography, intensity of surface and upper level jets and wave disturbances are likely to bring dissimilarities in the sensitivity of convection to the surface state between the two regions. The eastern African domain can also remotely influence convection in the western part of the region via the genesis of westward propagating AEWs (Laing and Carbone, 2008) and long-lived MCSs (Laing et al., 2012). Yet, notably few studies have investigated land–atmosphere interactions in eastern Sahel.

2.2 AMSR-E soil moisture

Soil moisture (SM) data from the Advanced Microwave Scanning Radiometer - Earth Observing System (AMSR-E, Jun 2002–Oct 2011) is analyzed in this study. The AMSR-E unit is carried on board of the polar orbiting AQUA satellite, measuring brightness temperatures in 12 channels, at 6 different frequencies (6.9–89 GHz) (Njoku and Jackson, 2003). Soil moisture derived from the lowest C-band frequency of 6.9 GHz is used here, as lower frequencies experience less signal contamination from vegetation and surface roughness, and are able to receive emission information from deeper soil layers (still just a few centimetres, Owe et al., 2008). The AQUA orbit is sun-synchronous with typically one overpass per pixel per day at either 13:30 or 01:30 local solar time (LST). In order to capture the surface moisture state shortly before afternoon convection activity, only data of ascending day orbit, i.e. 13:30 LST is used here. It is important to note, that the day overpass is prone to higher biases compared to the night overpass, because of the greater temperature differences between surface and canopy involved in the physics algorithm (Njoku and Jackson, 2003).

We utilize the Level 3 estimates of AMSR-E soil moisture derived with the Land Surface Parameter Model (LPRM, Owe et al., 2008) for JJAS 2002–2011. The product is available at $0.25^\circ \times 0.25^\circ$ spatial resolution. The LPRM is not valid for dense vegetation and water bodies. Therefore pixels with an optical depth > 0.8 are excluded. Water body and soil moisture quality masks were adopted from the materials of T12. Accordingly, pixels containing more than 5 % water are excluded, using water body classification of the 1 km Global Land Cover 2000 data set [Available online at <http://forobs.jrc.ec.europa.eu/products/glc2000/products.php>]. Application of the soil moisture quality mask, based on the correlation analysis between precipitation and soil moisture data sets, is intended to reduce the number of pixels covered with wetlands (for details see suppl. in T12).

Many days (40–50 %) do not contain soil moisture information due to satellite revisit times. Over a given longitude per day the number of overpasses below 40° N do not exceed one with occasionally daily or every third day sampling (see Fig. 1 in Njoku and Jackson, 2003). The latter significantly reduces the size of the collected rainfall event sample available for the analyses. The AMSR-E instrument is chosen because it documents a relatively long period and performs better than ASCAT (Dorigo et al., 2010) over sparsely vegetated and deserted areas. The AMSR-E product also proved to be accurate at the precipitation event scale in capturing rain-related soil moisture variability and timing, when compared with in situ data in the Sahel (Gruhler and Rosnay, 2008).

2.3 TMPA-v7 precipitation

The Tropical Rainfall Measuring Mission (TRMM) Multi-satellite Precipitation Analysis (TMPA) represents a partial-global coverage (50° S– 50° N) product of combined precipitation estimates. Three-hourly precipitation time-series of the TMPA product (1998–2015, Huffman et al., 2007) at $0.25^\circ \times 0.25^\circ$ horizontal resolution are used to estimate locations of afternoon convective precipitation events in the study.

The TMPA algorithm (Huffman et al., 2007) involves the following steps: (1) - merging multiple independent passive microwave sensors, (2) - their inter-calibration to the TRMM Combined Instrument (TCI) precipitation estimates, (3) - further

30 blending with preliminary calibrated infra-red products from geostationary satellites, and finally (4) scaling of the estimates to match monthly accumulated Global Precipitation Climatology Center (GPCC) rain gauge data.

In this study we utilize the product version 7 (TRMM-3B42), which includes several modifications to the algorithm and additional satellite data (Huffman and Bolvin, 2014). Consistent with the soil moisture record, only 10 years (2002–2011) of JJAS precipitation data is used. To ensure similar solar forcing on the surface and boundary layer, the 3-h precipitation time-series for the present application are adjusted to LST (based on longitude) by taking the closest 3-h UTC time step. It is important to note that any 3-h TMPA value is not referred directly to its nominal hour, but represents the average of the "best" 5 overpass data within a 3 hourly window, centered around the nominal hour, i.e. +/- 90 min range. Variable time of the TMPA "best" data average is not expected to significantly affect our SMPC results.

3 Methods

3.1 Description of statistical framework after T12

The SMPC in this study is referred to as the relationship between the afternoon convective rainfall and the antecedent soil 10 moisture conditions. Using the method of T12, we examine first whether afternoon rainfall is more likely over soils that are untypically (relative to its climatology) drier or wetter than their surrounding. Next, following the definition of G15, we assess whether afternoon rainfall is more likely on days when soils are untypically drier or wetter than their temporal mean. Subsequently, the higher than expected probability of convective rainfall events to occur over spatially drier or wetter soils is referred to as *spatial SMPC*, while the higher than expected likelihood of convective rain events to occur over temporally 15 wetter or drier soils quantifies *temporal SMPC*. The following paragraph describes criteria which are used to define a convective precipitation event, and evaluate soil moisture statistics antecedent to every event. The framework algorithm implemented in this study largely follows the method of T12 and is summarized in Fig. 2.

3.1.1 Definition of convective rainfall event (Fig. 2C)

We define a convective event location, L_{max} , as the location where accumulated afternoon precipitation between 15:00–21:00 20 LST exceeds a threshold of 6 mm. Then, locations of afternoon accumulated precipitation minima, L_{min} , are identified within a 5×5 pixel box ($1.25^\circ \times 1.25^\circ$)¹ centered at L_{max} (Fig. 2C). The choice of a later accumulation time than in T12 (i.e. 15:00–21:00 LST instead of 12:00–21:00 LST) ensures that the soil moisture measurement at 13:30 LST precedes precipitation without introducing additional filters. The twice larger afternoon accumulated rainfall threshold than in T12 yields qualitatively similar results, though leads to a slightly higher mean SMPC significance over the domain. According to additional sensitivity 25 tests, the choice of higher threshold values in the method mostly influences the amount of significant grid boxes linked to a reduction in the event sample size, yet does not qualitatively affect the dominant preference of the afternoon rainfall over specific soil moisture conditions (Petrova, 2017).

¹Following T12, a box size of $1.25^\circ \times 1.25^\circ$ is selected as minimum possible size to resolve soil moisture variability around the center of the box, taking into account the 50 km footprint of the AMSR-E soil moisture.

The following set of assumptions is used to improve the accuracy of the convective event sample. If one of the conditions is not fulfilled, an event is excluded from further calculation:

- 30 (1) - accumulated precipitation in the preceding hours (06:00–15:00 LST) in the entire $1.25^\circ \times 1.25^\circ$ box must be zero;
- (2) - elevation height difference within the event box must not exceed 300 m. The latter is done to minimize the effect of orographic uplifting on the rainfall variability. The resulting distribution of the orography mask is shown in Fig. 3a.
- (3) - number of identified L_{min} locations within one box must be 3 or more (for averaging reasons). In that case, all L_{min} locations will have the same afternoon accumulated precipitation value, which will most often be zero.
- (4) - if boxes overlap, the event with larger afternoon accumulated precipitation value is retained.

5 3.1.2 Soil moisture statistics in event locations (Fig. 2D - E)

Once events are identified, soil moisture anomaly S' measured prior to the precipitation event (at 13:30 LST) at L_{max} , $\overline{L_{min}}$ or any combination of the two is stored and analyzed. $S'^{\overline{L_{min}}}$ represents an average value of S' measured in every identified L_{min} location within a $1.25^\circ \times 1.25^\circ$ event box. S' has its climatological mean subtracted, calculated as linear averaging within a 21-day moving window centered on the particular day and across the entire multi-year record. To exclude contribution of a rain event onto the anomaly values, the year of the event is excluded from the climatological mean calculation. To estimate, if soils were anomalously dry or wet in the location of maximum rain (L_{max}) compared to the expected range (climatology) for that location, we store pre-event soil moisture anomaly at L_{max} , i.e. $Y_e: S'_e{}^{L_{max}}$ (Fig. 2D). To investigate if soils in the location of maximum rain (L_{max}) were drier than in the neighbour region(s) where it rained less or did not rain (L_{min}), we calculate the pre-event soil moisture gradient between L_{max} and $\overline{L_{min}}$ scaled per 100 m, i.e. $Y_e: \Delta(S'_e{}^{L_{max}}) = S'_e{}^{L_{max}} - S'_e{}^{\overline{L_{min}}}$ with the dimension of $m^3 m^{-3} 100 m^{-1}$, where Δ - stands for gradient.

For each Y_e parameter we define the control sample Y_c , represented by an array of corresponding Y values measured in the same L_{max} and L_{min} event locations in the same calendar month, but on the non-event years. The measure of coupling is then quantified by the magnitude of a difference between mean statistics of the event and control samples, $\delta_e = \text{mean}(Y_e) - \text{mean}(Y_c)$, and the measure of δ_e significance (Fig. 2E). Significance is represented by a percentile, P_e , of the observed δ_e in a bootstrapped sample of δ values that is observed by chance. For that Y_e and Y_c are pooled together and re-sampled without replacement 5000 times.

3.1.3 Definition of temporal and spatial SMPC (Fig. 2F)

Parameters of δ_e and P_e calculated for the soil moisture gradients $\Delta(S'_e{}^{L_{max}})$ prior to the event quantify preference of rain to occur over soils drier ($\delta_e < 0$, $P_e \leq 10\%$) or wetter ($\delta_e > 0$, $P_e \geq 90\%$) than its $1.25^\circ \times 1.25^\circ$ environment, and are referred to as negative or positive *spatial SMPC* respectively. The same parameters estimated for the temporal soil moisture anomaly $S'_e{}^{L_{max}}$ instead specify expressed preference of rain to occur over soils drier or wetter than its temporal mean, i.e. negative or positive *temporal SMPC* accordingly.

In this study, estimation of δ_e and its significance P_e for the spatial and temporal coupling is realized over $5^\circ \times 5^\circ$, $2.5^\circ \times 2.5^\circ$ and $1^\circ \times 1^\circ$ boxes. Aggregation of event statistics at a higher resolution than used in the global studies of T12 and

30 G15 results in a smaller event sample size per grid box, yet allows a reduction of the potential influence of meridional or zonal gradient in the parameter statistics, i.e. makes the spread in underlying surface and atmospheric moisture conditions across the box latitudes smaller (Section 4.2). The latter is valuable for the interpretation of obtained statistics in terms of land cover and atmospheric state. Hence, most of the study focuses on the smallest $1^\circ \times 1^\circ$ spatial grid.

3.2 Statistics of convective events

5 Application of the algorithm to the 10 years of JJAS AMSR-E soil moisture and TMPA precipitation time-series yields 10131 afternoon rainfall events. The distribution of identified events over the domain at $5^\circ \times 5^\circ$ and $1^\circ \times 1^\circ$ grid is shown in Figure 3b and 3c respectively. The signature of orography and large-scale dynamic effects on event occurrence becomes more evident at the finer event-aggregation scale, thus giving an advantage to the highest horizontal resolution. Figure 3c shows that most events occur between 10° N and 18° N, and the occurrence maxima are zonally aligned. Two maxima are found over the central Sahel, covering the area between 10° W– 15° E - aligned with the mean position of the AEJ core (Figure 1b). Another two maxima are evident at about 22° E and 30° E. Overall, the obtained distribution of identified rain events at $1^\circ \times 1^\circ$ grid resolution is consistent with the observed distribution of intense MCS over the region (Mathon and Laurent, 2001).

4 Results of spatial SMPC analysis

4.1 SMPC at 5° horizontal scale. Consistency to previous studies

15 We start our assessment by investigating the spatial soil moisture–precipitation coupling relationship. In agreement with the global-scale studies of T12 and G15, we find a dominantly negative spatial SMPC in the North African domain at the $5^\circ \times 5^\circ$ grid, i.e. a strong preference for convective rainfall events to occur over spatially drier soils (Fig. 4a). The majority of the $5^\circ \times 5^\circ$ boxes (72 %) have percentile values P_e lower than 10 %, implying a significant negative difference in the mean magnitude of soil moisture gradients $\Delta(S_e^{L_{max}})$ prior to the events relative to their typical (non-event) state. No significant positive difference between event and non-event conditions is found at the 5° scale (Table 1).

