

1 **Understanding the past to interpret the future: comparison of simulated**  
2 **groundwater recharge in the upper Colorado River basin using observed and**  
3 **GCM historical climate data**

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13 **Abstract** In evaluating potential impacts of climate change on water resources, water managers seek to understand

14 how future conditions may differ from the recent past. Studies of climate impacts on groundwater recharge often

15 compare simulated recharge from future and historical time periods on an average monthly or overall average annual

16 basis, or compare average recharge from future decades to that from a single recent decade. Baseline historical

17 recharge estimates, which are compared with future conditions, are often from simulations using observed historical

18 climate data. Comparison of average monthly results, average annual results, or even averaging over selected

19 historical decades, may mask the true variability in historical results and lead to misinterpretation of future

20 conditions. Comparison of future recharge results simulated using general circulation model (GCM) climate data to

21 recharge results simulated using actual historical climate data may also result in an incomplete understanding of the

22 likelihood of future changes from past conditions. In this study, groundwater recharge is estimated in the upper

23 Colorado River basin using the distributed-parameter Soil-Water Balance groundwater recharge model for the 1951–

24 2010 time period. Recharge simulations are performed using precipitation, maximum temperature, and minimum

25 temperature data from observed climate data and from 97 CMIP5 model projections. Results indicate average

26 monthly and average annual simulated recharge are similar using observed and GCM climate data. However, 10-

27 year moving average recharge results show substantial differences between observed and simulated climate data,

28 particularly during the 1970–2000 time period, with much greater variability seen for results using observed climate

29 data.  
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31 **Keywords:** Colorado River, groundwater recharge, climate change, groundwater management

## 32 **1. Introduction**

33 In order to prepare for possible changes in water resources in response to a changing climate, water managers must  
34 understand how future hydrologic conditions may differ from conditions in the recent past. Recently, simulations of  
35 future hydrologic conditions using downscaled climate data from one or more general circulation models (GCM)  
36 and multiple emission scenarios have become a common tool for understanding potential change in hydrologic  
37 systems (Holman et al. 2012). These published studies are often of surface-water systems (for example, see Barnett  
38 et al. 2007; Christensen et al. 2004; Kopytkovskiy et al. 2015; Miller et al. 2012; Vano et al., 2014). Impacts to  
39 groundwater also are expected in future climates (Green et al. 2011) and investigations of potential impacts to  
40 groundwater systems, especially changes to groundwater recharge in response to changing climate, also have used  
41 this approach. Studies comparing simulated groundwater recharge in future climates projected by GCMs to  
42 historical recharge have been reported for basins in Germany (Eckhardt and Ulbrich, 2003), British Columbia (Allen  
43 et al., 2010; Scibeck and Allen, 2006; Toews and Allen, 2009), Australia (Crosbie et al., 2010; Crosbie et al., 2011;  
44 Crosbie et al., 2013; McCallum et al., 2010), southern Canada (Jyrkama and Sykes, 2007), eastern Canada (Kurylyk  
45 and MacQuarrie, 2013), Africa (Mileham et al., 2009; Nyenje and Batelaan, 2009), England (Holman et al., 2009),  
46 the western United States (Meixner et al., 2016), and the upper Colorado River basin (Tillman et al., 2016). These  
47 studies typically simulate groundwater recharge averaged annually or monthly over future time periods using GCM-  
48 output climate data and compare these results to baseline simulated recharge over historical time periods using  
49 observed climate data, stochastic-weather-generator climate data, or GCM-simulated climate data. The time periods  
50 of comparison, both future and past, as well as the source of historical climate data, both actual observed and  
51 generated or simulated, vary among the studies. Although similar recharge-simulation results are expected using  
52 observed climate data and GCM-output climate data that are downscaled using the same observed data,  
53 understanding any differences in these historical results may be important when comparing with projected future  
54 recharge.

55 In this study, groundwater recharge is estimated in the upper Colorado River basin using the distributed-parameter  
56 Soil-Water Balance groundwater recharge model for the 1951–2010 historical time period. Recharge simulations  
57 are performed using precipitation, maximum temperature, and minimum temperature data from actual climate  
58 observations and from 97 downscaled Coupled Model Intercomparison Project phase 5 (CMIP5) model results over

59 the historical time period. Simulated historical recharge results are compared for the observed and modeled climate  
60 input datasets on an average monthly, average annual, and moving ten-year average basis. All SWB groundwater  
61 recharge modeling results for the UCRB described in this manuscript are available at the USGS ScienceBase web  
62 site (Tillman, 2016).

63

## 64 **2. Study area**

65 More than 3 million people in Mexico and 35 million people in the United States depend on the Colorado River to  
66 supply their domestic and industrial water needs (Bureau of Reclamation 2011; Colorado River Basin Salinity  
67 Control Forum 2013). The Colorado River also supplies irrigation water for over 1.8 million hectares of land in the  
68 United States and Mexico and hydroelectric power along the river and its tributaries generates about 12 billion  
69 kilowatt hours annually (Colorado River Basin Salinity Control Forum 2011). Miller et al. (2014) estimated that  
70 annual discharge of groundwater to rivers and streams (base flow) in the upper Colorado River basin (UCRB) can  
71 range from 21 to 58 percent of streamflow, with higher percentages during low-flow conditions. Recently, a study  
72 by Castle et al. (2014) using remotely sensed gravity observations from the NASA Gravity Recovery and Climate  
73 Experiment (GRACE) mission found that UCRB groundwater was depleted by more than 50 km<sup>3</sup> from December  
74 2004 to November 2013. Understanding groundwater-budget components, including groundwater recharge, is  
75 important to sustainably manage both groundwater and surface-water supplies in the Colorado River Basin. From  
76 headwaters in the Rocky Mountains through seven states and Mexico, the Colorado River traverses more than 2200  
77 km to discharge into the Gulf of California (fig. 1A). The Colorado River Basin drains parts of Wyoming, Utah,  
78 Colorado, New Mexico, Arizona, Nevada, California, and Mexico, and is divided into upper and lower basins at the  
79 compact point of Lee Ferry, Arizona, a location 1.6 km downstream of the mouth of the Paria River (fig. 1A;  
80 Anderson 2004). The UCRB is defined for this study as the 293,721 km<sup>2</sup> drainage area of the Colorado River basin  
81 above the Lee Ferry compact point and the Great Divide closed basin, as delineated by the Region 14 hydrologic  
82 unit code (HUC; see <http://water.usgs.gov/GIS/huc.html>). Major tributaries to the Colorado River in the upper basin  
83 include the Dolores, Green, Gunnison, San Juan, White, and Yampa Rivers (fig. 1B). Average annual precipitation  
84 ranges from less than 250 mm in low-elevation areas to more than 1000 mm in high elevation areas in the Southern  
85 Rocky Mountains (fig. 1C, PRISM Climate Group 2012). The UCRB varies in elevation from about 944 m near the  
86 Lees Ferry streamgauge to more than 4260 m in peaks in the Southern Rocky Mountains in the eastern part of the

87 UCRB (Liebermann et al. 1989). UCRB land cover is predominately shrub/scrub and evergreen forest (Fry et al.  
88 2011), with few high-density population centers (fig. 1D).

