

# 1 Recharge estimation and soil moisture dynamics in a 2 Mediterranean, semi-arid karst region

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## 8 9 **Abstract**

10 Knowledge of soil moisture dynamics in the unsaturated soil zone provides valuable information  
11 on the temporal and spatial variability of groundwater recharge. This is especially true for the  
12 Mediterranean region, where a substantial fraction of long-term groundwater recharge is expected  
13 to occur during high magnitude precipitation events of above-average wet winters. To elucidate  
14 process understanding of infiltration processes during these extreme events, a monitoring network  
15 of precipitation gauges, meteorological stations, and soil moisture plots was installed in an area  
16 with a steep climatic gradient in the Jordan Valley region. In three soil moisture plots, Hydrus-1D  
17 was used to simulate water movement in the unsaturated soil zone with soil hydraulic parameters  
18 estimated by the Shuffled Complex Evolution Metropolis algorithm. To generalize our results, we  
19 modified soil depth and rainfall input to simulate the effect of the pronounced climatic gradient  
20 and soil depth variability on percolation fluxes and applied the calibrated model to a time series  
21 with 62 years of meteorological data.

22 Soil moisture measurements showed a pronounced seasonality and suggested rapid infiltration  
23 during heavy rainstorms. Hydrus-1D successfully simulated short and long-term soil moisture  
24 patterns, with the majority of simulated deep percolation occurring during a few intensive rainfall  
25 events. Temperature drops in a nearby groundwater well were observed synchronously with  
26 simulated percolation pulses, indicating rapid groundwater recharge mechanisms. The 62-year  
27 model run yielded annual percolation fluxes of up to 66% of precipitation depths during wet years  
28 and of 0% during dry years. Furthermore, a dependence of recharge on the temporal rainfall  
29 distribution could be shown. Strong correlations between depth of recharge and soil depth were  
30 also observed.

# 1     **1     Introduction**

2     In the Mediterranean region, groundwater is the main source for domestic and agricultural water  
3     supplies (EUWI, 2007). Knowledge on the quantity of groundwater recharge is a prerequisite for  
4     sustainable water resources planning and effective water use. Small-scale differences in climate,  
5     geology, land use, topography and soil properties cause a high spatial and temporal variability of  
6     groundwater recharge making the assessment and predictions of recharge a challenge (e.g. Zagana  
7     et al., 2007). Karst areas are important in this respect, because during high intensity winter storms  
8     precipitation may rapidly infiltrate into exposed karst surfaces and induce high recharge rates (De  
9     Vries and Simmers, 2002), which are common in the Mediterranean area (Ford and Williams,  
10    2007). A rapidly increasing water demand in the last decades has led to a widespread  
11    overexploitation of groundwater resources (EUWI, 2007). Furthermore, the Mediterranean region  
12    has been identified as a “hot spot” of current and future climate change (Giorgi, 2006; IPCC,  
13    2013), imposing additional pressure on its limited water resources. Hence, more insights into  
14    processes of aquifer replenishment in Mediterranean karst regions are of vital importance.

15    A large variety of methods suitable for estimating recharge rates were developed in the last  
16    decades (De Vries and Simmers, 2002; Scanlon et al., 2002). Infiltration, percolation and recharge  
17    quantities in Mediterranean karst have mainly been approached from two sides: On the one hand,  
18    hydrologists and geomorphologists characterized the surface water balance on small plots by  
19    sprinkling experiments or by runoff measurements during natural rainstorms (e.g. Cerdà, 1998;  
20    Lavee et al., 1998). Large-scale experiments also included tracers and facilitated statements on  
21    runoff generation processes (e.g. Lange et al., 2003). However, these studies quantified infiltration  
22    by the difference between artificial/natural rainfall and measured overland flow but did not  
23    differentiate between recharge and evapotranspiration. On the other hand, hydrogeologists  
24    frequently assessed average recharge rates of entire karst catchments from spring discharge  
25    measurements or hydraulic head data. Methods include knowledge (GIS)-based mapping (Andreo  
26    et al., 2008), multiple linear regression (Allocca et al., 2014), conceptual models (e.g. Hartmann et  
27    al., 2013a), coupled water-balance groundwater models (Sheffer et al., 2010), and chloride mass  
28    balances (Marei et al., 2010; Schmidt et al., 2013). However, these studies treat karst systems as  
29    units, including both the unsaturated and the saturated zones, and are limited in temporal and  
30    spatial resolution. Studies on cave drips (Gregory et al., 2009; Arbel et al., 2010; Lange et al.  
31    2010) provided insights into the deeper unsaturated zone in terms of water storage, spatial  
32    variability of percolation and flow paths. Their data was also used to incorporate variability in  
33    recharge modelling (Hartmann et al., 2012). However, it was difficult to distinguish between  
34    processes in the unsaturated soil zone and in the underlying epikarst, and uncertainty remains  
35    regarding the representativeness of cave drip data with respect to infiltration processes. This is  
36    mainly due to the facts that the contributing areas of cave drips are unknown and caves might  
37    have developed their own hydraulic environments. Therefore cave drips are not necessarily  
38    representative for the bulk karst vadose zone (Lange et al., 2010).

39    Only limited knowledge on recharge dynamics is available for the carbonate Mountain Aquifer  
40    system shared between the West Bank and Israel, although it is of regional importance. First

1 recharge estimates were based on long-term spring discharge and groundwater well abstraction  
2 data (Goldschmidt and Jacobs, 1958). Later, groundwater flow models were used to establish  
3 empirical rainfall-recharge relationships (Baida and Burstein, 1970; Guttman and Zukerman,  
4 1995; Zukerman, 1999). Average recharge rates were assessed by a simple water balance  
5 approach (Hughes et al., 2008) and by a chloride mass balance (Marei et al., 2010). Sheffer et al.  
6 (2010) coupled a water budget model with a groundwater flow model for the entire western part  
7 of the Mountain Aquifer and used spring discharge and groundwater level data for calibration.  
8 They reported recharge rates ranging between 9% and 40% of annual rainfall and showed that the  
9 temporal distribution of rainfall within the winter season had considerable effects on overall  
10 recharge rates.

11 Observations of soil moisture may offer unique insights into near-surface hydrological processes,  
12 because water fluxes are susceptible to conditions and properties of the vadose soil zone across  
13 several scales (Vereecken et al., 2008). Yet, soil moisture is rarely measured in semi-arid areas  
14 and is seldom used for recharge estimation purposes. Scott et al. (2000) exemplified the potential  
15 of soil moisture time series to calibrate Hydrus-1D soil hydraulic parameters in southeastern  
16 Arizona. Their results demonstrated the high inter-annual variability of water fluxes in these  
17 environments where considerable percolation only occurs during above-average wet years.

18 The objective of this study is to investigate the spatial and temporal variability of soil water  
19 percolation, and hence groundwater recharge rates, for an Eastern Mediterranean carbonate  
20 aquifer. We use continuously recorded soil moisture data to calibrate one-dimensional water flow  
21 models (Hydrus-1D) with the Shuffled Complex Evolution Metropolis (SCEM) algorithm. The  
22 calibrated models are then used to assess spatial and temporal patterns of soil water percolation in  
23 a Mediterranean karst area, which is characterized by strong climatic gradients and variable soil  
24 depths.

25 A common challenge of hydrological research in semi-arid and developing regions is the lack of  
26 data. At the same time, sound knowledge on the often-limited water resources is of vital  
27 importance, especially in karst areas. This situation necessitates compromises. The calibrated soil  
28 hydraulic parameters of our model should be treated as effective parameters that represent both  
29 preferential and matrix flow components within a single, unimodal pore size distribution. They  
30 are site-specific and should not be used to characterize the physics of a porous medium with the  
31 given grain size distribution. Despite increasing work on (preferential) water transport in  
32 heterogeneous porous media, there is still no convincing integrated physical theory about non-  
33 Darcian flow at the scale of interest (Beven and German, 2013). And even if such a theory  
34 existed, measurement problems in natural clay soils would restrict its application to laboratory  
35 monoliths. From this perspective, the use of a simple model with a minimum number of calibrated  
36 parameter seemed to be a valid compromise to infer statements on groundwater recharge from a  
37 limited number of measurements in the unsaturated zone.

38

## 2 Study area

Our study area is located on the western margin of the Jordan Rift Valley 25 km northeast of Jerusalem (Figure 1). Precipitation shows a pronounced seasonality with cold fronts (mainly Cyprus lows) carrying moisture from the Mediterranean Sea during winter season from October to April (Goldreich, 2003). High rainfall intensities can occur mainly in autumn and spring from convective rainfall events originating from the South (Red Sea Troughs). The topographic gradient from the mountain range (highest elevation: 1016 m a.s.l.) in the west to the Jordan Valley in the east results in a strong precipitation gradient and arid conditions in the Jordan Valley (rain-shadow desert). Long-term average annual precipitation decreases from 532 mm in Jerusalem (810 m a.s.l.) to 156 mm in Jericho (290 m b.s.l.) (Morin et al., 2009). Mean annual potential evapotranspiration add up to 1350 mm in the mountains and 1650 mm in the Jordan Valley (Israel Meteorological Service – <http://www.ims.gov.il>).

Outcropping geological formations consist of carbonate rocks of the Upper Cretaceous age (Begin, 1975). They are composed of fractured and highly permeable layers of limestone and dolomite alternating with marl and chalk layers of low permeability, often considered partial aquicludes (Weiss and Gvirtzman, 2007). Senonian chalks form outcrops of low hydraulic conductivity in the southeast (Rofe and Raffety, 1963). Soil parent material consists of residual clay minerals from carbonate rock weathering and from the aeolian input of dust (silt and clay fraction) originating from the Sahara desert (Yaalon, 1997). Predominant soil types are Terra Rossa and Rendzina, both characterized by high clay contents. Rendzina soils contain carbonate in the soil matrix, are thinner and still show recent development, whereas Terra Rossa soils were formed under past climatic conditions (Shapiro, 2006). As a result of the diverse underlying carbonate rock with different degrees of weathering and due to heterogeneous topography, soil depth is highly variable. The slopes are covered by massive bedrock exposures, and loose rock fragments of different sizes alternate with soil pockets of variable dimensions, shapes, and depths (Figure 2). Soil development is intensified where dissolution cracks and karst fissures provide favourable drainage of the vadose soil zone to the underlying bedrock. In valley bottoms, fine textured alluvial soils (Vertisols) with soil depths up to several meters have developed. Shallow Brown Lithosols and loessial Arid Brown Soils dominate in the eastern, low-lying areas receiving less rainfall (Shapiro, 2006). In general, soils in the region have significantly been transformed by human activities such as land cultivation, terracing, and deforestation during the last 5000 years (Yaalon, 1997).