Figure 5 further compares the percentage of the domain area with significant negative and positive coupling identified in our study, T12 and G15 (see also Table A1). The differences arise due to disparities in the data sets and methodologies. The weakest negative and the strongest positive coupling signal corresponds to the estimates based on PERSIANN (Precipitation Estimation from Remotely Sensed Information using Artificial Neural Networks, Hsu et al. (1997)) data set from G15. This is possibly linked to the lower consistency between the PERSIANN precipitation and soil moisture variability in time (T12). On average, results based on different data set combinations summarized in Fig. 5 agree that afternoon precipitation occurs more often than expected over spatially drier soils in 42 % of the $5^\circ \times 5^\circ$ boxes, against only 4 % with a preference over spatially wetter soils (red and blue lines in Fig. 5 respectively).

The variability of spatial SMPC patterns among different data set combinations has shown to be quite strong over the globe and was not analyzed further in G15. We find, however, that in the North African domain, areas of significant negative spatial

coupling are fairly consistent. One of the most robust negative spatial SMPC signals is found in the south–western part of the domain (Fig. 4a,b). Fourteen out of 18 data set combinations summarized in Fig. 5, including this study, locate the cluster of the lowest percentiles roughly between 5–15° N and 10° W–10° E (Fig. 5, rhombs). This region occupies a relatively vast and flat area, associated with a reduced orographic forcing on convection development compared to the East, and a regional minimum in cold cloud occurrence (Laing and Carbone, 2008). The effect of large-scale structures like AEWs and AEJ on convection, on the contrary, is expected to be stronger in the western Sahel than further East. However, this does not exclude but may even favour higher sensitivity of convection triggering to soil moisture heterogeneities (Gantner and Kalthoff, 2010; Adler et al., 2011). Overall, the identified negative spatial SMPC relationship in the region is consistent with the recent observational- (Taylor and Ellis, 2006; Taylor, 2010; Lothon et al., 2011) and model-based (e.g. Gantner and Kalthoff, 2010; Garcia-Carreras et al., 2011; Birch et al., 2013; Taylor et al., 2013) studies in the Western Sahel.

Another cluster of the lowest percentiles and the largest differences in soil moisture state between event and non-event days δ_e is identified in the south–east of the domain (Fig. 4a,b). The proximity to the Ethiopian Highlands and the presence of extensive seasonally flooded regions in this area makes it generally difficult to isolate effect of surface state on convection. This also possibly led to less agreement in the spatial SMPC estimates identified in our study, G15, and T12 analyses (not shown). Unlike in the western Sahel, no accurate estimates of the SMPC exist in this eastern region.

4.2 Robustness of the negative SMPC at finer 2.5° and 1° scale

To identify factors and potential physical mechanisms that influence the magnitude and variability of the SMPC we reduce the event-aggregation scale to the finer 2.5° × 2.5° and 1° × 1° horizontal grid². In particular, aggregation of the convective rainfall events and corresponding soil moisture statistics over the smallest 1° × 1° grid boxes reveal more details on the effects of land surface conditions on the SMPC.

The percentile maps obtained for the finer scales are presented in Figures 4c and 4e. Despite the reduction in the amount of significant δ_e values, largely due to the decreased number of events in every box, negative spatial SMPC relationships remain dominant at the finer scales, and exhibit a similar spatial pattern as at the 5° × 5° grid. The featured regions of significant negative coupling now scale down to the territories of Burkina Faso, Benin, parts of Ivory Coast, Ghana and Mali (7–15° N, 10° W–7° E) in the West, as well as South Sudan (5–13° N, 24–34° E) in the East (Fig. 4e). In total, 42 % (21 %) of the boxes reveal significant negative difference δ_e for the 2.5° (1°) scale, versus initial 72 % at the 5° × 5° grid (Table 1).

The overall distribution of the difference δ_e does not change at the finer scales (Fig. 4d,f). However, multiple pixels with a positive δ_e emerge. For example, a small region enclosed between the Cameroon mountains and Jos Plateau (7° N; 8° E, Fig. 4f) now indicates a higher likelihood of rainfall to occur over spatially wetter soils. The relationship, though non-significant, is plausible. This area includes part of the Niger river valley and represents a prominent location of intense convection and a local maximum of the cold cloud occurrence, linked to the initiation of convection at the lee side of the high terrain (Laing and Carbone, 2008). The potential link between the land surface characteristics and SMPC parameter is explored in more detail in

²The SMPC statistic for the finer scales is calculated if at least 8 events in a box are present.

the following section. In total, 14 % of the $1^\circ \times 1^\circ$ boxes reveal a positive δ_e shift, against less than 6 % and 3 % for coarser $2.5^\circ \times 2.5^\circ$ and $5^\circ \times 5^\circ$ grids respectively.

4.3 Evidence for "wetland-breeze" mechanism in the SMPC statistics

In areas like the Cameroon Mountains where orography or floodplains have an effect on deep convection development, persistent wet and dry surface moisture patterns may pre-exist or develop, and therefore, lead to the occurrence of stronger than
 5 usual spatial soil moisture gradients. In the SMPC statistics, such gradients occur as extremes in a given distribution of soil moisture gradients $\Delta(S'_e{}^{L_{max}})$ within a $1^\circ \times 1^\circ$ box. Here, we define a gradient $\Delta(S'_e{}^{L_{max}})$ to be an extreme if it lies outside the following range: $(Q_{25} - 1.5 \times IQR)$ to $(Q_{75} + 1.5 \times IQR)$, where Q_{75} and Q_{25} are the third and first quartiles respectively, and the interquartile range (IQR) is the difference between them (Fig. 6a).

The distribution and magnitude of the extreme soil moisture gradients identified in the domain are shown in Figure 6c. We
 10 find that 28 % of all valid $1^\circ \times 1^\circ$ boxes contain extreme $\Delta(S'_e{}^{L_{max}})$. A large part of the extreme soil moisture gradients are located in the regions of significant negative coupling in the West and East. In these areas, extreme $\Delta(S'_e{}^{L_{max}})$ lead to an overestimation of the $\Delta(S'_e{}^{L_{max}})$ sample mean and therefore, δ_e magnitude. As a result, extreme gradients appear to predefine SMPC significance. Removal of extremes leads to a decrease in the number of boxes with significant negative spatial coupling by 30 %. However, in most cases the sign of the coupling remains unchanged (not shown).

15 Further analysis shows that extremes tend to cluster around major rivers and wetland areas in the East and West (Fig. 6b,c). Regional distribution of extreme $\Delta(S'_e{}^{L_{max}})$ is consistent with the distribution of natural wetland fraction presented earlier by Matthews and Fung (1987) and shown in Fig. 6b, and more recent estimates of inundation regions identified by Prigent et al. (2007). Strong positive soil moisture gradients are found around the Senegal river close to the coast, and on the lee side of the Cameroon Mountains. Strong negative soil moisture gradients are more numerous and seen all along the western flow of
 20 the Niger river, downwind of the permanent wetlands of Ez Zeraf Game Reserve and irrigated lands of the Gezira Scheme in Sudan. The scatter of the extremes in the East is likely related to the recurrent floods of the White Nile river. The reason for the absence of extremes around the Logone floodplains is obscure.

To verify the link between the location of extreme soil moisture gradients and flooded areas, we additionally identify $1^\circ \times 1^\circ$ boxes with the events, which L_{min} pixels were likely covered with a wetland or a floodplain, and hence were wetter than the
 25 neighbouring L_{max} location (Fig. 6c, crosses). This is done by calculating a linear regression function between climatology of one-day soil moisture drying rate (i.e. difference in soil moisture on the event day 0 and on day -1) and climatology of initial soil moisture value (i.e. soil moisture on day -1)³ (appendix Fig. A1). We consider the L_{min} location to be likely flooded, if (a) all initial soil moisture values are high, and (b) climatological values of drying rate are always small and do not vary much with the initial soil moisture (regression slope is close to zero). For the details the reader is referred to the Appendix A and Figure
 30 A1. From the Figure 6c it is seen that the distribution of grid boxes with the potentially flooded L_{min} locations conforms with the distribution of natural wetland fraction in Fig. 6b and includes the majority of extreme $\Delta(S'_e{}^{L_{max}})$ locations. This result supports the presence of the link between the afternoon rainfall maximum and flooded areas.

³The climatologies are calculated for every L_{min} location, for the same month as the event but for the non-event years.

The identified sensitivity of the afternoon rainfall to the strong negative soil moisture gradients around water bodies is in agreement with the results of the observational-based study of Taylor (2010). Analyzing 24 years of Meteosat brightness temperatures over the Niger Inland Delta, he found that convection was initiated more often over and to the east of the wetland in the morning hours. However, later in the day mesoscale convective systems tended to develop and propagate away from the wet areas towards drier soils, suggesting formation of deep convection and afternoon precipitation over negative soil moisture gradients. Similarly, observed by Alter et al. (2015), enhancement of rain to the East of irrigated land at 14° N, 33° E and its suppression over the Gezira Scheme itself is consistent with the location of negative (positive) extreme soil moisture gradients to the West (East) of the irrigated region (Fig. 6c). All the above supports the consistency of the observed afternoon rainfall intensification over the drier soils adjacent to flooded areas with "wetland–breeze" mechanism. The ability of the method to capture these effects is also noteworthy.

10 4.4 Effect of propagation of deep convective events on the SMPC statistics in eastern and western domains.

Another physical effect that may influence the SMPC relationship is related to the propagation and evolution of mesoscale convective systems (not accounted for by the current algorithm). Previous studies indicate that an opposite SMPC relationship might be expected at early versus late stages of MCS development (Taylor and Lebel, 1998; Taylor and Ellis, 2006; Taylor et al., 2010; Alonge et al., 2007; Clark et al., 2003; Gantner and Kalthoff, 2010). In this respect, a distinct strength or even sign of the spatial SMPC measure may result from separation of the rainfall events into those formed by a weaker and smaller MCSs - mostly found in the early afternoon - or by long-lived and organized MCSs - dominant during late afternoon hours (Mathon and Laurent, 2001). Differences in SMPC response to MCS life cycle are also expected to exist between the two regions of significant negative coupling, in the East and West. To characterize these differences, we analyze precipitation diurnal cycles averaged over event days in the East and West first (Fig. 7), and then estimate sensitivity of the spatial SMPC to varying rainfall accumulation times (Fig. 8).

The Hovmöller diagram of rainfall averaged over 1000 event days in the Western domain (black rectangle in Fig. 4) shows that intensification of the moist convection in the region is generally concentrated around main orographical features (Fig. 7a,c). The peak in precipitation occurs at similar times across the domain, and thereby does not reveal expressed signature of the system propagation. Most of the MCS are therefore expected to be shorter-lived and smaller, as their dissipation locations would be found close to their initiation (Mathon and Laurent, 2001).

In the East, on the contrary, the strong south–western propagation component of moist convection dominates the zonal progression of the most intense rainfall during diurnal cycle averaged over 754 event days (Fig. 7d). A large number of MCS initiate at the lee side of the Ethiopian Highlands and propagate westward undergoing cycles of regeneration and growing into a mature and organized MCS (Laing and Carbone, 2008; Laing et al., 2012). The emergence of an absolute rain rate maximum downwind of the permanent wetlands of the Ez Zeraf Game Reserve (at 30° E) during afternoon hours supports an influence of the flooded areas on moist convection intensification in the region (Fig. 7d,f). Consistent with the results of Taylor (2010) obtained for the Niger Inland Delta, the presence of wetlands in the Eastern domain is expected to increase the number of organized and long-lived propagating MCS in the late afternoon, originating from either locally triggered MCS, i.e. formed at

the dry land–wetland boundary, or from re-intensified pre-existing westward propagating systems. We may therefore expect a greater sample of long-lived and organized propagating MCS to be found in the late afternoon hours in the Eastern than in the Western domain. Accordingly, the response of the SMPC statistics to propagating MCS is expected to be stronger in the East compared to the West. Figure 8, which shows the change of the SMPC parameter between different rainfall accumulation time periods confirms this hypothesis. For this assessment an additional area in the North is considered (dashed rectangle in Fig. 4), as representative of a region where large-scale atmospheric and surface conditions differ from those of the East and West domains.