89

90 **Fig. 1** Location of the upper Colorado River basin study area within the southwestern United States (A), major  
91 tributaries to the Colorado River (B), average annual precipitation (C; PRISM Climate Group 2012), and major  
92 land-cover classifications (D; Fry et al. 2011).

93

94 Areas with the potential for recharge of groundwater supplies through infiltration of excess precipitation are present  
95 across most of the UCRB. Regional aquifers in the UCRB are composed of permeable, moderately to well-  
96 consolidated sedimentary rocks ranging in age from Permian to Tertiary (Robson and Banta 1995), although  
97 groundwater in shallow alluvial deposits may be locally important in some locations in the Southern Rocky  
98 Mountains (Apodaca and Bails 2000). At least three groups of regional, productive water-yielding geologic units  
99 have been identified in the UCRB (Robson and Banta 1995; Geldon 2003a,b; Freethey and Cordy 1991). Tertiary  
100 aquifers of limited extent in the northern and southeastern parts of the basin overlie Mesozoic aquifers that also are  
101 present throughout most of the study area. Deeper Paleozoic aquifers are present throughout much of the UCRB and  
102 may outcrop at land surface in uplifted areas. Major aquifers are each partially separated by confining units, and  
103 groundwater flows between the aquifers in areas where confining units are missing. Interconnection of the aquifers  
104 creates a regional groundwater-flow system (Geldon 2003a,b; Freethey and Cordy 1991). In his investigation of the  
105 hydrologic and groundwater-flow systems in the UCRB, Geldon (2003b) estimates about 8.14 km<sup>3</sup> of recharge to all  
106 groundwater systems in the area, excluding the upper San Juan basin which was not part of the study.

107

### 108 **3. Methods and data**

109 The Soil-Water Balance groundwater recharge model was used to simulate groundwater recharge in the UCRB for  
110 the water-year 1951–2010 time period at a daily time step. Recharge simulations were performed using both  
111 observed historical climate data and simulated historical climate data from CMIP5 GCM output.

112

#### 113 **3.1 Soil-water balance recharge model**

114 The Soil-Water-Balance (SWB) computer code (Westenbroek et al. 2010) estimates spatial and temporal variations  
 115 in groundwater recharge by calculating water balance components at daily time steps. SWB has been used in  
 116 several completed and ongoing regional groundwater studies in the U.S. including the High Plains Aquifer (Stanton  
 117 et al. 2011), the Lake Michigan Basin (Feinstein et al. 2010), basins in Wisconsin (Dripps and Bradbury 2009) and  
 118 Minnesota (Smith and Westenbroek 2015), the Northern Atlantic Coastal Plain Aquifer System (Masterson et al.  
 119 2013), the Ozark Plateau Groundwater Availability Study (see <http://ar.water.usgs.gov/ozarks/waterbud.html>), and  
 120 the Appalachian Plateaus Groundwater Availability Study (see  
 121 <http://va.water.usgs.gov/appalachianplateaus/waterbud.html>). SWB follows a modified Thornthwaite-Mather soil-  
 122 water-balance accounting approach (Thornthwaite 1948; Thornthwaite and Mather 1957) and recharge is estimated  
 123 separately for each grid cell within the model domain. Sources and sinks of water within each grid cell are  
 124 estimated based on climate data and landscape characteristics, and recharge is then estimated as the difference  
 125 between the change in soil moisture and these sources and sinks:

126

$$\begin{array}{l}
 127 \qquad \qquad \qquad \textit{water sources} \qquad \qquad \qquad \textit{water sinks} \\
 128 \text{ (rainfall + snowmelt + inflow) - (interception + outflow + AET) - } \Delta \text{ soil moisture = RECHARGE} \qquad (1) \\
 129
 \end{array}$$

130 Spatially gridded datasets required for SWB simulations include land cover, overland flow direction, hydrologic soil  
 131 group, available soil-water capacity, daily precipitation, daily maximum temperature, and daily minimum  
 132 temperature. Tabular information required by SWB include runoff curve numbers, vegetation rooting depths,  
 133 interception values, and maximum daily recharge values for each combination of hydrologic soil group and land-  
 134 cover type. Inflow to a cell is surface flow from adjacent cells, calculated using the National Resources  
 135 Conservation Service curve number rainfall-runoff relation. The direction of runoff from cell to cell is determined  
 136 using a flow-direction grid derived from a digital-elevation model (DEM). Interception is a user-specified amount  
 137 of precipitation that is trapped and used by vegetation. Outflow from a cell is calculated in the same manner as  
 138 inflow to the cell. There are several methods available for estimating potential evapotranspiration (PET) in the  
 139 SWB model, from which actual evapotranspiration (AET) is calculated. For the UCRB simulations, the Hargreaves-  
 140 Samani (1985) method is used as it produces spatially variable estimates of potential ET (PET) from spatially  
 141 varying minimum and maximum air temperature data for each daily time step:

142

$$143 \quad \text{PET} = 0.0135 \times \text{RS} \times (\text{T} + 17.8) \text{ with } \text{RS} = \text{K}_{\text{RS}} \times \text{RA} \times \text{TD}^{0.5} \quad (2)$$

