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A proxy-model comparison for mid-Pliocene warm period hydroclimate in the Southwestern US

Sofia Menemenlis^{a,*}, Sarah M. White^b, Daniel E. Ibarra^{c,d}, Juan M. Lora^a

^a*Dept. of Earth and Planetary Sciences, Yale University, New Haven, CT*

^b*Dept. of Geography, UC Berkeley, Berkeley, CA*

^c*Dept. of Earth and Planetary Science, UC Berkeley, Berkeley, CA*

^d*Dept. of Earth, Environmental and Planetary Sciences and the Institute at Brown for Environment and Society, Brown University, Providence, RI*

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*Corresponding author; current institution: Program in Atmospheric and Oceanic Sciences, Princeton University, Princeton, NJ
Email address: smenemenlis@princeton.edu (Sofia Menemenlis)

Abstract

Hydroclimate proxy reconstructions and paleoclimate models of the mid-Pliocene warm period provide insight into how, under a moderate greenhouse warming scenario, Earth-system feedbacks may impact regional hydroclimate. However, in the Southwestern United States there is discord between these two types of information: proxy data have been interpreted to indicate much wetter conditions, while the most recent generation of mid-Pliocene warm period climate models simulates drying. We use a water and energy balance framework to directly compare paleoclimate model output to a refined compilation of proxy records of the presence and areal extent of mid-Pliocene lakes. Within this framework, we quantify uncertainties in the proxy system model parameters and in the interpretation of available proxy records. We find that despite these significant uncertainties, most paleoclimate models simulate a regional balance between precipitation and evaporative demand that could not have sustained the extent of recorded lakes from this time. Moreover, the extensive lakes included as boundary conditions in mid-Pliocene warm period climate models are inconsistent with the regional climate simulated by those same models. This study identifies and quantifies the remaining unknowns in our picture of regional mid-Pliocene warm period hydroclimate, with implications for analyses of climate dynamics during this time.

1. Introduction

Climate models project that global warming and regional drying in the American Southwest will exacerbate ongoing problems of water scarcity, drought, and wildfire in coming decades (e.g., Seager and Vecchi, 2010; Williams et al., 2020), but natural variability on decadal timescales muddles our predictions of the timing and magnitude of anthropogenic changes. Past climate states, as understood through proxy reconstructions and paleoclimate models, provide additional insight into how regional hydroclimatic conditions change over long timescales, and into the forcings and mechanisms responsible for these changes. The mid-Pliocene warm period 3.3-2.9 Ma is the most recent example of the long-term Earth-system response to elevated warmth and near-modern CO₂ concentrations (Tierney et al., 2020); consequently, research efforts in proxy reconstruction and paleoclimate modeling have focused on this time period and region.

In contrast to climate model projections of future drying, published reconstructions of the mid and late Pliocene (3.6-2.6 Ma) hydroclimate indicate that the Western US was generally wetter than present by multiple measures. Compilations of Pliocene flora and faunal records, as well as stable isotope data, have been interpreted as evidence for higher-than-present mean annual precipitation (Molnar and Cane, 2007; Salzmann et al., 2008; Winnick et al., 2013). The presence and size of pluvial lakes record climate-driven changes in water availability over time, so records from outcrops, shorelines, and cores of these

23 lakes provide another archive of past climate conditions. Today, the internally-
24 draining Great Basin region in the Western United States includes a few mod-
25 estly sized lakes, but published records indicate more extensive lakes during
26 the Pliocene between 3.6-2.6 Ma (Pound et al., 2014). Under conditions of in-
27 creased warmth and higher evaporative demand, these lakes could have only
28 been sustained by higher-than-modern precipitation (Ibarra et al., 2018).

29 The Mediterranean-type climate of the western US is shaped by dynamical
30 processes associated with the subtropics and midlatitudes: subtropical highs
31 bring hot, dry summers, while extratropical storm tracks bring most of the re-
32 gion's precipitation during the winter (Seager et al., 2019). The climate mod-
33 eling studies invoked to explain wetter Pliocene conditions in the region have
34 predominantly focused on changes in the meridional Hadley circulation and
35 feedbacks from the tropical Pacific: Burls and Fedorov (2017), using a model
36 tuned to reproduce the reduced meridional and zonal temperature gradients
37 indicated by proxy reconstructions of the early Pliocene, demonstrate that a
38 weakened zonal-mean Hadley circulation would have increased precipitation
39 minus evaporation in subtropical regions between 10-30 degrees latitude, in-
40 cluding in parts of southwestern North America. And, following studies linking
41 reduced temperature gradients to a permanent El Niño-like state during the
42 Pliocene (e.g., Wara et al., 2005; Fedorov et al., 2006), others attribute wetter-
43 than-modern conditions in western North America to extratropical teleconnec-
44 tion patterns similar to those associated with modern El Niño events (Molnar
45 and Cane, 2007; Goldner et al., 2011; Winnick et al., 2013). Although this work
46 has been cited to explain mid-Pliocene warm period conditions and analogize
47 to future warming (Ibarra et al., 2018; Tierney et al., 2020), the magnitude of
48 tropical Pacific sea surface temperature (SST) gradients in the Pliocene is still
49 disputed: some analyses of proxy data argue for only modest reductions in the
50 zonal gradient (Zhang et al., 2014; O'Brien et al., 2014; Tierney et al., 2019),
51 while others argue for considerable reductions (Ravelo et al., 2014; Wycech
52 et al., 2020; White and Ravelo, 2020). Meanwhile, Pliocene climate models are
53 similarly equivocal: both older (Brierley et al., 2015) and more recent gener-
54 ations of models show a general decrease in El Niño-Southern Oscillation am-
55 plitude and a slight reduction in the zonal SST gradient, but do not agree on
56 whether there was a shift to an El Niño-like mean state (Brierley et al., 2015;
57 Oldeman et al., 2021).

58 Moreover, recent modeling studies highlight the influence of mid-Pliocene
59 ice, vegetation, and orography reconstructions, used as boundary conditions
60 in climate models (PRISM4; Dowsett et al., 2016), on terrestrial hydroclimate.
61 Feng et al. (2021) present sensitivity simulations demonstrating that Pliocene
62 ice sheets and vegetation cause wetter conditions in the Sahel and subtropi-
63 cal east Asia, particularly in Boreal Summer (June–September). This influence
64 does not extend to moister conditions in the US southwest. Pliocene boundary
65 conditions instead have a drying effect there: sensitivity simulations using the
66 UofT-CCSM4 Pliocene model demonstrate that Pliocene ice sheets and orogra-
67 phy, applied separately or in combination, alter Northern Hemisphere winter-
68 time stationary waves and divert North Pacific atmospheric rivers away from

69 western North America, leading to a regional decrease in extreme precipitation
70 and drier conditions in the annual mean (Menemenlis et al., 2021). Taken to-
71 gether, modeling studies of the mid-Pliocene hydrologic cycle indicate that the
72 hydroclimate of the Western United States is sensitive to dynamical changes in
73 tropical and extratropical circulation patterns, but that the interplay between
74 these processes during the mid-Pliocene warm period is still ambiguous.

75 The Pliocene climate was more stable than the Pleistocene, but the mid-
76 Pliocene nonetheless experienced variability due to changes in Earth’s obliqu-
77 uity and precessional cycle (Haywood et al., 2002). Orbital changes would
78 have influenced temperatures, precipitation patterns, and seasonality through
79 feedbacks from ice sheets and vegetation (Willeit et al., 2013; Haywood et al.,
80 2013a; Prescott et al., 2014). Given the sensitivity of major spatial features
81 of modeled midlatitude terrestrial hydroclimate to such processes (e.g., Feng
82 et al., 2021; Chan and Abe-Ouchi, 2020; Menemenlis et al., 2021), it is impor-
83 tant to consider how regional hydroclimate might have changed over orbital
84 timescales during the mid-Pliocene warm period. While the lack of precise tem-
85 poral constraints on terrestrial proxies poses a challenge for model-data compar-
86 ison (Haywood et al., 2013a; Salzmann et al., 2013), the timing of pluvial
87 lakes can be assessed with relative accuracy using a combination of paleomag-
88 netic and stratigraphic methods. In cases where proxy records are dated with
89 < 10 kyr accuracy, new climatic interpretations are made possible. For example,
90 Knott et al. (2018, 2019) use updated tephrochronology and $^{40}\text{Ar}/^{39}\text{Ar}$ dating
91 to constrain the ages of South Great Basin lake deposits; they argue that the
92 presence of lakes in Eureka Valley and Death Valley before ~ 3.3 Ma was a re-
93 sult of glacial conditions, and their subsequent absence was caused by warmer
94 and regionally drier interglacial conditions. Such analyses suggest that tempo-
95 ral constraints on proxy records could significantly affect our view of regional
96 hydroclimate conditions and model-data fit.