From Figure 8 it is seen that the earlier rainfall accumulation time periods, i.e. 12:00–18:00 UTC in the East and 15:00–21:00 UTC in the West and North, result in the strongest negative δ_e difference, and hence spatial SMPC relationship in all three domains. No positive δ_e values are found for these time periods, and the fraction of negative soil moisture gradients preceding rainfall events is relatively high, i.e. 62 %, 57 % and 55 % for East, West and North accordingly. Later accumulation times lead to a decrease in the magnitude and significance of the coupling parameter δ_e , and an increase in its spatial variability across the domains. These changes are associated with an increase in the amount of the positive soil moisture gradients in the regions (not shown).

Despite the similarities, differences in the SMPC response exist between the domains. In the East, the spatial SMPC shows the strongest sensitivity to the rainfall accumulation time and switches the sign to a positive one for the 18:00–24:00 UTC period. In accordance with Fig. 7d, the 18:00–24:00 UTC period reflects the afternoon progression of the mature MCS formed during early afternoon hours at the Ethiopian Highlands and around wetlands. According to Taylor and Lebel (1998) and Taylor et al. (2010), large and organized MCSs are expected to be more efficient in developing over wetter soils, associated to a well expressed BL moisture anomaly and higher MSE and CAPE and, at the same time, might get suppressed over drier surfaces (Clark et al., 2003). These observations are consistent with the identified here increase in fraction of positive $\Delta(S_e^{Lmax})$ in all the domains towards late afternoon hours, and the strongest SMPC response in the East.

Additional analysis reveals that the majority of large and negative soil moisture gradients in all domains are linked to the rainfall events that are identified during the first afternoon time step (i.e 12:00 UTC and 15:00 UTC for the East and West respectively) (not shown), and are therefore likely linked to weaker MCSs at the early stage of their development. The smaller and less organized MCSs have shown to be more sensitive to the thermally-induced surface convergence zones and are likely to develop over spatially drier soils, adjacent to the strong gradients (e.g, Gantner and Kalthoff, 2010). This knowledge is consistent with the strongest negative δ_e difference identified here and hence with the prominent negative SMPC relationship observed during early afternoon times in all three domains.

5 Results of temporal SMPC analysis

5.1 Co-variability of the spatial and temporal SMPC

To further explore the consistency of the identified negative spatial SMPC with the physical effects we analyze the temporal SMPC relationship. By analogy to the spatial SMPC, we compute the soil moisture anomaly S_e^{Lmax} prior to the event and its

5 difference δ_e to the typical state. Analysis of $S_e^{L_{max}}$ and its δ_e indicates a strong preference for rainfall events to occur over soils that are drier than their temporal mean (Fig. 9a) and drier than usual (Fig. 9b). The percentile values P_e lower than 10% are found in 67 % of the studied $1^\circ \times 1^\circ$ boxes (Table 1). The latter implies that a temporally negative SMPC dominates over the domain, which reaffirms the co-existence of the negative spatial and temporal coupling identified by G15, but at a finer 1° horizontal scale.

The question remains whether the two coupling relationships are independent of one another. To answer this question we calculate the Spearman rank correlation coefficient⁴ event-wise between the soil moisture anomaly $S_e^{L_{max}}$ and soil moisture gradients $\Delta(S_e^{L_{max}})$ in every $1^\circ \times 1^\circ$ box. The correlation map in Figure 9c shows that a high and significant correlation exists between $S_e^{L_{max}}$ and $\Delta(S_e^{L_{max}})$ anywhere in the domain. The mean correlation of 0.47 over the domain supports the existence of relatively strong and positive monotonic relationship between the magnitude of spatial soil moisture gradient and soil moisture anomaly measured in L_{max} . For comparison, the mean correlation estimated between soil moisture gradients and mean soil moisture anomaly over the 1.25° event box is small (0.13). All the above suggests that in the North African region the spatial and temporal SMPC relationships, as defined by the current framework, are not independent of each other.

15 The strong and positive correlation (*in time*) identified between the soil moisture anomalies and gradients also yields a regional co-variability of the SMPC patterns. The *spatial* correlation between the two coupling distributions is high (0.64). The largest magnitudes of both $S_e^{L_{max}}$ and $\Delta(S_e^{L_{max}})$ parameters and their corresponding δ_e measures are found in the southern part of the domain. These regions are generally characterized as the areas of higher BL moisture and rainfall frequency, and therefore higher variability of soil moisture in time and space.

Mechanistically, the presence of the temporally negative SMPC in the areas of the highest BL moisture in the domain (or lowest lifting condensation level (LCL), Fig. 10a), is consistent with a higher relevance of mechanisms associated with the BL growth for convection initialization in regions of higher CAPE and lower convective inhibition (CIN) (Klüpfel et al., 2011; Gantner and Kalthoff, 2010; Adler et al., 2011). In this way, larger negative deviations of the soil moisture amount from its climatological mean, i.e. δ_e , would lead to stronger than usual thermals, which can easier overcome CIN and release CAPE (Klüpfel et al., 2011).

25 Moreover, in combination with a strong negative spatial gradients, these strong thermals can initiate breeze-like circulations, creating more favourable conditions for bringing BL up to the LFC, especially over the southern regions, where BL moisture is in abundance. The relevance of drier surface conditions for moist convection development on event days over the wetter latitudes is supported by the observed significant increase of the LCL height (decrease in pressure) in the South on event days compared to the typical state (Fig. 10b, red shading). The higher LCL is associated with a decrease of BL relative and specific humidity (not shown) and supports the relevance of drier surface conditions for convection intensification as opposed to variations in BL water vapour amount prior to the events.

A different picture is observed over the drier latitudes of Northern Sahel at the Sahara margin. The estimated difference in LCL prior to the events relative to the typical state indicates that a significantly lower than usual LCL (higher pressure),

⁴Spearman correlation is a measure of monotonic relationship. Therefore, zero or low correlation value does not imply zero relationship between two variables.

associated to a significantly higher amount of water vapour in the BL (not shown), is present on the event days over the dry regions (Fig. 10b, blue shading). This result is consistent with the previously reported decisive role of low-level moisture on MCS evolution in the drier Sahelian regions (Klüpfel et al., 2012). At these latitudes the northward excursion of moist monsoon air has been shown to favour convective activity (Barthe et al., 2010; Cuesta et al., 2010).

5 Considering also the relatively large number of dry days (10 days on average) preceding rain events in the North, it is less likely that underlying surface heterogeneity caused by a previous rainfall could have an influence on convection development on the event day. In the case study of Klüpfel et al. (2012) MCS was initiated due to the arrival of the cold pool and convergence zone emanated by a remote convective system hundreds kilometers away. Similar mechanisms may play a role in moist convection development in Northern Sahel.

10 **6 Summary and discussion**

6.1 Role of thermally-induced circulations in moist convection development

The dominant negative spatial SMPC relationship observed over the North African region agrees on the sign of the SMPC suggested by previous case- and modelling studies (e.g. Taylor and Ellis (2006); Birch et al. (2013); Froidevaux et al. (2014)), and reconciles a number of physical effects. Following the sensitivity analyses, the two main factors are identified, which
15 directly influence the magnitude and significance of the negative spatial SMPC relationship. These are the time of the afternoon rainfall accumulation and the extreme soil moisture gradients. Our results show that the observed relationship between the spatial SMPC measure and the time of rainfall accumulation conforms well with the varying sensitivity of rainfall to the underlying soil moisture conditions for different stages of MCS development (Taylor and Lebel, 1998; Clark et al., 2003; Taylor and Ellis, 2006; Alonge et al., 2007; Taylor et al., 2010). This further suggests potentially higher relevance of drier
20 soils and soil moisture heterogeneity on rainfall when the rain systems are smaller and at early stage of their development, consistently to results of e.g. Gantner and Kalthoff (2010). All the above also emphasizes the importance of considering the MCS properties as an additional factor in the analysis of the SMPC (Ford et al., 2015).

Extreme soil moisture gradients are found to predefine the significance of the negative spatial coupling in 30 % of the domain grid boxes. Concurrently, the extremes tend to cluster in the direct vicinity of major flooded areas and the irrigated land. The
25 identified sensitivity of the afternoon rainfall to the strong soil moisture gradients adjacent to wetland areas shows consistency with the role of wetland–breeze mechanism in convection intensification over spatially drier soils. If this is true, our results demonstrate for the first time that the wetland–breeze mechanism could be a systematic feature in the North African region with further implications for the rainfall predictability.

This observed sensitivity of the SMPC measure to the flooded areas and to the MCS life cycle complies well with a potential
30 role of thermally-induced circulations in afternoon rainfall development over strong negative soil moisture gradients. Moreover, the identified preference of rainfall to occur simultaneously over temporally drier soils (negative temporal SMPC) and strong negative soil moisture gradients (negative spatial SMPC) might be considered as the most effective combination to initiate thermal circulations. This would imply a presence of a higher buoyancy and moisture flux at the event location, and hence a

higher probability of convection development. Schematic representation of the deep convection initiation by the breeze-like circulations formed under the co-existence of the two SMPC effects is illustrated in Figure 11.

6.2 Role of rainfall persistence

In the context of this study, the drying of the soil prior to the rainfall events might be considered as the primary process that underlies the magnitude of both SMPC relationships, and helps to explain the opposite sign of the temporal coupling identified in the North African region as compared to the temperate latitudes and wet climates (G15).

Consistently to the observed 2 to 4 day periodicity of rainfall in Western Africa (Laing et al., 2012; Taylor and Lebel, 1998), 2 to 3 dry days (rain < 1 mm) on average are found to precede each convective event day over southern latitudes, suggesting a strong drying of the upper soil layer in the event locations prior to the rain. The number of dry days reaches 10 over the dry and deserted regions in the North. Following the analysis of Schwendike et al. (2010) an almost complete recovery of the pre-rainfall surface moisture conditions may be expected in 2–3 days following the rainfall. Schematically, this typical variability of rainfall and soil moisture might be illustrated as a sequence of daily rain events separated by the periods of drying (Fig. 12a). From the Figure it is seen that prior to the rain events the soil dries out, and soil moisture reaches certain minimum value S_{min} . The climatology value S_{clim} of soil moisture in the same location, however, is expected to be higher than any S_{min} in most of the cases, as it includes all, dry and wet event days. Hence, when subtracted from the climatological value, a soil moisture measured prior to the event will very likely yield a negative anomaly - S_e^{Lmax} , especially when averaged over many events. Therefore, a negative correlation between soil moisture anomaly and rainfall might be expected. Though discussed in the framework of North Africa, similar behaviour might be expected in other water-limited regions of the world.

A different situation might occur in the wet temperate latitudes, where the variability of rainfall is to a large extent linked to fluctuations between passage of a cyclone and a blocking situation (Schär et al., 1999). Such a behavior might be illustrated as a multi-day sequence of rain events, associated with precipitation persistence as defined by the persistence in the weather regimes (Fig. 12b, see also Fig. 2 in Hohenegger et al. (2009)). During these periods soil moisture increases and remains relatively high. Hence, a higher fraction of events might be expected to occur over soils that are wetter than usual, resulting in a positive soil moisture anomaly S_e^{Lmax} prior to the event. The above relationship is consistent with the positive temporal SMPC, identified in G15. It is important to note, however, that the rainfall persistence may not be solely atmospherically driven, but may also reflect effects of the land surface (Salvucci et al., 2002; Guillod et al., 2014; Seneviratne et al., 2010).

The modulation of the SMPC sign depending on the large-scale weather regime was studied e.g. by Boé (2012) over France. The analysis showed that the synoptic blocking situations generally associated with drier conditions lead to a negative SMPC, while positive correlation of rainfall to drier soil conditions was observed in wet weather regime. Similarly, most pronounced effect of negative soil moisture gradients on convection initiation over Europe and a higher correlation of the gradients to land surface temperatures was observed for the period with less antecedent rainfall (Taylor, 2015).

7 Conclusions

In this study, we revisit the negative spatial and negative temporal SMPC relationships identified earlier by T12 and G15 at the $5^\circ \times 5^\circ$ horizontal grid. We use the probability-based approach of T12 and 10 years of satellite-based soil moisture and precipitation data (i) - to identify the potential link of the observed statistical relationships to the physical mechanisms and (ii) - to study the regional co-variability of the SMPC effects.