On the hillslopes, annual plants and Mediterranean shrubs (predominantly *Sarcopoterium spinosum*) are the dominant vegetation types. They are used for extensive grazing by goats and sheep. South-facing slopes show lower vegetation density and higher proportion of bare soil and rock outcrops than the north-facing slopes, where the presence of biogenic crusts was reported (Kutiel et al., 1998). Minor land use types consist of scattered built-up areas, olive plantations on terraced land and rainfed or partly irrigated agricultural land (annual and perennial crops, herbs and vegetables) in valley bottoms.

## 3 Material and methods

### 3.1 Hydrometeorological measurements

To capture the spatial variation of rainfall along the strong climatic gradient, we installed a rain gauge network (Figure 1) consisting of 14 tipping buckets (RG3-M) connected to a HOBO pendant event data logger (Onset Computer Corporation), recording 0.2 mm per tip. Daily cumulative precipitation was calculated from event data. All gauges were calibrated before employment, maintained, and cleaned twice a year before and after the rainfall season. Temperature was measured at four climatic stations (Thies GmbH and Onset Computer Corporation) at 10-minute intervals. Additional rainfall and climatic data was obtained from the Israel Meteorological Service database (<http://www.data.gov.il/ims>) for long-term analyses. Every 20 minutes, groundwater levels and temperatures were recorded in a well tapping the local perched spring aquifer using pressure transducers (Mini-Diver, Eijkelkamp). Moreover, we measured water levels in several ephemeral streams of Wadi Auja with pressure transducers (Mini-Diver, Eijkelkamp; Dipper-3, SEBA Hydrometrie) every 5 minutes. Irrigation experiments (Sohrt et al., 2014) demonstrated that infiltration rates at locations close to the soil moisture plots were considerably higher than measured rainfall intensities during our observation period.

### 3.2 Soil moisture measurements

Seven soil moisture plots were installed, each equipped with four capacitance soil moisture sensors (5TM/5TE, Decagon Devices Inc.), measuring soil moisture and soil temperature every 10 minutes. We paid attention that the plots did not receive lateral surplus water from upslope overland flow by placing them distant from rock outcrops and at locations with minimum slope. To minimize disturbance, we inserted the sensors vertically into the upslope wall of manually dug soil pits (depth between 50 cm and 100 cm). After installation, we refilled the pits with the parent soil material and compacted approximately to pre-disturbance bulk density. The probes were connected to data loggers (EM50, Decagon Devices Inc.), which were sealed by plastic bags and buried in the soil to avoid vandalism. We used the internal calibration function for mineral soils with a measurement accuracy of 4% of the volumetric water content (VWC). Further information on the performance of the employed sensors can be found in Kizito et al. (2008). Due to instrument malfunction and vandalism, we obtained continuous data of the entire measurement period (October 2011 to May 2013) from only three locations (SM-1–SM-3). Plot SM-1 is located at a gentle part of a slope, while SM-2 and SM-3 are located on rather flat topography. Further characteristics of the plots are summarized in Table 1.

The dielectric permittivity of water changes with temperature (e.g. Wraith and Or, 1999). Hence, measurement techniques of soil moisture based on the difference of dielectric permittivity between water and soil matrix are affected by this phenomenon. In our case, soil temperature was highly variable and changed by up to 20 °C within 24 hours due to a strong radiation input and partly uncovered soil. We corrected our soil moisture data applying multiple linear regressions against soil temperature as described by Cobos and Campell (2007).

### 3.3 Soil sampling and multistep outflow experiments

We took 35 undisturbed soil samples (height = 4 cm, diameter = 5.6 cm) with a volume of 100 cm<sup>3</sup> in the surrounding of the soil moisture plots in depths between 5 cm and 70 cm. They were analysed in the laboratory of the Forest Research Institute of Baden-Württemberg, Freiburg, Germany by means of multistep outflow (MSO) experiments (Puhmann et al., 2009). The setup of the MSO-experiments was based on the pressure cell method, where samples were equipped with microtensiometers, placed on porous ceramic plates and gradually saturated. Suctions of up to -500 hPa were gradually applied at the bottom of the ceramic plates. Cumulative outflow as well as the pressure head were continuously monitored and logged. Furthermore, samples were placed in a pressure plate apparatus to obtain points of the retention curves at -900 hPa. Mualem/van-Genuchten parameters were derived by means of an inverse parameter optimization procedure. We compared water retention and conductivity functions from the laboratory MSO-experiments with those derived through inverse modelling of our soil moisture plots.

### 3.4 Modelling of the soil zone

Water balance at the plot scale in absence of surface runoff can be described by:

$$\frac{ds}{dt} = P - E_a - L \quad \text{with} \quad E_a = E_i + E_s + E_t, \quad (1)$$

where  $ds/dt$  is the storage change over time,  $P$  is the precipitation,  $L$  is the percolation at the profile bottom and  $E_a$  is the evapotranspiration per time interval.  $E_a$  is composed of the terms  $E_i$  (evaporation of intercepted precipitation),  $E_s$  (soil evaporation) and  $E_t$  (plant transpiration).

For our three soil moisture plots, soil water content and water fluxes were simulated on a daily basis with Hydrus-1D (version 4.16; Šimůnek et al., 2013) for a period of 32 months. Hydrus-1D solves the Richards equation numerically for water transport in variable saturated media. Matric potential dependent water retention and hydraulic conductivity were calculated using the Mualem/van-Genuchten soil hydraulic model (van Genuchten, 1980). To reduce the effect of non-linearity of the hydraulic conductivity function close to saturated conditions, an air entry value of -2 cm as suggested by Vogel et al. (2001) was used. Interception by the plant canopy was calculated by an empirical equation including the leaf area index and daily precipitation values (see Šimůnek et al., 2013 for more details). Potential evapotranspiration was calculated by the Hargreaves-equation (Hargreaves and Samani, 1985). Originally developed for a lysimeter station in California, this method adequately reproduced potential evapotranspiration under semi-arid climates (Jensen et al., 1997; Weiß and Menzel, 2008). Potential evapotranspiration was split into potential evaporation from the soil surface and potential transpiration from plants according to Beer's law based on the time variable surface cover fraction. Both fluxes were reduced to actual values based on a root water uptake model (Feddes et al., 1978) applying plant parameters for grass and an energy balance surface evaporation model (Camillo and Gurney, 1986). In our study area, vegetation cover shows a strong seasonality due to the restricted water availability during the dry season. To account for this, time dependent plant growth data was implemented into the model with intra-annual variation of surface cover fraction. According to field observations, the

1 start of the growing season was set to mid November and the maximum vegetation density was  
 2 assumed for February/March shortly after the largest monthly precipitation amounts were  
 3 observed. The depth from which plants took up water was controlled by a root distribution  
 4 function. An exponential decrease of root density with soil depth was assumed, observed at the  
 5 study sites and often reported for the Mediterranean region (e.g. De Rosnay and Polcher, 1998; De  
 6 Baets et al., 2008). Temporal variations of rooting depth and root density were disregarded. With  
 7 these components, Hydrus-1D continuously computed water content and water fluxes at user  
 8 defined observation points (here: depths of the soil moisture probes) and at the lower profile  
 9 boundary. Model input data, selected parameter values and their ranges, and the corresponding  
 10 data sources and calculation methods are summarized in Table 2.

### 11 **3.5 Calibration procedure, uncertainty analysis and parameter sensitivity**

12 An increase of clay content and bulk density with depth was observed at all profiles and the  
 13 individual probes in various depths at our plots differed noticeably. As a result, a particular soil  
 14 material with singular soil hydraulic properties was independently assigned for each soil moisture  
 15 probe. Observed soil moisture data from two winter and one summer season (October 2011 to  
 16 April 2013) were used for calibration of Hydrus-1D. We individually determined soil hydraulic  
 17 parameters for every soil material by inverse modelling using the Shuffled Complex Evolution  
 18 Metropolis optimization algorithm (SCEM; Vrugt et al., 2003) and the Kling-Gupta efficiency  
 19 (KGE; Gupta et al., 2009) in a modified version from Kling et al. (2012) as the objective function:

$$20 \quad KGE = 1 - \sqrt{(r - 1)^2 + (\alpha - 1)^2 + (\beta - 1)^2} \quad (2)$$

21 with:

$$22 \quad r = \frac{Cov_{so}}{\sigma_s \sigma_o}, \alpha = \frac{\mu_s}{\mu_o} \text{ and } \beta = \frac{\sigma_s / \mu_s}{\sigma_o / \mu_o},$$

23 where  $r$  is the correlation coefficient between simulated and observed VWC ( $Cov_{so}$  is the  
 24 covariance between simulated and observed VWC),  $\alpha$  is a dimensionless measure for the bias ( $\mu_s$   
 25 and  $\mu_o$  are the mean simulated and observed VWC) and  $\beta$  is a dimensionless measure for  
 26 variability ( $\sigma_s$  and  $\sigma_o$  are the standard deviations of simulated and observed VWC). SCEM is  
 27 widely used to efficiently solve global optimization problems (e.g. Vrugt et al., 2005; Schoups et  
 28 al., 2005; Feyen, 2007; Hartmann et al., 2012) and to find optimal model parameter sets. As  
 29 algorithmic parameters for SCEM, 24 complexes/parallel sequences were selected (equal to the  
 30 number of parameters to be optimized), the population size was set to 144 and the number of  
 31 accepted draws to infer posterior distribution was set to 1000. The SCEM routine was run until  
 32 the scale reduction score (SR), a convergence criterion defined by Gelman and Rubin (1992), was  
 33 fulfilled. As proposed by Vrugt et al. (2003), a SR value of 1.2 was chosen, indicating that the  
 34 Markov chain had converged to a stationary posterior distribution for all parameters. Predicted  
 35 soil moisture ranges were used for parameter uncertainty assessment. They were determined by  
 36 running Hydrus-1D with 1000 parameter sets obtained through the SCEM algorithm after  
 37 reaching convergence.

38

### 1   **3.6    Spatial and temporal extrapolation of percolation**

2    To extrapolate our point measurements of soil water balance, we varied soil depth and climatic  
3    input parameters (precipitation and temperature) over ranges observed in our study area. We used  
4    the calibrated soil hydraulic parameters of our deepest (1 m) soil moisture plot (SM-1), which had  
5    sensors at 10, 25, 40 and 80 cm. Moreover, we assumed that the rooting depth was limited to the  
6    soil depth with no changes in the vertical root distribution or plant surface cover fraction. We cut  
7    off the profile according to the simulated soil depth, which reduced the number of independent  
8    soil layers when the depths fell below 60, 32.5 and 17.5 cm. For soil thicknesses exceeding 1 m,  
9    we extended the bottom layer. To simulate the range of climatic conditions with elevations  
10   between 400 and 1000 m a.s.l., we modified rainfall and air temperature according to calculated  
11   mean annual gradients based on observed rainfall and climatic data. We had three seasons of  
12   measured climate data, which we analysed separately due to seasonal differences in cumulative  
13   rainfall amount and distribution.