97 A key aim of the most recent set of coordinated mid-Pliocene warm period
98 modeling experiments (from the Pliocene Model Intercomparison Project ver-
99 sion 2, PlioMIP2) is to reduce uncertainty associated with the “time-slab” ap-
100 proach of PlioMIP1, which aimed to simulate the average of warm interglacial
101 periods from 3.264 – 3.025 Ma. PlioMIP2 targets a “time-slice” of 3.207-3.204
102 Ma during the KM5c interglacial at 3.205 Ma. Using an updated set of boundary
103 conditions, notably including closed Arctic Gateways, these simulations achieve
104 a closer match between the multi-model mean and reconstructed SSTs (Hay-
105 wood et al., 2020). However, while PlioMIP1 simulated wetter conditions in
106 the Western US, PlioMIP2 simulates drying across much of the US Southwest
107 (Haywood et al., 2020), largely driven by a decrease in cool-season precipita-
108 tion (November-April, see Figure S1). This leaves a discrepancy between the
109 general evidence for wetting in the Pliocene western US, with widespread lakes
110 in the region included as boundary conditions in PlioMIP2 models (Pound et al.,
111 2014; Dowsett et al., 2016), and the drying simulated by the PlioMIP2 multi-
112 model ensemble.

113 To understand the nature of and influences on Pliocene hydroclimate in the
114 US Southwest, we first must understand the (dis)agreement between existing

115 proxy reconstructions and paleoclimate model output. The “PMIP triangle” de-
116 scribes three broad causes of model-data discrepancies: uncertainties in proxy
117 data, climate model boundary conditions, and climate model physics (Haywood
118 et al., 2013b, 2016). We ask the following questions. Given unresolved orbital
119 variability in proxy data, does the mismatch between proxies and models stem
120 from differences in the time periods captured by each? Or do PlioMIP2 mod-
121 els, in their reliance on imperfect model boundary conditions and incomplete
122 model physics, underestimate precipitation in the region? Or, is there too much
123 uncertainty in the available evidence to discern either way? Building on Ibarra
124 et al. (2018), we use a proxy-system model approach to compare an updated
125 compilation of proxy evidence for mid-Pliocene lakes to climate model output
126 from PlioMIP2.

127 **2. Methods**

128 *2.1. Regional setting*

129 We focus on a 93,000 km² area of the Great Basin covering much of East-
130 ern California and some parts of Western Nevada (see Supplement Section 1.1).
131 This “South Great Basin” area, shown in Figure 1, encloses a group of contigu-
132 ous watersheds including Owens, China, Searles, Panamint, and Death Valleys,
133 which, under wetter (i.e., Last Glacial Maximum) conditions, formed an inter-
134 connected system of lakes and rivers (Reheis et al., 2014; Knott et al., 2019). At
135 present, the South Great Basin experiences an arid desert and steppe climate,
136 bordered to the northwest by the warm temperate climate of the Sierra Nevada.
137 Most precipitation arrives in the wintertime (Figure S1). The ensemble of ten
138 PlioMIP2 climate models used in this study predicts a 3.4° increase in tempera-
139 ture and a (mostly cool-season-driven) 0.13 mm d⁻¹ decrease in annual-mean
140 precipitation across this region (Figures S1 and S2). Since the mid-Pliocene
141 warm period, tectonic factors have evidently had a secondary influence to cli-
142 matic ones on the overall presence and size of lakes: Since 3 Ma, the Sierra
143 Nevada has risen and tilted toward the Pacific (Mix et al., 2019), deepening the
144 valleys of the South Great Basin to the east (note that this change is not captured
145 by the paleogeographic reconstruction used in PlioMIP2 boundary conditions).
146 Higher Pliocene elevations in the South Great Basin would have slightly affected
147 lake surface evaporation rates (Equation 5); our results are insensitive to these
148 elevation-driven changes in evaporation (Figure S3). And despite uplift of the
149 Sierra Nevada, precipitation isotope data indicate that deflection and blocking
150 leading to the Sierra Nevada rain shadow have remained fairly constant since
151 the Pliocene (Mix et al., 2019).

152 *2.2. Proxy compilation*

153 We assemble a compilation of South Great Basin lakes that existed between
154 3.21-3.20 Ma, drawing on previous compilations of Pliocene hydroclimate proxy
155 data, as well as recently published studies (Table S1). For each lake, we gather

156 paleoenvironmental and stratigraphic information from sediment cores, out-
 157 crops, and geophysical data, with particular attention to dating methods, from
 158 the original literature. To estimate lake areas, we use estimates from the origi-
 159 nal literature or from Pound et al. (2014), where possible. In some cases, we
 160 estimate lake area based on areal extent of the outcropping sediments. Lake ar-
 161 eas in the original literature and from Pound et al. (2014) are based on outcrops
 162 recording lake level highstands; to estimate lowstands for perennial lakes, we
 163 use the area of the modern playa, on the basis that continuous sedimentation
 164 around 3.2 Ma implies that a perennial lake present in the Pliocene (but not
 165 today) occupied the area of its modern playa, at a minimum. In the case of
 166 Amargosa Marsh, we take one half the highstand value as a conservative esti-
 167 mate of lowstand, since widespread outcrops in the region show no evidence of
 168 desiccation.

169 We present three scenarios—“wet,” “intermediate,” and “dry”—to represent
 170 a range of interpretations of lake areas during the 3.21-3.20 Ma period. The
 171 dry scenario is composed only of the lowstands of perennial lakes, defined as
 172 lakes with 1) a continuous stratigraphic record through the 3.21-3.20 Ma in-
 173 terval supported by unambiguous dating, and 2) no evidence of desiccation.
 174 The intermediate scenario includes perennial lakes at their highstands. The
 175 wet scenario includes perennial lakes at their highstand as well as “ephemeral”
 176 lakes, which meet condition 1 but not condition 2, and “potential” lakes. Poten-
 177 tial lakes include those for which there are no Pliocene outcrops or drill cores,
 178 but geophysical data indicate a deep basin fill of probable Pliocene age (e.g.,
 179 Owens Lake). Potential lakes also include those with outcrop evidence of lacus-
 180 trine conditions, but poor dating (e.g., Mono Lake). The difference between the
 181 wet and dry scenarios is thus a measure of uncertainty, reflecting uncertainty in
 182 dating and lake extent, as well as limitations of interpretation and extrapolation
 183 based on available data. Further details of dating and lake area estimation for
 184 each lake are given in Table S1.

185 2.3. Modeling framework

To directly compare between the proxy compilation and climate model data,
 we update the proxy-system model described in Ibarra et al. (2018). This model
 assumes a simplified steady-state balance between volumetric fluxes of water
 into and out of a system of pluvial lakes:

$$\frac{dV}{dt} = \text{Runoff} + \text{Precipitation} - \text{Evaporation} = 0$$

This water balance can be expressed in terms of basin area and lake area, as
 follows:

$$Pk_{run}(A_B - A_L) + PA_L = E_L A_L \quad (1)$$

where P is precipitation, k_{run} is the fraction of P converted to runoff, A_B is the
 area of a terminal basin, A_L is lake area, and E_L is lake evaporation. Rearrang-
 ing equation 1,

$$\frac{A_L}{A_B} (\%) = \frac{Pk_{run}}{E_L - P + Pk_{run}} \times 100. \quad (2)$$

We use the Priestley-Taylor equation (Priestley and Taylor, 1972) to determine E_L . The latent heat flux over the surface of a lake, expressed as LE_L , where L is the latent heat of evaporation and E_L is the rate of evaporation, is determined as follows:

$$LE_L = \alpha \left(\frac{\Delta}{\Delta + \gamma} \right) R_N. \quad (3)$$

The constant α is empirically determined. R_N is the net downward radiation flux at the surface, which can be expressed as the sum of the surface radiation fluxes $R_{s,i} - R_{s,o} + R_{l,i} - R_{l,o}$, where the subscripts s , l , i , and o denote shortwave, longwave, incoming, and outgoing radiation, respectively. Δ is the temperature-dependent slope of the saturation vapor pressure curve in kPa K^{-1} :

$$\Delta = \frac{4098 \left(0.6108 \times \exp \left(\frac{17.27T}{T+237.3} \right) \right)}{(237.3 + T)^2}, \quad (4)$$

where T is temperature. γ is the psychrometric constant, which is elevation-dependent and which we calculate following Allen et al. (1998):

$$\gamma = \frac{c_p p}{\epsilon \lambda} = 0.665 \times 10^{-3} p, \quad (5)$$

where c_p is the specific heat of water at constant pressure, ϵ is the ratio of the molecular weights of water vapor and dry air, λ is the latent heat of vaporization of water, and atmospheric pressure p depends on elevation z . Pressure p is approximated by:

$$p = p_0 \left(\frac{T_0 - \Gamma z}{T_0} \right)^{\frac{-gM}{R\Gamma}}, \quad (6)$$

186 where p_0 and T_0 are reference pressure and temperature, R is the ideal gas
 187 constant, g is gravitational acceleration, M is the molar mass of air, and Γ is the
 188 environmental lapse rate 0.0065 K m^{-1} .