We find that the negative spatial coupling dominates over the region of North Africa. The result is independent of the choice of the observational data sets and is robust with respect to the event aggregation (spatial) scale. Compared to the previously
5 used coarser $5^\circ \times 5^\circ$ grid, the interest of the finest considered $1^\circ \times 1^\circ$ event-aggregation scale is that it reveals links to the wetland areas and rivers which can not be captured at the coarser scale.

The co-variability analysis of the two SMPC relationships indicates that spatial and temporal effects of soil moisture on afternoon precipitation in the North African region do not only co-exist but are also dependent of one another. The latter suggests that if rain falls over temporally drier soils, it is likely to be surrounded by a wetter environment. This combination is
10 consistent with the relevance of processes associated with the dominance of sensible heat flux and boundary layer growth on convection initiation, and supports the role of mesoscale variability in surface soil moisture for deep convection development.

The identified negative sign of the temporal coupling in the semi-arid conditions of the Sahelian environment is not unexpected. The drying of the soil for several days prior to the rainfall events is likely to underlay the preference of rain to occur over temporally drier soils. This additionally may play a role in the opposite sign of the temporal coupling in the North African
15 region as compared to the positive relationship identified in wetter climates by G15. For the same reason, the predictability potential of the temporal effect on rainfall in the North African region is expected to be lower compared to the spatial effect.

Analysis of the spatial SMPC measure and factors which can influence its magnitude and variability in particular reveals two "hot-spot" regions, where predictability skill of spatial soil moisture variability on rainfall might be higher. These are the Western African domain ($7-15^\circ$ N, 10° W- 7° E), and South Sudan in the East ($5-13^\circ$ N, $24-34^\circ$ E). In the Western domain, the
20 negative spatial SMPC signal is indicated to be more robust. In the East, the spatial coupling is found to be largely modulated by the presence of wetlands and is susceptible to the amount of longer-lived propagating MCS. The analysis of the BL moisture conditions (here, LCL) preceding the rainfall events further supports potential relevance of spatially and temporally drier soils for convection development in the South of the domain, where BL moisture is in abundance. In the drier northern latitudes the variability of BL moisture, associated to intrusions of moisture from the south, seems to be more decisive.

Following our analysis, a number of potential improvements to the framework might be summarized. Apparent non-local effects of water bodies, that are originally excluded by the method, hints on the potential gaps in the filtering procedure and emphasizes potential role of moist convection evolution and propagation that are neglected by the method. The presence of wetland regions itself, as we have shown, complicates interpretation of the SMPC relationships. The uncertainty estimates of
5 the soil moisture parameter derived over the recursively flooded regions are still missing. In the future, dynamical wetland products like the Global Inundation Extent from Multi-Satellites (GIEMS, Prigent et al. (2012, 2007)) may be used to better isolate the effect of water bodies on moist convection development.

Notwithstanding these limitations, the present study demonstrates the ability of probability-based methods to identify characteristic features of physical effects. Considering continuous increase in availability, time span and quality of satellite data, development of similar statistical methods should be valuable. The region of the strong negative SMPC identified in the East would highly benefit from more modelling and observational analysis in the eastern Sahel. The knowledge on the regional variability of the SMPC presented here can be further explored in drought and climate change research, observational campaigns and for validation of GCMs.

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

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20 References

- Adler, B., Kalthoff, N., and Gantner, L.: The impact of soil moisture inhomogeneities on the modification of a mesoscale convective system: An idealised model study, *Atmospheric Research*, 101, 354–372, 2011.
- Alonge, C. J., Mohr, K. I., and Tao, W.-K.: Numerical studies of wet versus dry soil regimes in the West African Sahel, *Journal of Hydrometeorology*, 8, 102–116, 2007.
- 25 Alter, R., Im, E., and Eltahir, E.: Rainfall consistently enhanced around the Gezira Scheme in East Africa due to irrigation, *Nature Geoscience*, 8, 763–767, 2015.
- Amante, C. and Eakins, B.: ETOPO1 1 arc-minute global relief model: procedures, data sources and analysis, <http://scholar.google.com/scholar?hl=en&btnG=Search&q=intitle:ETOPO1+1+ARC-minute+global+relief+model:+procedures,+data+sources+and+analysis#0>, 2009.
- 30 Barthe, C., Asencio, N., Lafore, J.-P., Chong, M., Campistron, B., and Cazenave, F.: Multi-scale analysis of the 25–27 July 2006 convective period over Niamey: Comparison between Doppler radar observations and simulations, *Quarterly Journal of the Royal Meteorological Society*, 136, 190–208, 2010.
- Birch, C. E., Parker, D. J., O’Leary, A., Marsham, J. H., Taylor, C. M., Harris, P. P., and Lister, G. M. S.: Impact of soil moisture and convectively generated waves on the initiation of a West African mesoscale convective system, *Quarterly Journal of the Royal Meteorological*
- 35 *Society*, 139, 1712–1730, 2013.
- Boé, J.: Modulation of soil moisture–precipitation interactions over France by large scale circulation, *Climate Dynamics*, 40, 875–892, 2012.
- Clark, D. B., Taylor, C. M., Thorpe, a. J., Harding, R. J., and Nicholls, M. E.: The influence of spatial variability of boundary-layer moisture on tropical continental squall lines, *Quarterly Journal of the Royal Meteorological Society*, 129, 1101–1121, 2003.
- Couvreux, F., Rio, C., Guichard, F., Lathon, M., Canut, G., Bouniol, D., and A., G.: Initiation of daytime local convection in a semi-arid
- 5 region analyzed with Large-Eddy Simulations and AMMA observations, *Quarterly Journal of the Royal Meteorological Society*, 138, 56–71, 2012.
- Couvreux, F., Guichard, F., Gounou, a., Bouniol, D., Peyrillé, P., and Köhler, M.: Modelling of the thermodynamical diurnal cycle in the lower atmosphere: A Joint evaluation of four contrasted regimes in the tropics over land, *Boundary-Layer Meteorology*, 150, 185–214, <https://doi.org/10.1007/s10546-013-9862-6>, 2013.
- 10 Cuesta, J., Lavaysse, C., Flamant, C., Mimouni, M., and Knippertz, P.: Northward bursts of the West African monsoon leading to rainfall over the Hoggar Massif, Algeria, *Quarterly Journal of the Royal Meteorological Society*, 136, 174–189, 2010.
- Dee, D. P., Uppala, S. M., Simmons, a. J., Berrisford, P., Poli, P., Kobayashi, S., Andrae, U., Balmaseda, M. a., Balsamo, G., Bauer, P., Bechtold, P., Beljaars, a. C. M., van de Berg, L., Bidlot, J., Bormann, N., Delsol, C., Dragani, R., Fuentes, M., Geer, a. J., Haimberger, L., Healy, S. B., Hersbach, H., Hólm, E. V., Isaksen, L., Kållberg, P., Köhler, M., Matricardi, M., McNally, a. P., Monge-Sanz, B. M., Morcrette, J.-J., Park, B.-K., Peubey, C., de Rosnay, P., Tavolato, C., Thépaut, J.-N., and Vitart, F.: The ERA-Interim reanalysis: configuration and performance of the data assimilation system, *Quarterly Journal of the Royal Meteorological Society*, 137, 553–597, 2011.
- 15 Dirmeyer, P.: The terrestrial segment of soil moisture–climate coupling, *Geophysical Research Letters*, 38, 2011.
- Dirmeyer, P. and Wang, Z.: Intensified land surface control on boundary layer growth in a changing climate, *Geophysical Research Letters*, 45, 1290–1294, 2014.

- 20 Dirmeyer, P., Cash, B. a., Kinter, J. L., Stan, C., Jung, T., Marx, L., Towers, P., Wedi, N., Adams, J. M., Altshuler, E. L., Huang, B., Jin, E. K., and Manganello, J.: Evidence for enhanced land–atmosphere feedback in a warming climate, *Journal of Hydrometeorology*, 13, 981–995, 2012.
- Dorigo, W. a., Scipal, K., Parinussa, R. M., Liu, Y. Y., Wagner, W., de Jeu, R. a. M., and Naeimi, V.: Error characterisation of global active and passive microwave soil moisture datasets, *Hydrology and Earth System Sciences*, 14, 2605–2616, 2010.
- 25 Duvel, J.: Convection over tropical Africa and the Atlantic Ocean during northern summer. Part II: Modulation by easterly waves, *Monthly Weather Review*, 118, 1885–1868, 1990.
- Ek, M. and Mahrt, L.: Daytime evolution of relative humidity at the boundary layer top, *American Meteorological Society*, 122, 2709–2721, 1994.
- Eltahir, E. a. B.: A soil moisture-rainfall feedback mechanism: 1. Theory and observations, *Water Resources Research*, 34, 765–776, 1998.
- 30 Findell, K. and Eltahir, E.: Atmospheric controls on soil moisture-boundary layer interactions. Part II: Feedbacks within the continental United States, *Journal of Hydrometeorology*, 4, 570–583, 2003a.
- Findell, K. and Eltahir, E.: Atmospheric controls on soil moisture-boundary layer interactions. Part I: Framework development, *Journal of Hydrometeorology*, 4, 552–569, 2003b.
- Ford, T. W., Rapp, A. D., Quiring, S. M., and Blake, J.: Soil moisture–precipitation coupling: observations from the Oklahoma Mesonet and underlying physical mechanisms, *Hydrology and Earth System Sciences*, 19, 3617–3631, 2015.
- 35 Froidevaux, P., Schlemmer, L., Schmidli, J., Langhans, W., and Schär, C.: Influence of the background wind on the local soil moisture-precipitation feedback, *Journal of the Atmospheric Sciences*, 17, 782–799, 2014.
- Gallego-Elvira, B. and Taylor, C.: Global observational diagnosis of soil moisture control on the land surface energy balance, *Geophysical Research Letters*, pp. 2623–2631, 2016.
- Gantner, L. and Kalthoff, N.: Sensitivity of a modelled life cycle of a mesoscale convective system to soil conditions over West Africa, *Quarterly Journal of the Royal Meteorological Society*, 136, 471–482, 2010.
- 5 Garcia-Carreras, L., Parker, D. J., and Marsham, J. H.: What is the mechanism for the modification of convective cloud distributions by land surface–induced flows?, *Journal of the Atmospheric Sciences*, 68, 619–634, 2011.
- Goutorbe, J., Lebel, T., and Tinga, A.: HAPEX-Sahel: a large-scale study of land-atmosphere interactions in the semi-arid tropics, *Annales Geophysicae*, 12, 53–63, 1994.
- Gruhler, C. and Rosnay, P. D.: Evaluation of AMSRE soil moisture product based on ground measurements over temperate and semiarid regions, *Geophysical Research Letters*, 35, 2–7, 2008.
- 10 Guichard, F., Kergoat, L., Mougin, E., and Timouk, F.: Surface thermodynamics and radiative budget in the Sahelian Gourma: Seasonal and diurnal cycles, *Journal of hydrology*, 375, 161–177, 2009.
- Guillod, B., Orlowsky, B., and Miralles, D.: Reconciling spatial and temporal soil moisture effects on afternoon rainfall, *Nature communications*, 6, 1–6, 2015.
- 15 Guillod, B. P., Orlowsky, B., Miralles, D., Teuling, a. J., Blanken, P. D., Buchmann, N., Ciais, P., Ek, M., Findell, K. L., Gentine, P., Lintner, B. R., Scott, R. L., Van den Hurk, B., and I. Seneviratne, S.: Land-surface controls on afternoon precipitation diagnosed from observational data: uncertainties and confounding factors, *Atmospheric Chemistry and Physics*, 14, 8343–8367, 2014.
- Hohenegger, C., Brockhaus, P., Bretherton, C. S., and Schär, C.: The soil moisture–precipitation feedback in simulations with explicit and parameterized convection, *Journal of Climate*, 22, 5003–5020, 2009.