144

145 where PET is potential ET, RS is incoming solar radiation, T is mean air temperature in °C, KRS is a calibration  
 146 coefficient, RA is extraterrestrial radiation, and TD is the measured air temperature range (Hargreaves and Samani  
 147 1985). Extraterrestrial radiation is estimated as a function of the day of year and latitude following the method of  
 148 Allen et al. (2006). The computation of soil moisture in equation 1 requires several intermediary values. First, PET  
 149 is subtracted from precipitation (P) for all grid cells. If P – PET is negative (i.e., if P < PET), then there is a potential  
 150 deficiency of water. Accumulated Potential Water Loss (APWL) is computed as the running sum of daily P – PET  
 151 values during times when P < PET. Soil moisture is estimated using the current AWPL value in the Thornthwaite-  
 152 Mather relation that describes the nonlinear relation between soil moisture and APWL. Actual ET (AET) is then  
 153 equal to only the amount of water that can be extracted from the soil. If P – PET is positive (i.e., if P > PET), a  
 154 potential surplus of water exists and AET is equal to PET. Soil moisture is calculated by adding P – PET directly to  
 155 the previous day's soil-moisture value. If the new soil moisture value is less than the maximum water-holding  
 156 capacity of the soil (calculated as the product of the available soil water capacity and the root-zone depth), then the  
 157 Thornthwaite-Mather relation is used to back-calculate a reduced APWL. If the new soil moisture value is greater  
 158 than the maximum water-holding capacity of the soil, then soil moisture is capped at the maximum water-holding  
 159 capacity, excess soil-moisture becomes recharge, and AWPL is set to zero.

160 All spatially gridded input datasets were resampled to the same cell size and geographic coordinate system as the  
 161 1/8<sup>th</sup> degree climate data described below. For a detailed description of the source, manipulation, and resampling of  
 162 SWB model inputs for UCRB recharge simulations, and a sensitivity analysis of model results, see Tillman (2015).  
 163 See Westenbroek et al. (2010) for detailed explanations of SWB processes. Annual recharge simulated during the  
 164 1951-2010 historical time period by the SWB model over the same UCRB area as the Geldon (2003b) study is 9.1  
 165 km<sup>3</sup> and 8.6 km<sup>3</sup> (mean and median annual values), representing 11% and 6% percent differences, respectively, with  
 166 the Geldon (2003b) estimate.

167

168 Climate changes are expressed in SWB simulated recharge results (equation 1) through the computation of AET  
 169 (mean temperature) and through precipitation input. The SWB model does not include changes in land use over

170 time or simulate changes in stomatal conductance or leaf area in a CO<sub>2</sub> enriched atmosphere (Eckhardt and Ulbrich  
171 2003; Holman et al. 2012). Only direct impacts of climate change are evaluated in SWB recharge results.

172

### 173 **3.2 Climate data**

174 Groundwater recharge was simulated for the 1951–2010 time period on a daily time step using both observed and  
175 simulated precipitation and temperature climate data. Daily 1/8<sup>th</sup> degree gridded observed climate data were  
176 processed for the UCRB study area as described in Maurer et al. (2002) and obtained from  
177 [http://www.engr.scu.edu/~emaurer/gridded\\_obs/index\\_gridded\\_obs.html](http://www.engr.scu.edu/~emaurer/gridded_obs/index_gridded_obs.html). Simulated daily precipitation and  
178 temperature data for the UCRB study area were obtained from the downscaled climate and hydrology projections  
179 archive ([http://gdo-dcp.ucllnl.org/downscaled\\_cmip\\_projections/dcpInterface.html](http://gdo-dcp.ucllnl.org/downscaled_cmip_projections/dcpInterface.html)) as downscaled 1/8<sup>th</sup> degree bias-  
180 corrected spatially disaggregated (BCSD) climate projection datasets (Bureau of Reclamation 2013). For UCRB  
181 groundwater recharge simulations, simulated climate datasets were available for 97 climate projections from the  
182 Coupled Model Intercomparison Project phase 5 (CMIP5) multi-model archive (supplemental Table S1). Each of  
183 the 97 ensemble members were derived from a General Circulation Model (GCM) run using a given future-emission  
184 scenario, known as a Representative Concentration Pathway (RCP), with a unique initial condition. The four RCPs,  
185 developed at the request of the Intergovernmental Panel on Climate Change (IPCC), are for radiative forcing levels  
186 of 8.5, 6, 4.5, and 2.6 W/m<sup>2</sup> by the end of the century (Van Vuuren 2011). The four RCPs include one very high  
187 baseline (no climate policy) emission scenario (RCP8.5), two medium stabilization scenarios (RCP4.5 and RCP6),  
188 and one very low forcing level (RCP2.6; Van Vuuren 2011). Since GCMs are typically run at coarse spatial  
189 resolutions (e.g., ~100-200 km on a grid side) and at time scales of 100-years or longer, there is a need to post-  
190 process GCM-derived variables such as precipitation and temperature to finer spatial scales in order to conduct  
191 climate impact assessments. This post-processing step is commonly referred to as downscaling, and there is a  
192 continuum of downscaling methods ranging from statistical approaches to physically-based modeling. The 97  
193 projections used in this study were developed using a statistical downscaling method referred to as BCSD (Bias-  
194 Correction and Spatial Disaggregation; Wood et al. 2004). The BCSD method was used to develop monthly  
195 precipitation and temperature fields at 1/8° × 1/8° (latitude × longitude) spatial resolution from the GCM native-  
196 scales. The monthly precipitation and temperature fields were subsequently disaggregated to daily values using a  
197 historical resampling and scaling technique (Wood et al. 2002). These daily precipitation and temperature data for

198 the UCRB study area were obtained from the downscaled climate and hydrology projections archive  
199 ([http://gdo-dcp.ucllnl.org/downscaled\\_cmip\\_projections/dcpInterface.html](http://gdo-dcp.ucllnl.org/downscaled_cmip_projections/dcpInterface.html); Bureau of Reclamation 2013).

200

#### 201 **4. Results and discussion**

202 Daily simulated groundwater recharge for the UCRB for water years 1951–2010 was aggregated into monthly time  
203 periods for further analysis. Recharge results are presented as boxplots indicating 90<sup>th</sup>, 75<sup>th</sup>, 50<sup>th</sup> (median), 25<sup>th</sup>, and  
204 10<sup>th</sup> percentiles, as well as mean values, over the time period of analysis. Simulated recharge results using observed  
205 climate data and GCM-simulated climate data are presented separately. Although simulated recharge differences  
206 between GCM-climate data from different RCPs are not expected, GCM-climate results are presented by combining  
207 results from all RCPs and by presenting results for each RCP separately. In discussing results, differences presented  
208 in terms of percent (%) refer to percentage difference, which is the absolute value of the difference between two  
209 values divided by their mean.

