

1 **Spatial and temporal variability of groundwater recharge in a sandstone**
2 **aquifer in a semi-arid region**

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9

10 **Abstract**

11 With the aim to understand the spatial and temporal variability of groundwater recharge, a high-
12 resolution, spatially-distributed numerical model (MIKE SHE) representing surface water and
13 groundwater was used to simulate responses to precipitation in a 2.16 km² upland catchment on
14 fractured sandstone near Los Angeles, California. Exceptionally high temporal and spatial resolution
15 was used for this catchment modeling: hourly climate data, a 20x20 meter grid in the horizontal plane
16 and 240 numerical layers distributed vertically within the thick vadose zone and in the upper part of
17 the groundwater zone. The finest-practical spatial and temporal resolution were selected to
18 accommodate the large degree of surface and subsurface variability of catchment features. Physical
19 property values for the different lithologies were assigned based on previous on-site investigations
20 whereas the parameters controlling streamflow and evapotranspiration were derived from
21 calibration to continuous streamflow at the outfall and to average hydraulic heads from 17 wells.
22 Confidence in the calibrated model was enhanced by validation through, i) comparison of simulated
23 average recharge to estimates based on the applications of the chloride mass-balance method to data
24 from the groundwater and vadose zones within and beyond the catchment and, ii) comparison of the

25 water isotope signature (^{18}O and ^2H) in shallow groundwater to the variability of isotope signatures
26 for precipitation events over an annual cycle and, iii) comparison of simulated recharge time series
27 and observed fluctuation of water levels. The average simulated recharge across the catchment for
28 the period 1995-2014 is 16 mm y^{-1} (4% of the average annual precipitation), which is consistent with
29 previous estimates obtained by using the chloride mass balance method (4.2% of the average
30 precipitation). However, one of the most unexpected results was that local recharge was simulated
31 to vary from 0 to $> 1000 \text{ mm y}^{-1}$ due to episodic precipitation and overland runoff effects. This
32 recharge occurs episodically with the major flux events at the bottom of the evapotranspiration zone,
33 as simulated by MIKE SHE and confirmed by the isotope signatures, occurring only at the end of the
34 rainy season. This is the first study that combines MIKE SHE simulations with the analysis of water
35 isotopes in groundwater and rainfall to determine the timing of recharge in a sedimentary bedrock
36 aquifer in a semi-arid region. The study advances the understanding of recharge and unsaturated
37 flow processes and enhances our ability to predict the effects of surface and subsurface features on
38 recharge rates. This is crucial in highly heterogeneous contaminated sites because different
39 contaminant source areas have widely varying recharge and, hence, groundwater fluxes impacting
40 their mobility.

41 **Introduction**

42 Assessment of groundwater recharge is fundamental to create strategies for management of water
43 resources and to estimate volumetric groundwater flow through contaminated sites. Recharge rates
44 represent an indication of upper limit of the volume of precipitation that may be accessible for
45 sustainable use and can govern the volume of water available to transport contaminants. Its
46 importance is greater in semi-arid regions where dominance of evapotranspiration limits water
47 resources. In these regions, estimated recharge rates depend on the temporal and spatial resolution
48 of the investigation and the uncertainties associated with recharge values are usually large

49 (Scanlon, 2000; Xie et al., 2018; Crosbie et al., 2018). In favorable circumstances, geochemical-based
50 methods have proven to be especially useful for estimating recharge rates. In areas where the
51 geologic and anthropogenic sources of chloride in the subsurface are negligible, the distribution of
52 chloride in the vadose zone and groundwater has been used to calculate long-term site-wide
53 (Wood and Sanford, 1995; Gebru and Tesfahunegn, 2018; Jebreen et al., 2018) and location-specific
54 recharge values (Heilweil et al., 2006; Huang et al., 2018), to determine mechanisms of flow in the
55 vadose zone (Sukhija et al., 2003; Li et al., 2017), and to evaluate the effects of environmental
56 changes on recharge (Scanlon et al., 2007; Cartwright et al., 2007). Elevated tritium in precipitation
57 derived from atmospheric releases during nuclear tests in the 1960's and transported into the
58 subsurface has also been an invaluable tracer to determine modern recharge and mechanisms of
59 flow in both vadose and groundwater zones (Cook and Böhlke, 2000; De Vries and Simmers, 2002).
60 These geochemical and isotopic techniques are based on the interpretation of hydrologic process
61 influences on the distribution of tracers in the subsurface but cannot show the dynamic, short-term
62 temporal effects nor provide a continuous spatial representation of these processes at the
63 catchment scale.

64 Numerical hydrologic models that integrate surface water and groundwater flows have been
65 developed to simulate the spatial and temporal distribution of surface runoff, infiltration,
66 evapotranspiration and groundwater recharge. However, the application of nearly all such
67 simulation tools have been limited to humid regions (Wheater et al., 2007) with minimal
68 application to semiarid regions. Scanlon et al. (2006), in their review on recharge in semiarid areas,
69 reported only 7 papers providing a continuous spatial distribution of recharge, out of a total of 98
70 studies. However, these studies investigated large areas, from 1,039,647 km² (Flint and Flint, 2007)
71 to 60 km² (Flint et al., 2001), using a relatively coarse spatial resolution (from 72,900 m² - Flint and
72 Flint, 2007 to 900 m² - Flint et al., 2001). In the last decade, although modeling techniques have
73 advanced to include combined surface water-groundwater simulations, recharge in semiarid areas

74 has been represented with a GIS approach (Hernández-Marín et al., 2018) often using remote
75 sensing data (Wang et al., 2008; Coelho et al., 2017; Crosbie et al., 2015) or neglecting the surface
76 water component and focusing on unsaturated zone (Levy et al., 2017; Turkeltaub et al., 2015).

77 Among the commercially available models, the physically based MIKE SHE represents the land-
78 based hydrologic system, with an integration of the surface flows (i.e. precipitation, infiltration,
79 evapotranspiration and runoff) and subsurface flows (i.e., percolation into the vadose zone and
80 recharge across the water table) (Ma et al., 2016). However, the literature shows only two
81 applications of MIKE SHE to assess recharge in semiarid areas. Liu et al. (2007) analyzed the
82 recharge response associated with overland flow in an alluvial watershed (surface area: 91 km² -
83 cell size: 2,500 m²) in the Tarim Basin, China. Smerdon et al. (2009) distinguished and quantified
84 the contributions of three sources to the total recharge for a valley bottom aquifer in the Okanagan
85 Basin (Canada) (surface area: 130 km² - cell size: 10,000 m²).

86 In this study encompassing a 20-year period (1995-2014), we used MIKE SHE to simulate the
87 recharge and the other hydrologic processes in a small catchment (2.16 km²) located on an exposed
88 bedrock upland plateau (from 650 to 490 m asl) in the Simi Hills, near Los Angeles, California (Fig.
89 1). The area is semi-arid with potential evapotranspiration (CIMIS, 1999) exceeding the average
90 annual precipitation (396 mm as the recorded average annual precipitation over the 1995 -2014
91 period). The bedrock consists of sandstone with interbeds of shale and siltstone, densely fractured
92 with bedding parallel partings and vertical joints and faults (Cilona et al., 2015; Cilona et al., 2016;
93 Link et al., 1984; MWH, 2016) (Fig. 2). The hydrogeology of the site has been investigated
94 intensively over the past 20 years because of the chemical contamination (mainly Trichloroethene -
95 TCE) in groundwater (Pierce et al., 2018a; Pierce et al., 2018b; Sterling et al., 2005; MWH, 2009;
96 Cherry J.A., 2009) and construction and application of a 3-D flow model (FeFlow) has been an on-
97 going effort supporting characterization and corrective measures (AquaResource and MWH, 2007).

98 From the application at the site of the chloride mass balance (CMB), based on measurement of

99 chloride in atmospheric deposition, surface water and groundwater, Manna et al. (2016) estimated
100 a long-term average recharge of 19 mm y⁻¹, corresponding to the 4.2 % of the average precipitation
101 (455 mm for the period 1878-2014). More recently, Manna et al. (2017) analyzed porewater Cl
102 concentration profiles from the vadose and groundwater zones at 11 locations across the site. This
103 provided spatially variable, long-term recharge values ranging from 4 to 23 mm y⁻¹ and indicated
104 that, on average, 80% of the flow in the vadose zone occurs as intergranular flow in the rock matrix
105 and 20% as fracture flow. These chloride-based methods lump together hydrologic processes
106 providing long-term recharge estimates for only few locations across a large site. However, to
107 inform the 3-D groundwater flow model and to simulate plume fluxes, information about the spatial
108 and temporal distribution of recharge is needed.

109 In this study, we analyze the spatial and temporal variability of recharge in a catchment of the
110 contaminated site not only to constrain recharge values but also to uncover hydrologic processes
111 that cause the borehole-scale spatial variability observed in those previous studies (Manna et al.,
112 2016; Manna et al., 2017). The catchment was chosen because it is representative of the varied
113 surface and subsurface conditions found throughout the contaminated area and it is believed to be
114 minimally impacted during the calibration period by the surface water controls measures in place.
115 Given that the scope of the paper is to simulate the natural conditions, these initiatives are not
116 considered in our modeling. To better represent the large range of surface and subsurface features
117 and provide high-resolution representation of the spatial distribution of recharge, we used hourly
118 climate data, sub-hourly time step and a fine grid of 400 m² cells for a total of 5,420 cells. In
119 addition to the spatial variability, we also examined the seasonal dynamics of the hydrologic
120 processes by tracking vadose zone water budgets for representative cells of the model. This
121 analysis helped in understanding the transient conditions that determine the rates of the
122 hydrologic processes throughout the year. The model was calibrated using measurements of runoff
123 from instrumented outfall flows and quarterly observations of groundwater levels in 17 wells

124 distributed across the catchment for the simulated period. Unlike the previous applications of MIKE
125 SHE in the literature, the simulation results were also validated through comparison with transient
126 water levels from shallow wells, comparison with previous independent recharge estimates based
127 on application of the Chloride Mass Balance (Manna et al., 2016; Manna et al., 2017) and through
128 the analysis of water isotopes from rainfall and groundwater that indicated the timing of recharge.
129 Finally, we proposed a conceptual model for various recharge conditions in the fractured sandstone
130 aquifer based on the results of the MIKE SHE simulation along with findings of previous recharge
131 studies for the site (Manna et al., 2016; Manna et al., 2017). In particular, the MIKE SHE simulations
132 contributed to the conceptual model concerning the role of surface feature variability (e.g.
133 topography and vegetation) on the hydrological processes whereas the Cl-based studies informed
134 the flow mechanisms in the underlying portion of the system.