189 The Budyko relationship relates k_{run} to the prevailing long-term climatic
 190 conditions of a catchment. We use the following analytical formulation from Fu
 191 (1981); Zhang et al. (2004):

$$1 - k_{run} = \frac{ET}{P} = 1 + \frac{E_P}{P} - \left[1 + \left(\frac{E_P}{P} \right)^\omega \right]^{1/\omega}. \quad (7)$$

192 E_P is the potential evapotranspiration, which we initially calculate using
 193 the “energy-only” method in which $E_P = R_N/L_V$, where L_V is the latent heat
 194 of vaporization $2264.76 \text{ kJ kg}^{-1}$. This method has been shown to capture E_P
 195 reasonably well in modern climates (Roderick et al., 2014; Scheff and Frierson,
 196 2014), even as compared to more data-intensive methods for estimating E_P
 197 (e.g., Milly and Dunne, 2016). The amount of precipitation converted to runoff
 198 is a monotonic function of the aridity index E_P/P , with variations across space
 199 and time captured by the parameter ω . The Budyko relationship is most reli-
 200 able when considering large areas ($10,000 \text{ km}^2$ or larger) and using long-term
 201 climatological data (Donohue et al., 2007).

202 After finding modern values of A_L/A_B (see next section), we vary annual
203 mean surface temperature and precipitation over reasonable ranges, and calcu-
204 late how this impacts A_L/A_B . Precipitation appears directly in the right-hand
205 side of equation 2 and impacts k_{run} via equation 7. R_N and E_P both increase
206 with temperature, so changes in temperature affect E_L via equation 3, and k_{run}
207 via equation 7. Our approach using absolute changes in precipitation contrasts
208 with Ibarra et al. (2018), who use proportional changes in precipitation. This
209 modification does not affect our conclusions (Figure S4), but using absolute
210 precipitation changes more clearly conveys the magnitude of precipitation in-
211 creases that would be necessary to produce the A_L/A_B values inferred from
212 proxy reconstructions.

213 2.4. Modern lake areas from reanalysis

214 Modern lake areas are the baseline for our forward modeling. We use 41
215 years (1980-2020) of 0.5° latitude \times 0.625° longitude fields of observationally-
216 corrected precipitation, skin temperature, surface radiation, and topography
217 from the Modern-Era Retrospective Reanalysis for Research and Applications,
218 Version 2 (MERRA2; see Supplement Section 1.3). To verify that this product
219 captures the spatial heterogeneity within and around the South Great Basin, we
220 apply the proxy-system model described in the previous section to each grid cell,
221 using representative values of $\omega=2.6$ and $\alpha=1.6$. Although these calculations
222 do not account for uncertainties or spatial differences in model parameters, the
223 model reasonably predicts the presence and extent of lakes across the broader
224 Great Basin region (Figure 1).

225 We use a single A_L/A_B to represent the entire South Great Basin region. We
226 find P , k_{run} , and E_L for each reanalysis grid cell within the South Great Basin
227 region, then take their sum to find an overall basin-normalized lake area for the
228 entire region. This approach is a compromise between the small spatial scales
229 of lake basins and the coarse spatial scales of reanalysis and climate model data
230 (cf. Ibarra et al. (2018)). It implicitly accounts for overflow between grid cells,
231 and allows calculated lake areas to remain accurate even if individual drainage
232 regions within the South Great Basin were reconfigured over time, for example
233 by tectonic changes. However, the model does not see spatial differences in ω
234 or other parameters.

235 When applied to MERRA2 data, the model reproduces the combined area of
236 observed modern lakes in the South Great Basin. The modern A_L/A_B of peren-
237 nial lakes (including Owens Lake, which in the 1920s was desiccated by human
238 activity) is 0.5%, and the combined area of perennial and seasonal/ephemeral
239 lakes is 1.5% (see Supplement Section 1.2). The model parameters needed to
240 reproduce areas in this range are reasonable: the ω range of 2.4-3.1, with a
241 mean of 2.7, compares favorably to the US average of 2.6. And although direct
242 measurements of ω are sparse, this range encompasses empirically determined
243 ω values of 2.4 and 2.6 for two small watersheds on the eastern side of the
244 Sierra adjacent to the South Great Basin region (Greve et al., 2015, see Figure
245 3a).

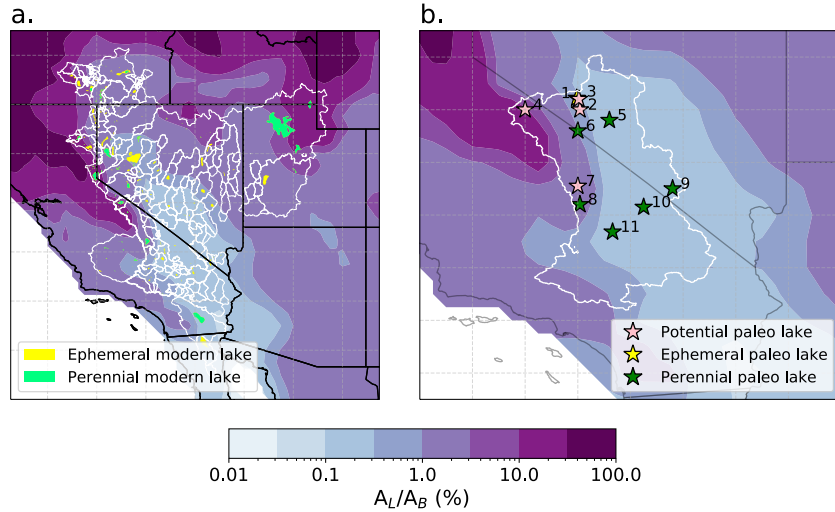


Figure 1: Modern A_L/A_B calculated for each grid cell using MERRA2 reanalysis data, with $\omega=2.6$ and $\alpha=1.26$. (a) Modern drainage areas indicated by thin white lines, and South Great Basin region indicated by thick white lines. Modern perennial lakes, including Owens lake, shown in green; modern ephemeral lakes shown in yellow. (b) Close-up of South Great Basin region, with Pliocene proxy sites marked by pink, yellow, and green stars respectively indicating potential, ephemeral, and perennial lakes. (1) Redlich Summit, (2) Columbus Salt Marsh (3) Rhodes Salt Marsh (4) Mono Lake, (5) Clayton Valley, (6) Fish Lake Valley, (7) Coso Basin, (8) Owens Lake, (9) Amargosa Marsh, (10) Copper Canyon Formation, (11) Searles Lake.

246 2.5. Parameter uncertainty

247 Each model parameter introduces some uncertainty to our estimate of A_L/A_B .
 248 To quantify the combined uncertainty due to these parameters, we estimate
 249 the uncertainty in each parameter, then propagate these uncertainties using a
 250 Monte-Carlo framework.

251 Lake areas are sensitive to the parameter ω . The Fu (1981) analytical for-
 252 mulation of the Budyko relationship (Equation 7) incorporates ω to account
 253 for the observation that runoff amounts are affected not only by P and E_p ,
 254 but also by changes in vegetation, soil properties (i.e., soil type and depth),
 255 topographic characteristics, fire, and the seasonality of P and E (Roderick and
 256 Farquhar, 2011, others). As ω increases (decreases), the amount of precipitation
 257 converted to runoff decreases (increases), and lake areas decrease (increase).
 258 Figure 2a shows how changing ω affects modern South Great Basin A_L/A_B as
 259 determined using MERRA2 reanalysis.

260 Because ω has no definitive physical meaning and is impacted by a suite of
 261 catchment hydrology and vegetation characteristics, Greve et al. (2015) treated
 262 the combined impact of all catchment characteristics bearing on ω as a stochas-
 263 tic process, and identified a right-skewed gamma distribution as representative
 264 of the range of empirical measurements of ω for catchments in the United States.
 265 We use the range of ω values within this distribution that produce reasonable

266 areas for the present-day South Great Basin (blue points in Figure 2b). Changes
267 in ω between the Pliocene and present-day might have had an effect on the
268 fraction of precipitation converted to runoff; we discuss this caveat in section 4.