210 Simulated UCRB monthly groundwater recharge (fig. 2) indicates substantial recharge in the March through June  
211 time frame, during snowmelt and spring precipitation, with little recharge during other months of the year. Mean  
212 recharge values for March–June account for over 88% of the mean annual recharge for the UCRB for simulations  
213 using both observed and GCM climate datasets. Importantly, differences between monthly recharge results  
214 simulated using the different climate datasets are not large during these months. During the high-recharge months  
215 of March through June, simulated mean recharge from the GCM climate data, whether results were grouped together  
216 or separated by RCP, differed from the Maurer et al. (2002) observed climate data by a maximum of 26%. The  
217 highest recharge months of April and May, accounting for over 64% of annual recharge, differed by 4% or less.  
218 Similar results are noted comparing median recharge values, with 32% or less difference between observed and  
219 GCM climate data results in March through June, and April and May results differing by 5% or less. This similarity  
220 in groundwater recharge results using the different climate datasets is consistent whether the GCM-climate data  
221 simulations are grouped together (fig. 2b), with 5820 results per month (60 years × 97 ensembles), or separated by  
222 RCP (fig. 2c-f), with 960–1860 results per month (fig. 2). Substantial differences of 66–72% between mean  
223 recharge results simulated with observed and GCM climate data are seen for July, but this month accounts for only  
224 1% of the mean annual recharge for the basin. Differences in median recharge results are greater than 100% in

225 August, but this month also contributes very little to total annual recharge (less than 1%). Variability of within-  
226 monthly results also are similar whether observed or GCM climate data were used in the recharge simulations.  
227 Differences in recharge results from observed or GCM climate data between the 75<sup>th</sup> percentile and median values  
228 and between 25<sup>th</sup> percentile and median values are  $\leq 25\%$  for the high recharge months of March through June, with  
229 the exception of June differences between 75<sup>th</sup> percentile and median values (36–45%). For the months contributing  
230 the majority of recharge, comparisons of mean or median simulated monthly recharge results from future climate  
231 scenarios to historical results using either observed or GCM climate data, either with all results grouped together or  
232 separated by RCP, would produce similar conclusions about changes to the UCRB groundwater system.

233

234 **Fig. 2** Statistics for monthly groundwater recharge in the upper Colorado River basin for water-years 1951–2010  
235 simulated with the Soil-Water Balance model using (a) observed climate data (Maurer et al., 2002) and CMIP5  
236 GCM climate data (b-f).

237 By most measures, annual groundwater recharge results for the UCRB also are similar for simulations using the  
238 different climate datasets (fig. 3). Mean annual recharge results differ between simulations using observed climate  
239 data and GCM climate data by 5% or less, while median values differ by 2% or less. The spread of recharge results  
240 between 25<sup>th</sup> percentile and median values are likewise similar for simulations using the different climate datasets,  
241 with differences of 22% or less. Variability between 75<sup>th</sup> and median values, however, is substantially greater for  
242 recharge results simulated with observed climate data compared with results simulated with GCM climate data (fig.  
243 3). While separate-RCP results differ from all-RCP results by 5% or less, recharge results using observed climate  
244 data differ by 42–43% from GCM results, depending on whether GCM results are grouped by RCP or not (fig. 3).  
245 This greater spread in higher-than-median recharge values may be important if comparing mean or median recharge  
246 results spanning time periods shorter than the full 60 years of historical simulations. Comparisons of climate change  
247 impacts on the central tendency (mean or median) of groundwater recharge to historical results averaged over the  
248 60-year record, however, would result in similar conclusions whether observed or GCM climate data were used in  
249 historical simulations.

250

251 **Fig. 3** Annual groundwater recharge in the upper Colorado River basin for water-years 1951–2010 simulated with  
252 the Soil-Water Balance model using observed climate data (Maurer et al., 2002; left boxplot) and CMIP5 GCM  
253 climate data (all other boxplots).

254

255 Simulated annual groundwater recharge during ten-year periods, moving every 5 years, also was analyzed for water  
256 years 1951–2010 (fig. 4). The ten-year moving period balances the need to smooth out variability in recharge from  
257 individual years, whose effects are integrated over time in groundwater systems (Green et al. 2011), with a desire to  
258 provide useful information to water managers over a reasonably short time frame in order to allow for mitigating  
259 action. Moving the ten-year period through time by five years eliminates the subjectivity of picking decade start and  
260 stop years that may encompass anomalously wet or dry periods. Comparing future and past recharge results over  
261 ten-year moving periods addresses the question “how might conditions in future decades differ from conditions  
262 experienced in decades since 1951?”

263 Annual simulated recharge results over moving ten-year periods are similar among results using GCM climate data  
264 (fig. 4). The mean of annual separate-RCP results are within 3% of combined RCP results, with median values  
265 within 4%. Comparison of simulated recharge results between simulations using observed and GCM climate data,  
266 however, reveals substantial differences in mean annual values during some decades (fig. 4). During the 1976–1985  
267 ten-year period, mean annual recharge simulated with GCM climate data differ from simulations with observed  
268 climate data by 23–25%, depending on whether results are separated by RCP or are grouped together. Differences  
269 of 18–20% are observed in the 1991–2000 time period. Differences between recharge simulations using GCM  
270 climate data and observed climate data are even more pronounced when medians are used as an indication of central  
271 tendency during the ten-year period. Median annual recharge values differ by more than 41% during the 1976–1985  
272 decade, with differences of 16–20% in both the 1981–1990 and 1991–2000 decades (fig. 4). The distribution of  
273 annual values is also noticeably different for recharge simulations using observed climate data versus GCM climate  
274 data. Differences between the 75<sup>th</sup> percentile minus median values of annual recharge are greater than 50% in over  
275 one-third of the decadal comparisons between results using simulated climate data and results using observed  
276 climate data. Difference in 25<sup>th</sup> percentile minus median values are 50% or greater in almost half of the decadal  
277 comparisons. The differences in distribution of simulated groundwater recharge using observed climate data and

278 GCM climate data are not as strongly observed in PET results (supplemental fig. S1) where temperature changes are  
279 expressed, but are evident in precipitation for the basin (supplemental fig. S2). The difference in variability in  
280 simulated groundwater recharge is related to the smaller sample size ( $n=10$ ) for the observed data decadal analyses  
281 (compared with  $n=160-970$  for the GCM analyses), but also is a result of variability in observed precipitation that is  
282 not captured by GCM simulated historical climate data. While an annual simulated recharge time period is too short  
283 for meaningful comparisons with future changes, the inability of recharge simulations using GCM climate data to  
284 capture much of the annual variability of recharge using observed climate data (fig. 5) may affect the interpretation  
285 of changes in future versus past conditions over even longer averaging periods. Conclusions about comparisons of  
286 future changes in simulated annual groundwater recharge in 10-year moving periods to historical results may depend  
287 upon whether the historical results were simulated using observed or GCM climate data. For example, simulated  
288 mean annual recharge in future decades appears to increase relative to historical recharge modeled using simulated  
289 climate data (fig. 6a). The same projected average annual recharge, however, appears to decline somewhat relative  
290 to historical recharge simulated with observed climate data (fig. 6b).