135

136 **The site MIKE SHE model**

137 The MIKE SHE model (Refsgaard, 1995) simulations were conducted at a sub-hourly time step
138 using hourly meteorological data measured from 1995 through 2014 on site and from stations
139 proximal to the study area. A portion of the rainfall is intercepted by the vegetation canopy, from
140 which evaporation occurs. The remaining water reaches the surface, where it may infiltrate,
141 evaporate or runoff downslope if depression storage is satisfied. Water infiltrating into the
142 subsurface may be evapotranspired back to the atmosphere or percolate down to the water table to
143 become groundwater recharge. Actual evaporation and transpiration were simulated based on the
144 Kristensen and Jensen Evapotranspiration Model (Kristensen and Jensen, 1975), which considers
145 potential evapotranspiration estimated using the FAO 56 Penman-Monteith method (Allen et al.,
146 1998), available soil moisture and the crop characteristics (depth of the evapotranspiration zone,
147 leaf area index and crop coefficient) in each grid cell (Table 1). When the rainfall exceeds the

148 infiltration capacity, water is ponded on the ground surface and is available for runoff. The
149 infiltration capacity in the model is dynamic and a function of the unsaturated hydraulic
150 conductivity (K_u) and the water content properties (i.e., saturation point, field capacity and
151 permanent wilting point) of the surficial media. To describe the relation between water content,
152 conductivity and matric potential, the Van Genuchten model is used (Van Genuchten, 1980). The
153 rate of runoff is simulated using a 2D diffusive wave approximation and is controlled by the
154 topographic slope, the surface roughness and detention storage. The latter is the volume of water
155 stored in surface depressions before runoff starts. The unsaturated zone flow is simulated as the
156 change in soil moisture, resulting from cyclical input (infiltration) and output (recharge and
157 evapotranspiration). It is modelled as a 1D column using the full Richards equations (Richards,
158 1931) with finite difference cells that have variable discretization from the top of the column
159 (ground surface) to the base of the column (the unsaturated/saturated zone interface). Given the
160 variable thickness of the vadose zone and the low water fluxes, the model was run several times to
161 set consistent initial conditions. Our analysis began when the simulation showed that the degree of
162 change in average recharge value from one run to the next was about 0.3% indicating near steady-
163 state conditions. Recharge was calculated anytime that infiltration water arrives at the water table.
164 Most precipitation events do not result in recharge because infiltration into the shallow subsurface
165 is intercepted and evapotranspired. The flow in the groundwater zone was represented using 3D
166 finite difference Darcy equation. A fixed head boundary applied along the lateral sides and the
167 bottom of the model domain (490 m asl) was used to simulate the flow to and from the deeper
168 groundwater system, not explicitly represented in the integrated model but which extends several
169 hundred meters (Fig. 3). These fixed heads are based on observed groundwater levels at the site
170 and simulations based on a detailed 3-D groundwater flow model system that includes the
171 catchment and a much larger domain beyond (AquaResource and MWH, 2007). The groundwater
172 contribution to streamflow is minimal and intermittent ($\sim 0.1 \text{ mm y}^{-1}$ for the period of 1995-2014)

173 and only occurs at the farthest downstream location of the catchment where the groundwater table
174 rises close to the ground surface.

175 *Climate data*

176 Hourly rainfall data were collected from two stations within the catchment boundaries: the Sage
177 Ranch station, managed by Ventura County watershed
178 (<http://www.vcwatershed.net/hydrodata/php/getstation.php?siteid=272#top>) and the Simi Hills-
179 Rocketdyne Lab, managed by Boeing Inc. The annual precipitation ranges from 99 mm (2014) to
180 976 (1998), with an average value of 396 mm y⁻¹. The seasonal precipitation regime is
181 Mediterranean, with 77% of the total precipitation occurring from December to March.

182 Daily maximum and minimum air temperature observations were obtained from two climate
183 stations of the NOAA network: from 1995 to 1998 data were gathered from the Cheeseboro station
184 (<https://www.ncdc.noaa.gov/cdo-web/datasets/GHCND/stations/GHCND:USR0000CCHB/detail>)
185 and from 1998 to 2015 from the Van Nuys station ([https://www.ncdc.noaa.gov/cdo-](https://www.ncdc.noaa.gov/cdo-web/datasets/GSOM/stations/GHCND:USW00023130/detail)
186 [web/datasets/GSOM/stations/GHCND:USW00023130/detail](https://www.ncdc.noaa.gov/cdo-web/datasets/GSOM/stations/GHCND:USW00023130/detail)), respectively 6 km SW and 18 km E
187 of the study site. Temperatures were adjusted using a dry (10 °C km⁻¹) and wet (5.5 °C km⁻¹)
188 adiabatic lapse rate based on the elevation change between the SSFL site and the collecting station.
189 July, August and September are the warmest months with an average daily maximum temperature
190 of 30.5, 31 and 30.4 °C, respectively whereas February and December are the coldest with an
191 average daily maximum temperature of 17 and 17.4 °C, respectively. Annual average temperature is
192 16.7°C.

193 *Surface and subsurface parameters*

194 The MIKE SHE model was developed employing a 20 by 20 m finite-difference horizontal-plane grid
195 to represent the surface physical features, a fine vertical discretization of the vadose zone with 240
196 numerical layers ranging from 0.1 to 1 m thickness and 2 groundwater zone layers, with thickness

197 variable from 5 to 185 m, to represent vertical variability at, and just below, the position of the
198 water table (Fig. 3). This resolution was selected as a compromise between representation of
199 spatial variability at a more detailed scale and reasonable computational time. Maps of topography,
200 vegetation, surficial geology and land use were used to assign surface parameters (Fig. 1, Fig. 2 and
201 Fig. 4). High resolution topographic data (2 feet interval elevation contours) were obtained based
202 on an aerial survey of the site in 2010. These topography data were used to define the ground
203 surface elevations (Fig. 1).

204 The surface and subsurface hydrogeologic units include alluvium, fractured weathered and
205 unweathered bedrock comprised of sandstone, siltstone and shale beds of varying thickness, grain
206 size and cementation (Fig. 2 and Fig. 3). The physical properties of these units, derived from
207 previous on-site investigations (Allegre et al., 2016; Quinn et al., 2015; Quinn et al., 2016) and
208 adjusted by calibration, are summarized in Table 2. In particular, our model uses three separate
209 sets of Van Genuchten parameters to represent the pressure saturation-hydraulic conductivity
210 relationships. The parameters used reflect our understanding that the rock matrix transmits the
211 largest volume of recharge (80%), while recharge through the fractures is minimal (20%) (Manna
212 et al., 2017). Therefore, the relationships used are biased towards the matrix response. These
213 values were further calibrated using the groundwater level responses and the streamflow. Further
214 rock core samples indicate a high moisture content (~80%) (Cherry et al., 2009) indicating that K_u
215 is often close to K_s and the hydraulic conductivity-saturation curve reflects this understanding.

216 Four land use classes were identified and delineated based on aerial imagery and local land cover
217 datasets (Davis et al., 1998): developed areas (roads, building, parking lots); chaparral (chamise,
218 scrub oak), coastal scrub (Black sage) and exposed bedrock (areas without vegetation) (Fig. 4). The
219 first category represents only 5% of the study catchment whereas the two vegetation classes
220 (chaparral and coastal sage scrub) cover 83% of the area. The remaining 12% is represented by
221 bedrock outcrop. This latter category was subdivided into two classes: non-massive bedrock and

222 massive bedrock based on physical appearance. Massive bedrock areas were identified based on
223 rock masses that have resisted erosion over the decades and are presumed to be poorly-fractured
224 and/or well cemented such that local infiltration through these rock units is very low. These cell
225 assignments were identified using topography and imagery analysis. First, we used the minimum
226 downslope elevation change approach to identify topographic ridges; this algorithm calculates the
227 minimum elevation drop to a downslope neighbor. In a second stage, we isolate from the land use
228 map the exposed bedrock areas. Vegetation, indeed, generally does not grow on well cemented
229 rock. Finally, massive bedrock areas were assigned cells with downslope elevation change greater
230 than 1.25 meters in areas without vegetation.

231 Values of Leaf Area Index, depth of the root zone, surface roughness and Manning's number were
232 assigned to each land use class-specific parameters, based on the calibration process, with final
233 values similar to those available in the literature (Canadell et al., 1996; Scurlock et al., 2001; Chin et
234 al., 2000) (Table 1). To calculate the actual evapotranspiration, a crop coefficient varying monthly
235 between 0.53 and 1.02 has been used. This estimate are based on i) reference crop
236 evapotranspiration rates (RET) for Zone 9 of the Reference Evapotranspiration Zones map of the
237 California Irrigation Management Information System, that corresponds with the area the site is
238 within (ITRC, 2003), ii) a 'Pasture and Misc. grasses' land class chosen as representative of the site
239 and iii) a reduction of 8% to account for bare spots in vegetation and reduced vigor (ITRC, 2003).

240 *Unsaturated zone water budgets*

241 To assess the temporal variability of infiltration, evapotranspiration, change in storage and
242 recharge, we extracted the simulated unsaturated zone water budgets for two locations
243 representing the span of variability of the catchment. The two locations were selected based on
244 surface geology (Fig. 2) and land use category (Fig. 4): UZ1 represents an area of outcropping
245 bedrock without vegetation and UZ2 represents a cell with alluvium and vegetation cover. The

246 average infiltration value over the simulated period at the two locations (UZ-1: 87 mm y⁻¹; UZ-2:
247 395 mm y⁻¹) matches the average infiltration value for all the cells of the catchments with same land
248 use and surface geology. For these cells, we extracted the weekly time series of infiltration,
249 evapotranspiration, storage variations and flux at the bottom of the ET zone (i.e., drainage). The
250 latter indicates the volume of water that infiltrates into the vadose zone and will eventually become
251 recharge upon reaching the water table. The analysis of the seasonal variability of these fluxes
252 provided insights about their transient nature and about the effect of the surface variability on the
253 hydrologic processes in the unsaturated zone.

254 *Approach for model calibration*

255 In the model calibration procedure, the simulation results were compared to observed processes
256 and, to obtain acceptable matches, 10 parameters were available to adjust: surface roughness,
257 detention storage, imperviousness, rooting depth, leaf area index (LAI), crop coefficient,
258 unsaturated hydraulic conductivity and water content parameters of alluvium and weathered
259 bedrock. These were tested against an objective function of streamflow and groundwater level
260 measurements. An objective function is a measure of overall model fit of simulated to observed
261 values of groundwater levels and streamflow.

262 For the streamflow calibration, we compared the surface runoff generated by MIKE SHE to the data
263 collected at the catchment outfall between 2009 to 2011. This time interval had minimal
264 occurrence of substantial anthropogenic activities and was representative of natural hydrologic
265 conditions, as reported also by Manna et al. (2016).

266 For the calibration to groundwater levels, quarterly manually measured water level data were used.
267 Excluded from the calibration data were: i) wells with screened interval below the bottom of the
268 model domain (490 m a.s.l.), and ii) wells where the water table is strongly influenced by
269 subsurface complexity not represented in the saturated zone portion of the MIKE SHE model. After

270 these exclusions, water level data from 17 wells being used with water depths ranging from 25 to
271 137 meters bgs (Fig. 1, 2 and 4). The number of measurements in the time series at each well
272 varies from 1 (RD-130) to 139 (WS-09B) measurements. In the calibration procedure, average
273 values were used for comparison with average simulated values.

