

1 **Resilient California water portfolios require infrastructure investment**  
2 **partnerships that are viable for all partners**

3 **Andrew L. Hamilton<sup>1,2\*</sup>, Harrison B. Zeff<sup>1,2</sup>, Gregory W. Characklis<sup>1,2</sup>, Patrick M. Reed<sup>3</sup>**

4 <sup>1</sup>Department of Environmental Sciences and Engineering, Gillings School of Global Public  
5 Health, University of North Carolina at Chapel Hill, Chapel Hill, NC, USA

6 <sup>2</sup>Center on Financial Risk in Environmental Systems, Gillings School of Global Public Health,  
7 UNC Institute for the Environment, University of North Carolina at Chapel Hill, Chapel Hill,  
8 NC, USA

9 <sup>3</sup>Department of Civil and Environmental Engineering, Cornell University, Ithaca, NY, USA

10 \*Corresponding author: Andrew Hamilton ([andrew.hamilton@unc.edu](mailto:andrew.hamilton@unc.edu))

11 **Key Points**

- 12 • Water portfolio planning frameworks need to account for heterogeneity across  
13 participating partners, not just “average” performance  
14 • Exploratory modeling shows how most infrastructure partnerships are highly unequal in  
15 their water deliveries and financial risks  
16 • Viable partnership design is shown to be especially difficult for dry hydrologic scenarios,  
17 or when building conveyance without storage  
18

## 19 **Abstract**

20 Water scarcity is a growing problem around the world, and regions such as California are  
21 working to develop diversified, interconnected, and flexible water supply portfolios. To meet  
22 their resilient water portfolio goals, water utilities and irrigation districts will need to cooperate  
23 across scales to finance, build, and operate shared water supply infrastructure. However,  
24 planning studies to date have generally focused on partnership-level outcomes (i.e., highly  
25 aggregated mean cost-benefit analyses), while ignoring the heterogeneity of benefits, costs, and  
26 risks across the individual investing partners. This study contributes an exploratory modeling  
27 analysis that tests thousands of alternative water supply investment partnerships in the Central  
28 Valley of California, using a high-resolution simulation model to evaluate the effects of new  
29 infrastructure on individual water providers. The viability of conveyance and groundwater  
30 banking investments are as strongly shaped by partnership design choices (i.e., which water  
31 providers are participating, and how do they distribute the project's debt obligation?) as by  
32 extreme hydrologic conditions (i.e., floods and droughts). Importantly, most of the analyzed  
33 partnership structures yield highly unequal distributions of water supply and financial risks  
34 across the partners, limiting the viability of cooperative partnerships. Partnership viability is  
35 especially rare in the absence of groundwater banking facilities, or under dry hydrologic  
36 conditions, even under explicitly optimistic assumptions regarding climate change. These results  
37 emphasize the importance of high-resolution simulation models and careful partnership structure  
38 design when developing resilient water supply portfolios for institutionally complex regions  
39 confronting scarcity.

## 40 **1 Introduction**

41 In May 2021, California Governor Newsom announced a \$5.1 billion package for  
42 “immediate drought response and long-term water resilience investments” (Office of Governor  
43 Gavin Newsom, 2021). This follows the administration’s Water Resilience Portfolio Initiative  
44 (WRPI), an ambitious blueprint for bolstering the state’s water security (Newsom et al., 2020).  
45 The WRPI recommends a suite of actions to overcome challenges such as population growth,  
46 groundwater overdraft, and aging infrastructure, as well as climate change, which is already  
47 making droughts more frequent and severe (AghaKouchak, Cheng, Mazdidasni, & Farahmand,  
48 2014; AghaKouchak et al., 2021; Berg & Hall, 2017; Diffenbaugh, Swain, & Touma, 2015).  
49 Focal point recommendations in the WRPI include (1) expanding, improving, and diversifying  
50 the state’s water storage and conveyance infrastructure, (2) developing flexible institutions for  
51 water sharing (e.g., groundwater banking), and (3) preparing for more climate change-related  
52 extreme droughts and floods. The WRPI envisions a future in which separate agencies, utilities,  
53 and stakeholder groups collaboratively develop and manage a shared network of water  
54 infrastructure that bridges local, regional, and statewide scales. However, at present, it is not  
55 clear that planners have the tools they need to build this “cohesive, resilient ‘*water system of*  
56 *systems*’ across California” (Newsom et al., 2020). In this work, we show that traditional water  
57 supply planning tools are unsuitable for the task at hand. Exploratory modeling contributes new  
58 insights for designing and evaluating collaborative water investment partnerships under  
59 uncertainty. These insights have broad relevance beyond California, including the entire Western