14   Using a 62-year record of rainfall and temperature (1951–2013) available for Jerusalem (Israel  
15   Meteorological Service – [www.data.gov.il/ims](http://www.data.gov.il/ims)), we assessed the annual variability of water  
16   balance components at the location of our three soil moisture plots. Rainfall and temperature data  
17   from Jerusalem station were corrected for elevation differences between the Jerusalem station  
18   (810 m a.s.l.) and the three plots based on calculated elevation gradients.



## 1    **4    Results**

### 2    **4.1    Hydrometeorological conditions**

3    The three years of high resolution measurements of precipitation and meteorological parameters  
4    revealed considerable interannual variability and a strong elevation gradient, especially in terms  
5    of rainfall. Mean seasonal precipitation at the Kafr Malek station (830 m a.s.l.) situated close to  
6    the Mediterranean Sea–Dead Sea water divide was 526 mm (380–650 mm), while mean seasonal  
7    rainfall at the Auja Village station (270 m b.s.l.) in the Jordan Valley accounted for 106 mm (97–  
8    120 mm) leading to seasonal rainfall gradients between 6.4% to 7.2% per 100 m elevation  
9    difference (Figure 3). Mean rainfall intensity for the single stations was between 0.8 mm/h and  
10    1.5 mm/h, while maximum intensities exceeded values of 10 mm/h at some stations for only few  
11    time intervals during the complete observation period. Convective rainfall events with high  
12    intensities presumably from Red Sea Troughs were observed only during a short time period in  
13    spring 2011 with cumulative amounts below 40 mm. Mean annual temperature was 7 °C higher at  
14    Auja Village whereas relative humidity, wind speed, and net solar radiation were slightly higher at  
15    the more elevated station. Stations from the Israel Meteorological Service with long-term records  
16    at locations in Jerusalem and the Jordan Valley showed similar characteristics. Three major runoff  
17    events resulted from storms with large precipitation amounts and periods of high intensity. Runoff  
18    coefficients were smaller than 5% for single events and less than 2% for the entire season.

### 19    **4.2    Soil moisture dynamics**

20    Observed soil moisture at all soil profiles (Figure 4) showed a strong seasonality where the annual  
21    course can be divided into distinct phases. At the beginning of the rainy season, the previously dry  
22    (8% to 17% VWC) soil profile was stepwise wetting up starting from the upper to the lower  
23    sensors. During rainfall events with high amounts and intensities, the soil moisture data showed  
24    rapid infiltration of water into the deeper portions of the profile. Particularly at plot SM-1,  
25    saturated conditions started from the bottom probe close to the soil-bedrock interface, where these  
26    conditions persisted for several hours up to two days. During the strongest rainfall events also  
27    upper soil layers reached saturation, however for much shorter periods (Figure 4b). At plot SM-3  
28    we found indications of soil saturation from the bottom up to the surface during two events for a  
29    period of 8 and 16 hours, respectively. At the end of the rainy season, the soil dried out within a  
30    few weeks and the soil moisture content further declined at a low rate during the whole dry  
31    summer period.

32

### 33    **4.3    Modelling of the soil zone**

#### 34    **4.3.1    Parameter optimization, uncertainty analysis and model validation**

35    Soil hydraulic parameters were optimized for the three soil moisture plots individually, using the  
36    Shuffled Complex Evolution Metropolis algorithm. Between 20,000 and 36,000 model runs were

1 conducted until the convergence criterion was fulfilled. The calibrated parameter sets used for  
2 further assessment of the plot scale soil water balance, are given in Table 3, and their distributions  
3 are illustrated in Figure 5. All models were generally able to reproduce the observed temporal soil  
4 moisture patterns with KGE values between 0.82 and 0.94 (Figure 6). However, differences in  
5 predictive capacities at distinct water content levels could be observed, which varied between the  
6 single plots (Figure 6 and Figure 7). In general, the model tended to overestimate water contents  
7 close to saturated conditions except for deeper sections at plot SM-1 where an underestimation of  
8 simulated water contents was observed.

9 Parameter uncertainty was assessed by simulation of water contents using parameter sets obtained  
10 with SCEM after fulfilling the convergence criterion. The 95% soil moisture confidence interval  
11 showed a narrow band around the optimum model (Figure 8 exemplary for plot SM-1). At all  
12 sensors the difference between simulated volumetric water content for the best parameter set and  
13 the 95% confidence interval remained below 4%, i.e. less than the measurement error of the  
14 sensors.

15 Water retention and conductivity functions from the laboratory MSO-experiments are given in  
16 Figure 9. In comparison with the functions from inversely calibrated parameter sets with Hydrus-  
17 1D, they show similar characteristics at lower matric potential with an increasing deviation at  
18 higher matric potentials. Residual water contents from the MSO-analyses were generally higher  
19 than the calibrated Hydrus-1D parameter for our soil moisture plots.

20 Water temperature in a groundwater well near soil moisture plot SM-3 (cf. Figure 1) indicated  
21 five distinct recharge events lowering the mean groundwater temperature from 19 °C by 0.7–4 °C  
22 (Figure 7). The events coincided with the main peaks of modelled percolation from the soil  
23 moisture monitoring sites. During these events, mean daily air temperature was less than 6 °C.  
24 Although the well was strongly influenced by irregular pumping for water supply (visible as  
25 minor water level fluctuations in Figure 7), major recharge events induced sudden rises of the  
26 piezometric water level.

### 27 **4.3.2 Plot scale water balance**

28 Modelled fluxes of the various water balance components showed high temporal variability  
29 (Figure 8) and considerable differences in annual values between single years (Table 4).  
30 Evaporation and transpiration started shortly after the first rainfall events of the winter season  
31 when the water content in the upper soil layer began to increase. Percolation from the bottom of  
32 the soil zone only started after the cumulative rainfall during winter season exceeded a certain  
33 threshold. This threshold was found to be ca. 240 mm at plot SM-1, 200 mm at plot SM-2, and  
34 150 mm at plot SM-3. This threshold was not a fixed value but varied from year to year  
35 depending on the precipitation distribution over the winter season. In case of the season 2010/11  
36 with below-average rainfall, evapotranspiration during dry spells reduced the soil water storage  
37 and rainfall amounts of the following events were too low to exceed field capacity and to generate  
38 percolation at SM-3. Interception, soil evaporation and transpiration were highly variable during

1 the winter season and depended on the length of dry spells between rainfall events.  
2 Evapotranspiration almost ceased within a few weeks after the last rainfall events of the winter  
3 season. Mean overall losses through evapotranspiration and interception accounted for 73% of  
4 rainfall. Values slightly above 100% for the dry year 2010/11 resulted from elevated moisture  
5 conditions at the beginning of the simulation period. Percolation strongly varied from negligible  
6 amounts during the dry year 2010/2011 to values ranging between 28% and 45% of cumulative  
7 rainfall during 2011/12 and 2012/13, respectively. The largest proportion of percolation was  
8 calculated during a few strong rainstorms. On all three plots, more than 50% of the total  
9 percolation of the three years simulation period occurred within a time period of five to ten days.

### 10 **4.3.3 Spatial extrapolation of deep percolation**

11 During the hydrological year 2010/11, cumulative rainfall was below average with totals ranging  
12 between 275 and 425 mm (Figure 10) and a maximum daily amount below 50 mm. In this season,  
13 percolation was only simulated for soils with depths up to 60 and 110 cm, respectively. Modelled  
14 percolation increased to a maximum proportion of 40% for shallow soils with depths of 10 cm  
15 receiving the highest rainfall input. For the following above-average wet year 2011/12, seasonal  
16 rainfall ranged between 450 and 725 mm. Then simulated percolation rates reached up to 69% of  
17 rainfall and declined to values close to 0% only under conditions of lowest rainfall amount and  
18 soil depths greater than 160 cm. The third simulated year can be regarded as a year with average  
19 rainfall conditions (sums of 400 to 600 mm). Percentages of percolation were comparable to the  
20 previous year although cumulative rainfall was considerably less. This could be attributed to  
21 higher rainfall intensities during 2012/13 when daily rainfall amounts exceeded twice 80 mm and  
22 four days of rainfall accounted for almost 50% of the seasonal amount.

### 23 **4.3.4 Temporal extrapolation of deep percolation**

24 Modelling water balance components for 62 years (1951–2013) resulted in strong differences of  
25 simulated seasonal soil water percolation reflecting the high variability of rainfall input (Figure  
26 11). Mean annual rainfall was calculated for the three plots to range between 408 and 537 mm  
27 (standard deviation: 128–168 mm) and mean percolation fluxes between 82 and 150 mm  
28 (standard deviation: 93–141 mm). Percolation at the three plots varied between 0% and 66% of  
29 cumulative seasonal rainfall with an average between 16% and 24%. Other seasonal fluxes varied  
30 much less during the simulation period. The coefficient of determination between seasonal sums  
31 of simulated percolation and rainfall ranged between 0.82 and 0.88 on the three plots.

32

## 5 Discussion

### 5.1 Soil moisture dynamics

The observed seasonal dynamics of soil moisture, dominated by short wetting phases during and a rapid decrease after the rainfall season, were comparable with those reported in other studies in the Mediterranean region (Cantón et al., 2010; Ruiz-Sinoga et al., 2011). At all soil moisture plots, our soil moisture data suggested fast infiltration into deeper sections of the soil profile during rainfall events with high intensities and amounts (e.g. plot SM-1 in Figure 4b). The time lag between the reaction of the uppermost and the lowermost probe was often less than two hours, indicating flow velocities of exceeding 840 cm per day, despite of high clay content. These fast reactions suggest concentrated infiltration and preferential flow within the vadose soil zone as reported for the Mediterranean region by e.g. Cerdá et al. (1998), Öhrström et al. (2002) and Van Schaik et al. (2008). Brilliant Blue patterns from infiltration experiments conducted in the vicinity of our plots highlighted the influence of outcrops on infiltration by initiating preferential flow at the soil-bedrock interface. In the remaining soil preferential flow was less distinct, but vertical flow velocities of 0.08 cm/min suggested also here macropore flow (Sohrt et al., 2014). Hence, a certain fraction of preferential flow is ubiquitous and may further be enhanced by a high stone content in the soil and by bedrock outcrops in the vicinity, as observed particularly at SM-1. In general bedrock and stones may have multiple effects on infiltration, water retention and water movement in the soil (Cousin et al., 2003).

A noticeable difference between the plots was observed during rainfall events of high magnitude. At SM-1 (Figure 4b), the bottom probe suggested soil saturation for periods between 2 and 90 hours. Durations were apparently linked to the depth of the event precipitation (24 to 191 mm) and to the duration of the event (16 to 72 h). The upper probes showed saturation only during the largest rainfall events and for a much shorter duration. Volumetric soil moisture at 10 cm always remained below 30%. We observed a similar behaviour at SM-3 but not at SM-2. We hypothesize that these phases of saturation were caused by impounded percolation water due to limited conductivity of the soil-bedrock interface. Differences between our plots could be attributed to the variable permeability of the underlying Cenomanian dolomite (SM-1 and SM-3) and Turonian limestone (SM-2). While both formations are known to have high permeability (Keshet and Mimran, 1993), we observed Nari Crust (Dan, 1977) in the vicinity of SM-1, which may have reduced hydraulic conductivity. Sprinkling experiments on the same geological material type had already documented soil saturation and subsequent overland flow generation (Lange et al., 2003).