274 The calibration process proceeded in an iterative manner. After each calibration run, the two
275 calibration targets were examined with a variety of metrics. For the streamflow, we analyzed mean
276 error for simulated and observed average annual flow; mean error, root mean squared error,
277 correlation and Nash Sutcliffe Efficiency for the simulated and observed average monthly and daily
278 flows. An additional qualitative measure of the correlation between precipitation and streamflow
279 event was provided by the analysis of the graphical of plots of observed and simulated daily
280 streamflow hydrographs.

281 For the groundwater levels, the metrics were mean error, mean absolute error, root mean squared
282 error and normalized root mean squared error for the simulated and observed average water
283 levels. In addition, residual plots of simulated and observed water levels provided a quantitative
284 and qualitative assessment of the residual error present at the observation well throughout the
285 domain. Spatial patterns of groundwater level residual were compared against other spatial data
286 (e.g. hydraulic conductivity, boundary conditions, land uses, surface geology) to evaluate potential
287 correlations and adjustments that could improve the calibration.

288 Following an assessment of these calibration targets, the ten model parameters were adjusted for
289 better calibration metrics. In instances where the results were not consistent with the site
290 conceptualization, consideration was given as to whether an alternative conceptualization would
291 explain the results predicted by the model. Testing of alternative conceptualizations through
292 manual simulations was chosen over the alternative method of optimization of a single
293 conceptualization using software such as PEST (Doherty, 2004) given the uncertainty in how to

294 parameterize models in these semi-arid environments. Given the structural changes
295 (representation of the unsaturated flow, representation of impervious areas) that were made to the
296 model during the several simulations, it was not possible to carry out an exhaustive optimization or
297 sensitivity analysis. However, through the calibration process we gained semi-quantitative
298 information about the model sensitivity to each parameter which is presented in the results section.

299

300 *Approach for model validation*

301 To obtain confidence about the reasonableness of the results, simulation results from the calibrated
302 model were tested by a validation procedure, which included comparison to previous independent
303 recharge estimates based on chloride, and timing of recharge from isotopic data ($^{18}\text{O} - ^2\text{H}$) and from
304 analysis of observed fluctuations of water level hydrographs, not used in the calibration. The
305 premise of the validation is that the calibrated model must provide results consistent with the
306 validation information, that are entirely independent of the parameter assignments made in the
307 calibration.

308 Manna et al. (2016) estimated an average long-term recharge of 19 mm y^{-1} for the same catchment
309 using the chloride mass balance (CMB) method, based on the average Cl concentration measured in
310 the atmospheric deposition, comprised of rainfall and dry fallout (2.6 mg L^{-1}), surface water at the
311 catchment outfall (4 mg L^{-1}) and groundwater (52.5 mg L^{-1}). Since chloride concentration in
312 groundwater is proportional to the concentrating effect of water loss due to evapotranspiration, it
313 can be used as a proxy to determine the range of variability in recharge. Chloride concentration in
314 shallow groundwater monitoring wells ranges across the area from 17 to 162 mg/L corresponding
315 to recharge values of 43 and 5 mm y^{-1} , respectively. Manna et al (2017) also provided insights
316 regarding spatial variability of recharge within the catchment based on analysis of Cl profiles in
317 porewater from the vadose zone and groundwater which indicate a range of recharge from 4 to 21

318 mm y⁻¹ corresponding to <1 – 4.7% of the average annual precipitation for 4 locations located
319 within the catchment area. Although the recharge values obtained from the CMB method integrate
320 hydrologic processes occurring over longer time, from decades to millennia, they represent a
321 reasonable assessment of long-term, site-wide and location-specific average values and are
322 valuable for validation purposes.

323 Samples of rainfall and groundwater were analyzed for water isotopes (¹⁸O – ²H) . These water
324 isotopes are commonly used to assess evaporative processes and to determine sources and origins
325 of different groundwaters. Typically, the water isotope values vary seasonally over the annual cycle,
326 so that the groundwater composition reflects the season with most of the recharge. In this study,
327 we compared the isotopic signature of groundwater to that of precipitation for an entire
328 hydrological year to determine whether the timing of recharge indicated by the model is consistent
329 with the isotopic signature for the same period of the year. The available isotope data for rainfall
330 were determined for the period October 1994 to June 1995 collected at two rain gauge stations
331 (B/886 and RMDf), 5 km from the studied watershed and analyzed in the same year by an
332 automated gas-source mass spectrometer at the University of California Berkeley. The groundwater
333 samples were collected from monitoring wells in the studied catchment in two rounds of sampling:
334 the first in 2003-2004 and the second in 2013 (Fig. 1).

335 Furthermore, to test the ability of the model to simulate unsaturated zone flow processes and to
336 reproduce the transient recharge conditions, we compared the simulated time series of recharge,
337 obtained from MIKE SHE, with quarterly water level measurements at five locations not used in the
338 calibration process. The depth to groundwater at these wells ranges between 2 and 60 m with
339 seasonal fluctuations due to the recharge events. The recharge time series is obtained, extracting
340 the average, catchment-wide, monthly recharge values.

341

342 **Simulation results**

343 *Model calibration and sensitivity*

344 Streamflow measured at the outfall occurs in response to rainfall; however, some precipitation
345 events are followed by very low or no measurable flow (Fig. 5). This is evident for precipitation
346 events from April to June 2009, October and November 2010 and May and June 2011. In all these
347 cases, the surface runoff, generated by the precipitation events, infiltrates into the subsurface
348 without reaching the surface outfall (Fig. 5). These hydrologic dynamics are well simulated by MIKE
349 SHE. The comparison between the observed and the simulated hydrographs shows a good
350 correlation for the calibration period ($R^2=0.97$; average difference 4.7%). The average simulated
351 flow is 48 mm y^{-1} , about 14.5% of the average precipitation for the 2009-2011 period (331 mm)
352 and is almost coincident with the measured flow (46.2 mm y^{-1}) (Fig. 5). This value reflects the
353 precipitation conditions of the 2009-2011 period and is lower than the average runoff over the
354 entire simulated interval (110 mm y^{-1} , 28% of the annual precipitation). Monthly and daily Nash
355 Sutcliffe Efficiency (NSE) values of 0.94 and 0.87 were achieved respectively, indicating good fit to
356 observed flows (NSE=1 corresponds to a perfect match).

357 In addition to the surface water leaving the catchment, the model was also calibrated to the
358 observed average groundwater head data (Fig. 6). A good match was obtained for the 17 locations,
359 with almost all values falling within the 10 m confidence interval bands, with a correlation
360 coefficient of 0.96 and a mean absolute error of 4.5 m (Fig. 6). This good correlation provides
361 confidence about the spatial distribution of model parameters.

362 Of the 10 adjusted parameters, unsaturated hydraulic conductivity and water content parameters
363 of alluvium and weathered bedrock had the strongest effect on the calibration and are, therefore,
364 well constrained by the measured streamflow and groundwater levels. These geologic features
365 represent the upper layers of the model domain and variations in their physical and hydraulic

366 properties control the rate of infiltration, evapotranspiration, drainage and, therefore, recharge. A
367 third parameter important in the calibration was the detention storage. This is because a
368 substantial amount of water from precipitation, especially at the beginning of the rainy season,
369 infiltrates without generating runoff events at the outfall (Fig. 5). This volume of water is controlled
370 not only by the properties of unsaturated zone (Table 2) but also by the value of detention storage
371 assigned to each land use class (Table 1). Conversely, alterations in rooting depth, LAI and crop
372 coefficient only resulted in small changes in streamflow. This is because significant runoff events
373 tend to occur during brief high-intensity precipitation events with a magnitude that far exceeds the
374 relative amount of evapotranspiration, which might occur during these events. For the same reason,
375 though, these factors had a relatively greater effect on the volume of water available for drainage
376 and subsequent recharge.

377

378 *Spatial variability*

379 To study the spatial variability of the water budget components, average annual maps of infiltration
380 (Fig. 7a), evapotranspiration (Fig. 7b) and recharge (Fig. 7c) for the period 1995 – 2014 were
381 created.

382 Average infiltration for the catchment is 254 mm y⁻¹, corresponding to 64% of the total
383 precipitation but single cell values span over three orders of magnitude from 9 to > 1000 mm y⁻¹
384 (Fig. 7a). Low infiltration values are found in developed/paved (average 51 mm y⁻¹) and massive-
385 bedrock (average 14 mm y⁻¹) cells. Due to the low infiltration capacity, more runoff is generated in
386 these cells and, thus, infiltration is higher in nearby cells that receive the surface water. Where
387 these neighboring cells are covered by alluvium at the surface, infiltration is even higher. On
388 average, cells with alluvium at the surface have an infiltration value of 332 mm y⁻¹, 25% more than
389 those where bedrock outcrops. Higher infiltration is also displayed in depressed areas such as

390 those along the main drainages and where closed topographic depressions occur. These cells collect
391 most of the surface runoff creating conditions for focused infiltration and recharge.

392 Only a small portion of water that enters the subsurface reaches the water table because the
393 majority is lost due to evapotranspiration (Fig. 7b). The average evapotranspiration estimated
394 using MIKE SHE is 265 mm y^{-1} , a value slightly higher than the average infiltration. This excess of
395 ET over infiltration is attributed to canopy interception and evaporation of temporarily ponded
396 surface water. When removing these two water-loss processes, the average evapotranspiration is
397 237 mm y^{-1} , which corresponds to 60% of the annual precipitation and to 94% of the total
398 infiltration. Transpiration is the main process of ET contributing to about 70% of the total ET. This
399 result is expected considering the considerable depth of the roots (up to 5 meters for Chaparral)
400 and the fact that vegetation covers 83% of the catchment area. Single cell values of ET span over
401 three orders of magnitude, from 50 to $>1000 \text{ mm y}^{-1}$. Since the actual evapotranspiration depends
402 strongly on the availability of subsurface water, the spatial variability mimics the infiltration
403 pattern and the two factors are strongly correlated ($R^2=0.84$). Therefore, low ET is associated with
404 developed (asphalt, buildings) and massive bedrock areas and high ET values are found along the
405 main surface drainages where infiltration is high and locally available for evapotranspiration. The
406 presence of alluvium at the surface increases the ET values on average by 25%; for example,
407 average ET in cells with chaparral and alluvium is 400 mm y^{-1} whereas where chaparral is rooted in
408 weathered bedrock is $\sim 300 \text{ mm y}^{-1}$.

409 A map of the spatial distribution of the average annual recharge is shown in Fig. 7c. The average
410 recharge value for the catchment is 16 mm y^{-1} equal to 4.1 % of the precipitation and 6.5 % of the
411 infiltration. The range of variability of recharge is over three orders of magnitude and spatially
412 variable depending on topography, surface geology and land use. It is noteworthy that 79% of the
413 catchment has recharge less than 10 mm y^{-1} and 90% less than 30 mm y^{-1} , which indicates that the
414 largest volumes of recharge are focused in small portions of the site. The recharge map (Fig. 7c)

415 shows the influence of the surface parameters on recharge estimates. Recharge is high along the
416 main drainage because of the contribution of surface water flowing from the surrounding slopes
417 and enhanced infiltration where the topographic slope decreases abruptly. Relatively higher
418 recharge values are also observed in areas with alluvium at the surface because the infiltration and
419 retention capacities are higher and, therefore, water can seep from the overburden into the bedrock
420 once the evapotranspiration demand and driving forces are met. Recharge is also higher in cells
421 without vegetation cover, compared to other cells with equivalent topographic slope and surficial
422 geology, because the evapotranspiration in these areas is lower.