269 The Priestley-Taylor equation includes an empirically determined parameter
270 α . This constant encodes the extent to which characteristics of the atmospheric
271 boundary layer and overlying atmosphere cause actual evapotranspiration to di-
272 verge from potential evapotranspiration (Priestley and Taylor, 1972), and over
273 water surfaces is determined by the relative transport efficiency of turbulent
274 heat and water vapor (Assouline et al., 2016). Priestley and Taylor (1972) esti-
275 mated $\alpha=1.26$ for saturated surfaces; subsequent observational and theoretical
276 studies confirmed that, for advection-free situations over water, $1.20 < \alpha < 1.30$
277 (Brutsaert, 2005; Assouline et al., 2016). Figure 2a demonstrates that the im-
278 pact of variations in α between 1.2-1.3 is small compared to ω .

279 We parameterize how R_N and E_P scale with temperature. R_N increases
280 with greenhouse warming as surface long-wave radiation increases. Previous
281 studies analyzing climate model output from Coupled Model Intercomparison
282 Project (CMIP) Phase 3 and 5, and from paleoclimate models, find that R_N in-
283 creases by approximately 0.9-1.6% per K warming (Fu and Feng, 2014; Roderick
284 et al., 2014; Ibarra et al., 2018). E_P increases with temperature at a higher rate
285 than R_N . Although R_N contributes to the increase in E_P , the direct physical
286 effects of warming are more consequential: warmer near-surface air tempera-
287 tures increase both saturation vapor pressure and the slope of the saturation
288 vapor pressure curve Δ , resulting in higher rates of evaporation off of a well-
289 watered surface (Scheff and Frierson, 2014; Roderick et al., 2014; Fu and Feng,
290 2014). In their analysis of CMIP5 Representative Concentration Pathway (RCP)
291 8.5 models, Scheff and Frierson (2014) approximate the rate of annual-mean
292 E_P increase per degree of warming as ~ 1.5 -4% globally, with lower (higher)
293 values in warmer (cooler) climates. They estimate that in the western United
294 States, E_P scales with temperature at a rate of ~ 1.8 -2.8% across models, with a
295 multi-model mean value of approximately ~ 2.25 % (Scheff and Frierson (2014)
296 Figure 11). We use these ranges—0.9–1.6% and 1.8%–2.8% for R_N vs. T and
297 E_P vs. T respectively—to represent uncertainty.

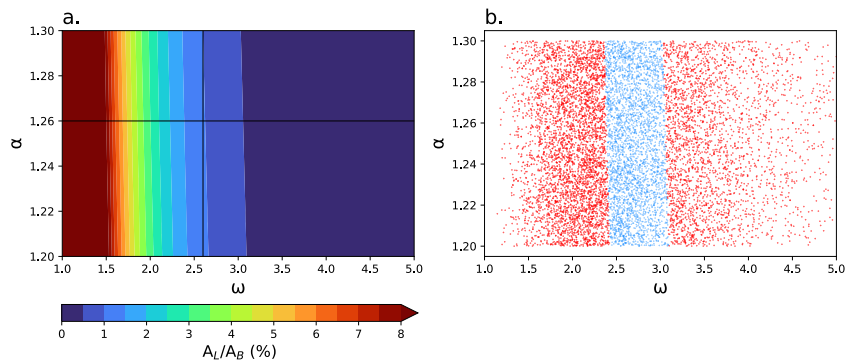


Figure 2: (a) Sensitivity of lake areas to variations in α and ω . (b) 10^4 samples of (ω, α) combinations, sampling ω from the Greve et al. (2015) gamma distribution and α from a uniform distribution between 1.2-1.3. Blue points indicate combinations that produce modern A_L/A_B between 0.5-1.5% of the South Great Basin.

298 3. Results

299 3.1. Lake areas from proxies

300 Our dry, intermediate, and wet scenarios yield lake areas (A_L/A_B) of 1.4%,
 301 3.6%, and 18.7%, respectively. This range of areas suggests a slightly to much
 302 wetter water balance than the modern, where perennial lakes cover 0.5% of the
 303 basin and perennial and ephemeral lakes combined cover 1.5% of basin area.
 304 We note that our dry scenario does not unequivocally represent an absolute min-
 305 imum; since some lakes are reconstructed from cores taken at a single location,
 306 it is possible that they could have recorded lakes with areas smaller than those
 307 included in our dry estimate. This is nevertheless unlikely given the presence of
 308 a number of perennial 3.2 Ma lakes in areas that are presently dry. In fact, our
 309 dry scenario is likely conservative, since 1) we omit Owens Lake despite Owens
 310 Valley being the probable source of water to the Coso and Searles Basins, which
 311 hosted lakes at 3.2 Ma, and 2) the estimate of lake area in Coso Basin is based
 312 on areal extent of outcropping Pliocene sediments, which are tilted and partly
 313 covered by younger sediments. Conversely, our wet scenario is likely generous,
 314 since it includes lakes that may not have existed at 3.2 Ma (potential lakes), and
 315 it assumes that all lake highstands occurred simultaneously. Further details of
 316 dating, lake area estimation, and paleoenvironment for each lake can be found
 317 in Table S1.

318 With our re-compilation effort, we find that the Pound et al. (2014) com-
 319 pilation of Pliocene lakes overestimates lake area in some regions and under-
 320 estimates it in others. Specifically, Pound et al. (2014) include several lakes in
 321 the southern South Great Basin that only existed in the Pleistocene (e.g., Lake
 322 Manix), and omit perennial Pliocene lakes further north and east (e.g., Amar-
 323 gosa Marsh). Additionally, Mono Lake was previously double-counted as both

324 Mono Lake and Lake Russell (the name given to the Pleistocene pluvial lake
325 that existed in the Mono Basin). These discrepancies yield large differences in
326 calculated lake area for the South Great Basin: the dry and wet scenarios from
327 the Pound et al. (2014) compilation are 19.7% and 20.5%, respectively.

328 3.2. Forward modeling and comparison to climate models

329 The observed Pliocene changes in A_L/A_B could have been accomplished by
330 a number of possible combinations of ΔT and ΔP , and uncertainty in our proxy
331 model parameters increases the spread of possibilities. To compare proxy and
332 model results and illustrate the associated uncertainty, we take the following
333 approach. For each combination of possible Pliocene changes in climatological-
334 mean surface temperature (ΔT) and precipitation (ΔP) shown in Figure 3, we
335 perform 10^3 computations of Equation 2, each time selecting ω , α , R_N , and
336 E_P by sampling from the distributions of the regionally valid combinations of
337 α and ω and from uniform distributions of the empirically-determined ranges
338 of changes in R_N (0.9–1.6%) and E_P (1.8%–2.8%) with temperature. Over
339 the space of $(\Delta T, \Delta P)$ we contour the mean values of A_L/A_B from our Monte-
340 Carlo sampling (Figure 3). Additional contours of the 25th and 75th percentiles
341 (thin solid), and of 5th and 95th percentiles (dotted), for each of the three
342 scenarios illustrate the uncertainty introduced by the proxy-system model used
343 to translate from $(\Delta T, \Delta P)$ in each climate model to A_L/A_B . The uncertainties
344 in our proxy-system model are smaller than the uncertainties captured by the
345 differences between dry, intermediate, and wet scenarios.

346 Pliocene minus pre-industrial ΔT and ΔP from those PlioMIP2 models with
347 surface temperature and precipitation fields available are shown as colored dots
348 in Figure 3 (See also Supplement Section 1.3 and Table S2). The multi-model
349 mean, indicated by a star, best matches an A_L/A_B value of 0.6%. Assuming that
350 the probable range of ω remains constant between the Pliocene and modern, as
351 in Figure 3, the majority of PlioMIP2 models predict an A_L/A_B in the South
352 Great Basin lower than the dry scenario of our proxy compilation. Only three
353 models—CESM2, COSMOS, and MIROC4m—fall between the 5th and 95th per-
354 centile contours for the dry scenario. None of the models falls within the 5th
355 and 95th percentile contours for the intermediate scenario.

356 One caveat to this comparison is the possibility of a systematic change in ω
357 across the Southern Great Basin region over time. However, ω would need to
358 decrease to 2.1 for the multi-model mean $(\Delta T, \Delta P)$ to predict A_L/A_B equal
359 to the dry proxy scenario, and to 1.6 to equal the intermediate scenario (Figure
360 S3 c-d). Could a widespread change in catchment conditions have driven such
361 decreases in ω between the mid-Pliocene warm period and the pre-industrial?
362 Vegetation reconstructions from 3.6–2.6 Ma (Molnar and Cane, 2007; Salz-
363 mann et al., 2008; Winnick et al., 2013) suggest expanded woodland and for-
364 est ecosystems in place of today’s dry shrublands. If these records accurately
365 capture conditions during the PRISM4 time-slice, one might expect greener
366 and leafier vegetation, increased leaf area and photosynthetic rates, and con-
367 sequently higher evapotranspiration resulting in *increased* ω values (Donohue

368 et al., 2007; Roderick and Farquhar, 2011). Elevated CO_2 has the effect of de-
 369 creasing leaf transpiration per unit area, so may have the inverse effect and
 370 thus *decrease* ω in humid areas. In water-limited areas, however, leaf area per
 371 unit ground area also increases, in which case elevated CO_2 would likely have
 372 little effect on ω . (Roderick and Farquhar, 2011). Finally, more frequent fires
 373 or a more even distribution of precipitation across the domain or across seasons
 374 would increase ω . On balance, known features of the mid-Pliocene world sug-
 375 gest an ambiguous, perhaps positive, influence on ω —rather than the decrease
 376 in ω that would bring models into closer agreement with proxy reconstructions.
 377 It is therefore unlikely that a change in the range of ω values across the South
 378 Great Basin would narrow the discrepancy between A_L/A_B indicated by our
 379 proxy compilation and by climate models.