### 5.2 Simulation of the plot scale water balance

The cumulative distribution functions of the parameters suggested narrow ranges and hence good identifiability for most model parameters (Figure 5). Nevertheless, measured soil moisture fell outside the 95% uncertainty band especially during high and low moisture conditions (Figure 7). This may indicate limitations of our simplified model, which is based on a unimodal pore-size distribution. By definition, our inversely estimated model parameters are effective parameters that

1 describe both, preferential and matrix flow. Compared to values of saturated hydraulic  
2 conductivity ( $K_s$ ) of a clay-rich soil matrix from established pedotransfer functions (e.g. Carsel  
3 and Parish, 1988), our  $K_s$  values are high (Table 3). Radcliffe and Šimůnek (2010) analysed data  
4 from the UNSODA soil hydraulic database (Nemes et al., 2001). They found decreasing  $K_s$  with  
5 increasing clay content but also a significant increase in parameter spread. This was attributed to a  
6 larger effect of soil structure. This effect will become more evident when moving from the scale  
7 of small soil cores to the plot scale, reflecting a common phenomenon of changing parameter  
8 values with changing spatial scale (e.g. Blöschl and Sivapalan, 1995). From this perspective, our  
9 estimated effective low alpha values describe the small pores of the soil matrix, while the high  
10 effective  $K_s$ -values represent the effect of preferential flow. Although clay content and bulk  
11 density slightly increased with soil depth at our plots, no clear pattern of calibrated soil hydraulic  
12 parameters could be observed. The expected decrease of  $K_s$  was apparently compensated by other  
13 factors such as the observed increasing stoniness of the soil with depth, which could lead to  
14 enhanced preferential flow at the soil-rock interface (Sohrt et al. 2014) or by water uptake by  
15 plants that was limited to the upper soil zone. Furthermore, persistent saturated conditions during  
16 major rainstorms as discussed in the previous section could not be simulated, as a percolation  
17 impounding soil-rock interface was not implemented in the model and a free drainage had to be  
18 assumed. Still, the conductivity and retention function derived from the MSO experiments  
19 showed an overall good agreement with those calibrated with the help of Hydrus-1D and SCHEM  
20 (Figure 9). We believe that this is another independent proof for the reliability of our simplified  
21 model. As discussed earlier, an increasing deviation of the respective functions with increasing  
22 matric potential could be addressed to the different measurement scales, where the MSO  
23 experiments represent mainly the soil matrix, while the parameter calibrated with Hydrus-1D  
24 comprise also preferential flow pathways at the plot scale. A bimodal pore-size distribution  
25 (Durner, 1994) may better represent the heterogeneous pore structure of our clay-rich soil, but  
26 at the cost of in a larger number of calibration parameter with presumably reduced parameter  
27 identifiability and higher model uncertainties.

28 Originally, Mualem (1976) set the parameter  $L$  to a fixed value of 0.5 for all soil types. Later, the  
29 physical interpretation of the parameter  $L$  representing tortuosity and pore connectivity was  
30 increasingly questioned and  $L$  was rather treated as an empirical shape factor for the hydraulic  
31 conductivity function in the Mualem/van-Genuchten model (Schaap and Leij, 2000). Schaap and  
32 Leij (2000) observed that fixed positive values of  $L$  can lead to poor predictions of the unsaturated  
33 hydraulic conductivity and that  $L$  was often negative for fine textured soils. Peters et al. (2011)  
34 analysed persistent parameter constraints for soil hydraulic functions and concluded that the  
35 conservative constraint of  $L > 0$  is too strict and that physical consistency of the hydraulic  
36 functions is given for:

$$37 \quad L > \frac{-2}{m} \quad \text{with } m = 1 - \frac{1}{n} \quad (3)$$

38 This constraint ensures monotonicity of the hydraulic functions. The requirement of Eq. (3) is  
39 fulfilled for all  $L$ -values of the parameter sets shown in Table 1.

1 Simulated mean evapotranspiration at our plots over the three-years simulation period accounted  
2 for 73% of rainfall, i.e. very close to the long-term average calculated by Schmidt et al. (2014) for  
3 the same area. Our values also fall into the range of Cantón et al. (2010), who derived annual  
4 effective evapotranspiration rates of more than 64% of annual rainfall based on eddy covariance  
5 measurements in southeastern semi-arid Spain. Our simulated percolation rates ranged between  
6 0% and 45% of precipitation (arithmetic mean: 28%) indicating strong inter-annual variability and  
7 a strong dependency on depth and temporal distribution of precipitation. During the entire three-  
8 year period, more than 50% of overall percolation fluxes occurred during less than 10 days of  
9 strong rainfall. These findings are supported by the response of groundwater temperatures  
10 observed in a nearby well indicating the arrival of groundwater recharge flux at the water table  
11 (Figure 7). Tracer experiments in a similar setting demonstrated that percolating water can pass  
12 the vadose soil and the epikarst at flow velocities of up to 4.3 m/h (Lange et al., 2010). Regarding  
13 the initiation of percolation at the basis of the soil profiles, we found seasonal rainfall thresholds  
14 of ca. 150 mm for the shallow and 240 mm for the deep soil moisture plots. Cave drip studies in  
15 the region (Arbel et al., 2010; Lange et al., 2010; Sheffer et al., 2011) measured similar thresholds  
16 for the initiation of percolation through the epikarst (100 to 220 mm).

17 In contrast to humid environments, lateral subsurface flow on rocky semi-arid hillslopes rarely  
18 develops, since they consist of individual soil pockets that are poorly connected due to frequent  
19 bedrock outcrops. Soil moisture seldom exceeds field capacity given that evapotranspiration  
20 exceeds precipitation depth throughout most of the year (Puigdefabregas et al., 1998).  
21 Furthermore, highly permeable bedrock favours the development of vertical structural pathways  
22 in karst areas (shafts beneath dolines and sinkholes). In the epikarst a lateral concentration of the  
23 percolation water from the soil zone toward such highly permeable pathways can take place  
24 (Williams, 1983). Despite this secondary concentration we can conclude that one-dimensional  
25 modelling of the soil water balance is a reasonable approach to understand percolation fluxes and  
26 subsequent groundwater recharge.

27

28 Nevertheless, we cannot exclude that frequently outcropping bedrock may affect water  
29 redistribution by surface runoff or by preferential infiltration along the soil-rock interface. The  
30 importance of these effects on percolation rates and groundwater recharge on the regional scale is  
31 subject to current research. During heavy storm events, overland flow generation cannot be  
32 excluded (Lange et al., 2003), but surface runoff typically accounts for only a few percent of  
33 annual rainfall (Gunkel and Lange, 2012). A second limitation of our investigations of plot scale  
34 percolation fluxes is the assumption of an identical vegetation cover at the single sites along the  
35 climatic gradient and a constant vegetation cycle throughout years of different seasonal rainfall  
36 depths. Although different plant species and vegetation cycles may alter soil moisture conditions  
37 prior to rainfall events, we could show that the event rainfall amount is the main factor that  
38 influences percolation rates.

39

### 5.3 Spatial and temporal extrapolation of deep percolation

Water balance modelling for variable soil depths and rainfall gradients revealed considerable differences for the three winter seasons. During the very dry year 2010/11, soil moisture exceeded field capacity only at locations with relatively shallow soils. During the wet years of 2011/12 and 2012/13, field capacity was exceeded several times at all plots and soils even reached saturation during strong rainfall events. This may lead to substantial percolation and groundwater recharge to local aquifers. These findings are in close agreement with discharge measurements at Auja spring, a large karst spring in the Jordan Valley, where 7 and 8 million m<sup>3</sup> were measured for the winter seasons 2011/12 and 2012/13 respectively, but only 0.5 million m<sup>3</sup> for the 2010/11 season (Schmidt et al., 2014).

A high temporal variability in percolation fluxes is also apparent from the long-term modelling of water balance components (Figure 11). For the 62-year simulation period, we calculated seasonal percolation rates between 0% and 66% (average: 20% to 28%) for our plots. The highest value was modelled for the extremely wet winter season 1991/92 (five times the mean annual percolation of 150 mm). For a slightly shorter time period, Schmidt et al. (2014) calculated an average recharge rate of 33% for the Auja spring catchment applying a conceptual reservoir model. They found that recharge of only five individual years accounted for one third of the total recharge of the 45-year period. In our study seven individual years provided one third of the total recharge. Furthermore, we compared seasonal percolation of our sites with recharge estimations from perched aquifers feeding small karst springs (Weiss and Gvirtzman, 2007) and the entire carbonate aquifer (Guttman und Zukerman, 1995) (Figure 12). Although our results plotted within the range of these large-scale recharge estimates, we want to emphasize that our calculations display point percolation fluxes. Even in years with below-average rainfall, a certain rise in the groundwater table and spring flow can be observed (season 2010/11 in Figure 7h; EXACT, 1998; Schmidt et al., 2014). Then recharge presumably occurs on areas with strongly developed epikarst and shallow or missing soil cover.

Our long-term point calculations suggest substantial differences in percolation fluxes between years of similar rainfall depths. Simulated percolation for plot SM-1 during the seasons 1976/77 and 2004/05 accounted for 16% and 35% of seasonal rainfall, respectively, although both seasons had very similar above-average rainfall (578 and 569 mm). These results are in line with findings of Sheffer et al. (2010) and Abusaada (2011) about the importance of temporal rainfall distribution on groundwater recharge.

### 5.4 Implications for recharge in Mediterranean karst areas

The steep climatic gradient, the hydraulic properties and characteristics of the carbonate rocks, the heterogeneous soil cover and the high temporal variability of precipitation on event and seasonal scales are dominating hydrological characteristics in our study area. Similar settings can be found across the entire Mediterranean region. Despite recent advances in the determination of groundwater recharge in karst areas, the assessment of the spatial and temporal distribution of recharge is still a challenge. Modelling approaches including hydrochemical and isotopic data

1 (Hartmann et al., 2013b) require additional information from springs (time series of discharge and  
2 water chemistry) for model parameter estimation, which are rarely available. Moreover, the exact  
3 delineation of the contributing recharge area is often a problem. Although simulated percolation  
4 fluxes from plot-scale soil moisture measurements cannot be directly transferred to the regional,  
5 i.e. catchment scale, they can still provide insights into the various processes responsible for the  
6 temporal and spatial variability of groundwater recharge as well as information on the relative  
7 importance of different process parameters.