423

424 *Temporal variability*

425 The seasonal variability of the hydrologic processes was examined analyzing unsaturated water
426 budgets at two locations with different land use and surficial geology (UZ-1 and UZ-2 in Fig.1)
427 Among the 20 years, we show the monthly average daily values from 2005 to 2007. This time span
428 features a wet year (2005 – 978 mm), a dry year (2007 – 149 mm) and one year with average
429 precipitation (2006 - 331 mm) and therefore is reasonably representative of the simulated period.

430 For areas with bedrock outcrop not covered by vegetation (UZ-1 in Fig. 1), the infiltration ranges
431 from 0 to 2.5 mm d⁻¹ (Fig. 8). The infiltration pattern shows null or minimal values during the
432 summer and positive events during the wet season. Water that enters the subsurface between April
433 and January replenishes the water content in the ET zone and becomes available for evaporation
434 but not for drainage. Evaporation is null during the summer because of the lack of precipitation
435 and because all the water stored in the first 20 cm of bedrock has been taken up by evaporation in
436 the previous months. Downward flux at the bottom of the ET zone (i.e. drainage) only happens
437 episodically when the water content in the ET zone is above the field capacity, at the end of the wet

438 season (i.e., March and April) or occasionally after exceptionally high-intensity precipitation events
439 (i.e., January 2005).

440 For areas with alluvium at surface (UZ-2 in Fig. 1) the infiltration has the same pattern but a
441 different order of magnitude (from 0 to 30 mm d⁻¹) due to the higher infiltration capacity of the
442 alluvium (Fig. 8). Here, the available water capacity of the ET zone is greater because of the
443 different physical properties (e.g. larger porosity) of the soil and the greater depth of the ET zone.
444 Therefore, almost all the infiltration water is taken up by the evapotranspiration. Unlike areas
445 without vegetation, evapotranspiration is not directly related to precipitation events and occurs
446 more continuously throughout the year. This is because alluvium stores a greater volume of water
447 in the ET zone that is nearly completely consumed by ET. A drainage flux is observed only during
448 high-intensity precipitation events that create near-saturation conditions such that water cannot be
449 held by tension in the shallow unsaturated zone and downward flow is initiated.

450 For both cases, drainage is not steady throughout the year but occurs episodically, controlled by
451 antecedent soil water content in the ET zone and by the intensity of precipitation. During drier-
452 than-average years, such as 2007, drainage occurs in areas without vegetation, whereas no
453 drainage is observed in cells with vegetation cover. After crossing the bottom of the ET zone, water
454 arrives at the water table with a time lag depending on the magnitude of the flux and on the
455 physical properties and the thickness of the vadose zone.

456 *Model validation*

457 The ability of the model to simulate transient hydrologic conditions was investigated through the
458 comparison between well hydrographs at five locations and the temporal variability of recharge
459 (Fig. 9). The spatially-average recharge rates obtained from MIKE SHE (monthly time-step) range
460 from 0.95 mm (November 2014) to 9.1 mm (March 2005). The latter is the response to the
461 extraordinary rainy season that occurred between December 2004 and March 2005 (903 mm)

462 whereas the first is due to dry conditions of the recent drought in California. The range of depth to
463 groundwater from 1995 to 2014 at the five locations considered is 2.8 – 14.4 m at RD-09, 17.8 – 30
464 m at RD-35A, 16.2 – 28.7 at RD-73, 37.7 – 50.8 m at RD-36B and 33.1 – 60.1 at WS-09B. The shape
465 of these hydrographs depends on surface (surface geology, topographic slope, land use) and
466 subsurface (mechanisms of flow in the vadose zone) conditions. For our validation purpose, it is
467 noteworthy that, at all the locations, the hydrographs show a good match with the recharge time
468 series such that the peaks in recharge coincide with water table rises. The greatest rises overlap the
469 two highest recharge periods (1998 and 2005), whereas a constant declining trend is observed
470 from 2011 to 2014 in response to drier conditions (Fig. 9). The good correlation suggests that, at
471 this scale, the equivalent porous media approach used is reasonable to simulate average responses
472 in groundwater because, although the bedrock has many interconnected fractures, it is only a minor
473 contributor to recharge.

474 The average recharge value for the catchment from the simulation is 16 mm y^{-1} and is consistent
475 with previous recharge estimates obtained for the site using the CMB method (19 mm y^{-1} – 4.2% of
476 the average precipitation, Manna et al., 2016; 16 mm y^{-1} – 3.5% of the average precipitation, Manna
477 et al., 2017). The frequency distribution of recharge values from the MIKE SHE simulation (92% of
478 the domain has average recharge lower than 40 mm y^{-1}) also corresponds well to the range of
479 variability based on chloride (from 0 to 43 mm y^{-1}) reported by Manna et al. (2016) and Manna et
480 al. (2017).

481 For additional information on recharge processes, we analyzed water isotopes obtained from
482 rainfall and groundwater samples (Fig. 10). The samples show a substantial isotopic range from
483 one precipitation event to another over the one-year collection period. ^{18}O varies between -2.8 and
484 -12.1‰ for B/886 and -2.8 and -11.7‰ for RDMF and ^2H varying between -11 and -89‰ for B/886
485 and -12 and -85‰ for RDMF (Table 3). This large range of values is probably due to the two

486 different trajectories of the precipitation events in southern California, one originating in the Pacific
487 and one over the Gulf of Mexico, as found by Friedman et al. (1992). The volume weighted mean
488 values for the two stations are -8.2 and -54.2‰ for B/886 and -8.2 and -56.2‰ for RDMF and are
489 consistent with global-scale maps of water isotopes for precipitation in southern California (Bowen
490 and Revenaugh, 2003).

491 Unlike rainfall, groundwater samples fall within a narrower range: from -6.5 to -7.5‰ for ^{18}O and
492 from -40.2 and to -52.2‰ for ^2H . All the samples are aligned along the local meteoric water line
493 (Fig. 10) indicating little if any evaporation from standing water on surface. This lack of
494 concentration effect on the isotopes is apparently in contrast to the chloride data. Manna et al.
495 (2016) found that Cl concentrations in groundwater are, on average, 20 times greater those from
496 atmospheric deposition because of the strong influence of evapotranspiration. The common
497 explanation for the lack of evaporation effects on the water isotopes in groundwater is that the
498 transpiration is the main evapotranspiration process (Clark, 2015; Cook and Böhlke, 2000).

499 Although transpiration through the vegetation causes a concentration effect on Cl, it does not cause
500 fractionation of the water isotopes and therefore the groundwater samples are not enriched (Clark,
501 2015; Cook and Böhlke, 2000).

502 The lack of evaporative water isotope signature associated with high groundwater Cl concentration
503 can also be explained by recharging water that crosses the ET zone mobilizing precipitated salts but
504 without any evaporation. This hypothesis supports the results of the MIKE SHE simulations, which
505 show that throughout the year there are only episodic fluxes at the bottom of the ET zone (Fig. 9). A
506 relevant observation that corroborates this hypothesis is that the isotopic composition of
507 groundwater is similar to that found in rainfall samples collected at the end of the wet season
508 (March and June) or, on occasion, with high-intensity precipitation events (January - 203 mm)
509 (Table 3). This similarity can be attributed to the preponderance of recharge occurring at these

510 times and thereby resulting in the groundwater values being different from the weighted mean
511 precipitation by 1.2‰ ¹⁸O and 3‰ ²H.

512 This proposed model of episodic flow through the unsaturated ET zone is also corroborated by the
513 evidence presented by Manna et al., (2017) that, on average, 20% of the flow in the vadose zone
514 occurs as fast flow through the interconnected fractured network.

515

516 **Discussion and conceptual model for recharge**

517 To summarize the findings of this study, and its relationship to the literature and to the previous
518 recharge studies at the site (Manna et al., 2016; Manna et al., 2017), we propose the following
519 process-based conceptual model for site recharge (Fig. 11).

520 The average recharge value is 16 mm y⁻¹ which is consistent with previous estimates at the site, and
521 with those obtained for other sandstone aquifers in semi-arid areas in the United States (4% -
522 Heilweil et al., 2006) and other studies in semi-arid regions around the world (0.2 – 35 mm y⁻¹ equal
523 to 0 – 5% of the average precipitation, Scanlon et al., 2006). Recharge varies greatly across the
524 catchment as a function of topography, surface geology, and land use. High recharge occurs where
525 most runoff water seeps into the subsurface, creating conditions for focused recharge. This
526 condition happens where closed depressions occur and where sloped topography abruptly
527 transitions to flat along the main surface drainages (Fig. 11a). In most areas, alluvium covers the
528 fractured porous bedrock, thus enhancing infiltration and temporary storage of infiltrated water.
529 Generally, in semiarid regions, high recharge values along a valley, at the edge of the slope referred
530 to as Mountain Front Recharge (MFR) (Wilson and Guan, 2004). However, our catchment is located
531 on the top of a ridge standing 300 m above the surrounding valleys (Manna et al., 2016) and, thus,
532 our case study represents groundwater recharge on the mountain block rather than MFR.
533 Nonetheless, it is interesting that the processes observed in our small catchment are similar to

534 those described for aquifer-scale recharge studies (Aishlin and McNamara, 2011; Carling et al.,
535 2012; Manning and Solomon, 2003; Bresciani et al., 2018) and defined as MFR.

536 Infiltration from April to December (dry season) contributes to replenish the water content in the
537 ET zone and remains available for evapotranspiration (Fig. 11b). Conversely, during the wet season,
538 infiltration crosses the bottom of the ET zone (i.e. drainage) and migrates deeper through the
539 vadose zone. This happens when the soil is above the field capacity (FC), which is more frequent at
540 the end of the wet season in March or April and/or during high-intensity precipitation events, (Fig.
541 11c). This recharging water quickly crosses the ET zone, as shown by the ET zone water budgets
542 extracted from MIKE SHE (Fig. 9), and by the lack of evaporative signature in isotope composition
543 (Fig. 10).

544 The occurrence of this fast/preferential flow out of the ET zone is also corroborated by the analysis
545 of vertical chloride porewater concentration profiles in the unsaturated zone (Manna et al., 2017).
546 The Cl concentration is high in the ET zone (up to 10,000 mg L⁻¹) and considerably lower in deeper
547 vadose and groundwater zones (average 49 mg L⁻¹). The higher Cl concentrations in the shallow
548 subsurface is the effect of strong evapotranspiration that takes up water but not chloride, whereas
549 the lower concentration below is due to fast/preferential flow of water that escapes the
550 concentrating effect of water loss in the shallower zone. Case studies showing similar results for
551 water that crosses the ET zone preferentially in time and space to become potentially recharge have
552 been also reported in literature (Kurtzman et al., 2016), also referred to as selective recharge (Gat
553 and Tzur, 1967; Florea, 2013; Krabbenhoft et al., 1990) . The occurrence of these fluxes has been
554 also analyzed in function of precipitation characteristics and antecedent water content with rainfall
555 intensity being the main factor (Allocca et al., 2015; Crosbie et al., 2012; Nasta et al., 2018; Taylor et
556 al., 2013).