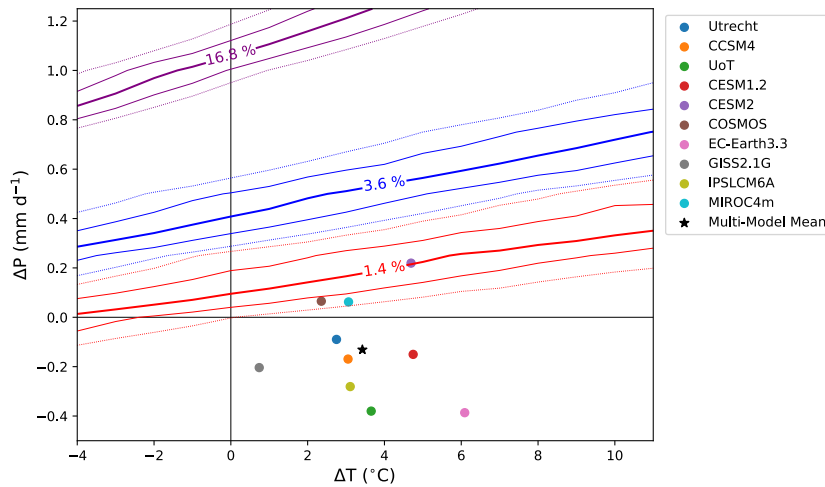


Figure 3: Contours of A_L/A_B for the South Great Basin from our proxy compilation on corresponding Pliocene–pre-industrial changes in surface temperature and precipitation. Red, blue, and purple contours represent A_L/A_B from our dry, intermediate, and wet scenarios respectively. Thick solid lines contour the mean, thin solid lines contour the 25th and 75th percentiles, and dotted lines contour the 5th and 95th percentiles. Colored dots mark $(\Delta T, \Delta P)$ for each PlioMIP2 climate model, and the black star marks the multi-model mean.

380 4. Discussion

381 The considerable differences between our dry, intermediate, and wet scenar-
 382 ios interpreted from proxy records testify to remaining uncertainties in proxy
 383 reconstructions of the presence and extent of mid-Pliocene lakes. We highlight
 384 the need for additional high-resolution (sub-orbital scale) proxy records of mid-
 385 Pliocene lakes, in particular from less complex regions of the domain considered
 386 here such as Fish Lake Valley, Clayton Valley, and Mono Basin. Nonetheless, only

387 three PlioMIP2 climate models produce mid-Pliocene temperature and precipi-
388 tation changes that fall within the 90% confidence interval of the dry scenario,
389 and all fall well below the 90% confidence interval of the intermediate scenario.
390 This indicates that uncertainties in the interpretation of proxy data, though sub-
391 stantial, are not solely responsible for model-data discrepancy in this region.
392 Our analysis therefore suggests that even during the narrow KM5c interglacial
393 time-slice, the water balance in the South Great basin was at least slightly wetter
394 than models predict, and potentially much wetter.

395 From these results we draw two conclusions about model boundary condi-
396 tions. First, and paradoxically, the widespread lakes used in PlioMIP2 boundary
397 conditions (Pound et al., 2014) could not have existed under the dry conditions
398 simulated by PlioMIP2 models themselves. This is both because the lakes used
399 in the PRISM4 reconstruction are too large and numerous, particularly in the
400 southern part of our study area (Figure S5), and also because the regional cli-
401 mate simulated by PlioMIP2 models is too dry (Figure 3). This inconsistency
402 is notable not only from the perspective of model evaluation, but also because
403 these boundary conditions can in turn impact the regional modeled climate.

404 Second, we speculate that errors in more remote boundary conditions may
405 contribute to over-drying of the US Southwest. In the UofT-CCSM4 model,
406 PlioMIP2 ice and orography boundary conditions trigger a wintertime station-
407 ary wave response that causes un-intuitive drying in the western United States
408 (Menemenlis et al., 2021). Most other PlioMIP2 models also show winter-time
409 drying in the South Great Basin (Figure S1), and may be subject to similar mech-
410 anisms. PlioMIP1, which used different boundary conditions from PlioMIP2,
411 did not see drying in this region. Importantly, boundary conditions for ice and
412 orography are themselves subject to large uncertainties. This warrants further
413 refinement of the paleoenvironmental reconstructions used for ice sheet and to-
414 pography boundary conditions, as well as further exploration of their impact on
415 modeled midlatitude dynamics.

416 The KM5c marine isotope stage was selected as the time period within the
417 mid-Pliocene most analogous to future climate because it experienced orbital
418 forcing similar to present-day (Haywood et al., 2013a). In this time slice, cli-
419 mate models simulate drier conditions than indicated by proxy data. These
420 proxy data, in turn, may be interpreted to suggest a water balance either simi-
421 lar to or much wetter than the present-day. It is therefore not obvious whether
422 the mechanisms invoked to explain much wetter Pliocene conditions in this
423 region—namely, tropical ocean feedbacks and teleconnections related to a per-
424 manent El Niño-like state—are accurate analogs for the dynamical response to
425 future global warming. Uncertainties at all three corners of the PMIP triangle re-
426 main considerable. Co-ordinated advancements in terrestrial proxy reconstruc-
427 tion, experimental design, and climate modeling will make possible a more com-
428 plete description of the regional hydroclimate response to mid-Pliocene warm
429 period conditions.

430 5. Conclusions

431 In the American Southwest, we compared mid-Pliocene warm period cli-
432 mate model output to lacustrine proxy records from the same period, with an
433 emphasis on quantifying the uncertainties in the interpretation of proxy records.
434 We assembled a refined compilation of proxy-recorded lakes, calculating areas
435 for a conservative “dry”, an “intermediate,” and a generous “wet” scenario. We
436 used a proxy-system modeling framework to directly compare between basin-
437 normalized lake areas from our proxy compilations, and paleoclimate model-
438 predicted temperature and precipitation in the region. We found that most of
439 the climate models simulate drier conditions than any of the three proxy-based
440 scenarios, and that both models and proxies point to drier conditions in this
441 region than previously thought. While these results cast doubt on the extent to
442 which some analyses of the Pliocene hydrologic cycle are relevant to the narrow
443 3.204–3.207 Ma time slice targeted by PlioMIP2 models, we also highlight sub-
444 stantial remaining uncertainties in our picture of regional hydroclimate during
445 this time.

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Supplement to ``A proxy-model comparison for mid-Pliocene warm period hydroclimate in the Southwestern US''

Sofia Menemenlis¹, Sarah M. White², Daniel E. Ibarra^{3,4}, Juan M. Lora¹

¹ *Dept. of Earth and Planetary Sciences, Yale University, New Haven, CT*

² *Dept. of Geography, UC Berkeley, Berkeley, CA*

³ *Dept. of Earth and Planetary Science, UC Berkeley, Berkeley, CA*

⁴ *Dept. of Earth, Environmental, and Planetary Sciences and the Institute at Brown for Environment and Society, Brown University, Providence, RI*

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1. Supplementary text

1.1 Defining the South Great Basin boundary

The South Great Basin perimeter is defined by the outer boundaries of inwardly-draining basins from the HydroBASINS database (Lehner and Grill, 2013; available from hydrosheds.org). At the northeastern boundary of the basin, we include a sub-region of an adjacent basin, since in this area a recent USGS groundwater model (Brooks et al., 2014) predicts subsurface water to flow toward Clayton Valley, where one of our Pliocene proxy sites is located (see Figure 1 in main text).

1.2 Finding modern lake areas

We draw on published maps, crowdsourced Open Street Map data (Open Street Map, 2021; Open Street Map data copyrighted OpenStreetMap contributors and available from openstreetmap.org), and Google satellite imagery (Google, 2021; Google Satellite Imagery ©2021 TerraMetrics, Map data ©2021 Google) to produce GIS shapefiles of modern perennial and seasonal lakes in the South Great Basin. The two largest modern perennial lakes are Mono Lake and Owens Lake. Although Owens Lake is dry today, it existed perennially until the 1920s, when the city of Los Angeles diverted water for human use (Reheis, 1997; Smith and Street-Perrott, 1983). There are also a number of smaller perennial lakes on the western side of the South Great Basin. Smith (1984) mapped present-day lakes and seasonal playas; we use their Figure 1 as an initial reference for the locations of seasonal playas, then trace more precise shapes using Open Street Map and Google satellite data. In addition to those mapped in Smith (1984), we include several additional playas known to contain water on a seasonal basis.