291

292 **Fig. 4** Annual groundwater recharge in the upper Colorado River basin over 10-year periods, moving every 5 years  
293 between water-years 1951 and 2010, simulated with the Soil-Water Balance model using (a) observed climate data  
294 (Maurer et al., 2002) and CMIP5 GCM climate data (b-f).

295 **Fig. 5** Percentage change of annual recharge (observed historical climate data) or mean annual recharge (for CMIP5  
296 simulated climate data) from 1951–2010 mean annual values in the upper Colorado River basin.

297 **Fig. 6** Comparison of ten-year averages, moving every five years, of simulated annual groundwater recharge in the  
298 upper Colorado River basin using projected climate data to modeled recharge using (a) simulated historical climate  
299 data and (b) observed historical climate data. Results presented as changes in recharge relative to historical average  
300 of ten-year means from 1951–2010.

301

## 302 **5. Summary and Conclusions**

303 Groundwater recharge in the upper Colorado River basin was simulated using the Soil Water Balance model with a  
304 daily time step for the water year 1951–2010 time period. Historical SWB recharge simulations were performed  
305 using both observed climate data and GCM-output climate data from 97 CMIP5 projections. Mean and median  
306 results for monthly and average annual time periods were similar for recharge simulations using observed or GCM  
307 climate data, with an increase in variability noted in observed annual results. Substantial differences in mean and  
308 median annual averages between simulated recharge using observed versus GCM climate data were seen in several  
309 moving ten-year time periods. Investigating potential changes in future groundwater recharge requires an  
310 understanding of the historical conditions with which they are compared. The likelihood of future UCRB  
311 groundwater recharge differing from that of the last 60 years is dependent upon changes in future climate, as well as  
312 potentially the choice of historical climate dataset used in recharge simulations and the time period of comparison.

313  
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320 coordinating support and led development of software infrastructure in partnership with the Global Organization for  
321 Earth System Science Portals.