- 20 Hsu, K.-I., Gao, X., Sorooshian, S., and Gupta, H. V.: Precipitation estimation from remotely sensed information using artificial neural networks, *Journal of Applied Meteorology*, 36, 1176–1190, 1997.
- Huffman, G. J. and Bolvin, D. T.: TRMM and other data precipitation data set documentation, Tech. Rep. May, Mesoscale Atmospheric Processes Laboratory, NASA Goddard Space Flight Center, ftp://precip.gsfc.nasa.gov/pub/trmmdocs/3B42_3B43_doc.pdf, 2014.
- Huffman, G. J., Bolvin, D. T., Nelkin, E. J., Wolff, D. B., Adler, R. F., Gu, G., Hong, Y., Bowman, K. P., and Stocker, E. F.: The TRMM
25 multisatellite precipitation analysis (TMPA): quasi-global, multiyear, combined-sensor precipitation estimates at fine scales, *Journal of Hydrometeorology*, 8, 38–55, 2007.
- Janicot, S. and Thorncroft, C.: Large-scale overview of the summer monsoon over West Africa during the AMMA field experiment in 2006, *Annales Geophysicae*, 26, 2569–2595, 2008.
- Klüpfel, V., Kalthoff, N., Gantner, L., and Kottmeier, C.: Evaluation of soil moisture ensemble runs to estimate precipitation variability in
30 convection-permitting model simulations for West Africa, *Atmospheric Research*, 101, 178–193, 2011.
- Klüpfel, V., Kalthoff, N., Gantner, L., and Taylor, C. M.: Convergence zones and their impact on the initiation of a mesoscale convective system in West Africa, *Quarterly Journal of the Royal Meteorological Society*, 138, 950–963, 2012.
- Kohler, M., Kalthoff, N., and Kottmeier, C.: The impact of soil moisture modifications on CBL characteristics in West Africa: A case-study from the AMMA campaign, *Quarterly Journal of the Royal Meteorological Society*, 136, 442–455, 2010.
- 35 Koster, R. D., Dirmeyer, P. a., Guo, Z., Bonan, G., Chan, E., Cox, P., Gordon, C. T., Kanae, S., Kowalczyk, E., Lawrence, D., Liu, P., Lu, C.-H., Malyshev, S., McAvaney, B., Mitchell, K., Mocko, D., Oki, T., Oleson, K., Pitman, A., Sud, Y. C., Taylor, C. M., Verseghy, D., Vasic, R., Xue, Y., and Yamada, T.: Regions of strong coupling between soil moisture and precipitation., *Science (New York, N.Y.)*, 305, 1138–40, 2004.
- Laing, A. and Carbone, R.: The propagation and diurnal cycles of deep convection in northern tropical Africa, *Quarterly Journal of the Royal Meteorological Society*, 109, 93–109, 2008.
- Laing, A. G., Trier, S. B., and Davis, C. a.: Numerical simulation of episodes of organized convection in tropical Northern Africa, *Monthly Weather Review*, 140, 2874–2886, 2012.
- 5 Lohou, F., Kergoat, L., Guichard, F., Boone, A., Cappelaere, B., Cohard, J.-M., Demarty, J., Galle, S., Grippa, M., Peugeot, C., Ramier, D., Taylor, C. M., and Timouk, F.: Surface response to rain events throughout the West African monsoon, *Atmospheric Chemistry and Physics*, 14, 3883–3898, 2014.
- Lothon, M., Campistron, B., Chong, M., Couvreux, F., Guichard, F., Rio, C., and Williams, E.: Life cycle of a mesoscale circular gust front observed by a C-band Doppler radar in West Africa, *Monthly Weather Review*, 139, 1370–1388, 2011.
- 10 Mathon, V. and Laurent, H.: Life cycle of Sahelian mesoscale convective cloud systems, *Quarterly Journal of the Royal Meteorological Society*, 127, 377–406, 2001.
- Mathon, V., Laurent, H., and Lebel, T.: Mesoscale convective system rainfall in the Sahel, *Journal of applied meteorology*, 41, 1081–1092, 2002.
- Matthews, E. and Fung, I.: Methane emission from natural wetlands: Global distribution, area, and environmental characteristics of sources,
15 *Global Biogeochemical Cycles*, 1, 61–86, 1987.
- Miralles, D. G., van den Berg, M. J., Teuling, a. J., and de Jeu, R. a. M.: Soil moisture-temperature coupling: A multiscale observational analysis, *Geophysical Research Letters*, 39, 2012.
- Mohr, K. and Thorncroft, C.: Intense convective systems in West Africa and their relationship to the African easterly jet, *Quarterly Journal of the Royal Meteorological Society*, 132, 163–176, 2006.

- 20 Nicholson, S. E.: The West African Sahel: A review of recent studies on the rainfall regime and its interannual variability, *ISRN Meteorology*, 2013, 1–32, 2013.
- Nicholson, S. E.: Evolution and current state of our understanding of the role played in the climate system by land surface processes in semi-arid regions, *Global and Planetary Change*, 133, 201–222, 2015.
- Njoku, E. and Jackson, T.: Soil moisture retrieval from AMSR-E, *IEEE Transactions on Geoscience and Remote Sensing*, 41, 215–229, 25 2003.
- Owe, M., de Jeu, R., and Holmes, T.: Multisensor historical climatology of satellite-derived global land surface moisture, *Journal of Geophysical Research*, 113, F01 002, 2008.
- Petrova, I.: Understanding soil moisture - precipitation coupling on mesoscales using observations over North Africa, *Reports on Earth System Science / Max Planck Institute for Meteorology*, 191, 2017.
- 30 Prigent, C., Papa, F., Aires, F., Rossow, W. B., and Matthews, E.: Global inundation dynamics inferred from multiple satellite observations, 1993–2000, *Journal of Geophysical Research: Atmospheres*, 112, 2007.
- Prigent, C., Papa, F., Aires, F., Jimenez, C., Rossow, W. B., and Matthews, E.: Changes in land surface water dynamics since the 1990s and relation to population pressure, *Geophysical Research Letters*, 39, 2012.
- Redelsperger, J.-L., Thorncroft, C. D., Diedhiou, A., Lebel, T., Parker, D. J., and Polcher, J.: African Monsoon Multidisciplinary Analysis: 35 An international research project and field campaign, *Bulletin of the American Meteorological Society*, 87, 1739–1746, 2006.
- Rochetin, N., Couvreur, F., and Guichard, F.: Morphology of breeze circulations induced by surface flux heterogeneities and their impact on convection initiation, *Quarterly Journal of the Royal Meteorological Society*, 143, 463–478, 2017.
- Salvucci, G. D., Saleem, J. a., and Kaufmann, R.: Investigating soil moisture feedbacks on precipitation with tests of Granger causality, *Advances in Water Resources*, 25, 1305–1312, 2002.
- Schär, C., Lüthi, D., and Beyerle, U.: The soil-precipitation feedback : A process study with a regional climate model, *Journal of Climate*, 12, 722–741, 1999.
- Schwendike, J., Kalthoff, N., and Kohler, M.: The impact of mesoscale convective systems on the surface and boundary-layer structure in 5 West Africa: Case-studies from the AMMA campaign 2006, *Quarterly Journal of the Royal Meteorological Society*, 136, 566–582, 2010.
- Seneviratne, S. I., Corti, T., Davin, E. L., Hirschi, M., Jaeger, E. B., Lehner, I., Orlowsky, B., and Teuling, A. J.: Investigating soil moisture-climate interactions in a changing climate: A review, *Earth-Science Reviews*, 99, 125–161, 2010.
- Taylor, C. and Lebel, T.: Observational evidence of persistent convective-scale rainfall patterns, *Monthly Weather Review*, 126, 1597–1607, 1998.
- 10 Taylor, C. M.: Feedbacks on convection from an African wetland, *Geophysical Research Letters*, 37, 2010.
- Taylor, C. M.: Detecting soil moisture impacts on convective initiation in Europe, *Geophysical Research Letters*, 42, 4631–4638, 2015.
- Taylor, C. M. and Ellis, R. J.: Satellite detection of soil moisture impacts on convection at the mesoscale, *Geophysical Research Letters*, 33, L03 404, 2006.
- Taylor, C. M., Parker, D. J., and Harris, P. P.: An observational case study of mesoscale atmospheric circulations induced by soil moisture, *Geophysical Research Letters*, 34, L15 801, 2007.
- Taylor, C. M., Harris, P. P., and Parker, D. J.: Impact of soil moisture on the development of a Sahelian mesoscale convective system: a 5 case-study from the AMMA Special Observing Period, *Quarterly Journal of the Royal Meteorological Society*, 136, 456–470, 2010.
- Taylor, C. M., Gounou, A., Guichard, F., Harris, P. P., Ellis, R. J., Couvreur, F., and De Kauwe, M.: Frequency of Sahelian storm initiation enhanced over mesoscale soil-moisture patterns, *Nature Geoscience*, 4, 430–433, 2011.

- Taylor, C. M., de Jeu, R. a. M., Guichard, F., Harris, P. P., and Dorigo, W. a.: Afternoon rain more likely over drier soils., *Nature*, 489, 423–426, 2012.
- 10 Taylor, C. M., Birch, C. E., Parker, D. J., Dixon, N., Guichard, F., Nikulin, G., and Lister, G. M. S.: Modeling soil moisture-precipitation feedback in the Sahel : Importance of spatial scale versus convective parameterization, *Geophysical Research Letters*, 40, 6213–6218, 2013.
- Timouk, F., Kergoat, L., Mougin, E., Lloyd, C., Ceschia, E., Cohard, J.-M., de Rosnay, P., Hiernaux, P., Demarez, V., and Taylor, C.: Response of sensible heat flux to water regime and vegetation development in a central Sahelian landscape., *Journal of Hydrology*, 375, 178–189, 15 2009.
- van Heerwaarden, C. C. and Guerau de Arellano, J. V.: Relative humidity as an indicator for cloud formation over heterogeneous land surfaces, *Journal of the Atmospheric Sciences*, 65, 3263–3277, 2008.
- Wolters, D.: Effects of soil moisture gradients on the path and the intensity of a West African squall line, *Quarterly Journal of the Royal Meteorological Society*, 136, 2162–2175, 2010.

Table 1. General statistics of the average event number, percentile P_e and δ_e difference estimated at various scales for three soil moisture parameters: soil moisture gradient $\Delta(S_e'^{Lmax})$, temporal soil moisture anomaly $S_e'^{Lmax}$, and (not presented in the methodology) soil moisture variance over the 1.25° box, $\sigma S_e^{1.25}$. Percentiles $P_e < 10\%$ ($> 90\%$) indicate significant negative (positive) δ_e difference, and hence negative (positive) SMPC relationship.

<i>Parameter</i>	<i>Scale</i>	$\overline{Num_e}$	$P < 10, [\%]$	$P > 90, [\%]$	$\delta_{ev} < 0, [\%]$	$\delta_{ev} > 0, [\%]$
$\Delta(S_e'^{Lmax}) :$	$5 \times 5^\circ$	309	72	0	92	2.8
	$2.5 \times 2.5^\circ$	84	42	1.4	73	5.6
	$1 \times 1^\circ$	17	21	1.3	43	14
$S_e'^{Lmax} :$	$1 \times 1^\circ$	-	67	0.8	92	8
$\sigma S_e^{1.25} :$	$1 \times 1^\circ$	-	33	3	78	22

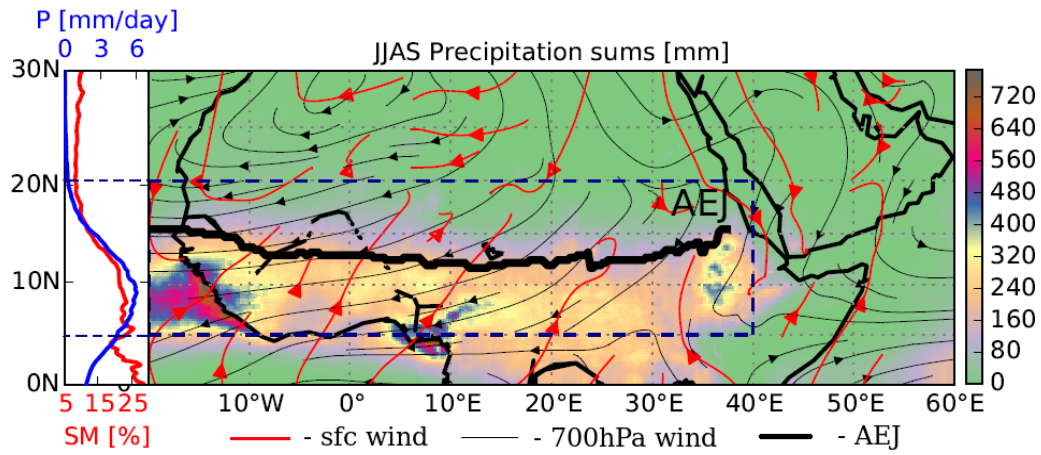


Figure 1. JJAS TMPA precipitation (shading), mean surface (red streamline) and 700 hPa (black streamline) ERA-Interim (1979 - present, Dee et al. (2011)) wind climatology averaged over 2002 - 2011 period. The black thick line shows mean location of the African Easterly Jet (AEJ). Inset plot to the left shows zonal means of daily AMSR-E soil moisture (red) and TMPA precipitation (blue) climatology calculated over a study domain (dashed rectangle, 5° - 20°N; 20°W - 40°E).