1.3 Reanalysis and paleoclimate model data

MERRA2 reanalysis data (Gelaro et al., 2017) is available at disc.gsfc.nasa.gov. We use precipitation, skin temperature, surface radiation, and topography fields from the following datasets: “MERRA-2 const_2d_asm_Nx: 2d, constants V5.12.4”, “MERRA-2 tavg1_2d_flux_Nx: 2d, 1-Hourly, Time-Averaged, Single-Level, Assimilation, Surface Flux Diagnostics V5.12.4”, and “MERRA-2 tavg1_2d_rad_Nx: 2d, 1-Hourly, Time-Averaged, Single-Level, Assimilation, Radiation Diagnostics V5.12.4” (GMAO, 2015a-c). MERRA2 corrects modeled precipitation with observational land surface precipitation data (Reichle et al., 2017). This approach also lessens bias in surface radiation terms as precipitation affects the partitioning between sensible and latent heating, particularly in areas where surface latent heating is moisture-limited (Draper et al., 2018).

We use output from 10 models participating in the Paleoclimate Model Intercomparison Project, Version 2 (PlioMIP2, Haywood et al. 2020). The modeling groups have archived monthly output for precipitation and surface temperature at the PlioMIP2 data repository at the University of Leeds. To find Pliocene minus pre-industrial anomalies in precipitation and surface temperature, we took the difference between the “Eoi400” and “E280” runs for each model. For additional detail regarding PlioMIP2 boundary conditions, experimental design, and results, see Dowsett et al. (2016), Haywood et al. (2016), and Haywood et al. (2020). For a summary of and references to the models used in this study, see Table S2.

For consistency, we interpolate all reanalysis and paleoclimate model data to a uniform 0.25° latitude by 0.25° longitude grid. Note that the PlioMIP2 models were run with different spatial resolutions (Table S2).

2. Supplementary figures

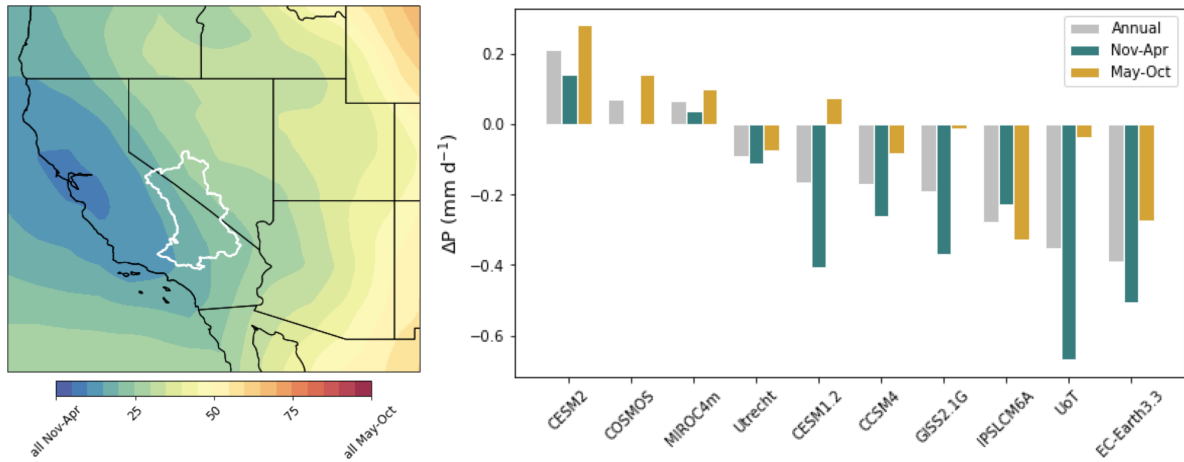


Figure S1. Contours: seasonality of precipitation in multi-model-mean pre-industrial simulations. Colors indicate the percent contribution of warm-season (May-October) precipitation to the annual-mean. Bar graph: Pliocene minus pre-industrial precipitation for the South Great Basin, annually and for each half-year, for each Pliocene MIP2 model used in our study.

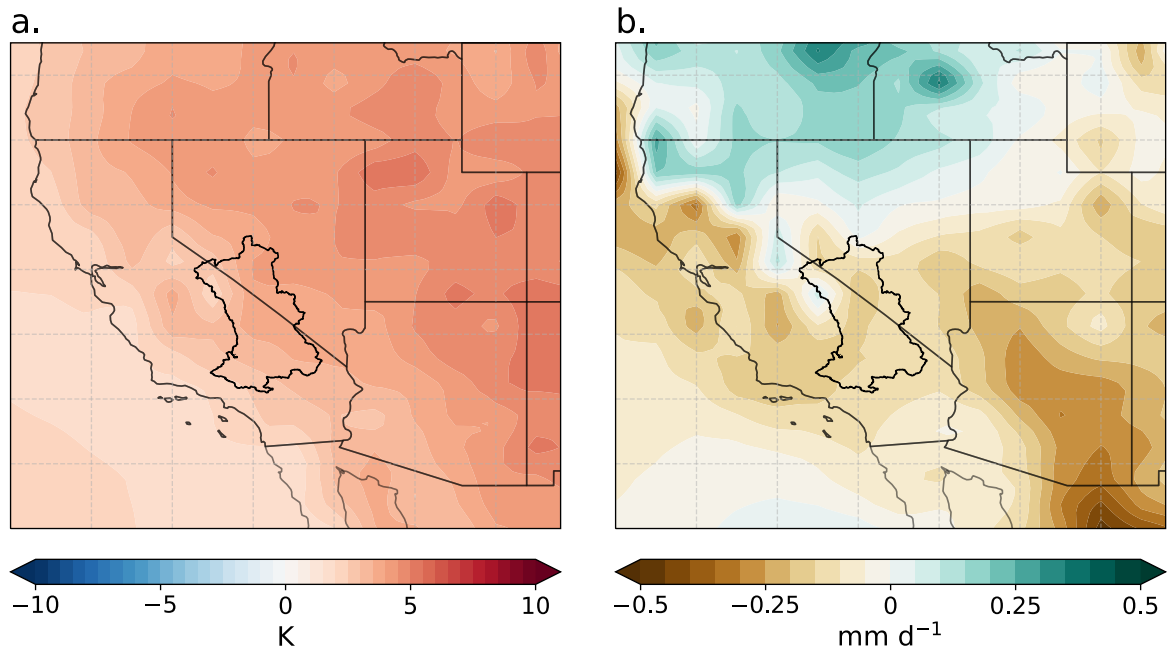


Figure S2. Pliocene minus pre-industrial change in (a) surface temperature, and (b) precipitation, for the ensemble mean of the models used in our analysis.