322

323

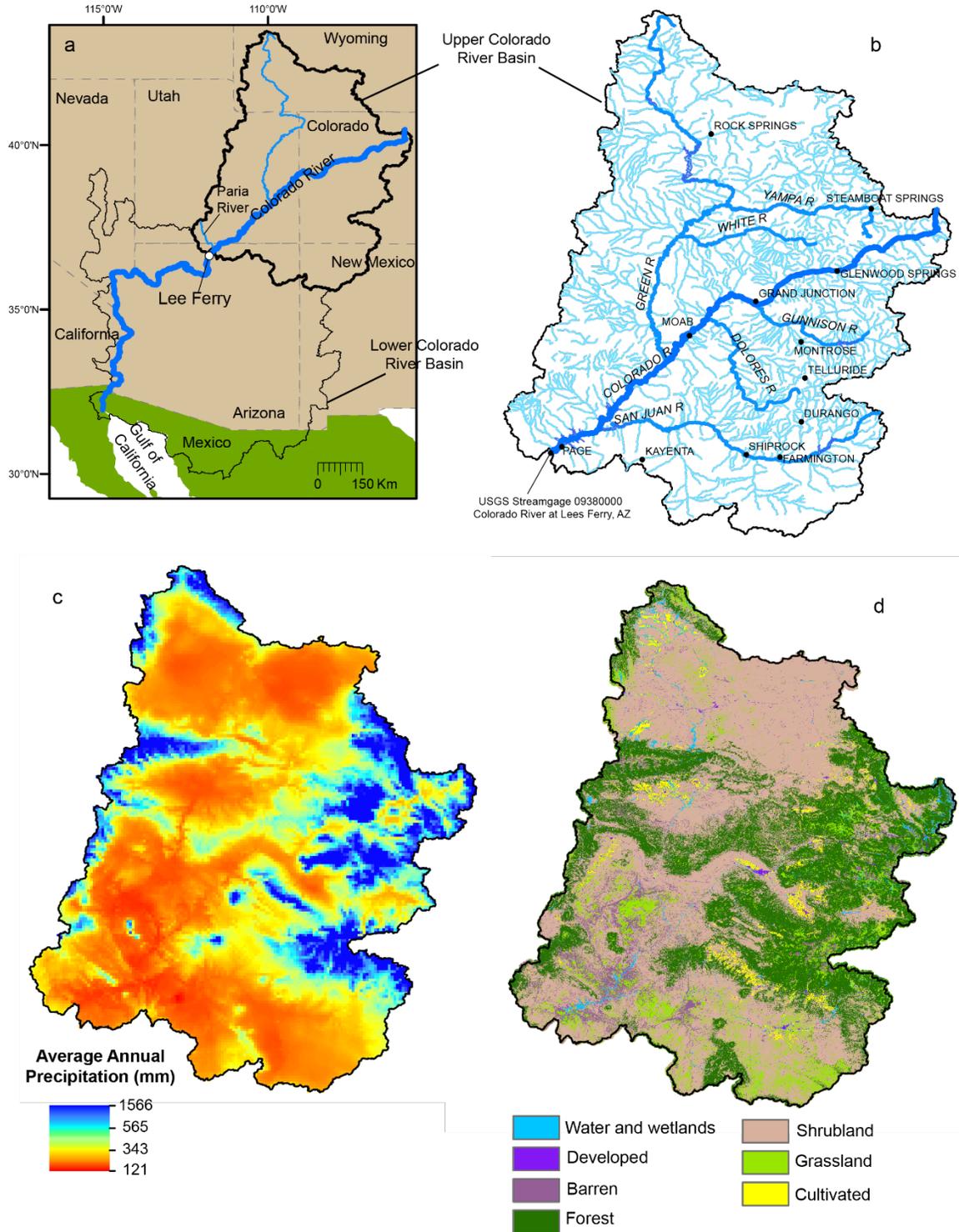
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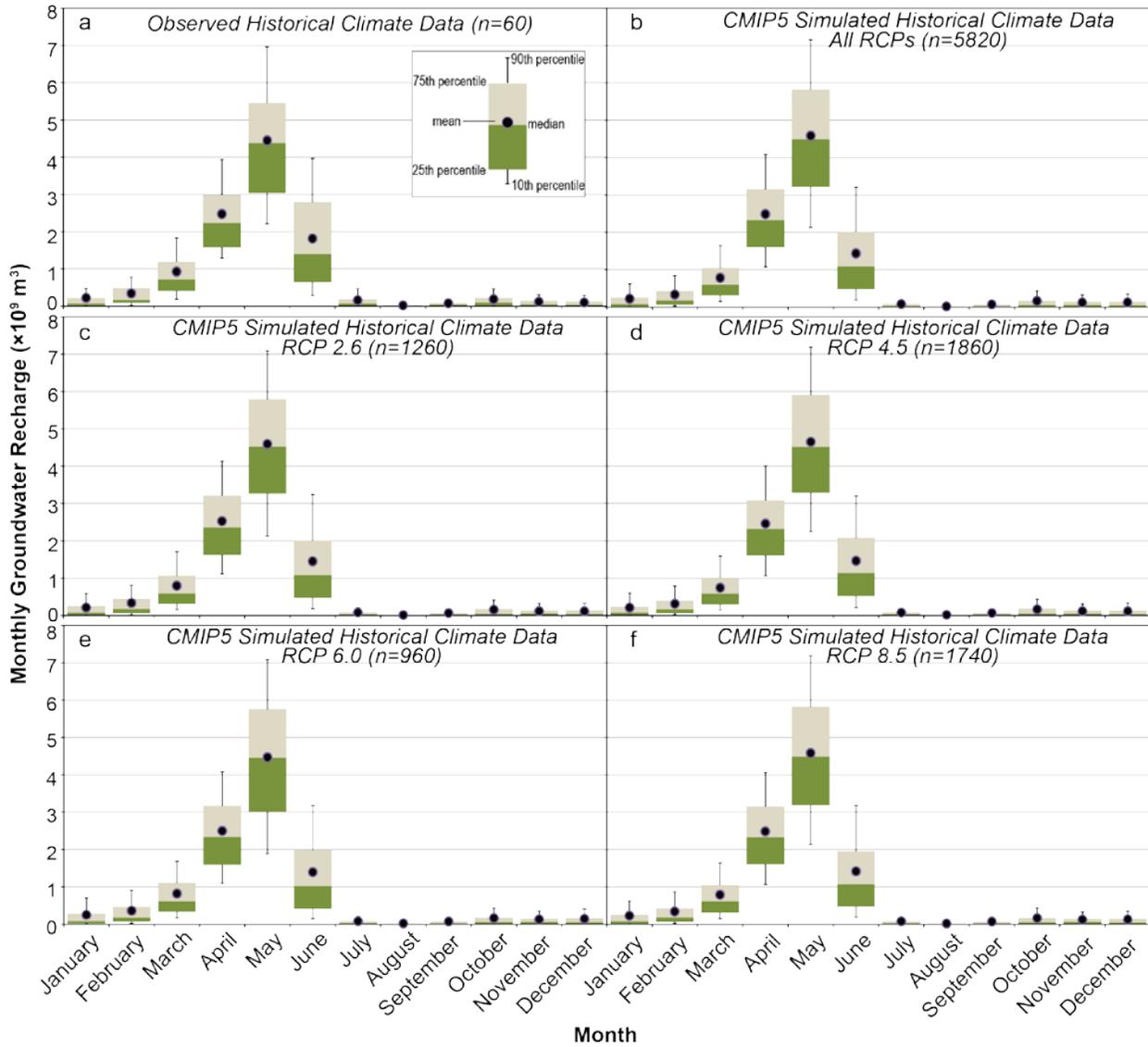
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476 **Fig. 1** Location of the upper Colorado River basin study area within the southwestern United  
 477 States (a), major tributaries to the Colorado River (b), average annual precipitation (c; PRISM  
 478 Climate Group 2012), and major land-cover classifications (d; Fry et al. 2011).

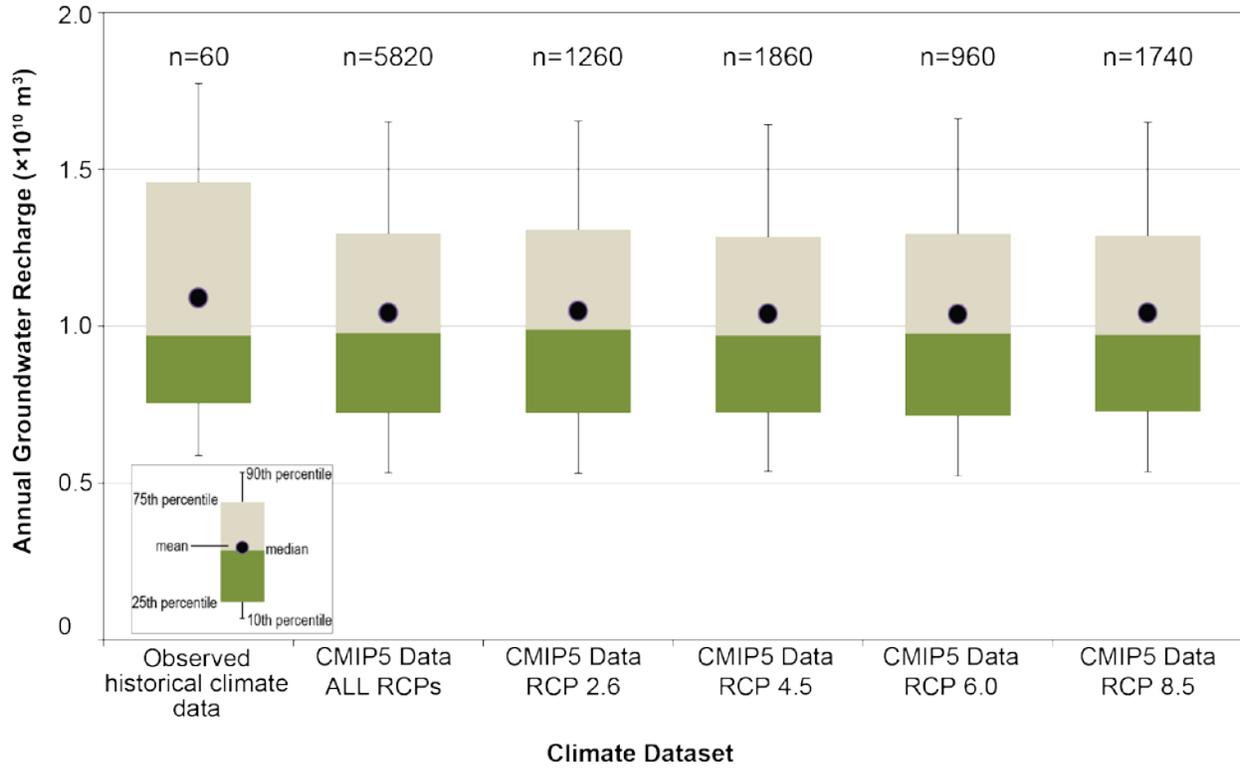
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481 **Fig. 2** Statistics for monthly groundwater recharge in the upper Colorado River basin for water-  
 482 years 1951–2010 simulated with the Soil-Water Balance model using (a) observed climate data  
 483 (Maurer et al., 2002) and CMIP5 GCM climate data (b-f).