557 Upon reaching the deeper vadose zone, water is redistributed between intergranular matrix flow
558 and fracture flow due to wettability and saturation concepts. The fractures and the matrix pores
559 drain the water from the ET zone. Active flow through the fractures is possible under conditions
560 such as ponding or intense precipitation, when a continuous slug of water lets i) the advective front
561 move ahead into the fracture (1 in Fig. 11c); ii) the matrix water flow into the fractures (2 in Fig.
562 11c). Otherwise, water is drawn from the fractures into the unsaturated matrix blocks (3 in Fig.
563 11c) and contributes to the slow vertical intergranular matrix flow (4 in Fig. 11c). According to
564 Manna et al. (2017), the first two mechanisms are much less frequent and contribute, on average, to
565 only 20% of the total recharge. It is most likely that conditions for flow in the fractures occur
566 episodically in areas of the site with high infiltration (topographic low and alluvium at the surface)
567 where temporary perched systems are observed.

568

569 **Conclusions**

570 For the upland bedrock catchment, the surface water-groundwater numerical model (MIKE SHE),
571 using a fine numerical grid (20 ×20 m) with calibration to streamflow and groundwater levels,
572 simulated the spatial and temporal variability of recharge across a 2.16 km² catchment in southern
573 California, USA. This is the first study that combined MIKE SHE simulations supported by analysis
574 of water isotopes and chloride mass balance to assess recharge in a sedimentary bedrock aquifer in
575 a semi-arid region. The calibrated simulations, indeed, were judged to be reliable and strongly
576 reflective of the natural system, based on the validation comparisons to mean recharge obtained
577 independently from the chloride mass balance method (Manna et al., 2016; Manna et al., 2017) and
578 to the timing of major recharge events indicated by water isotopes and water level fluctuations. The
579 simulations showed that major flux events at the bottom of the evapotranspiration zone, that result
580 in recharge tens of meters below the surface, occur episodically mostly at the end of the rainy

581 season and that recharge varies across the catchment between 0 and 1000 mm y⁻¹. The fine
582 numerical grid in the horizontal plane allowed meaningful examination of recharge spatial
583 variability. A substantially coarser grid would obscure influences of key surface features on the
584 hydrologic processes.

585 The results obtained from the catchment-scale simulations are being used to specify rules for
586 recharge to be assigned to the upper boundary condition of a 3-D numerical EPM groundwater flow
587 model (FeFlow), covering the studied catchment and a much larger area beyond (52 km²). The
588 modeled groundwater domain has many contaminant plumes and recharge is key to determine the
589 fluxes available to transport contaminants.

590 The aim of the MIKE SHE model is to represent the natural hydrologic conditions, after site
591 industrial operations ceased more than a decade ago. During historical operations from 1950's
592 through mid-2000's, use of imported and pumped groundwater likely caused increases to
593 infiltration and recharge locally in some areas. These conditions are beyond the scope of this paper
594 but worth further consideration in a follow-on study as it relates to when contaminant releases
595 occurred and may provide insights regarding how contaminant migration rates may have been
596 influenced. Future modeling efforts will also evaluate the effect on recharge of the surface water
597 control systems currently in place on the site. These storm water management measures aim to
598 limit the volume of water leaving the catchment and, therefore, will likely influence the natural
599 rates of the other hydrologic processes.

600

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607

608

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790 *Table 1 Land use class-specific parameters to model runoff and evapotranspiration.*

791

Land Use Class	Surface roughness (Manning's n)	Detention storage (mm)	Leaf Area Index	Depth of the evapotranspiration zone (m)
Developed*	0.04	1	-	0.2
Coastal Scrub	0.2	7.5	1.8 - 3	1.8 - 3
Chaparral	0.2	7.5	2.8 - 4.5	3.1 - 5
Exposed Bedrock/ Massive bedrock*	0.05	3	-	0.2

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795 *Table 2 Physical properties of the different hydrogeologic units.*

Hydrogeologic unit	Lithology	K_s (m s ⁻¹)	Saturation (θ_s)	Field capacity (θ_{fc})	Residual Water content (θ_r)	Van Genuchten parameters		
						α	n	l
Alluvium		1×10^{-6}	0.4	0.25	0.05	0.021	1.61	0.5
Weathered bedrock		2×10^{-7}	0.2	0.11	0.01	0.033	1.49	0.5
Unweathered bedrock	Shale/Siltstone	4.1×10^{-10} to 2.3×10^{-7}	0.13	0.1	0.025	0.01	1.23	0.5
Unweathered bedrock	Sandstone	1×10^{-10} to 1×10^{-5}	0.13	0.09	0.01	0.01	2	0.5

Unweathered bedrock	Fault zone	1×10^{-9} to 1×10^{-6}	0.13	0.1	0.025	0.01	2	0.5
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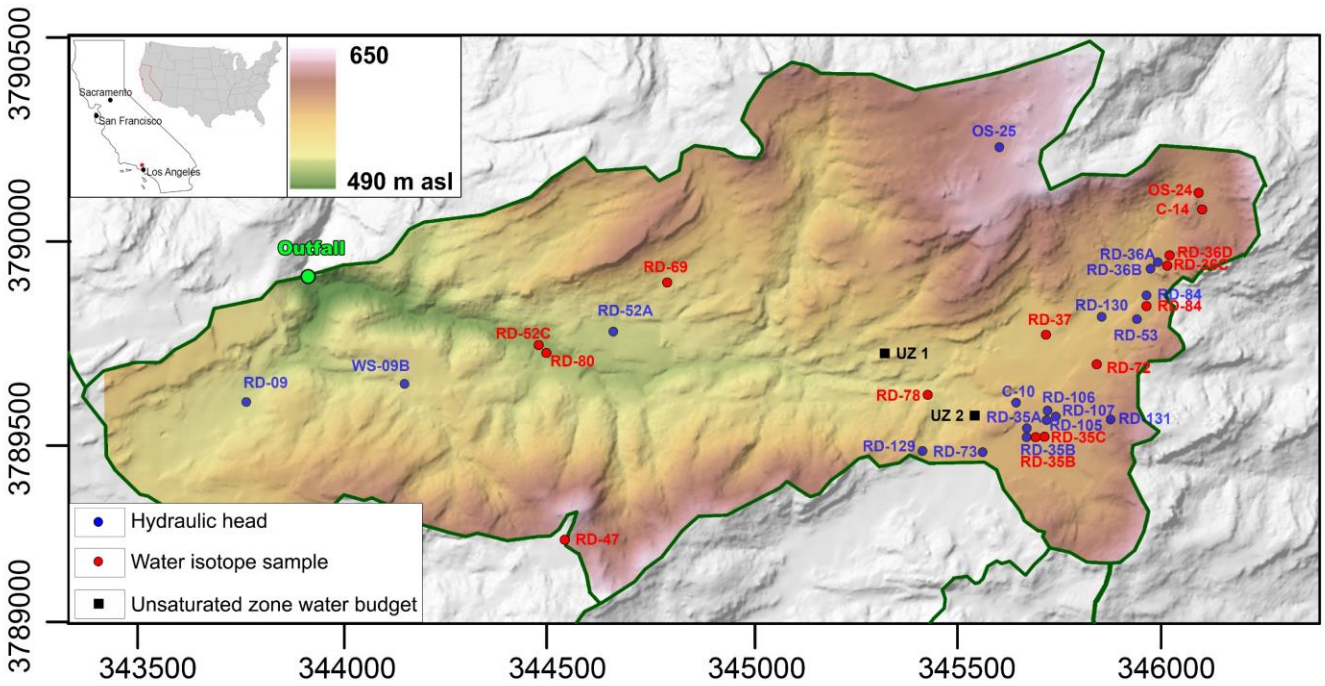
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799 *Table 3 Stable isotope composition of rainfall.*

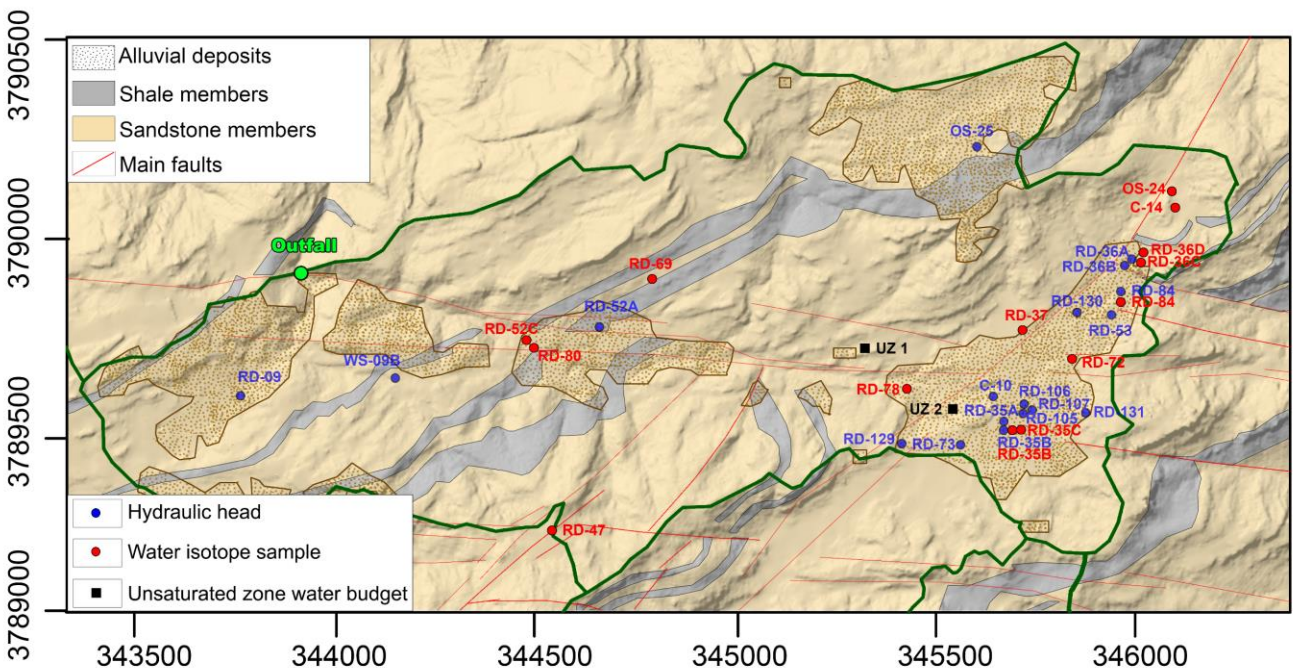
Date	B/886 Rain Gauge			RMDF Rain Gauge			Average		
	$\delta^{18}\text{O}$	$\delta^2\text{H}$	Rainfall (mm)	$\delta^{18}\text{O}$	$\delta^2\text{H}$	Rainfall (mm)	$\delta^{18}\text{O}$	$\delta^2\text{H}$	Rainfall (mm)
4/10/1994	-4	-19	3				-4.0	-19.0	3
25/11/1994	-5.2	-18	6	-5.1	-16	6	-5.2	-17.0	6
13/12/1994	-5.4	-23	9	-5.4	-25	9	-5.4	-24.0	9
24/12/1994	-10.3	-77	18	-10.1	-69	18	-10.2	-73.0	18
4/1/1995	-10.3	-75	94	-9.9	-69	121	-10.1	-72.0	108
11/1/1995	-6	-33	205	-7.4	-45	202	-6.7	-39.0	203
13/01/1995	-4.4	-19	20	-4.2	-20	18	-4.3	-19.5	19
16/01/1995	-2.8	-11	12	-2.8	-12	10	-2.8	-11.5	11
26/01/1995	-12.1	-89	152	-11.7	-85	150	-11.9	-87.2	151
7/3/1995	-6.8	-43	119	-6.4	-40	109	-6.6	-41.5	114
13/3/1995	-7.5	-44	NA	-7.8	-45	NA	-7.7	-44.5	NA
24/3/1995	-5.8	-22	NA	-5.5	-19	NA	-5.7	-20.5	NA
18/5/1995				-6.4	-42	34	-6.4	-42.0	34
22/6/1995	-8.6	-62	14	-8.6	-57	14	-8.6	-59.5	14
Volume weighted mean and total rainfall	-8.2	-54.2	650	-8.2	-56.2	691	-8.3	-55.2	689