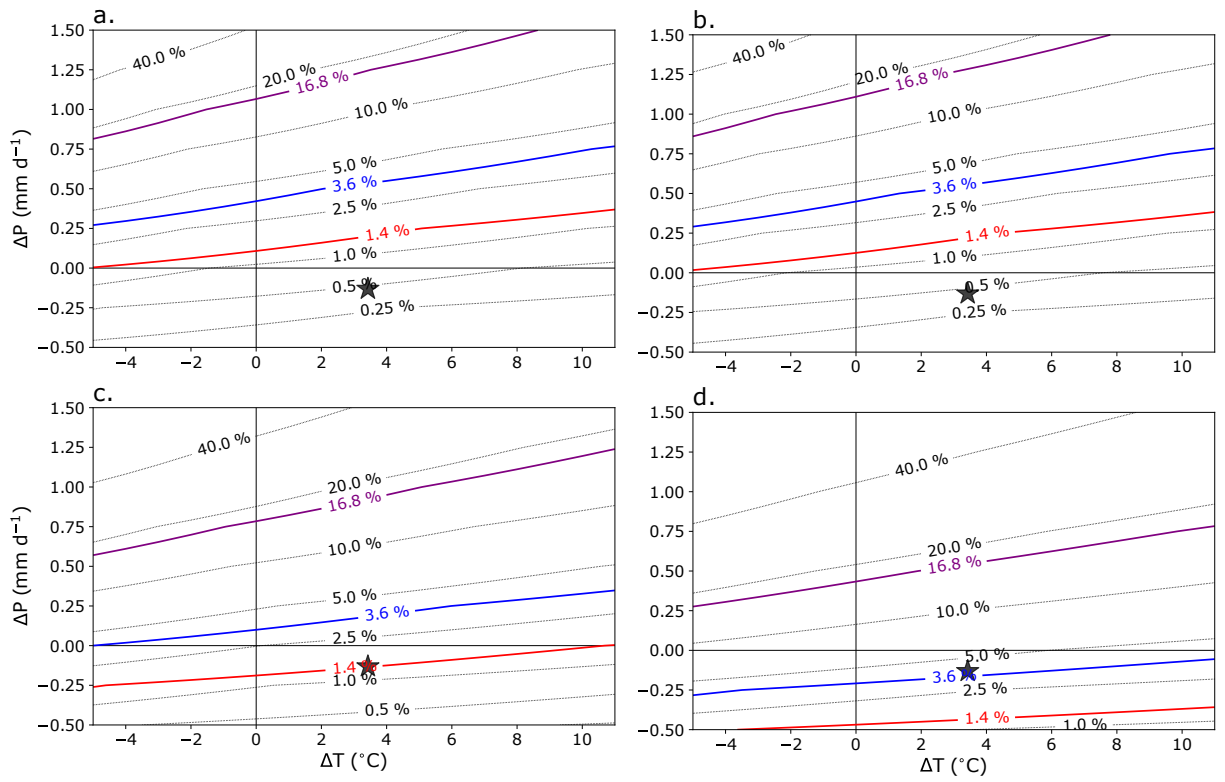


Figure S3. (a) Contour plot of South Great Basin lake areas calculated using the following parameters: $R_N/T = 1.25\%/K$, $E_P/T = 2.3\%/K$, $\omega = 2.7$, $\alpha = 1.25$. (b) Same as (a), but incorporating the on evaporation (via the psychrometric constant, Equation 5 in main text) of increasing Pliocene elevations by 1000 m against modern. (b) Same as (a), but with $\omega = 2.1$. This is the ω value at which the multi-model mean $(\Delta T, \Delta P)$ would correspond with the dry proxy scenario. (c) Same as (a), but with $\omega = 1.6$. This is the ω value at which the multi-model mean $(\Delta T, \Delta P)$ would correspond with the intermediate proxy scenario.

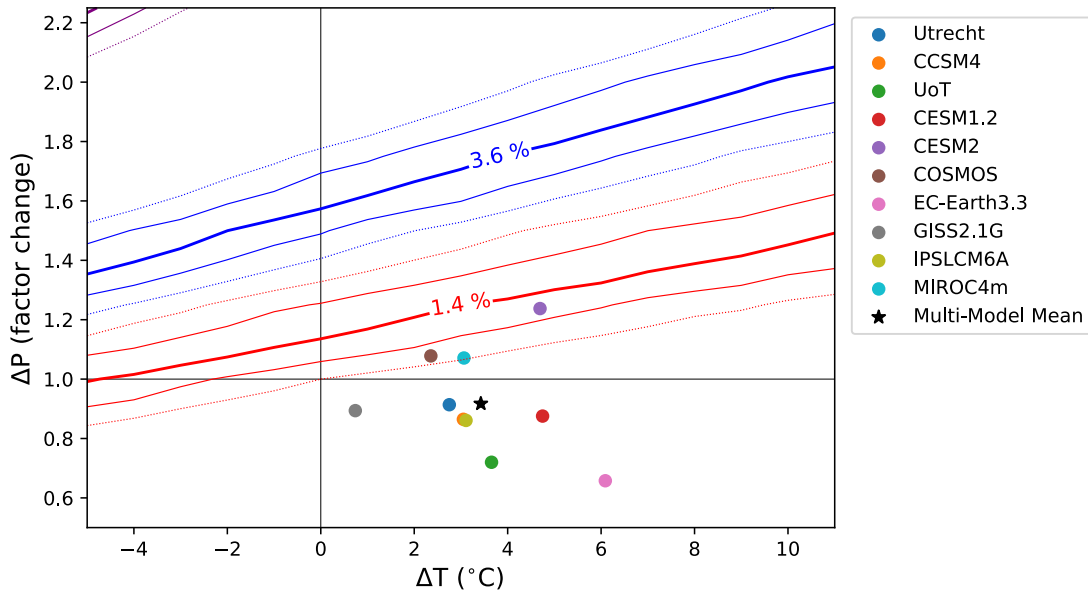


Figure S4. Similar to Figure 3, but adjusting precipitation by a scaling factor rather than by absolute values. Analogous to modeling results presented by Ibarra et al. (2018).

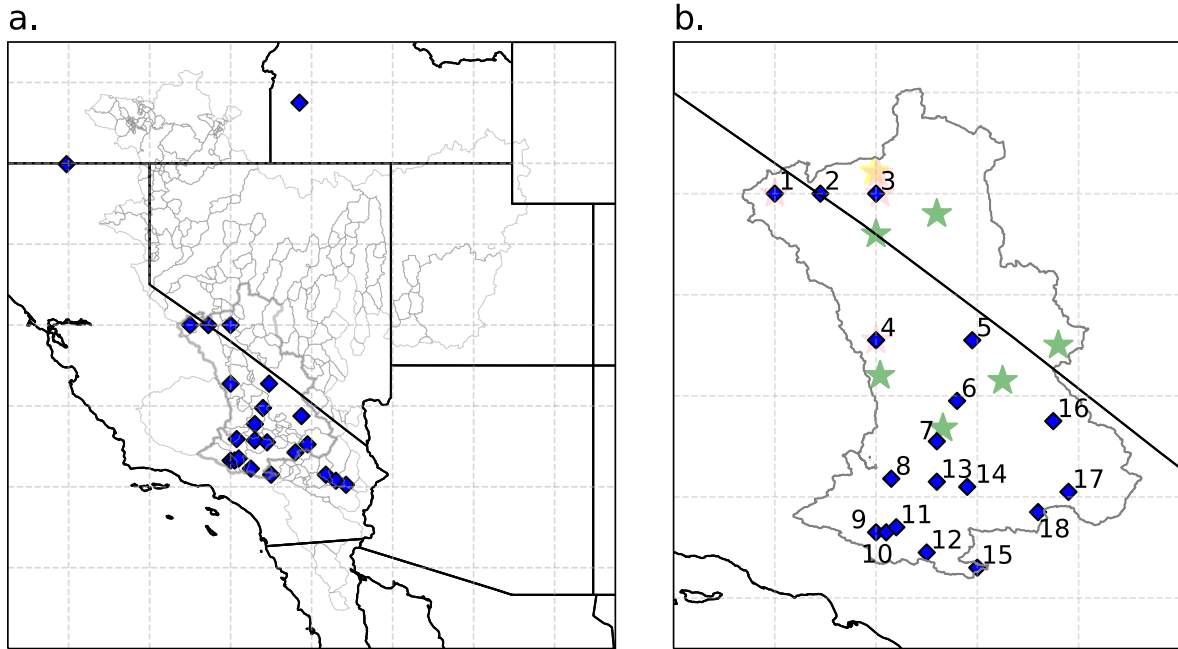


Figure S5. Blue diamonds indicate proxy locations from Pound et al. (2014). From (b): (1) Lake Russel (2) Mono Lake (3) Rhodes-Clayton-Fish Valleys, NV (4) Owens Lake (5) Lake Manly (6) Panamint Lake (7) China-Searles lake (8) Kochn Lake 1 (9) Harper 1 (10) Harper 2 (11) Harper 3 (12) Harper 4 (13) Kochn Lake 2 (14) Kochn Lake 3 (15) South Mojave (16) Lake Tecopa (17) Lake Mojave (18) Lake Manix. Stars show our proxy sites as in Figure 1 in the main text.

3. Supplementary tables

Table S1. South Great Basin lake areas from proxies. Dates based on paleomagnetic data are updated to the most recent paleomagnetic timescale (Gradstein et al., 2012), and dates based on correlations to known tephtras are updated to the most recent ages (Knott et al., 2018). Minimum lake areas are not calculated for potential or ephemeral lakes, since they are only included in the wet map. "Sed."=sediment, "correl."=correlated, 'paleomag.'=paleomagnetic data. Ages based on extrapolation assume a constant sedimentation rate from some datum to the bottom of basin fill, determined from geophysical data. Ages of tuffs: Bishop Tuff=0.772 +/- 0.008 Ma, Huckleberry Ridge Tuff=2.101 +/- 0.007 Ma, Putah Tuff=3.31 Ma, Mesquite Springs=3.32 Ma, Zabriskie Wash=3.335 +/- 0.002 Ma (Knott et al., 2018). Ages of paleomagnetic subchrons and transitions: Kaena=3.032-3.116 Ma, Mammoth=3.207-3.330 Ma, Gauss-Gilbert=3.596 MA (Ogg et al., 2012). NALMA = North American Land Mammal Ages, based on widely recognized type fossils. Most recent timescale from Bell et al. (2004). Hemphillian/Blancan boundary=4.6-5.2 Ma (Bell et al., 2004), 4.98-5.0 in eastern NV (Lindsay et al., 2002). Top of Blancan=2.5-1.9 Ma (Bell et al., 2004).