## 8 **6 Conclusions**

9 This study contributes to the assessment of percolation rates based on soil moisture measurements  
10 along a steep climatic gradient in a Mediterranean karst area. We showed that point measurements  
11 of soil moisture together with numerical modelling of the water flow in the unsaturated soil zone  
12 may help to understand dominant percolation mechanisms. We found an accentuated annual  
13 variability of percolation fluxes and a strong dependency on soil thickness, temporal distribution  
14 and seasonal depth of rainfall. To extrapolate our findings, we varied soil depth and climatic input  
15 parameters (precipitation and temperature) over ranges observed in our study area. Furthermore,  
16 we used a 62-year time series (1951–2013) of climatic input to run our calibrated models.  
17 Although our calculations are based on plot scale measurements, the results closely match long-  
18 term observations and their patterns of event and seasonal variability. They also reflect the  
19 thresholds for the initiation of groundwater recharge reported by other studies in the same region  
20 based on different approaches. Our results suggest that groundwater recharge is most prominent  
21 when single rainfall events are strong enough to exceed field capacity of soil pockets over a wide  
22 range of soil depths. Hence, the temporal distribution of rainfall has a strong effect on event and  
23 seasonal recharge amounts.

24 Our results corroborate the statement of De Vries and Simmers (2002) about the dependence of  
25 groundwater recharge in semi-(arid) areas on high intensity rainfall events. The use of empirical  
26 rainfall-recharge relationships can lead to large errors, since recharge rates are sensitive with  
27 respect to highly variable rainfall distributions and characteristics, which are most probably  
28 affected by predicted climate change in the Mediterranean (Giorgi and Lionello, 2008; Samuels et  
29 al., 2011; Reiser and Kutiel, 2012).

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1 **Tables**

2 **Table 1.** Soil moisture plot characteristics.

<b>Plot</b>	<b>Elevation</b> (m a.s.l.)	<b>Average annual rainfall<sup>a</sup></b> (mm)	<b>Soil depth</b> (cm)	<b>Sensor depths</b> (cm)	<b>Vegetation</b>	<b>Texture<sup>c</sup></b>
<b>SM-1</b>	830	526	100	10, 25, 40, 80	Mediterranean shrubs; annual plants	Sand: 20% Silt: 40% Clay: 40%
<b>SM-2</b>	660	340 <sup>b</sup>	50	5, 10, 20, 35	Annual plants	Sand: 32% Silt: 33% Clay: 35%
<b>SM-3</b>	440	351	60	5, 10, 20, 35	Annual plants	Sand: 46% Silt: 24% Clay: 30%

3 <sup>a</sup> Mean rainfall based on three winter seasons (2010-2013).

4 <sup>b</sup> Rainfall at plot SM-2 is estimated by inverse distance weighted interpolation with elevation as  
5 additional predictor.

6 <sup>c</sup> Textural characteristics were determined in the laboratory by sieving (particle size >0.063 mm) and  
7 sedimentation method (particle size <0.063 mm)

1 **Table 2.** Parameters and value ranges for Hydrus-1D modelling.

	<b>Parameter</b>	<b>Value/Range</b>	<b>Unit</b>	<b>Source / calculation method</b>	
	<b>Soil hydraulic parameter</b>				
	$\Theta_r$	Residual soil water content <sup>b</sup>	0 – 0.3	m <sup>3</sup> /m <sup>3</sup>	Calibrated <sup>a</sup>
	$\Theta_s$	Saturated soil water content <sup>b</sup>	0.3 – 0.6	m <sup>3</sup> /m <sup>3</sup>	Calibrated <sup>a</sup>
	$\alpha$	Van Genuchten parameter related to air entry suction	0.0001 – 0.1	1/mm	Calibrated <sup>a</sup>
	$n$	Van Genuchten parameter related to pore size distribution	1.01 – 3	-	Calibrated <sup>a</sup>
	$K_s$	Saturated hydraulic conductivity	5 – 10000	mm/day	Calibrated <sup>a</sup>
	$L$	Van Genuchten parameter related to tortuosity	-2 – 2	-	Calibrated <sup>a</sup>
	<b>Meteorological parameter</b>				
	$P$	Daily precipitation		mm	Measured time series <sup>c</sup>
	$T_{max}$	Daily maximum temperature		°C	Measured time series <sup>d</sup>
	$T_{min}$	Daily minimum temperature		°C	Measured time series <sup>d</sup>
	$R_a$	Extraterrestrial solar radiation (for Hargreaves equation only)		MJ/m <sup>2</sup>	Calculated according to Allen et al. 1998
	<b>Vegetation parameter</b>				
	$D_r$	Rooting depth	0.5 – 1	m	Estimated based on field observations
	SCF	Surface Cover Fraction	0.1 – 1	m/m	Estimated based on field observations
	LAI	Leaf Area Index		m/m	Calculated according to Šimůnek (2013)
	$P_0$	Fedde's parameter	-100	mm	Hydrus-1D internal database (grass)
	$P_{opt}$	Fedde's parameter	-250	mm	Hydrus-1D internal database (grass)
	$P_{2H}$	Fedde's parameter	-3000	mm	Hydrus-1D internal database (grass)
	$P_{2L}$	Fedde's parameter	-10000	mm	Hydrus-1D internal database (grass)
	$P_3$	Fedde's parameter	-80000	mm	Hydrus-1D internal database (grass)
	$r_{2H}$	Fedde's parameter	5	mm/day	Hydrus-1D internal database (grass)
	$r_{2L}$	Fedde's parameter	1	mm/day	Hydrus-1D internal database (grass)
	$\alpha_i$	Interception constant	1	mm	Estimated
	$D_s$	Depth of soil profile	0.5 – 1	m	Measured at experimental plots

2 <sup>a</sup> Parameter calibrated for each soil material with SCEM algorithm and Kling-Gupta efficiency as  
3 optimization criterion.

4 <sup>b</sup> The upper parameter limit of  $\Theta_r$  and the lower parameter limit of  $\Theta_s$  were obtained from the lowest  
5 respectively highest measured volumetric soil moisture value of each layer in the respective soil  
6 moisture plot.

7 <sup>c</sup> Rainfall at plot SM-2 is estimated by inverse distance weighted interpolation with elevation as  
8 additional predictor.

9 <sup>d</sup> Maximum and minimum daily air temperature at the soil moisture plots is estimated by calculation of  
10 an elevation-temperature gradient based on meteorological stations in the Jordan Valley and the  
11 mountains.

1 **Table 3.** SCEM optimized hydraulic parameter sets for the different plots and probe depths.

<b>Plot</b>	<b>Layer</b>	<b><math>\Theta_r</math></b> (m <sup>3</sup> /m <sup>3</sup> )	<b><math>\Theta_s</math></b> (m <sup>3</sup> /m <sup>3</sup> )	<b><math>\alpha</math></b> (1/mm)	<b>n</b> (-)	<b>K<sub>s</sub></b> (mm/day)	<b>L</b> (-)	<b>KGE</b> (-)
<b>SM-1</b>	1 (-10 cm)	0.01	0.41	0.004	1.23	427	2.0	0.91
	2 (-25 cm)	0.12	0.49	0.026	1.30	8159	-2.0	0.94
	3 (-40 cm)	0.11	0.59	0.018	1.54	9468	-2.0	0.90
	4 (-80 cm)	0.10	0.59	0.028	1.36	8732	0.1	0.82
<b>SM-2</b>	1 (-5 cm)	0.00	0.49	0.041	1.18	126	-2.0	0.89
	2 (-10 cm)	0.05	0.40	0.002	1.23	5094	0.6	0.90
	3 (-18 cm)	0.12	0.59	0.012	1.37	9288	2.0	0.87
	4 (-55 cm)	0.13	0.51	0.013	1.43	2679	1.0	0.90
<b>SM-3</b>	1 (-5 cm)	0.00	0.60	0.008	1.23	482	-2.0	0.91
	2 (-10 cm)	0.00	0.56	0.004	1.23	9908	-1.2	0.92
	3 (-20 cm)	0.05	0.46	0.003	1.22	9976	1.2	0.91
	4 (-35 cm)	0.11	0.60	0.001	1.66	5751	2.0	0.94

2

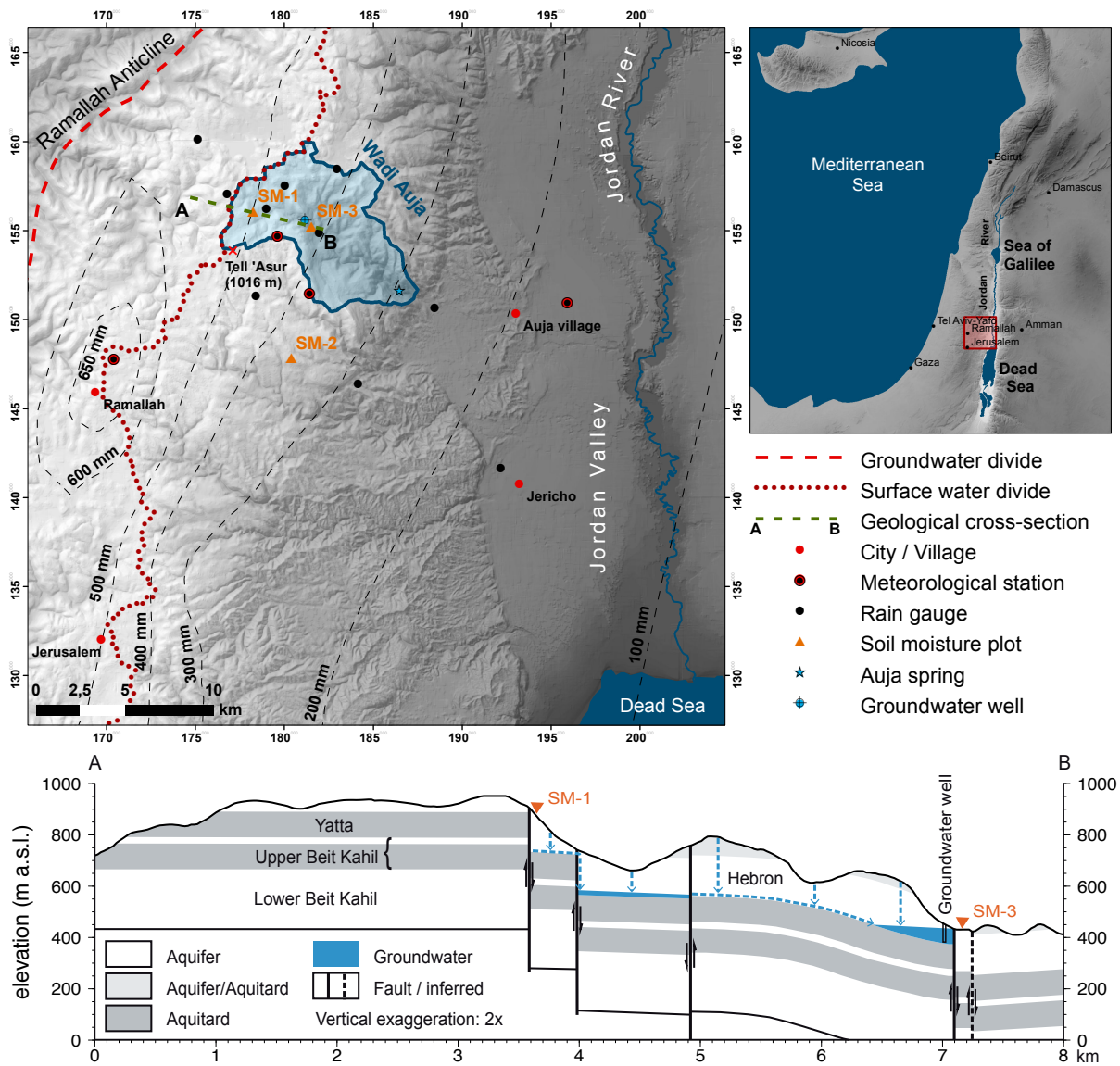
1 **Table 4.** Cumulative sums of the simulated water balance components in mm and % for the three  
 2 consecutive hydrological years 2010-2013 at the individual soil moisture plots.