60 U.S. and other water-scarce regions around the world seeking to develop more resilient water  
61 supply systems.

62 Water supply planning analyses generally rely on simulation models to evaluate the  
63 impacts of alternative policies and investments. However, modern supply networks in regions  
64 such as California are extremely complex, both in terms of the engineered system of reservoirs,  
65 canals, and groundwater recharge facilities, as well as the institutional systems of environmental  
66 regulations, water rights, and groundwater banking arrangements (Escriva-Bou, Mccann, et al.,  
67 2020). Exacerbating these complexities, atmospheric rivers deliver a large fraction of the state's  
68 annual precipitation during a few short events (Dettinger, Ralph, Das, Neiman, & Cayan, 2011;  
69 Gershunov, Shulgina, Ralph, Lavers, & Rutz, 2017), introducing strong interdependencies  
70 between floods and droughts. This makes it critical to resolve daily-scale dynamics (Hanak,  
71 Jezdimirovic, et al., 2018; Kocis & Dahlke, 2017; Malek et al., 2022; Zeff et al., 2021), while  
72 simultaneously multi-decadal simulations are needed to properly evaluate the impacts of long-  
73 term infrastructure investments, the slow dynamics of groundwater storage change (Manna,  
74 Walton, Cherry, & Parker, 2019), and the deep uncertainties in climatic, economic, and  
75 regulatory changes. Lastly, it is critical that planning models resolve a wide range of spatial  
76 scales and system actors in order to evaluate how water moves through statewide infrastructure  
77 networks in response to local actions by individual water utilities, irrigation districts, and water  
78 storage districts (Zeff et al., 2021) (hereafter referred to collectively as "water districts"). To  
79 date, much of this complexity is beyond the reach of the state's primary water resources planning  
80 models (e.g., CalSim (California Department of Water Resources, 2017; Draper et al., 2004),  
81 CalLite (Islam et al., 2011), CALVIN (Draper, Jenkins, Kirby, Lund, & Howitt, 2003)).

82 California has over \$33 billion of water-related expenditures per year, 85% of which are  
83 funded by local agencies (Hanak, Chappelle, et al., 2018). Despite recent high-profile  
84 commitments to water infrastructure from state and federal governments (Office of Governor  
85 Gavin Newsom, 2021; The White House, 2021), individual water districts are responsible for the  
86 vast majority of the funding. Water districts typically finance large capital projects through debt  
87 that must be repaid over decades using water sales revenues. A key question is whether the  
88 additional water gained from a project will generate sufficient revenues to cover debt payments  
89 without requiring budget cuts, unpopular customer rate hikes, or even bankruptcy (Chapman &  
90 Breeding, 2014; Jeffrey Hughes et al., 2014; Leurig, 2010). However, benefits can be difficult to  
91 assess in light of system complexities and uncertainties as described above. Moreover, as the  
92 state's water supply portfolio becomes more diverse, flexible, and inter-connected, it becomes  
93 increasingly difficult for water districts to evaluate individual capital projects due to interactions  
94 across the infrastructural and institutional networks (Haines, 2018).

95 Cooperative finance and operation of infrastructure can benefit water districts through  
96 economies of scale, reduced redundancy, and increased flexibility (Escriva-Bou, Sencan, Hanak,  
97 & Wilkinson, 2020; Jeff Hughes & Fox, 2019; Riggs & Hughes, 2019). Larger and more diverse  
98 coalitions may also be better positioned to raise capital and harness state and federal subsidies  
99 (Cypher & Grinnell, 2007; Hansen, Mullin, & Riggs, 2020; Newsom et al., 2020). However,  
100 cooperation introduces significant new complexities related to the heterogeneity of outcomes  
101 across the participants. Capital projects that look favorable at the partnership level may yield

102 poor results for individual partners due to differences in water rights, groundwater recharge  
103 capacities, and locations within the infrastructure network. Some partners may also bear an  
104 outsized share of the risk in unfavorable future scenarios (i.e., losses under climate change)  
105 (Gold, Reed, Trindade, & Characklis, 2019; Gorelick, Zeff, Hughes, Eskaf, & Characklis, 2019;  
106 Herman, Zeff, Reed, & Characklis, 2014). An additional challenge is assigning the share of  
107 project debt to be borne by each partner; standard approaches for apportioning cost shares to  
108 multiple beneficiaries are unsuitable when there is significant uncertainty or a large number of  
109 potential partners (De Souza, Medellín-Azuara, Lund, & Howitt, 2011; Giglio & Wrightington,  
110 1972). Recent water portfolio planning studies (San Joaquin River Restoration Program, 2011;  
111 Sunding, 2015; U.S. Bureau of Reclamation, 2017, 2020) have focused on expected  
112 costs/benefits at highly aggregated levels, typically under minimal uncertainty, while  
113 vulnerability assessments under broader uncertainty have focused on individual water districts  
114 (Groves et al., 2015; Lempert & Groves, 2010; Tariq, Lempert, Riverson, Schwartz, & Berg,  
115 2017) or aggregate regional outcomes (Connell-Buck, Medellín-Azuara, Lund, & Madani, 2011;  
116 Schwarz et al., 2019, 2018; Selmon et al., 2019). There is little research to date on disaggregating  
117 costs, benefits, and risks to design robust partnerships that are broadly satisfactory to all partners  
118 (Herman, Reed, Zeff, & Characklis, 2015; Jafino, Kwakkel, & Taebi, 2021).