Site name	Category	Age min (Ma)	Age max (Ma)	Lat	Lon	Lake area (km ²)	Min lake area (km ²)	Modern setting, proxy record type	Lake setting, stratigraphy, etc.	Basis for categorization	Age constraints	Reference for ages and lake setting	Reference for lake area
Redlich Summit	Ephemeral lake	2	6	38.22	-118	9687		Dry; outcrops	Modern divide between Columbus and Rhodes Salt Marshes. Pliocene lake sed: green mudstone, deltaic sandstone, diatomaceous siltstone, nearshore sandstone.	At least two desiccation surfaces indicate lake was ephemeral.	Interbedded tephra correl. to those at 5.9 Ma and 2.2-2.5 Ma, diatom assemblages	Reheis et al., 2002	Pound et al., 2014 (area of their "Rhodes-Clayton-Fish" valley, minus areas of Clayton Valley and Fish Lakes)
Columbus Salt Marsh	Potential lake	0	7.8	38	-118			Playa; sed. core and geophys. data	If Redlich Summit was wet, then Columbus Salt Marsh was also wet	Unclear if lake existed during KM5c; lake inferred from geophys. data, extrapolation, and lake at	Extrapolation (Bishop Tuff to bottom of basin fill)	Uranium Resources Inc. report, Reheis et al., 2002	Counted with Redlich Summit

										Redlich Summit			
Rhodes Salt Marsh	Potential Lake	2	6	38.2	-118			Brine pools; outcrops	If Redlich Summit was wet, then Rhodes Salt Marsh was also wet. Limited outcrops of lake sediments are similar to Redlich Summit.	Unclear if lake existed during KM5c; detailed regional stratigraphy is lacking, few outcrops	Shared tephras with Redlich Summit	Reheis et al., 2002	Counted with Redlich Summit
Mono Lake	Potential Lake	0	4	38	-119	1125		Lake; outcrops	Volcanically active basin since 4 Ma. Pliocene lake sed: diatomite, oolitic sand, fine-grained sand and silt, with freshwater fish, gastropods, mollusks, microfossils, and a few mammals.	Unclear if lake existed during KM5c; dates are old with large uncertainty, detailed stratigraphy is lacking.	Fish correlated to Pliocene Glens Ferry Fm., K-Ar ages on volcanics at/near top of lake sediments (3.3 Ma, 3.4 Ma, 3.6 Ma)	Gilbert et al., 1968 (volcanics); Miller and Smith 1981 (fish)	Pound et al., 2014
Clayton Valley	Perennial lake	1.8	3.4	37.8	-117	663	90	Small playa before mining; outcrops, sed. cores	Most paleoenvironmental information is from sed. cores. Pliocene lake sed: green clay with organic-rich layers, fine to coarse silt and sand, rare halite.	Evidence of lacustrine conditions continuously through the mid-Pliocene.	Extrapolation (Bishop Tuff and Huckleberry Ridge tuff to bottom of basin fill)	Coffey et al., in review	Rush 1968 report; min from modern playa

Fish Lake Valley	Perennial lake	2.85	3.4	37.6	-118	450	188	Small playas; outcrop	Part of the modern drainage to Columbus Salt Marsh. Pliocene lake sed: gypsiferous green mudstone, siltstone, sandstone, conglomerate.	Evidence of lacustrine conditions continuously through the mid-Pliocene	Interbedded tephra correl. To Zabriskie Wash and Putah Tuffs, paleomag. (Kaena and Mammoth subchrons)	Reheis et al., 1991; Reheis et al., 2002; updated tephra correlations and ages from Knott et al., 2019	Reheis et al., 1993; min from modern playa
Owens Lake	Potential lake	0.8	4.5	36.55	-118	1500		Lake (before 20 th century draining) ; sed. core and geophys. data	Drains into Searles Lake via Owens River during wet intervals. Bottom of oldest sed. Core is ~800 ka; core is underlain by deep basin fill.	Unclear if lake existed during KM5c; lake inferred from geophys. Data, extrapolation, and filling of Searles Lake	Extrapolation (Bishop Tuff to bottom of basin fill)	Smith and Bischoff, 1997	Pound et al., 2014
Coso Basin	Perennial lake	2	6	36.2	-118	143	143	Dry; outcrops	Immediately south of Owens Lake. Regional geophysical data imply either 1) Coso and Owens were separate Pliocene lakes, or 2) Coso Lake was ancestral to Owens Lake.	Evidence of lacustrine conditions continuously through the mid-Pliocene.	18 K-Ar ages and 2 fission track ages on volcanics, NALMA (Hemphillia and Blancan fossils)	Bacon et al., 1982. Kamola and Walker [1999]'s age model places lake pre-KM5c, but based on less evidence than Bacon et al.	Estimated from areal extent of Coso Fm; an underestimate since Coso Fm. is partly buried, so min area is not calculated.
Amargosa Marsh	Perennial lake	2.4	3.35	36.5	-116	1250	625	Small playas; outcrops, sed. cores	Sed. indicates marshland, spring-fed ponds, and playas, with gastropods,	No evidence for dessication from numerous sed. cores and outcrops.	Tephra: K-Ar date near top of sed is 2.1 +/- 0.4 Ma, tephra near base of	Hoover 1989, Hay et al., 1986, Knott et al., 2018	Hoover 1989. Min extent is 1/2 max, from widespread evidence for

									ostracodes, and bivalves.		sed. correl. to tuff of Zabriskie Wash		wet conditions from sed. cores
Copper Canyon Fm, Death Valley	Perennial lake	3.15	4.33	36.1 5	-117	1	1	Dry; outcrops	Copper Canyon Fm. is only accessible by guided tour, so is excluded from most discussions of Pliocene lakes in the region	Evidence of fresh to saline spring-fed lake continuously through the mid-Pliocene	Ar-Ar dates on basalts bracketing lake sed. (4.33 Ma, 3.1 Ma), paleomag. (Gilbert- Gauss, Mammoth and Cochiti subchrons)	Nyborg 2011	Estimated from areal extent of Copper Canyon Fm.
Searles Lake	Perennial lake	2.5	3.4	35.6 8	-117	800	288	Playa and brine pools; sed. core	Modern terminus of Owens River. Paleoenviro nmental information from sediment core KM-3, first described by Smith et al., 1983	Evidence of lacustrine conditions continuously through the mid-Pliocene	Paleomag. (Mammoth subchron), tephra correl. to tuffs of Mesquite Springs and Zabriskie Wash	Smith et al., 1983; Knott et al., 2018	Pound et al., 2014 (their "China- Scarles lake"). Min from modern Searles playa (133 km ²) + adjacent China basin playa (155 km ²)

Table S2. Details of PlioMIP2 models used in this analysis.

Group	Model	Atmosphere Resolution	Ocean Resolution	South Great Basin ΔT_s (C)	South Great Basin ΔP (mm d⁻¹)	Reference
University of Utrecht, the Netherlands	CESM 1.0.5 (CCSM4) – Utrecht version	2.5° × 1.9°	~1°	2.75	-0.09	Baatsen et al., 2021, in prep.
National Center for Atmospheric Research, USA	CCSM4	~1°	~1°	3.05	-0.17	Feng et al. 2020
University of Toronto, Canada	CESM 1.0.5 (CCSM4) – UoT version	~1°	~1°	3.65	-0.38	Chandan and Peltier 2017, 2018
National Center for Atmospheric Research, USA	CESM1.2	~1°	~1°	4.75	-0.15	Feng et al. 2020
National Center for Atmospheric Research, USA	CESM2	~1°	~1°	4.69	0.22	Feng et al. 2020
Alfred Wegener Institute, Germany	COSMOS	3.75° × 3.75°	3.0° × 1.8°	2.36	0.07	Stepanek et al., 2020
Stockholm University, Sweden	EC-Earth3.3	1.125° × 1.125°	1.0° × 1.0°	6.09	-0.39	Zheng et al., 2019
Goddard Institute for Space Studies, USA	GISS2.1G	2.0° × 2.5°	1.0° × 1.25°	0.74	-0.20	Kelley et al., 2020
Institut Pierre-Simon Laplace	IPSLCM6A	2.5° × 1.26°	1.0° × 1.0°	3.11	-0.28	Lurton et al., 2020
Japan Agency for Marine-Earth Science and Technology, Japan	MIROC4m	~2.8° × 2.8°	0.5° – 1.4° × 1.4°	3.06	0.06	Chan and Abe-Ouchi, 2020

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