Plot	Year	Rainfall	Interception	Evaporation		Transpiration		Bottom flux		
		(mm)	(mm)	(%)	(mm)	(%)	(mm)	(%)	(mm)	(%)
SM-1	2010/2011	381	62	16	99	26	209	55	13	3
	2011/2012	650	59	9	93	14	209	32	294	45
	2012/2013 <sup>a</sup>	547	39	7	102	19	179	33	224	41
SM-2	2010/2011	248	53	21	81	33	117	47	0	0
	2011/2012	418	55	13	89	21	159	48	118	28
	2012/2013 <sup>a</sup>	346	33	10	84	24	127	37	101	29
SM-3	2010/2011	237	47	20	119	50	84	35	2	1
	2011/2012	436	53	12	120	27	130	30	135	31
	2012/2013 <sup>a</sup>	380	30	8	111	29	105	28	125	33

3 <sup>a</sup> The hydrological year 2012/2013 was modelled until 30<sup>th</sup> of April 2013.

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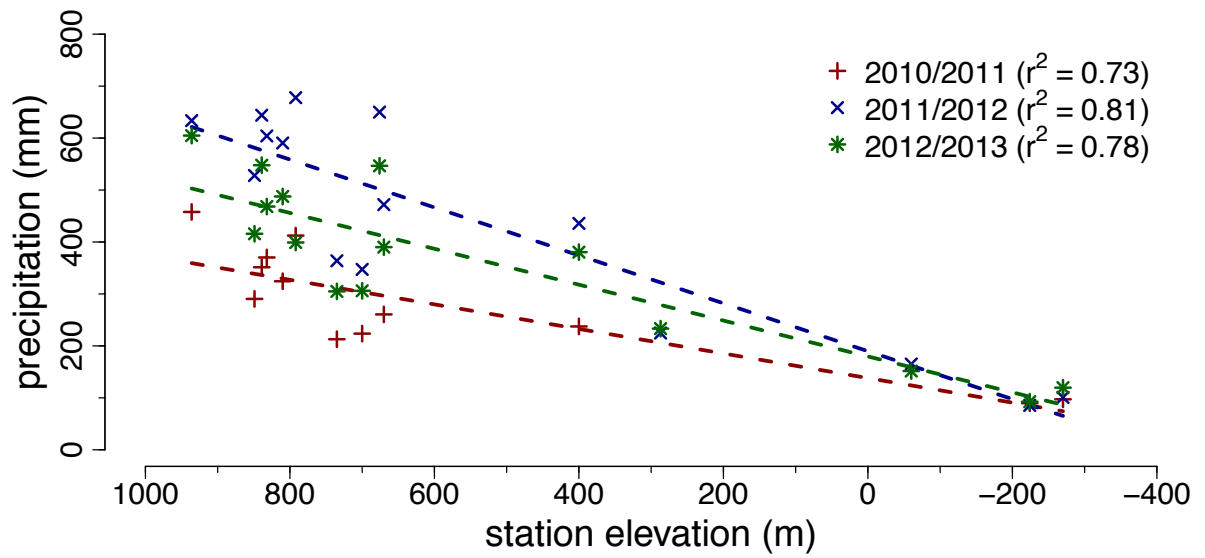
1 **Figures**



2  
3 **Figure 1.** Study area with location of meteorological stations, rain gauges, soil moisture plots  
4 (SM-1, SM-2, SM-3) and isohyets of long-term average annual rainfall ( $\geq 20$  years) according to  
5 data from ANTEA (1998). Coordinates in the detailed map are in Palestinian Grid format. In the  
6 upper slope sections of Wadi Auja a local perched spring aquifer formed which is tapped by an  
7 abstraction well.

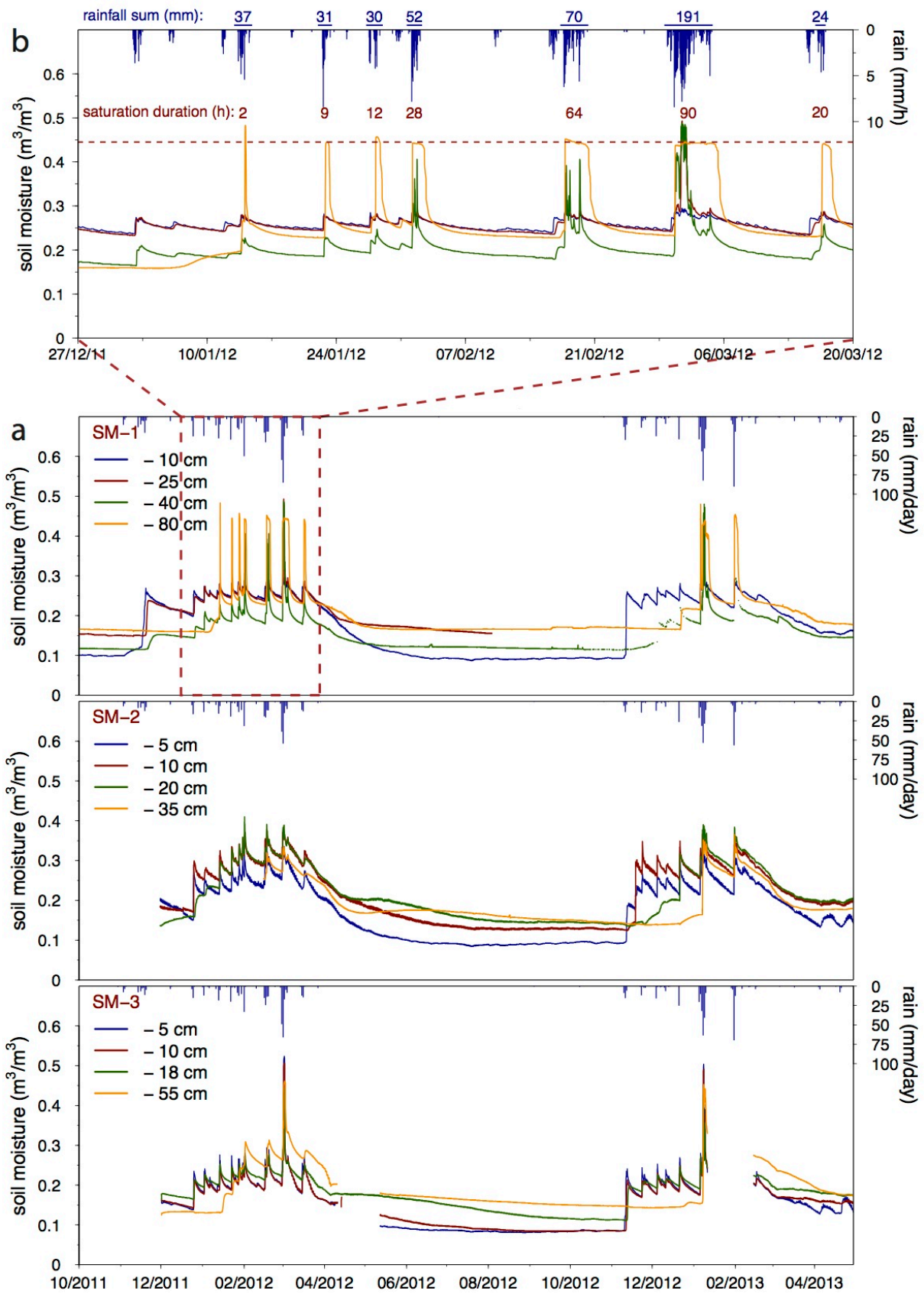


1  
2 **Figure 2.** Typical hillslopes in the study area. The image shows the plain of Ein Samia with semi-  
3 arid climatic conditions, where the valley bottom is used for partly irrigated agriculture and the  
4 hillslopes are used as extensive grazing land for goats and sheep. Wadi Auja ephemeral stream  
5 enter the plain from the left (West) and drains to the right (East) in direction to the Jordan Valley.



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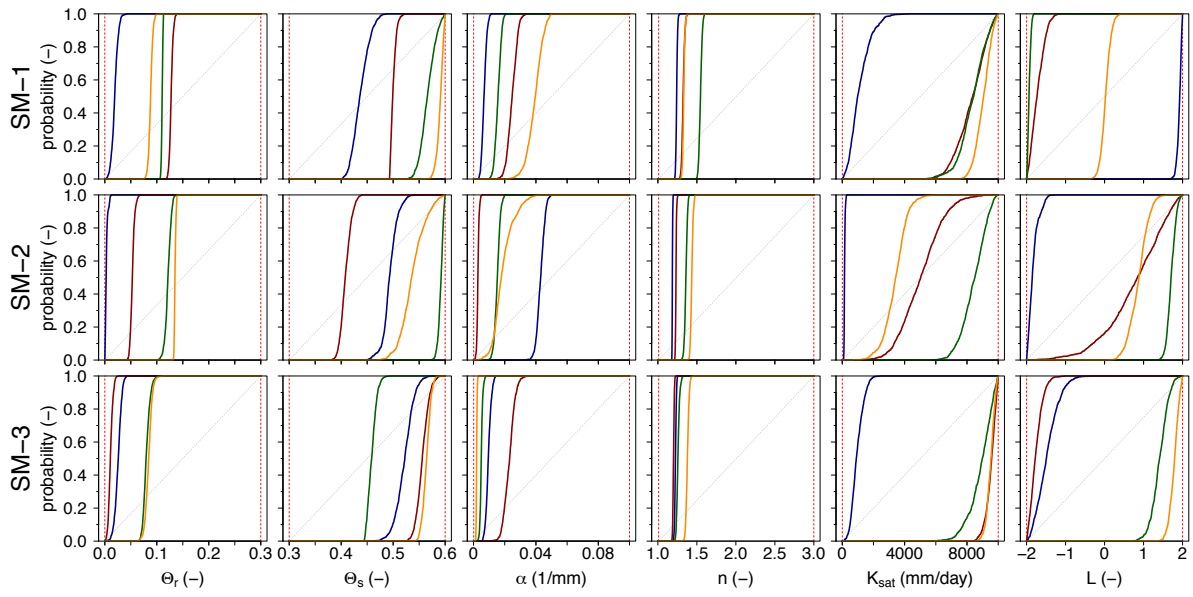
2 **Figure 3.** Correlation between average annual rainfall and station elevation for the individual  
 3 hydrological years during the observation period 2010-2013.