119 In this work, we contribute an exploratory modeling framework (see detailed review by  
120 Moallemi, Kwakkel, de Haan, & Bryan, 2020) for testing thousands of candidate infrastructure  
121 partnership structures across multiple future hydrologic scenarios, using a flexible and high-  
122 resolution water resources simulation model, the California Food-Energy-Water System  
123 (CALFEWS (Zeff et al., 2021)). Each partnership structure is assessed in terms of aggregate  
124 performance as well as its impacts on individual partners, in order to search for investments that  
125 are viable for all partners across multiple plausible futures. We take an explicitly optimistic view  
126 of uncertainty in this study by modeling outcomes under present-day demands, institutions, and  
127 regulatory contexts. Similarly, we assume a limited degree of hydrologic uncertainty by focusing  
128 on stationary hydro-climatic variability, rather than explicitly focusing on the more severe floods  
129 and droughts expected with climate change (Gonzalez et al., 2018). A major aim of this  
130 exploratory analysis is to show that, even under these strongly optimistic assumptions, traditional  
131 planning frameworks are unlikely to produce viable investment partnerships.

132 Our results focus on the southern Central Valley, a water-stressed and agriculturally-  
133 productive region that relies heavily on overdrafted aquifers to meet its irrigation and drinking  
134 water demands, particularly during drought, and may face severe cutbacks under the Sustainable  
135 Groundwater Management Act (Faunt & Sneed, 2015; Hanak et al., 2019; Levy et al., 2021;  
136 Newsom et al., 2020). Water districts are mobilizing to develop new infrastructure (e.g., canals,  
137 groundwater recharge facilities) and flexible institutions (e.g., water trading, groundwater  
138 banking) in order to balance supplies and demands (Escriva-Bou, Sencan, et al., 2020; Hanak,  
139 Jezdimirovic, et al., 2018; Jezdimirovic, Hanak, & Escriva-Bou, 2020; Scanlon, Reedy, Faunt,  
140 Pool, & Uhlman, 2016). Thus, the region is emblematic of the challenges and opportunities  
141 facing water supply organizations throughout the Western U.S. and other water-stressed regions  
142 of the world, and can advance our ability to develop resilient and sustainable water portfolios

143 capable of managing intensifying scarcity and climate-related uncertainty (AghaKouchak et al.,  
144 2021; Famiglietti, 2014; Jiménez Cisneros et al., 2015; Lall et al., 2018).