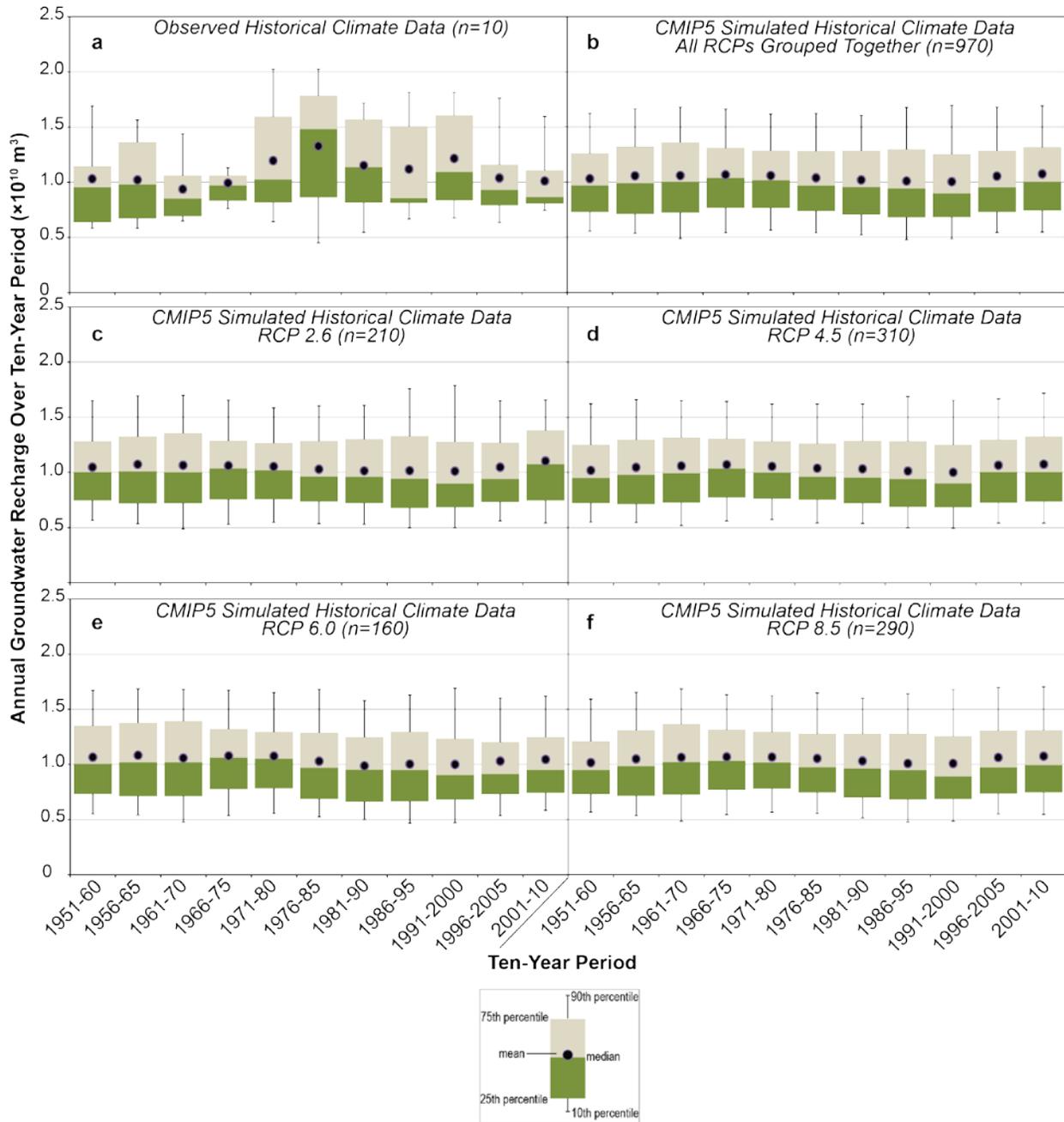
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486 **Fig. 3** Annual groundwater recharge in the upper Colorado River basin for water-years 1951–  
 487 2010 simulated with the Soil-Water Balance model using observed climate data (Maurer et al.,  
 488 2002; left boxplot) and CMIP5 GCM climate data (all other boxplots).