1  
2 **Figure 4.** Observed volumetric soil moisture at different depths of the three experimental plots  
3 during the complete monitoring period (a) and details on the winter season 2011/2012 for plot  
4 SM-1 (b).

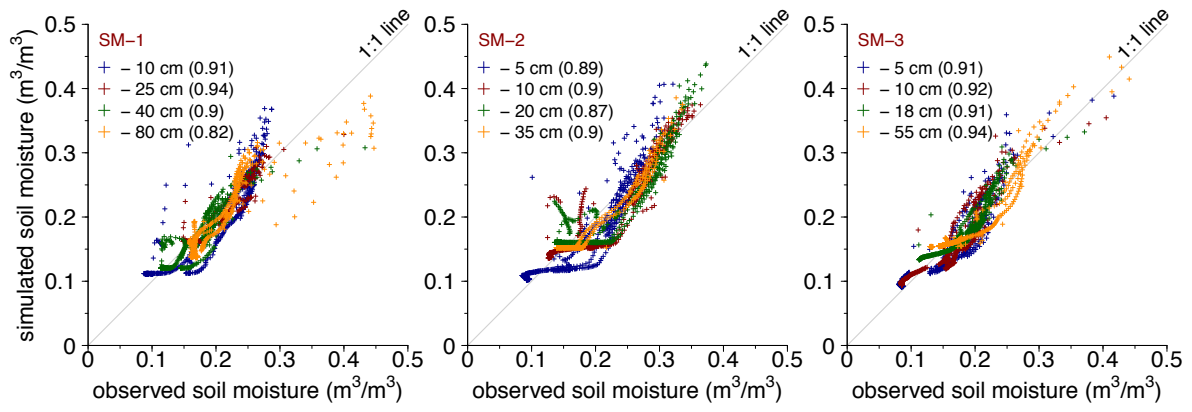


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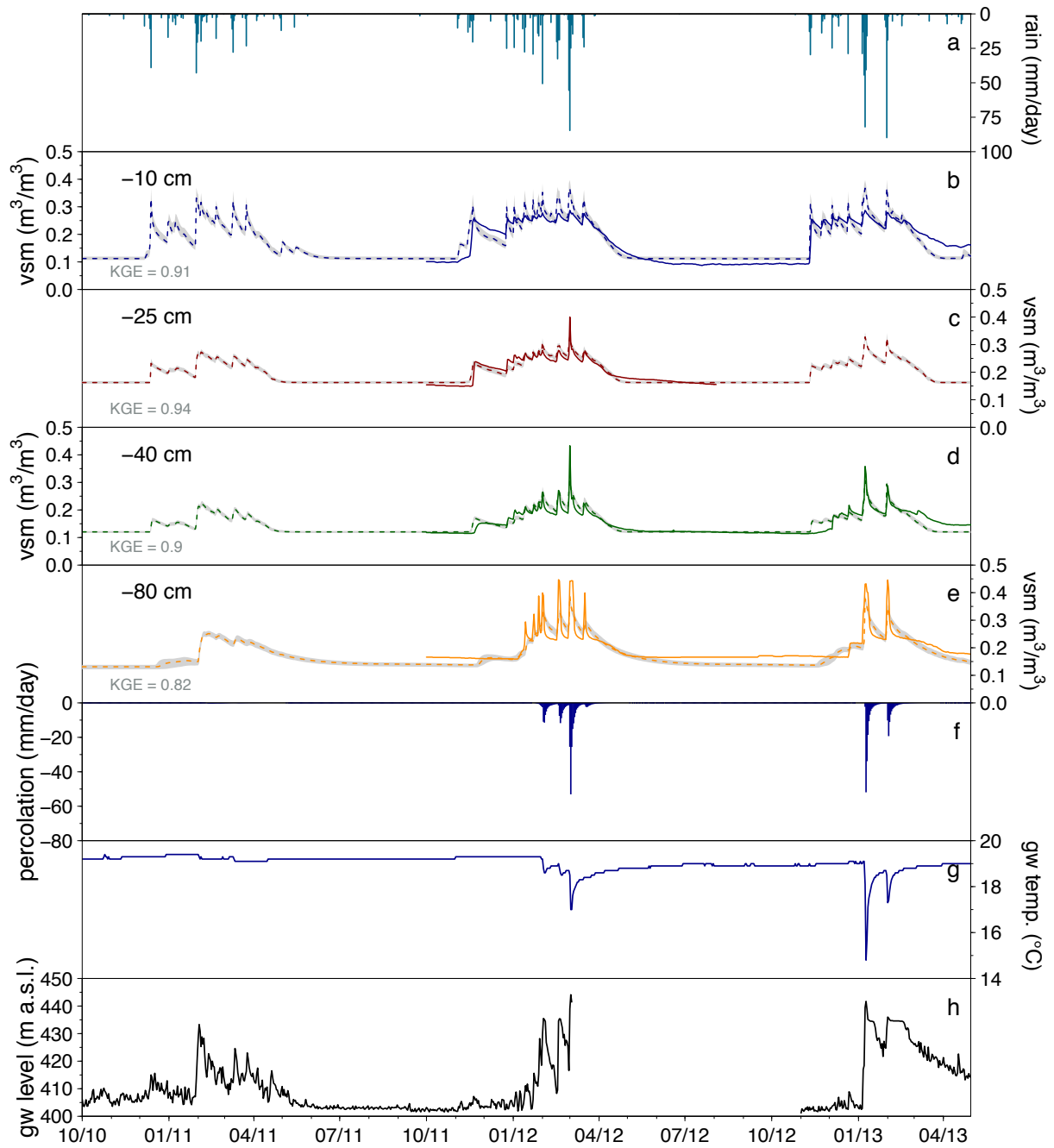
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3 **Figure 5.** Observed volumetric soil moisture at different depths of the three experimental plots  
4 during the complete monitoring period (a) and details on the winter season 2011/2012 for plot  
5 SM-1 (b).



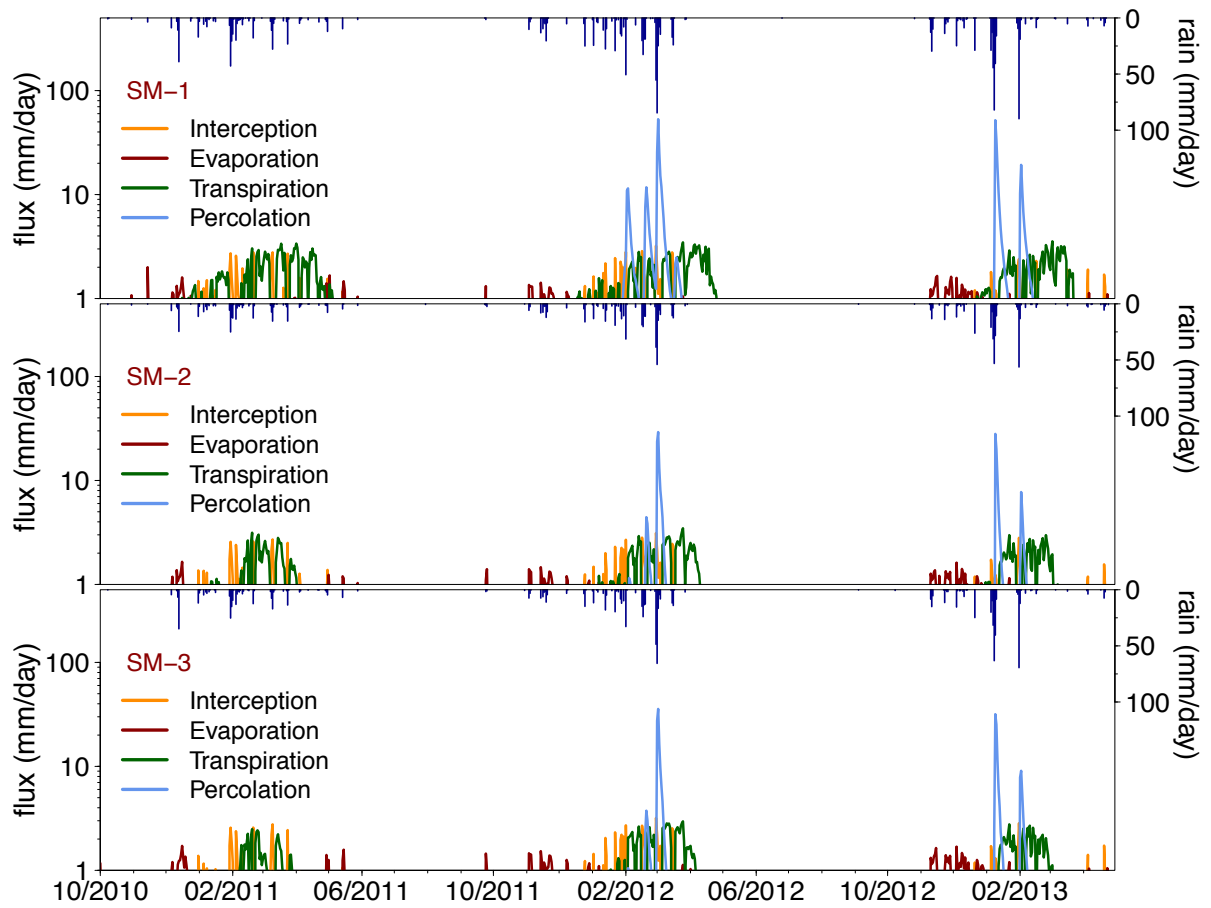
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2 **Figure 6.** Observed volumetric soil moisture at different depths of the three experimental plots  
 3 during the complete monitoring period (a) and details on the winter season 2011/2012 for plot  
 4 SM-1 (b).