## 145 **2 Methods**

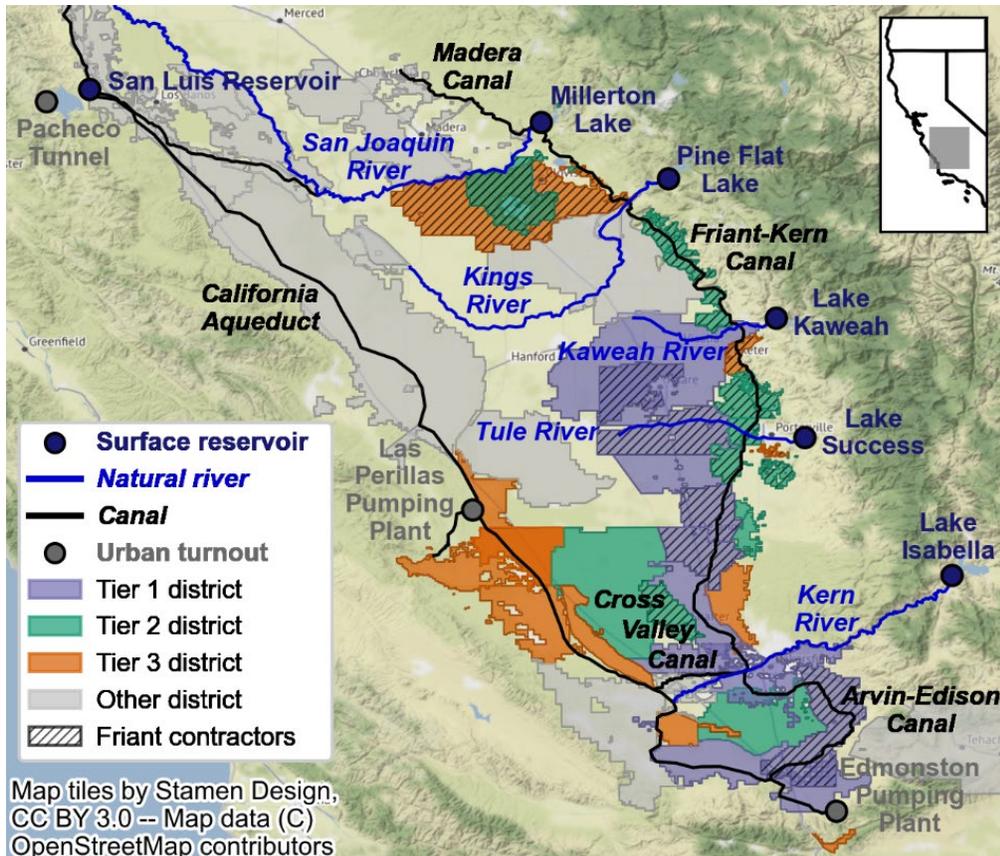
### 146 2.1 Overview

147 The Friant-Kern Canal is a major water conveyance system in the southeastern Central  
148 Valley. As part of the Central Valley Project, it diverts San Joaquin River water from Lake  
149 Millerton to the “Friant contractors” to the south (Figure 1). The canal also conveys water from  
150 other local rivers to a wider array of water districts and groundwater banks willing to take  
151 surplus deliveries during high-flow periods. However, widespread subsidence caused by  
152 groundwater overdraft has damaged the canal and reduced its capacity by 60% along critical  
153 reaches (Faunt & Sneed, 2015; Friant Water Authority, 2019). Water districts relying on the  
154 canal, especially the Friant contractors, are advocating for its rehabilitation. Simultaneously,  
155 many water districts are working to build more groundwater recharge and banking facilities  
156 (Dahlke et al., 2018; Hanak, Jezdimirovic, Escriva-Bou, & Ayres, 2020; Jezdimirovic et al.,  
157 2020; Scanlon et al., 2016). Building conveyance and groundwater recharge simultaneously  
158 could yield synergistic benefits when the additional deliveries due to canal expansion, which  
159 primarily occur during high-flow periods, exceed immediate irrigation demands and require  
160 storage until drier periods (Alam, Gebremichael, Li, Dozier, & Lettenmaier, 2020; Hanak,  
161 Jezdimirovic, et al., 2018; Kocis & Dahlke, 2017; Wendt, van Loon, Scanlon, & Hannah, 2021).  
162 Local farms, water districts, and politicians have lobbied for external support, but subsidies are  
163 unlikely to fund the projects in full (Whisnand, 2021). Thus, water districts that stand to benefit  
164 will need to collaborate to raise the remaining capital, creating the challenge of defining,  
165 designing, and evaluating these investment partnerships.

166 In this study, an exploratory ensemble modeling approach is used to evaluate thousands  
167 of plausible partnership structures. Each candidate partnership structure contains a subset of 41  
168 water districts (Tiers 1-3 in Figure 1) that could potentially benefit from rehabilitating the canal  
169 and/or adding a new groundwater bank along the canal. Our exploratory sampling design assigns  
170 to each partner an ownership share that controls its access to capacity in the new infrastructure  
171 and its share of debt payments. Each candidate partnership structure is applied to three different  
172 capital projects (canal rehabilitation, groundwater bank development, and both). Each of the  
173 capital projects are evaluated using three different 50-year hydrologic scenarios (wet, average,  
174 and dry) that are selected from a larger stochastic streamflow ensemble that captures the  
175 system’s internal hydro-climatic variability. All partnership structure, capital project, and  
176 hydrologic scenario combinations are simulated using the CALFEWS model (Zeff et al., 2021).

177 Water portfolio investments are assessed using two key performance metrics. First is the  
178 captured water gain in billions of liters per year, GL/year (or thousands of acre-feet per year,  
179 kAF/year). This metric measures the expected “new water” delivered by the investment, defined  
180 as the difference in average annual water deliveries to partner districts with and without the new  
181 infrastructure under a particular hydrologic scenario. Each capital project represents a large up-  
182 front investment that requires annual debt payments, with each partner’s share of debt equal to its  
183 ownership share. The second measure of performance is the cost of gains in dollars per million

184 liters, \$/ML, (or dollars per acre-foot, \$/AF), defined as the annual debt payment divided by the  
 185 captured water gain. Effective partnerships will reliably provide their partners with significant  
 186 water gains at relatively low cost.