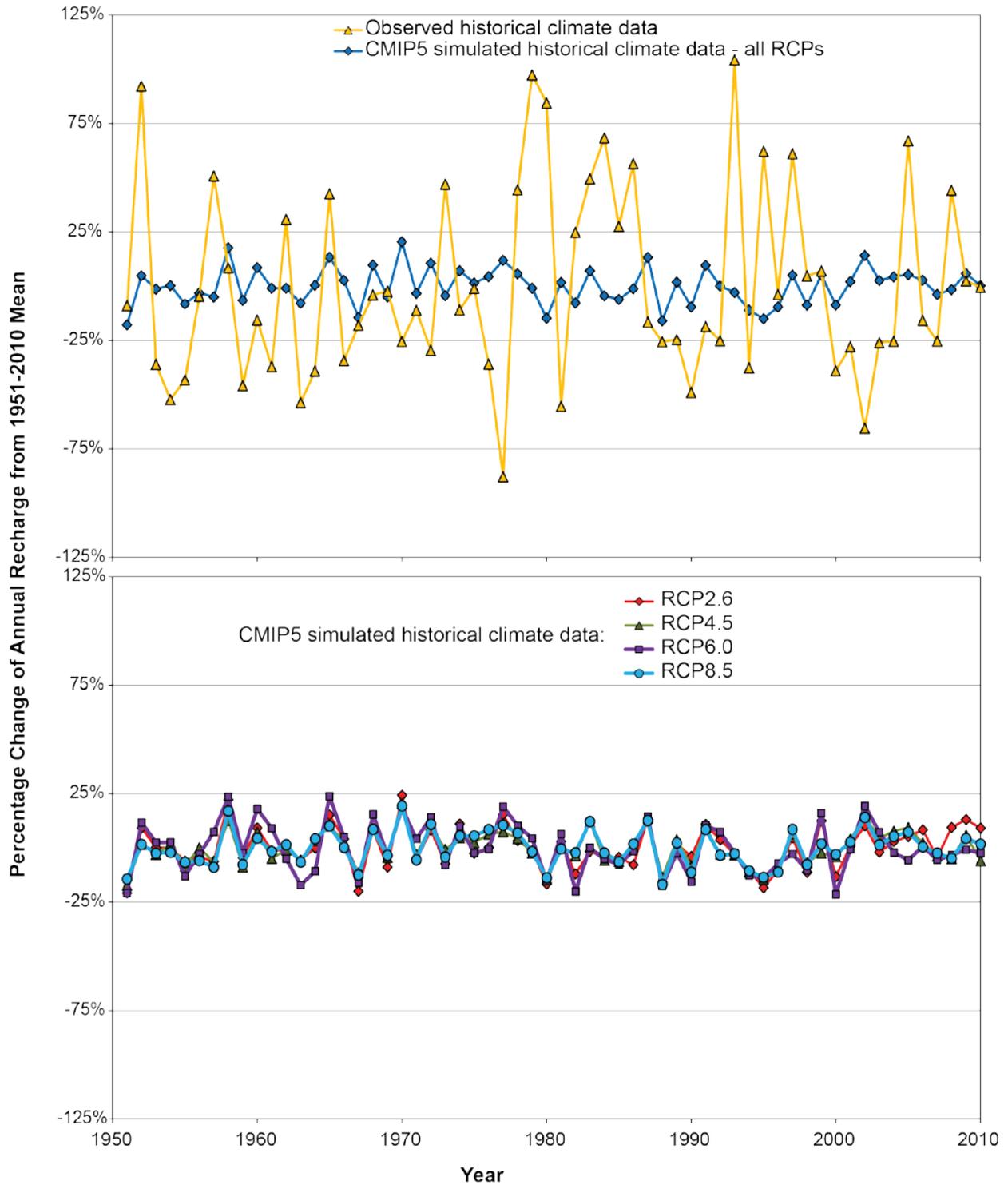
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491 **Fig. 4** Annual groundwater recharge in the upper Colorado River basin over 10-year periods,  
 492 moving every 5 years between water-years 1951 and 2010, simulated with the Soil-Water  
 493 Balance model using (a) observed climate data (Maurer et al., 2002) and CMIP5 GCM climate  
 494 data (b-f).

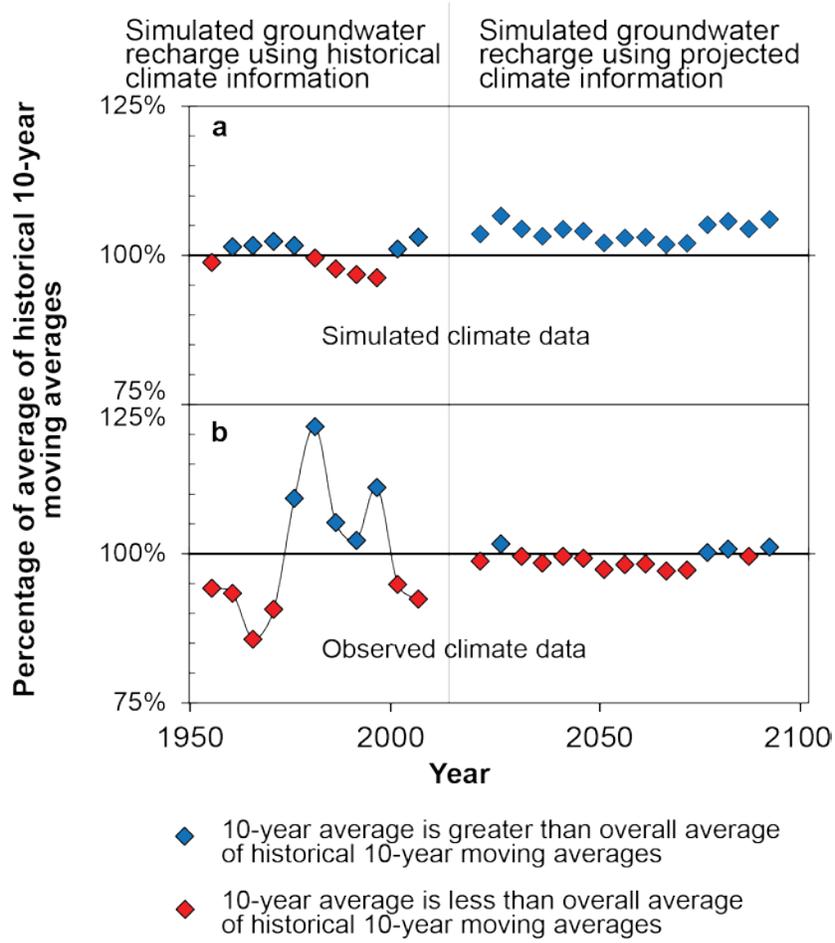
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497 **Fig. 5** Percentage change from 1951–2010 mean annual values of simulated annual recharge  
498 using observed historical climate data and mean simulated annual recharge using CMIP5 climate  
499 data in the upper Colorado River basin.

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502 **Fig. 6** Comparison of ten-year averages, moving every five years, of simulated annual  
 503 groundwater recharge in the upper Colorado River basin using projected climate data to modeled  
 504 recharge using (a) simulated historical climate data and (b) observed historical climate data.  
 505 Results presented as changes in recharge relative to historical average of ten-year means from  
 506 1951–2010.

507