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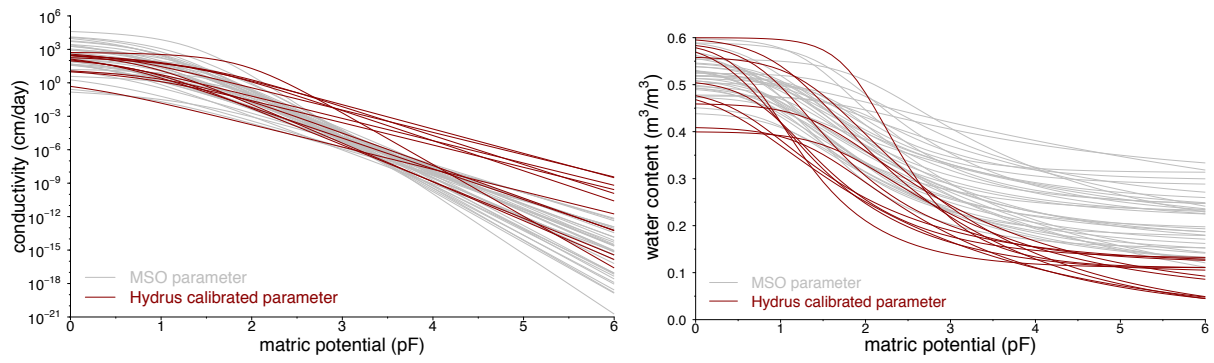
2 **Figure 7.** Time series of rainfall (a), simulated and observed volumetric water content for soil  
 3 moisture plot SM-1 (b-e), Hydrus-1D simulated percolation (f), water temperature (g) and  
 4 piezometric water levels (h) in a nearby groundwater well. The grey shaded area represents the  
 5 95% confidence interval of soil moisture based on model parameter sets obtained using SCEM  
 6 after fulfilment of the convergence criterion.



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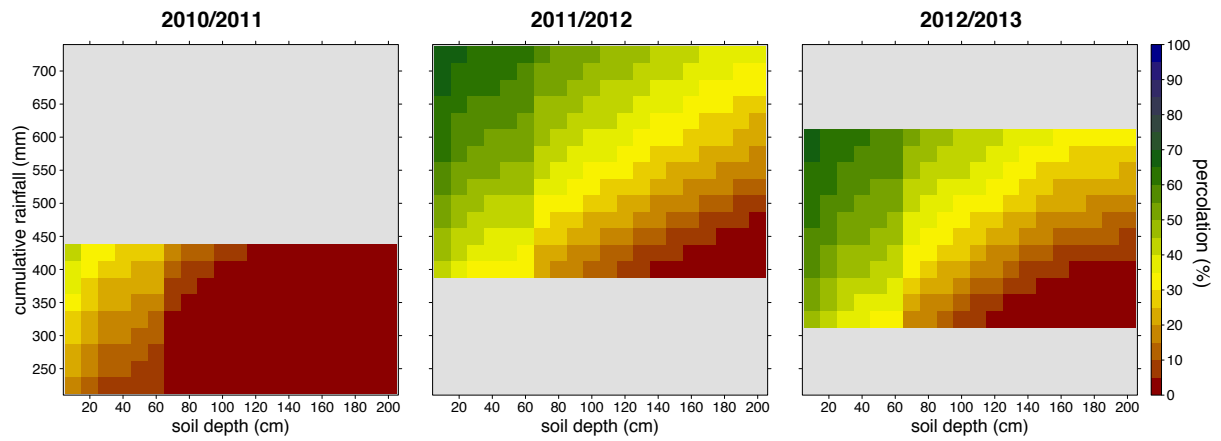
**Figure 8.** Simulated daily water fluxes at the single soil moisture plots for the simulation period 2010-2013.

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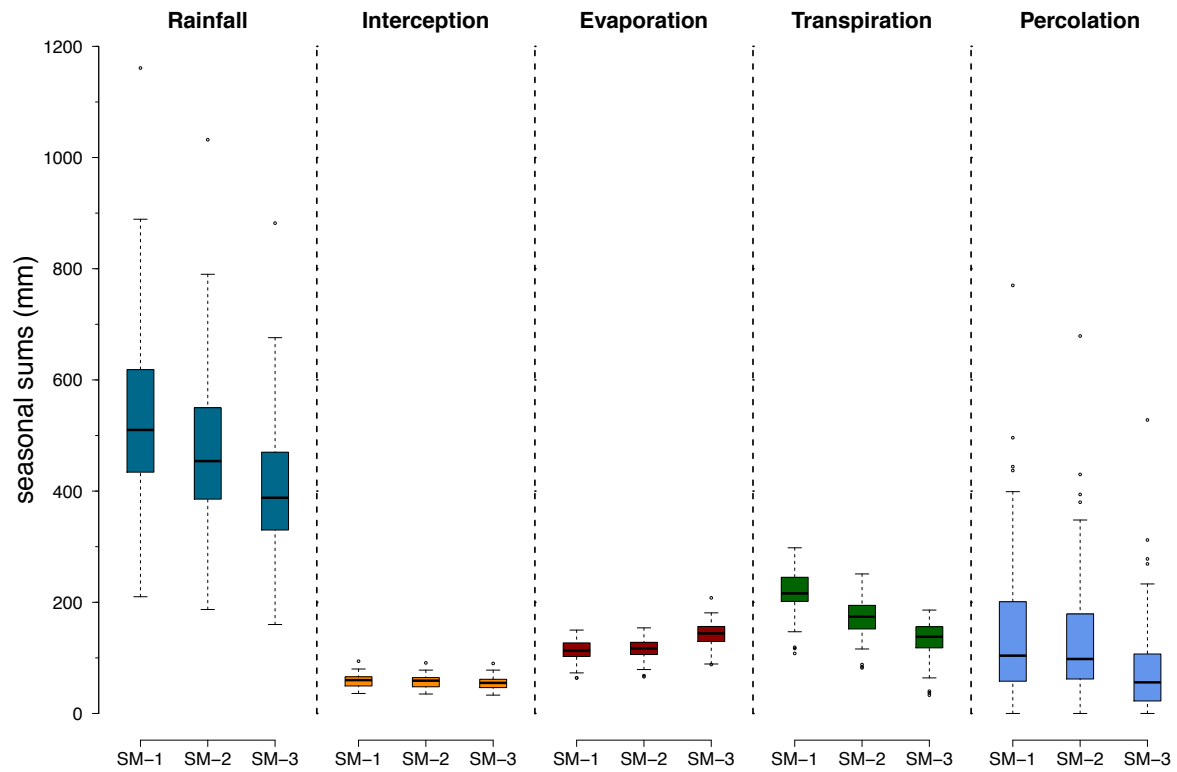
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3 **Figure 9.** Comparison of the water retention and conductivity functions of the Mualem/Van  
4 Genuchten parameter sets derived from MSO experiments with those inversely calibrated with  
5 Hydrus-1D and SCEM using observed soil moisture time series.



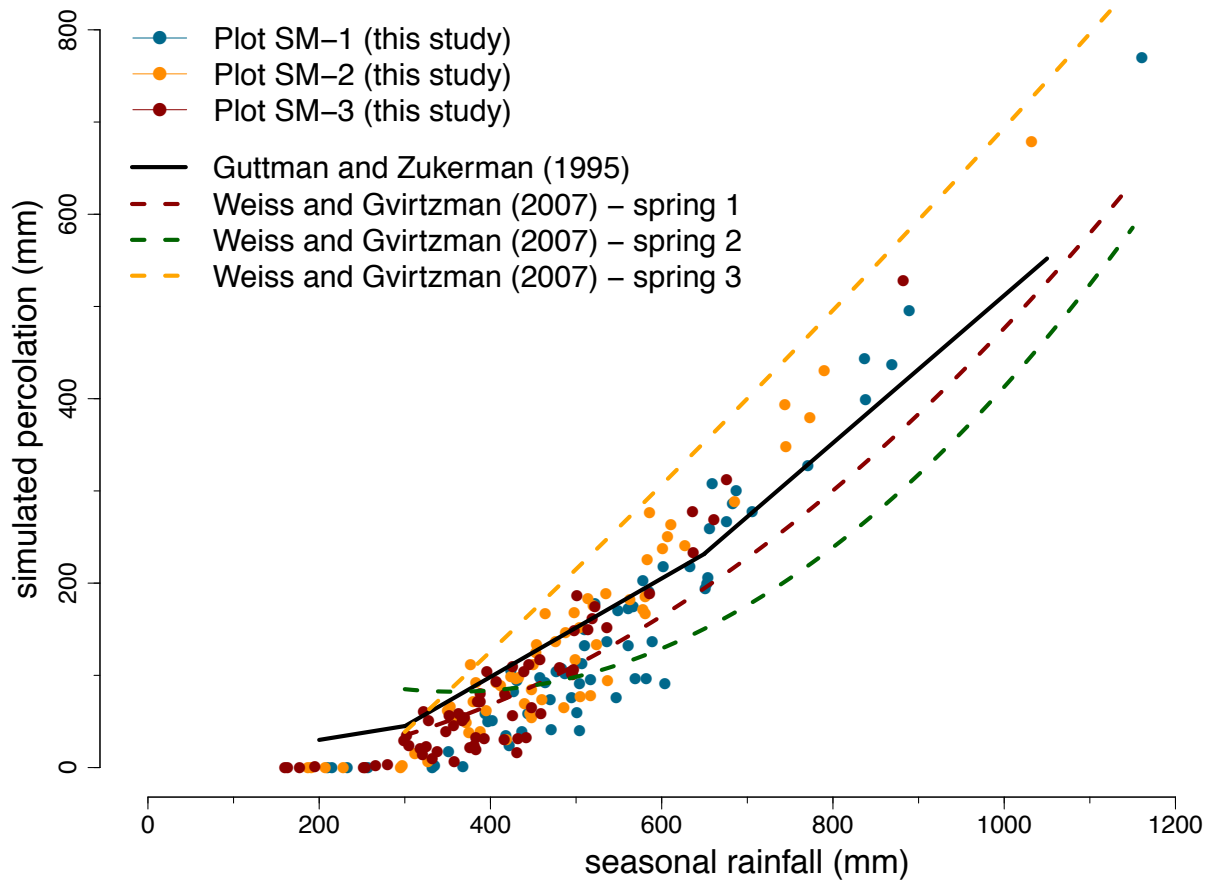
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2 **Figure 10.** Simulated percolation versus soil depth and rainfall amounts along the climatic  
 3 gradient for three consecutive winter seasons with different rainfall depths and distribution  
 4 patterns. Simulations were based on calibrated soil hydraulic properties of plots SM-1. The grey  
 5 shaded areas display rainfall depths, which have not been reached in the study area within  
 6 altitudes of 400 to 1000 m a.s.l. according to calculated rainfall gradients. The points represent the  
 7 plot scale simulated percolation fluxes using optimal parameter sets for the single plots SM-1,  
 8 SM-2 and SM-3.



1

2 **Figure 11.** Seasonal sums of simulated water balance components for the period 1951 to 2013  
 3 using the calibrated soil hydraulic parameters of the various plots. Rainfall and temperature data  
 4 were obtained from the nearby Jerusalem central station (<http://www.data.gov.il/ims>) and  
 5 corrected for the single locations by applying a simple elevation gradient-based correction factor.



1

2 **Figure 9.** Simulated seasonal percolation at the plot scale (SM-1, SM-2, SM-3) for the period  
 3 1951-2013 in comparison to rainfall-recharge relationships for the carbonate aquifer (Guttman  
 4 and Zukerman, 1995) and three small karst springs emerging from local perched aquifers (Weiss  
 5 and Gvirtzman, 2007).