188 **Figure 1.** Geography of water supply in the southern Central Valley. Five major reservoirs store  
 189 runoff from the Sierra Nevada mountains in the east and release it into the region’s major rivers,  
 190 where much of the flow is withdrawn by water districts. Millerton Lake diverts San Joaquin  
 191 River water into the Madera Canal and Friant-Kern Canal (FKC) as part of the Central Valley  
 192 Project (CVP). Water districts receiving CVP contract water from the FKC are shown as “Friant  
 193 Contractors”. The CVP and State Water Project also pump water from the Sacramento-San  
 194 Joaquin River Delta to San Luis Reservoir, where it is routed via a series of pumps and canals to  
 195 water districts in the valley and urban districts along the coast. Water districts are grouped into  
 196 three tiers based on the results of the experiment, with Tier 1 districts having the highest  
 197 potential to benefit from new infrastructure and Tier 3 districts having the lowest potential. Note  
 198 that three coastal urban districts from the experiment are not shown (one Tier 2 and two Tier 3).

199 **2.2 CALFEWS simulation model**

200 The California Food-Energy-Water System (CALFEWS) is an open-source,  
 201 Python/Cython-based model for simulating the movement of water supplies within California,  
 202 with a focus on the Central Valley (Zeff et al., 2021). CALFEWS operates across multiple  
 203 scales, from statewide representation of major inter-basin transfer projects to distributed local

204 representation of district-level conjunctive surface and groundwater management. The model has  
205 been found to reproduce historical reservoir storages, canal flows, surface water deliveries, and  
206 groundwater banking accounts with a high degree of fidelity.

207 CALFEWS has three major advantages when compared to more commonly-used water  
208 supply models in California (e.g., CalSim (California Department of Water Resources, 2017;  
209 Draper et al., 2004), CalLite (Islam et al., 2011), CALVIN (Draper et al., 2003)). Firstly, it  
210 models dynamics at a finer spatio-temporal resolution (water district representation, daily time  
211 step), while still allowing for simulations to be run at a statewide scale over many decades. This  
212 gives CALFEWS an unprecedented ability to track the impacts of district-level decision-making  
213 on statewide water supply projects, and the impacts of short-lived high-flow periods (e.g.,  
214 atmospheric rivers) on long-lived infrastructure investments. Secondly, CALFEWS uses a rules-  
215 based representation of system dynamics, in contrast to the mathematical programming  
216 techniques used by the aforementioned models. This rules-based approach can more flexibly  
217 represent system complexities such as adaptive operations, environmental regulations,  
218 groundwater banking arrangements, and distributed district-level decision-making. Lastly,  
219 ensembles of CALFEWS simulations can be dispatched in parallel on high-performance  
220 computing infrastructure, enabling high-throughput exploratory modeling approaches (see  
221 Section 2.5).

### 222 2.3 Infrastructure project alternatives

223 The first infrastructure project considered in this study is the rehabilitation of the Friant-  
224 Kern Canal. Widespread groundwater overdraft in the region has damaged the canal via  
225 subsidence (Faunt & Sneed, 2015). In certain sections near the Tule River, canal capacity has  
226 been reduced by almost 60% (Friant Water Authority, 2019; U.S. Bureau of Reclamation, 2020).  
227 Water districts and government representatives have recently been negotiating a partnership to  
228 rehabilitate the canal. Although the new design specifications are uncertain, we assume that the  
229 entire length of the canal will be returned to its original design capacity and that the cost borne  
230 by the districts involved in the partnership will be \$50 million. This is based on a recent funding  
231 agreement for the project, in which the Friant contractors have agreed to pay \$50 million out of  
232 an estimated \$500 million total (Whisnand, 2021). The remainder will come from federal and  
233 state funding sources as well as legal settlement agreements associated with environmental  
234 damages and land subsidence. The partnership design efforts in this study are focused on the \$50  
235 million share currently allocated to the Friant contractors. This is an unusually favorable case  
236 study for infrastructure investment partnerships due to the unusually high level of outside  
237 funding available; for other capital projects with lower subsidy levels, viability will generally be  
238 more difficult to achieve.

239 The second capital project is a jointly managed groundwater bank along the Friant-Kern  
240 Canal in the vicinity of the Tule River. This project is hypothetical and is not based on any  
241 particular existing or planned groundwater bank, but water districts throughout the region have  
242 been investing in new recharge facilities and banking relationships (Escriva-Bou, Sencan, et al.,  
243 2020; Hanak et al., 2020). The partnership's share of the total capital cost (i.e., after any  
244 subsidies) is assumed to be \$50 million, in line with cost estimates of other large groundwater

245 banks currently under development (Jezdimirovic et al., 2020). There are three main parameters  
246 controlling the function of groundwater recharge and recovery facilities in the CALFEWS model  
247 (Zeff et al., 2021): the baseline recharge capacity, the baseline recovery capacity, and the storage  
248 volume of infiltration basins. The two “baseline” values refer to initial capacities at the  
249 beginning of the recharge or recovery season; both capacities will decrease over the season with  
250 extended use. The uncertainty bounds for these parameters are set based on the ranges of pre-  
251 existing groundwater banks in the region: infiltration capacity between 0 and 1.5 GL/day (1.2  
252 kAF/day), infiltration pond storage volume between 0 and 1.5 GL (1.2 kAF), and recovery  
253 capacity between 0 and 0.9 GL/day (0.7 kAF/day). For comparison, the upper limits of these  
254 ranges are 50%, 50%, and 88% of the estimated parameters for the largest groundwater bank in  
255 the region, the Kern Water Bank (Kern Water Bank Authority, 2021).