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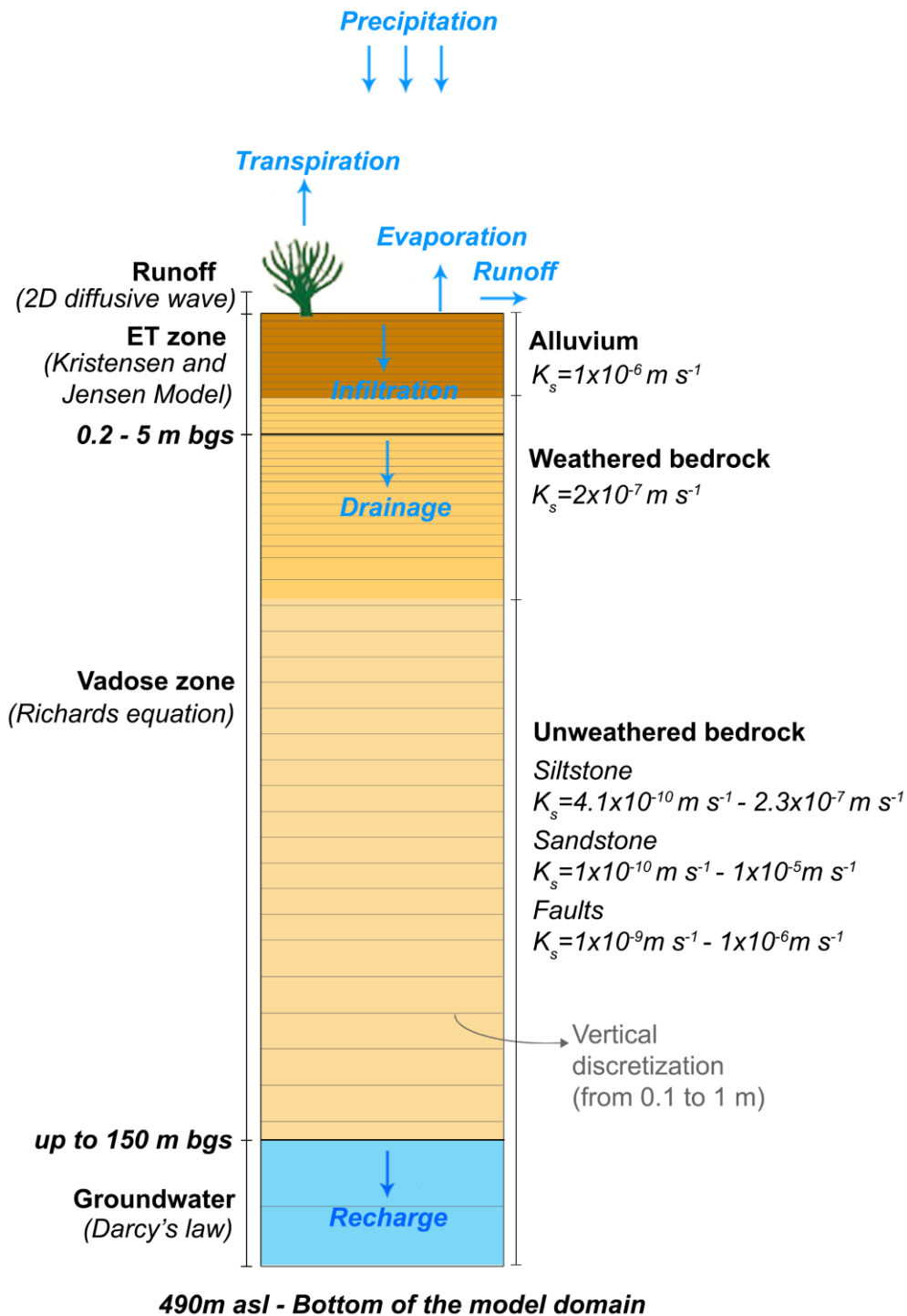
Figure 1 Topographic map of the study area and location of the wells used for calibration (blue), water isotopes sampling (red). In black the two cells where unsaturated zone water budgets were analyzed.



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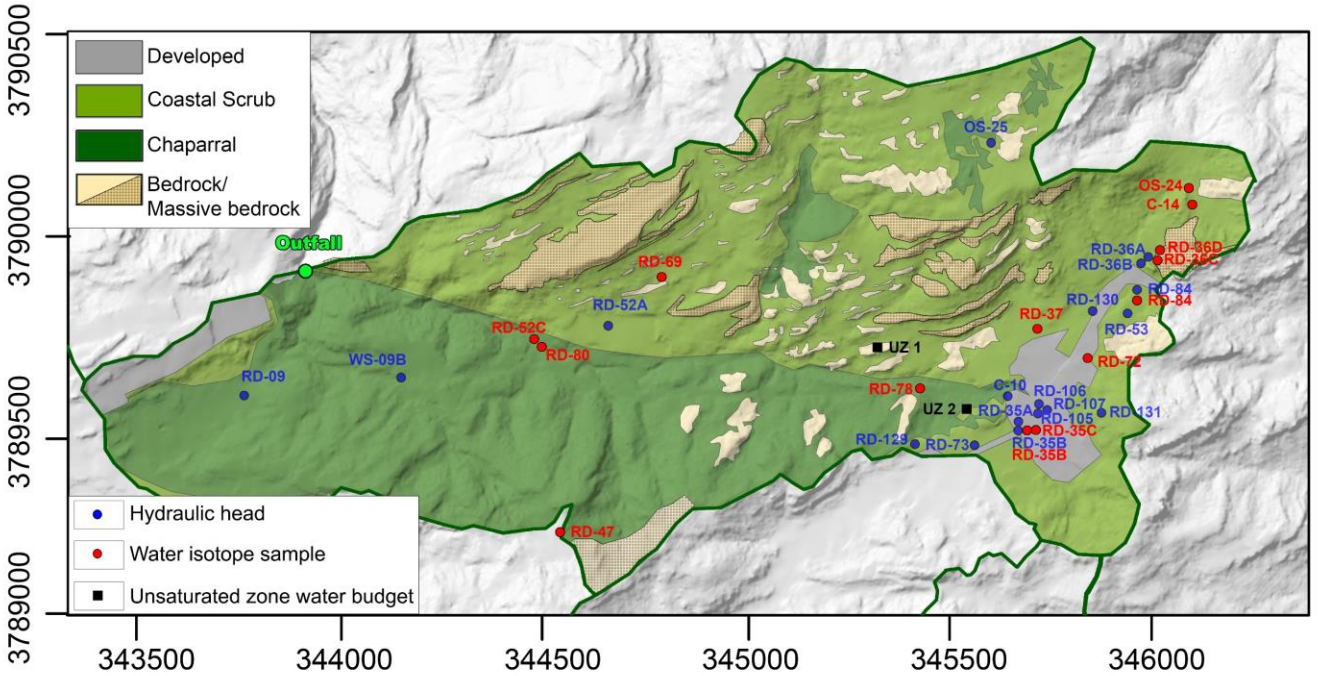
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Figure 2 Geologic map of the study area and location of the wells used for calibration (blue), water isotopes sampling (red). In black the two cells where unsaturated zone water budgets were analyzed.



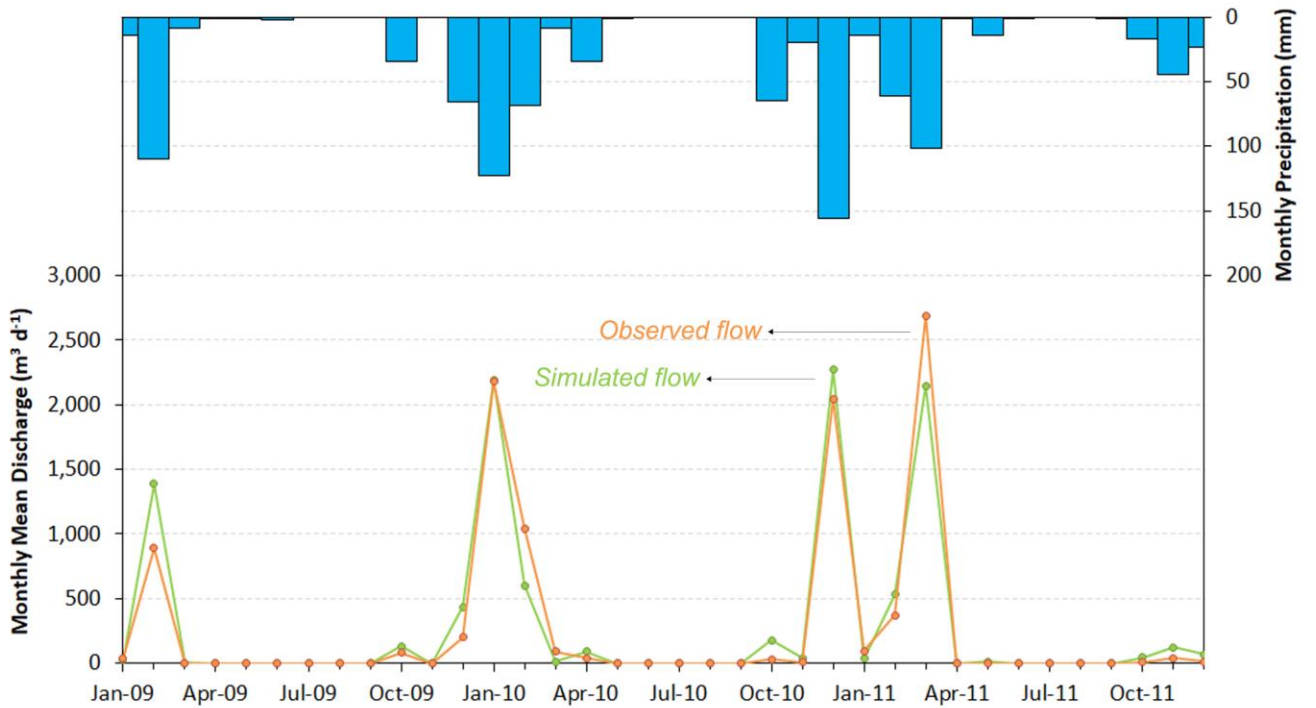
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808 Figure 3 Description of the vertical MIKE SHE model domain