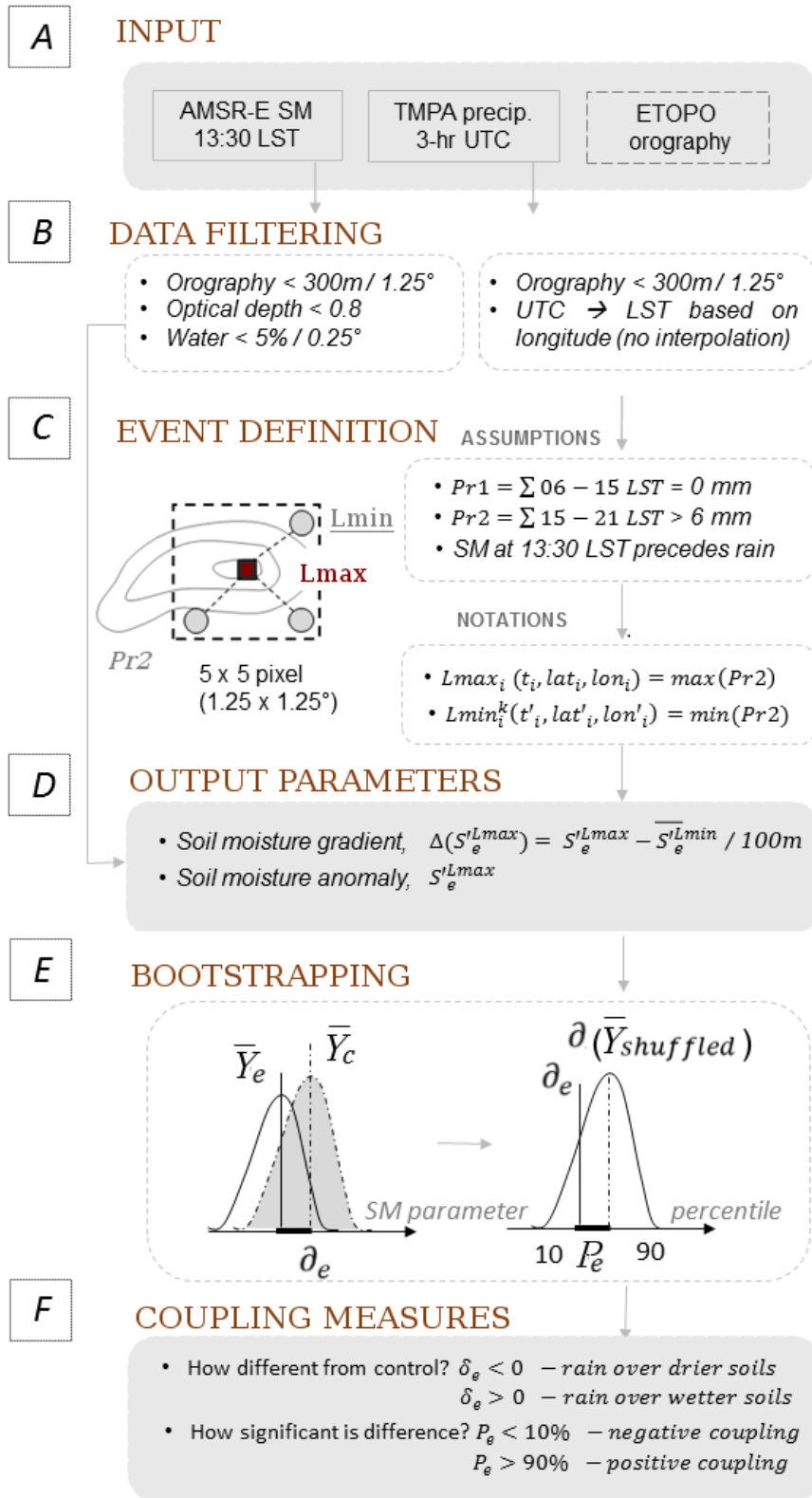


Figure 2. Schematic of data post-processing and statistical framework protocol implemented in the study.

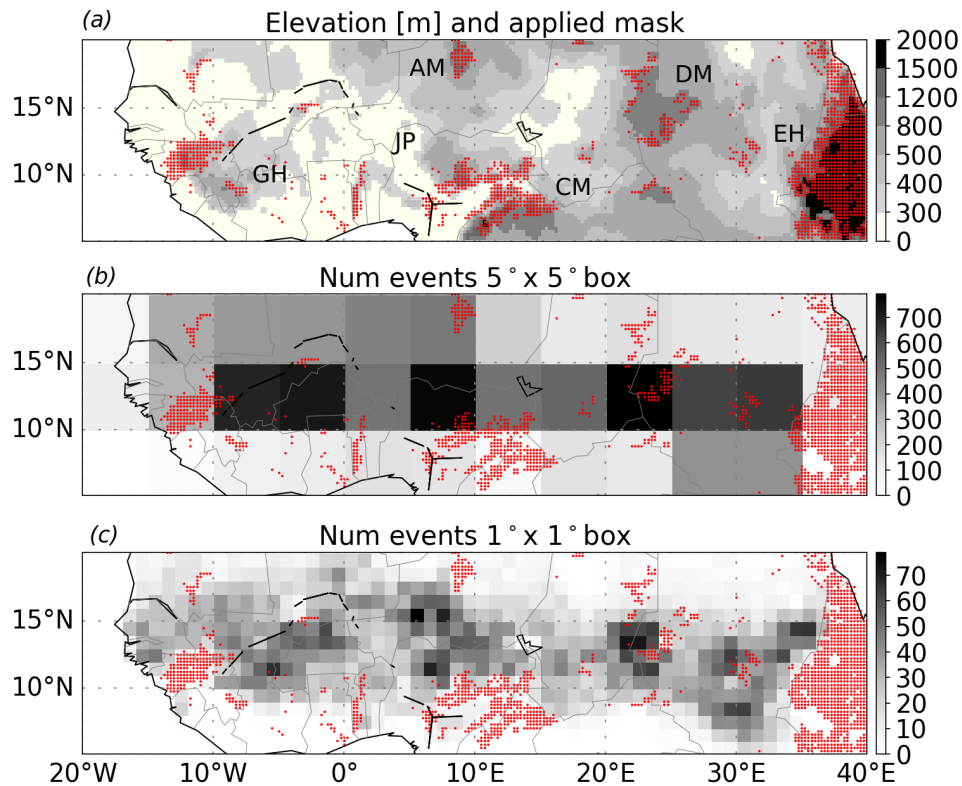


Figure 3. (a) - Elevation map based on 1 Arc-Minute Global Relief Model data ETOPO1 (Amante and Eakins, 2009) (grey shading) and orography mask used in the study (red dots). Main orographic features of the region are: AM - Air Mountains, DM - Darfur Mountains, EH - Ethiopian Highlands, CM - Cameroon Mountains, JP - Jos Plateau, GH - Guinea Highlands. (b-c) - Number of events in every (b) $5^\circ \times 5^\circ$ and (c) $1^\circ \times 1^\circ$ box (gray shading) and applied orography mask at the 0.25° horizontal resolution (red dots).

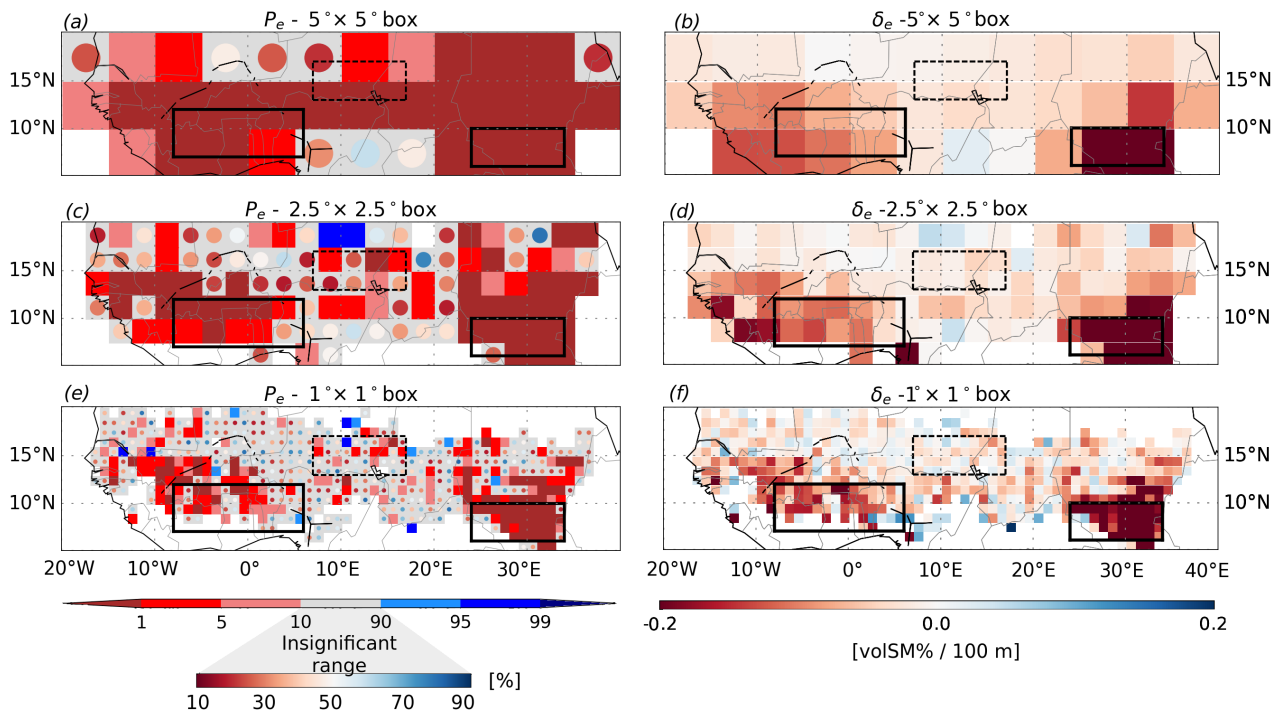


Figure 4. Distribution of percentiles P_e (left) of the observed δ_e difference (right), estimated over (a-b) $5^\circ \times 5^\circ$, (c-d) $2.5^\circ \times 2.5^\circ$ and (e-f) $1^\circ \times 1^\circ$ boxes. Percentiles $<10\%$ indicate significant negative coupling, i.e. rain over spatially drier soils, and percentiles $>90\%$ - significant positive coupling, i.e. rain over spatially wetter soils. The percentile values lying outside the significance range (i.e. in-between 10 % and 90 % percentiles) are illustrated by circles. Black rectangles on the maps indicate featured domains selected for an in-depth analysis.