256 Access to both pieces of infrastructure is restricted to the set of water districts investing  
257 in the partnership. Additionally, each partner is assigned priority access to a fraction of the new  
258 capacity that is proportional to its ownership share in the project (e.g., a district with a 20%  
259 ownership share will have priority access over 20% of the new capacity). If a district is not using  
260 its priority capacity at any given time, access is opened up to the broader set of partners. Note  
261 that the canal rehabilitation project does not impact non-partners’ access to the existing capacity.  
262 It only restricts access to the expanded capacity at the top of the canal.

263 Both capital projects are assumed to be financed with revenue bonds that require equal-  
264 sized annual payments over a 50-year period with 3% interest. This is a conservative assumption  
265 (in the sense of not over-estimating costs) because recent revenue bonds issued by California  
266 water districts have generally carried between 2.5-3.5% interest with maturities of 25 years or  
267 fewer (California State Treasurer’s Office, 2020), and a 50-year maturity bond would carry an  
268 additional premium in practice. Under these assumptions, partnerships would make annual debt  
269 payments of approximately \$1.943 million for either canal rehabilitation or groundwater bank  
270 development, and \$3.887 million for both projects. Each partner’s share of the overall debt  
271 payment is proportional to its ownership share.

## 272 2.4 Hydrologic scenarios

273 The hydrologic scenarios used in this work are generated using the California and West  
274 Coast Power System (CAPOW), an open-sourced, Python-based model for simulating the  
275 operation of the U.S. west coast bulk electric power system. CAPOW has a major focus on the  
276 impact of hydrometeorological variables (streamflow, temperature, wind speed, insolation) on  
277 system reliability and pricing (Su, Kern, Denaro, et al., 2020). As such, a major component of  
278 the model is its stochastic engine, which uses a hybrid vector autoregressive-bootstrapping  
279 approach to generate daily synthetic records of hydrometeorological variables at many locations  
280 across the west coast. These synthetic records are shown to maintain historical correlation  
281 patterns across space (i.e., northern vs. southern California) and time (i.e., intra- and interannual  
282 autocorrelation), while overcoming the limitations of the limited length of historical hydro-  
283 climatic observations. Although CAPOW is trained on historical data, it enables the generation  
284 of statistical replicate time series of synthetic hydro-climatic scenarios, better capturing a wide  
285 variety of plausible futures as well as extremes beyond the limited number of observed rare