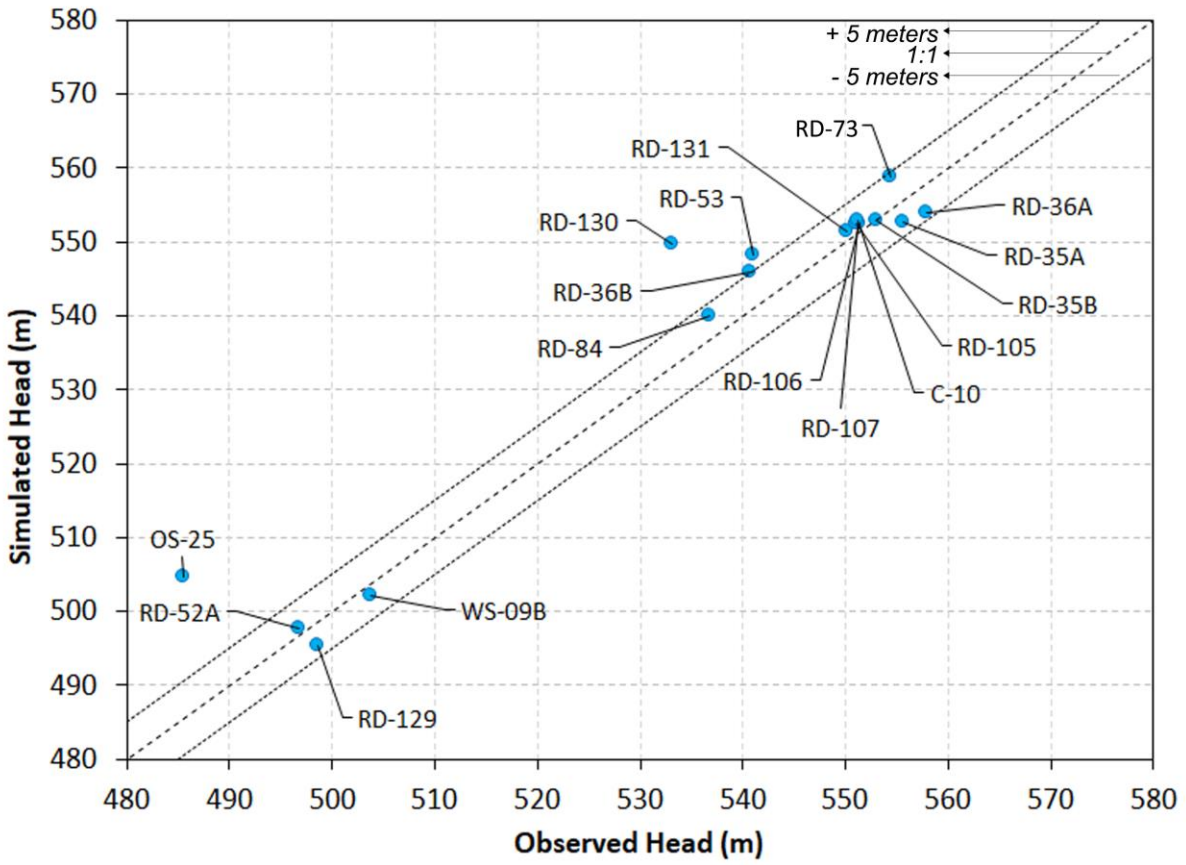
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810 Figure 4 Land use map and location of the wells used for calibration (blue), water isotopes sampling (red). In black the
 811 cells where unsaturated zone water budgets were analyzed.



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813 Figure 5 Monthly precipitation values and comparison between simulated (green) and observed (red) runoff flow at the
 814 outfall of the catchment from January 2009 to December 2011.

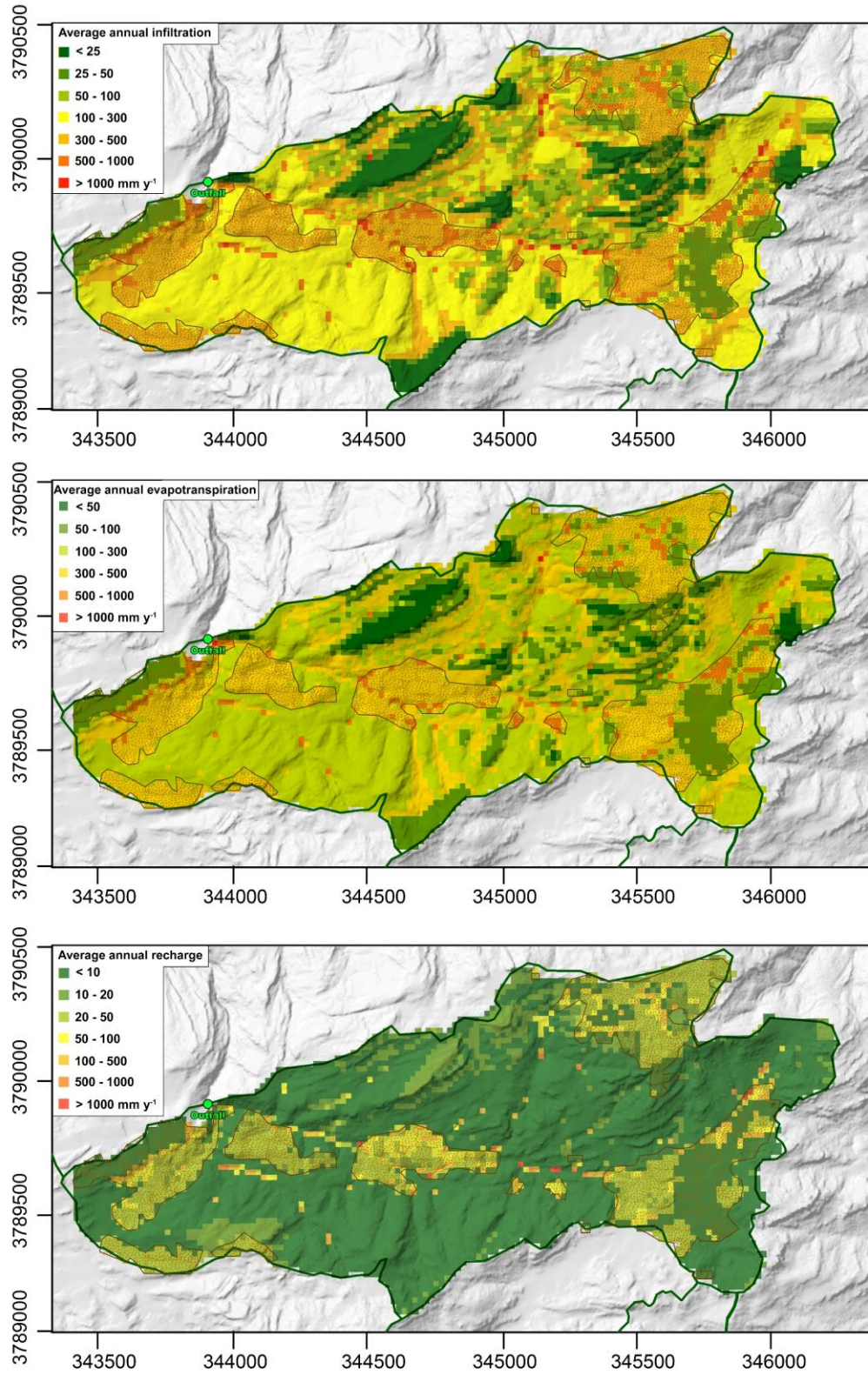


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816 *Figure 6 Comparison between simulated and observed groundwater head data for the 17 wells.*