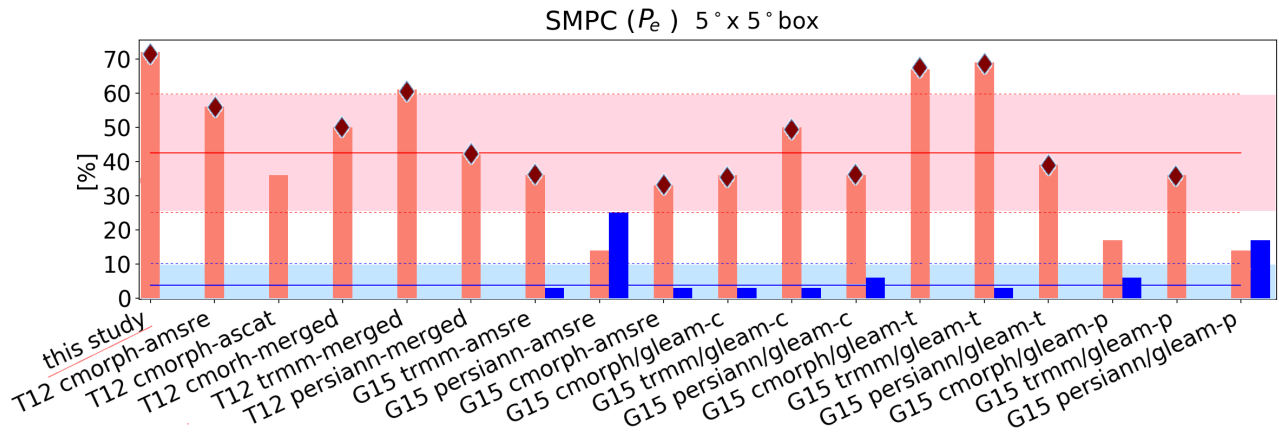
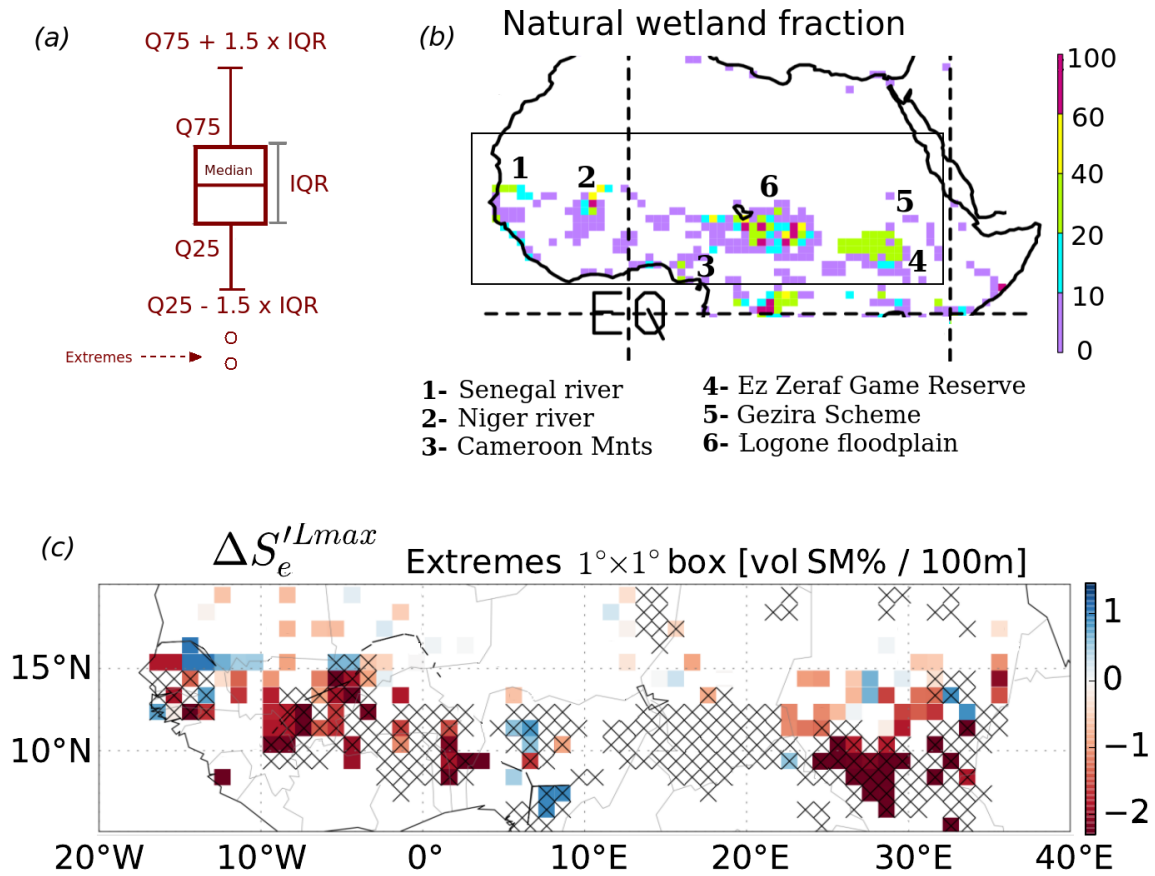


Figure 5. Percentage of $5^\circ \times 5^\circ$ grid boxes with significantly negative ($P_e < 10\%$, in red) and positive ($P_e > 90\%$, in blue) spatial SMPC over the North African domain in this study and previous studies of T12 and G15. Different data set combinations used in T12 and G15 are listed. Mean and st.dev. of the negative (positive) SMPC fractions across the experiments are shown as a red (blue) solid line and shading accordingly. Following visual inspection, the experiments, in which significant negative SMPC relationship exists in the western region of the North African domain are marked with a *rhom*b.



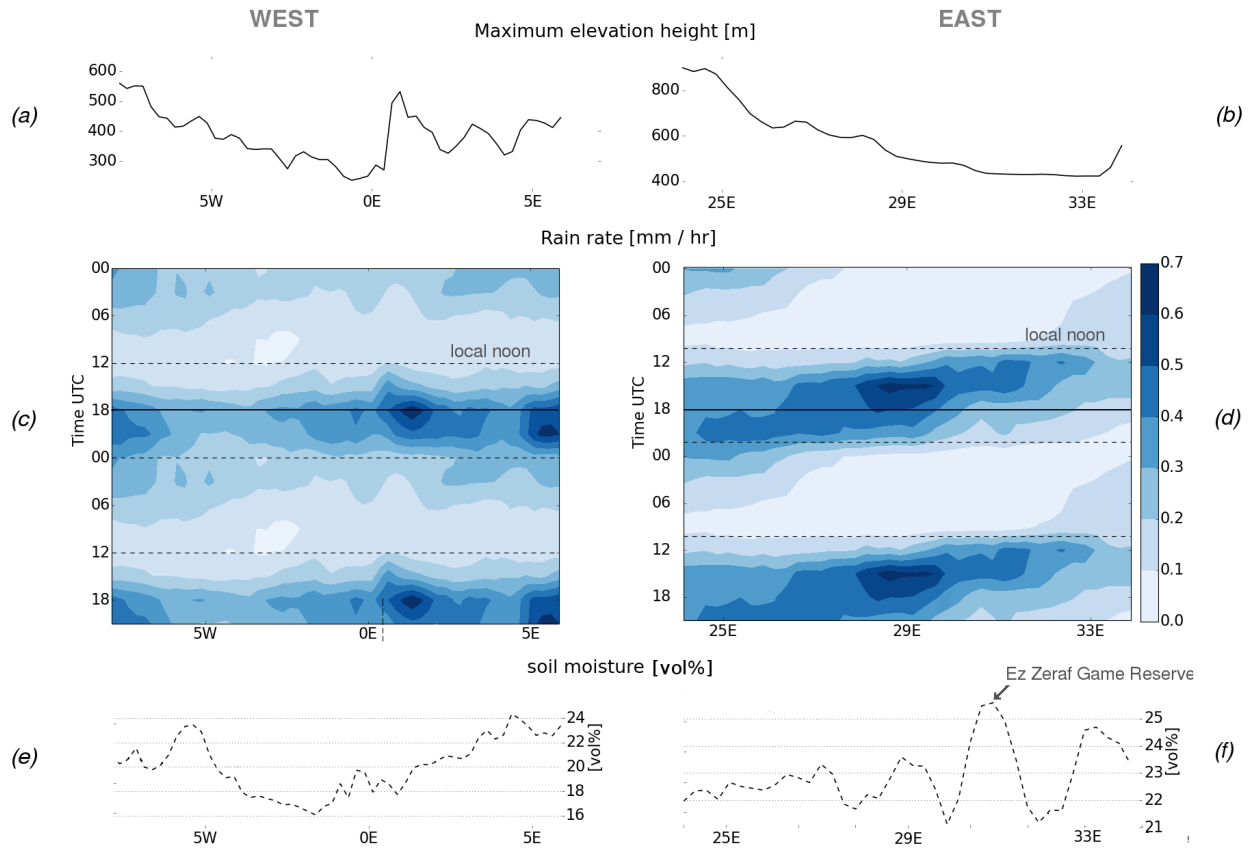


Figure 7. (a),(b) - Longitudinal cross-sections of maximum elevation height in the Western and Eastern domains respectively, (c),(d) - diurnal cycles of the rain rate averaged over event days and domain latitudes, and (e),(f) - Longitudinal cross section of soil moisture averaged over domain latitudes. Location of the Ez Zeraf Game Reserve permanent wetlands is marked by an arrow. All the times are given in UTC. Note, the UTC+2 hour difference to LST in the East.

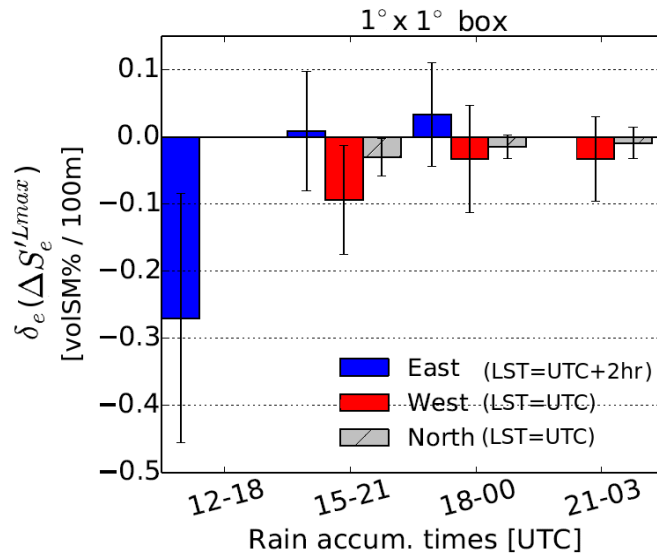


Figure 8. Value of the spatial coupling measure δ_e calculated for various afternoon rainfall accumulation times, and averaged over selected domains, i.e East ($6 - 10^\circ$ N, $24 - 34^\circ$ E), West ($7 - 12^\circ$ N, 8° W - 6° E) and North ($14 - 17^\circ$ N, $7 - 14^\circ$ E). Locations of the domains are shown in Fig. 4. Error bars indicate one std.dev. of δ_e values in every domain. Note, that all times are indicated in UTC.

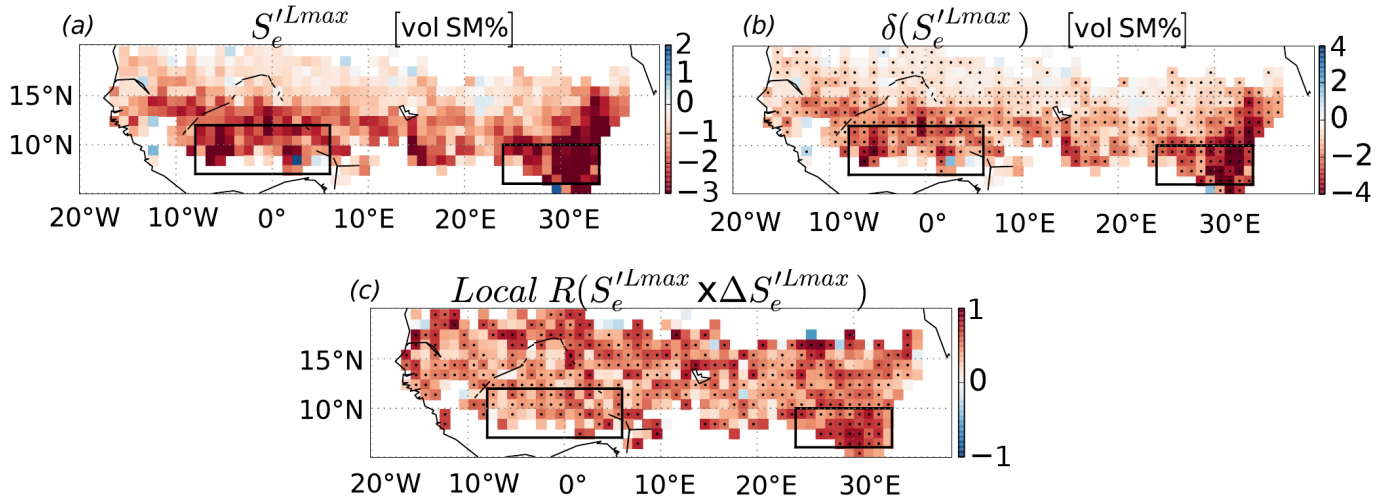


Figure 9. Distribution of the (a) pre-event soil moisture anomaly value $S_e'^{Lmax}$ in event locations and (b) its departure from the typical non-event conditions, δ_e , averaged over $1^\circ \times 1^\circ$ boxes; and (c) *Local* Spearman rank correlation coefficient calculated event-wise between soil moisture anomaly $S_e'^{Lmax}$ and spatial soil moisture gradients $\Delta S_e'^{Lmax}$ in every 1° box. Significant δ_e values with percentiles P_e below 10 % (above 90 %) and correlation coefficients with p -values lower than 0.05 are indicated by black dots.