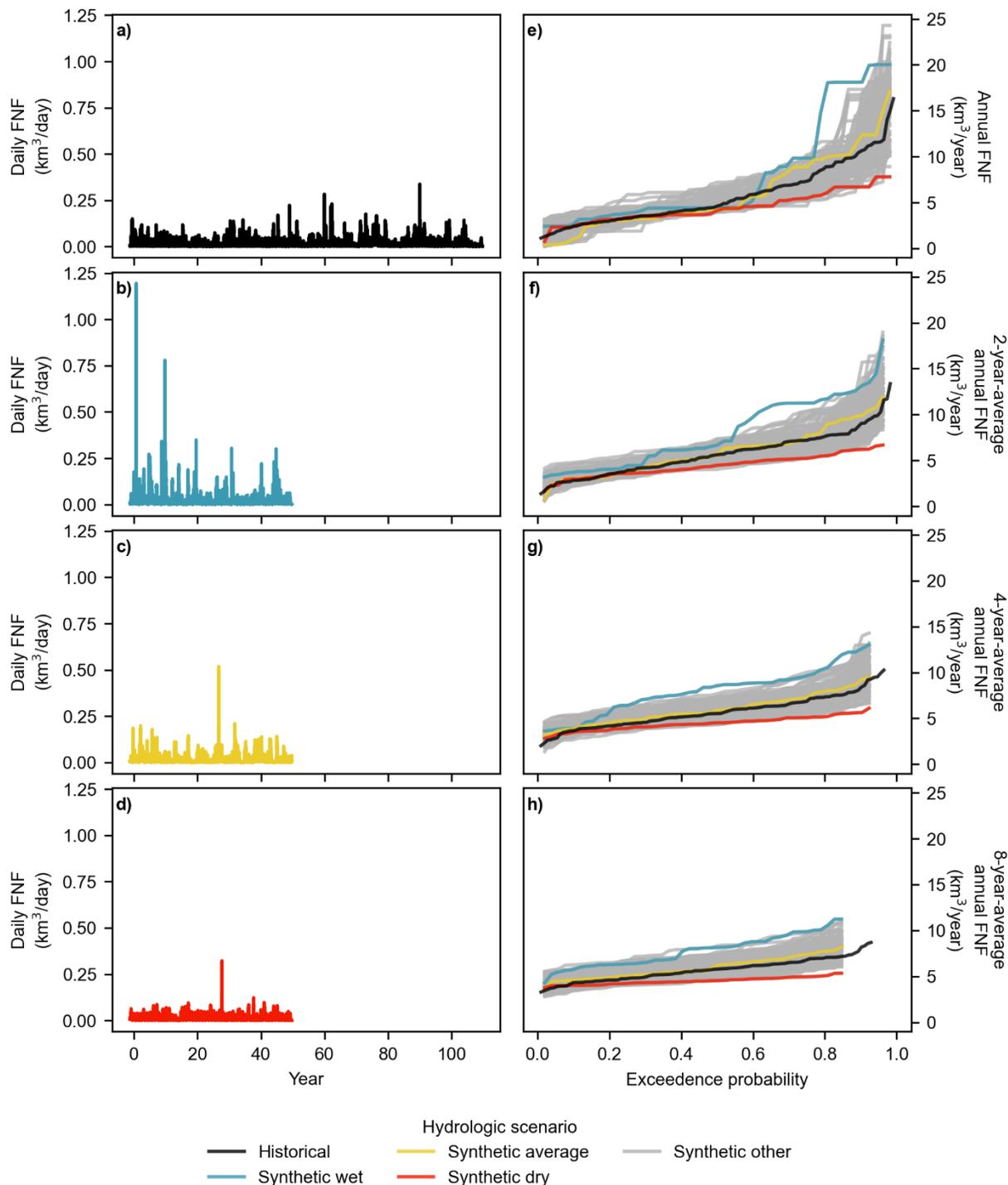
286 events in the modern historical record. This makes it a valuable tool for planning over the  
287 medium term (e.g., initiating capital investments in the next decade), where interannual  
288 variability is a major contributor to overall hydrologic uncertainty (Doss-Gollin, Farnham,  
289 Steinschneider, & Lall, 2019; Lehner et al., 2020; Su, Kern, Reed, & Characklis, 2020).

290 For the present work, the CAPOW stochastic engine is used to generate daily synthetic  
291 full natural flow records at fifteen major water supply reservoirs in California. These full natural  
292 flow records can be used to statistically reconstruct more detailed time series of streamflows,  
293 gains, and snowpacks (Zeff et al., 2021). We generate 101 alternative 50-year time series, which  
294 span a wide range of hydrologic outcomes across extreme quantiles of interest. Figure 2 and  
295 Supporting Information (SI) Figure S1 compare each 50-year synthetic record with the 111-year  
296 historical record in terms of 1-, 2-, 4-, and 8-year flow duration curves of full natural flow across  
297 the major South San Joaquin River Basin (SSJRB) reservoirs (Millerton, Pine Flat, Kaweah,  
298 Success, and Isabella), and the major Sacramento River Basin (SRB) reservoirs (Shasta,  
299 Oroville, New Bullards Bar, and Folsom), respectively. These curves demonstrate that the  
300 stochastic engine can accurately represent regional flows in terms averages, extremes, and  
301 persistence of multi-year wet and dry periods, while also providing a wider range of extremes for  
302 risk assessment.

303 These 101 scenarios are then sorted in terms of total 50-year full natural flow across the  
304 SSJRB reservoirs. The time series with the highest, median, and lowest flows are selected to  
305 represent “wet”, “average”, and “dry” hydrologic scenarios, respectively. The “average” scenario  
306 is found to track the historical flow duration curve well in the SSJRB (Figure 2), while the “wet”  
307 and “dry” scenarios are significantly wetter and drier, especially with respect to normal and wet  
308 periods (e.g., mid- to high-range exceedance). With respect to the most extreme droughts (e.g.,  
309 low exceedance), the “average” scenario is found to be quite extreme, while the “dry” scenario is  
310 similar to the historical record. The “wet” and “dry” scenarios are less extreme in the context of  
311 the SRB (SI Figure S1), and the “average” scenario appears to be wetter than average. This is  
312 due to the imperfect correlation in hydrologic conditions across the state; the wettest years in  
313 northern California are not necessarily the wettest years in the southern Central Valley, and vice  
314 versa. The SSJRB reservoirs are used for scenario selection because they have the most direct  
315 influence on the Friant-Kern Canal.