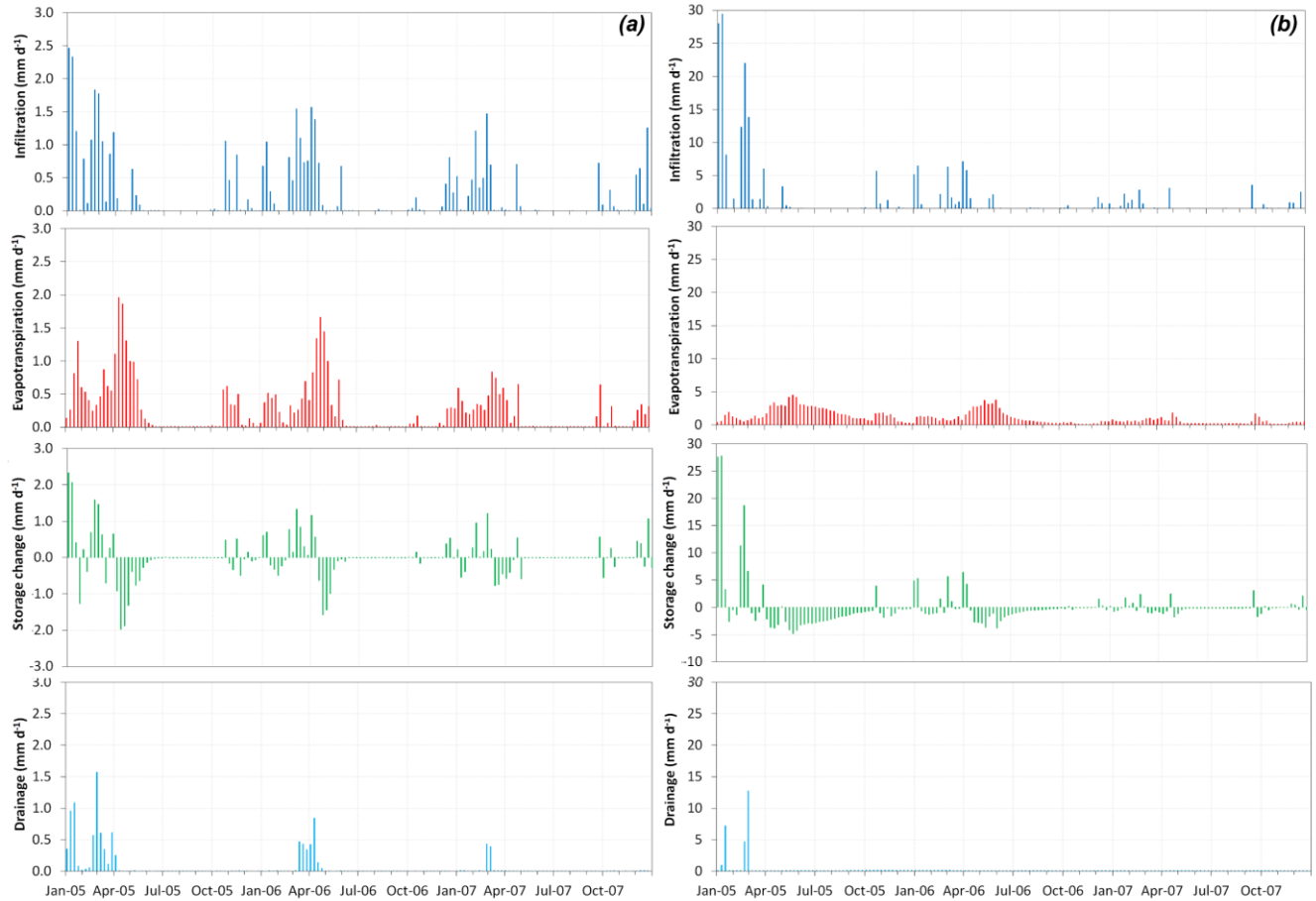
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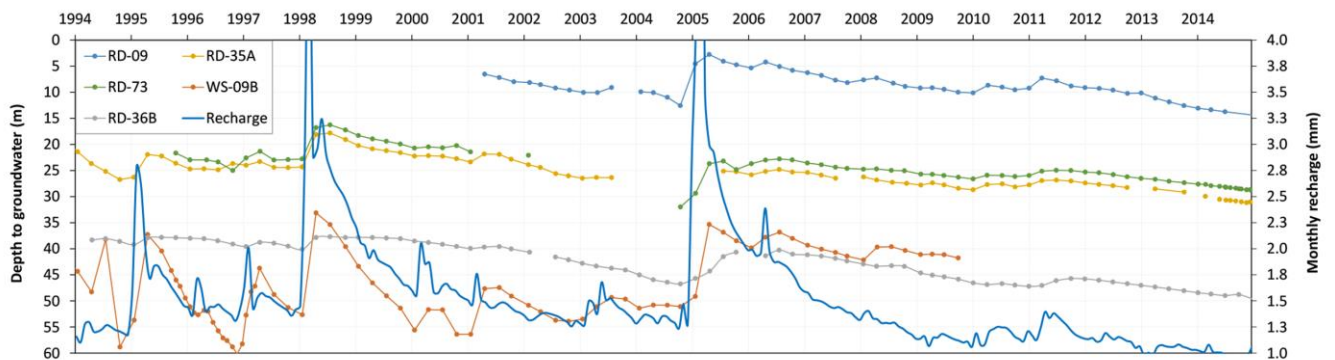
820 *Figure 7. Distribution of average annual infiltration (a), evapotranspiration (b) and recharge (c). Dashed polygons represent*
 821 *areas with alluvium at the surface.*



822

823 *Figure 8 Unsaturated zone water budget for ET zone from January 2004 to December 2007 for two cells representative of the*
 824 *domain: (a) UZ-1 area with outcropping bedrock without vegetation; (b) UZ-2 area with alluvium deposit covered by*
 825 *vegetation.*

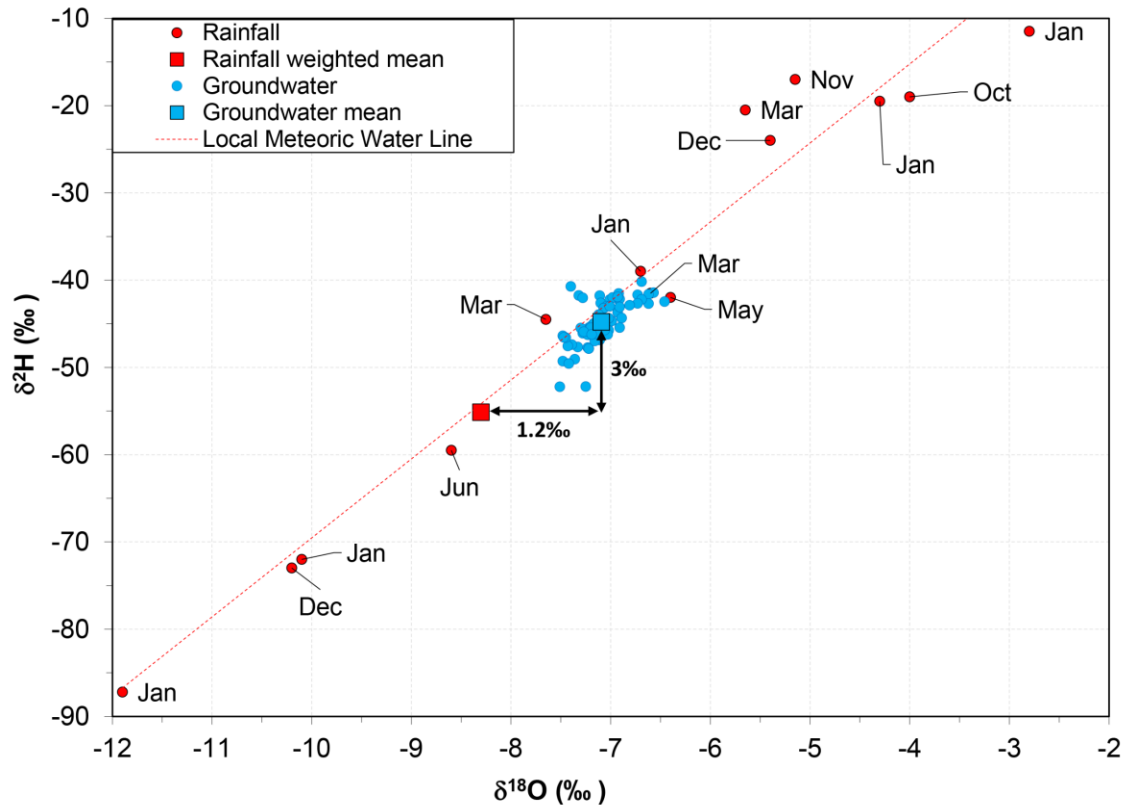
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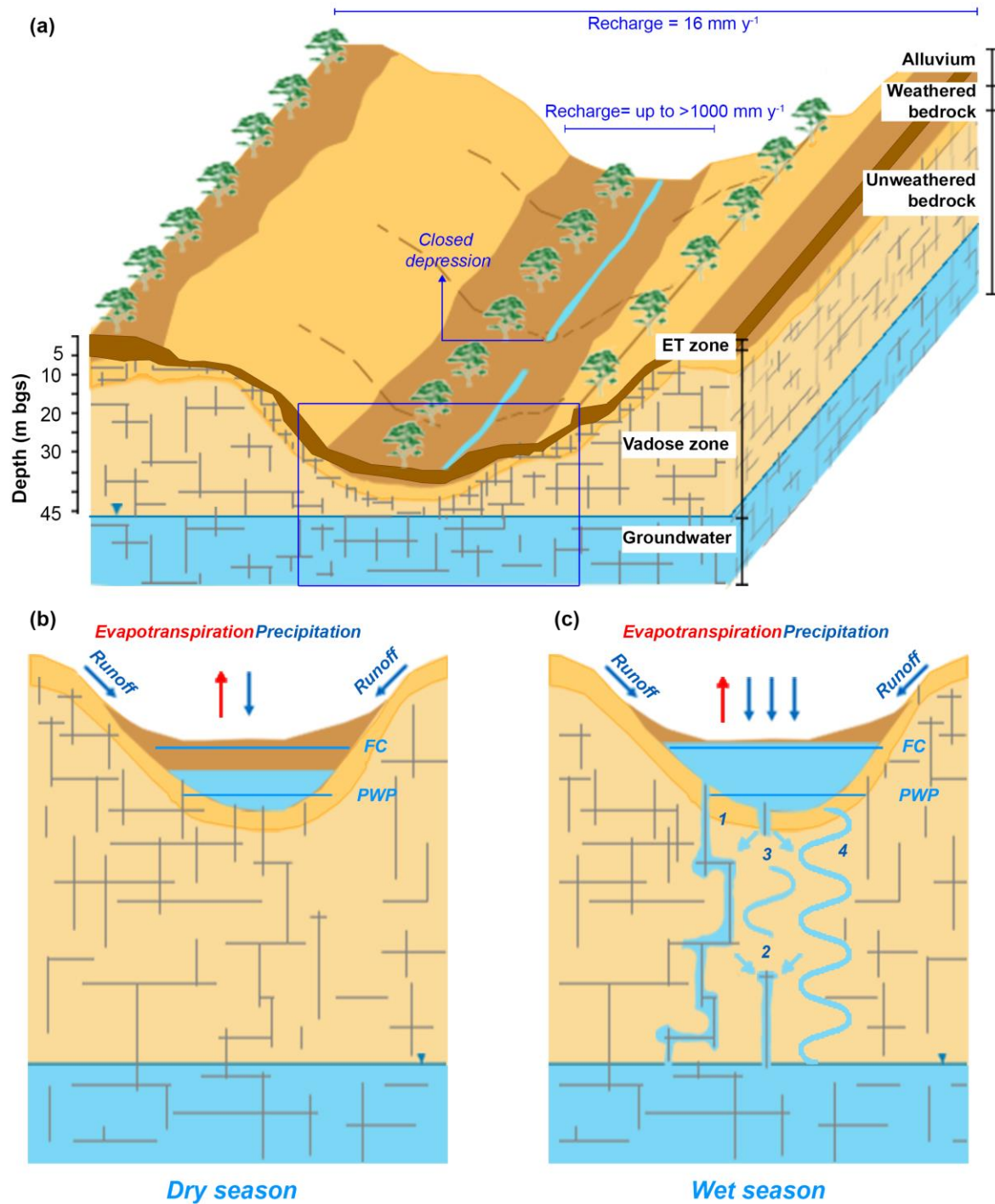
828 *Figure 9 Comparison between the monthly recharge time series and the depth to groundwater at five locations across the*
 829 *catchment.*

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832 *Figure 10 Water isotopes plot for rainfall samples collected at two rain gauge stations and groundwater samples from 16*
 833 *wells of the catchment.*



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835 Figure 11 Conceptual model for recharge at the site. (a) Spatial 3-D conceptual model of the catchment showing where high
 836 recharge occurs. 2-D schematic of the unsaturated zone hydrologic process during (b) dry season and (c) wet season. During
 837 the dry season water content is between the field capacity (FC) and the permanent wilting point (PWP) and therefore is
 838 consumed by evapotranspiration. Conversely, during the wet season, water content is above the FC and seeps into the
 839 underlying bedrock. Numbers describe mechanisms of flow in the vadose zone: 1 is fracture flow; 2 is water flowing from
 840 matrix into fractures; 3 is water flux from fractures into matrix; 4 is intergranular matrix flow.

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