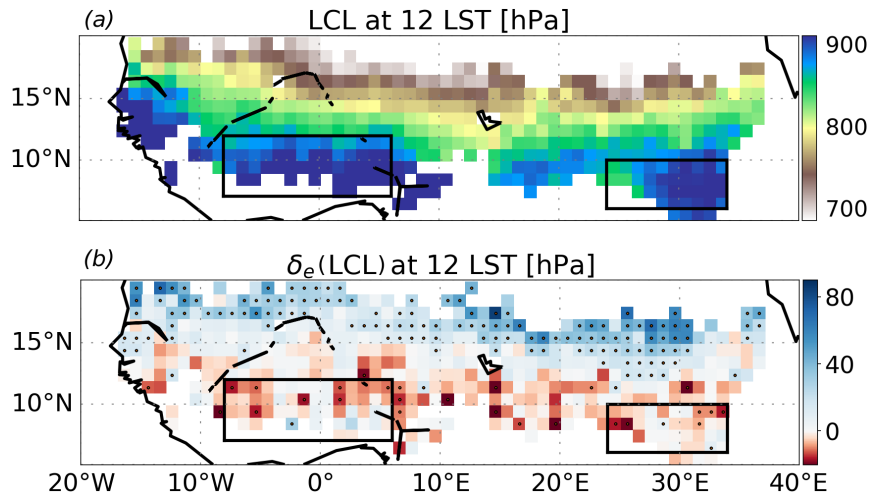


Figure 10. (a) - Lifting condensation level (LCL) value derived from the 6-hourly ERA-Interim temperature, specific humidity profile and surface pressure data on event days at 12:00 LST and averaged over the $1^\circ \times 1^\circ$ box. (b) - Corresponding δ_e difference of the mean LCL prior to the events relative to their typical state (climatology). The dot indicates significant δ_e values with percentiles P_e below 10 % (above 90 %). The positive (negative) δ_e values indicate lower (higher) than usual LCL (relative to the ground level).

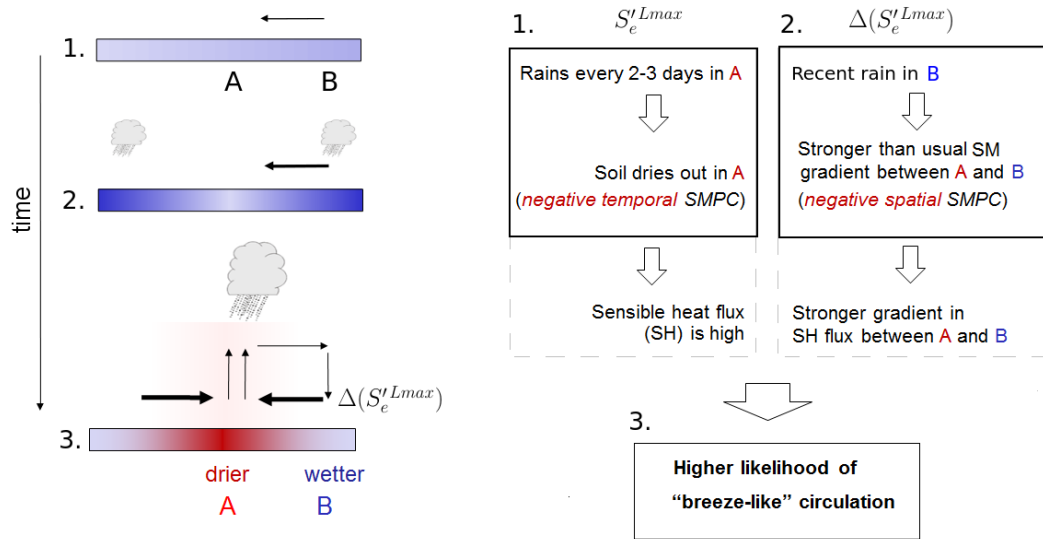


Figure 11. Conceptual diagram, illustrating intensification of moist convection by the initiated "breeze-like" circulations under favourable conditions of co-existing negative spatial and negative temporal SMPC effects. On the one hand, typically observed 2 to 4 day periodicity of rainfall in Western Africa leads to a strong drying of the upper soil layer in the location A prior to the rain, and therefore increases sensible heat and buoyancy flux locally. Simultaneously, recent rainfall in B produces wet soils - a potential moisture supply area for the location A. Strong spatial gradients in soil moisture between locations A and B together with a relatively strong buoyancy flux in A can favour formation of thermally-induced circulations under benign wind conditions. Considering a relatively weak mean surface wind of 2-3 m/s observed prior to the rainfall events over Southern latitudes, the mesoscale circulations are likely to be initiated.

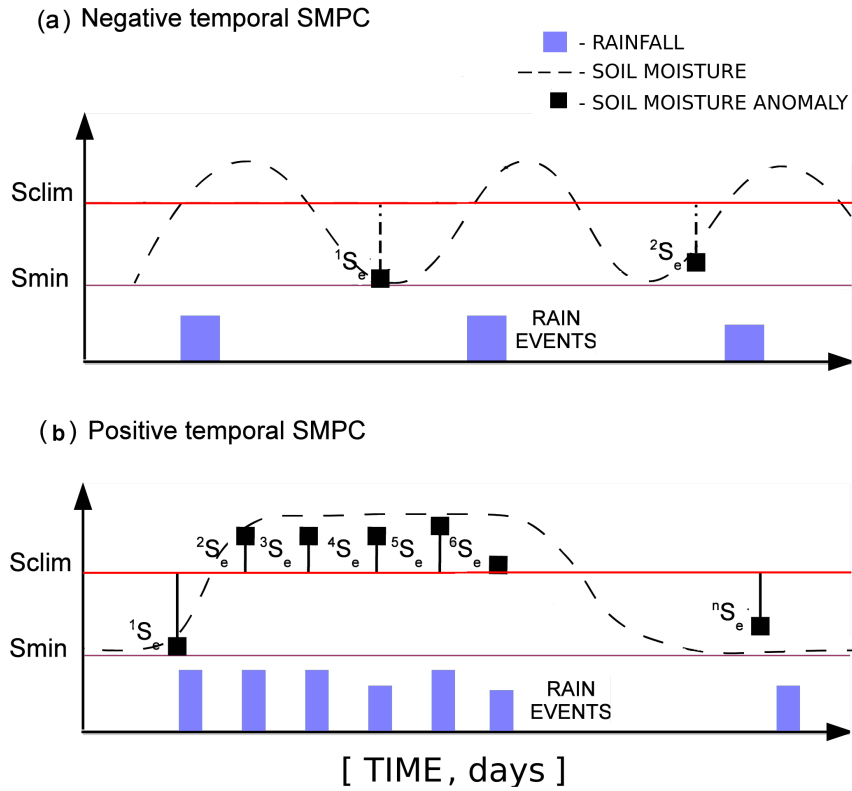


Figure 12. Conceptual diagrams of the relationship between daily rainfall occurrence and associated to it surface moisture variability in time as representative for (a) West Africa and temporally negative SMPC, and (b) Central and Northern Europe and positive temporal SMPC.

20 **Appendix A: : Additional material to the Section 4.3**

Method used for identification of potentially flooded locations

685 The following paragraph describes methodology used to identify events, which L_{min} locations are likely flooded. Following this methodology, a climatology of soil moisture drying rates is computed in every L_{min} location first. Climatology is calculated from values measured in the same L_{min} location, in the same month as the event but during the non-event years. Soil moisture drying rate is computed as the difference between soil moisture at 13:30 LST on the event day 0 and the previous day -1. Days with non-zero precipitation between two soil moisture measurements are excluded.

690 As a next step, we compute a climatology of soil moisture values on day -1 to estimate a potential range of soil moisture conditions in every L_{min} location. Finally, for every L_{min} location we identify a linear regression function which fits best into a scatter plot relationship between drying rates and initial soil moisture values. Based on the slope of the linear regression and a climatological range of drying rates we stratify L_{min} locations as being potentially "always dry", "normal" or "always wet" (Fig. A1). Locations, which are potentially "always dry" may be representative for rocky sand areas, and hence will most often
695 show low soil moisture values and drying rates close to zero. The "always wet" cases should indicate mostly high soil moisture values, but at the same time small drying rates under condition that water supply is present. The "normal" case is expected to show a clear relationship between the drying rate and initial soil moisture content. The higher the soil moisture is the larger the drying rate is expected to be.

Because the identification procedure requires selection of thresholds, the distribution of the potentially flooded locations will
700 slightly vary depending on the selected threshold of the regression slope, minimum drying rate value or minimum absolute soil moisture. For the calculation of the potential wetland locations presented in Fig. 6c we chose two thresholds: the slope had to be larger than -0.15 and the soil moisture values - larger than 20 %.

Table A1. Percentage of $5^\circ \times 5^\circ$ grid boxes with significantly negative ($P_e < 10\%$) and positive ($P_e > 90\%$) spatial SMPC over the North African domain in this study and previous studies of T12 and G15. Different data set combinations used in T12 and G15 are listed. *Merged* represents an integrated product composed of grid boxes in which either AMSR-E and ASCAT soil moisture data set was identified to be "best" following a quality-control check. Following visual inspection, the experiments, in which significant negative SMPC relationship exists in the western region of the Sahelian domain are indicated with a * - sign. For more details on various data set combinations the reader is referred to the original papers.

#	<i>Experiment</i>	<i>Frac. $P < 10\%$, [%]</i>	<i>Frac. $P > 90\%$, [%]</i>
1*	This study (trmm-amsre)	72	0
2*	T12 cmorph-amsre	56	0
3	T12 cmorph-ascats	36	0
4*	T12 cmorph-merged	50	0
5*	T12 trmm-merged	61	0
6*	T12 persiann-merged	42	0
7*	G15 trmm-amsre	36	3
8	G15 persiann-amsre	14	25
9*	G15 cmorph-amsre	33	3
10*	G15 cmorph-gleam(c)	36	3
11*	G15 trmm-gleam(c)	50	3
12*	G15 persiann-gleam(c)	36	6
13*	G15 cmorph-gleam(t)	67	0
14*	G15 trmm - gleam(t)	69	3
15*	G15 persiann-gleam(t)	39	0
16	G15 cmorph-gleam(p)	17	6
17*	G15 trmm-gleam(p)	36	0
18	G15 persiann-gleam(p)	14	17
	<i>mean</i>	42	4
	<i>std.dev</i>	10	1

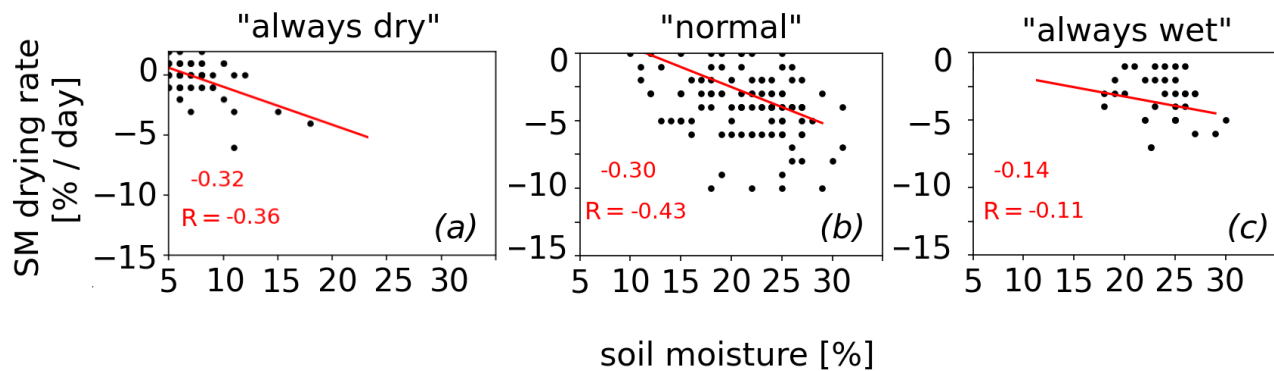


Figure A1. Examples of three possible relationships between climatology of soil moisture drying rate and absolute soil moisture in L_{min} locations. The red numbers indicate values of a slope and Spearman correlation respectively.