## 316 2.5 Exploratory modeling framework

317 This study employs a random sampling framework to develop 3,000 plausible candidate  
318 partnership structures. The sampling design requires a three-step process to generate each  
319 partnership structure. First, the size of the partnership  $np$  is randomly drawn from a Poisson  
320 distribution with a mean of 8 partners, excluding zero. This distribution, which creates  
321 partnership structures ranging from approximately one to twenty partners, was selected to strike  
322 a balance between relatively dense sampling at smaller partnership sizes (which are more likely  
323 to be viable) and broader exploration of larger partnership structures. Second,  $np$  partners are  
324 randomly selected without replacement from the set of 41 candidate water districts. Third, the  
325 ownership shares of the  $np$  partners are sampled from a uniform distribution and normalized to  
326 sum to one.



328 **Figure 2.** Comparison of synthetic and historical full natural flow (FNF) for the Tulare Lake  
 329 Basin. Panels (a-d) show the total full natural flows for the five major reservoirs of the South  
 330 San Joaquin River Basin (Millerton, Pine Flat, Kaweah, Success, and Isabella), under the  
 331 historical record and the wet, average, and dry synthetic scenarios. Panels (e-h) show the full  
 332 natural flow duration curves for each time series over 1-, 2-, 4-, and 8-year periods.

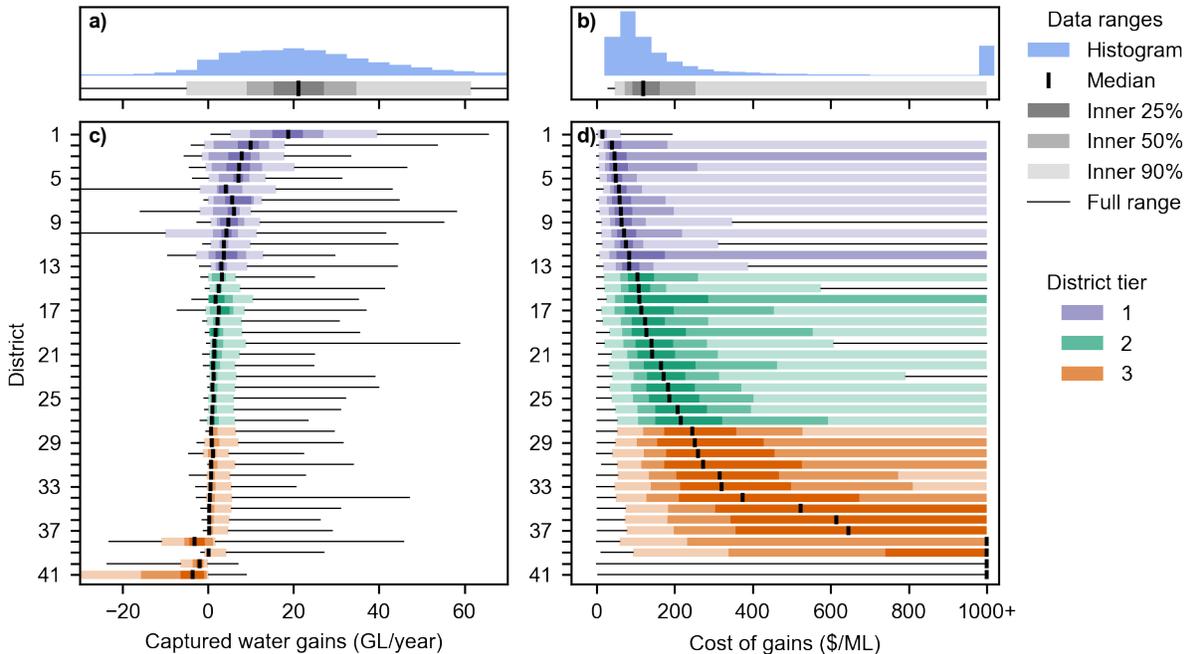
333 Each of the 3,000 sampled partnership structures is combined with each of three capital  
334 projects: canal rehabilitation, groundwater bank development, and both. Three groundwater  
335 recharge and recovery parameters are randomly sampled from uniform distributions across their  
336 uncertainty bounds (Section 2.3). This results in 9,000 partnership structure/capital project  
337 combinations, to which we add the “Friant-16” canal rehabilitation partnership representing  
338 business-as-usual planning (see Section 3.4), and the baseline case with no new infrastructure.  
339 Lastly, each of these combinations is combined with each of the three stochastic hydrologic  
340 scenarios: dry, average, and wet. In total, this represents 27,006 partnership structure/capital  
341 project/hydrologic scenario combinations.

342 Each of these 27,006 combinations is evaluated using the CALFEWS model. Simulations  
343 are dispatched in parallel across 800 cores using the Longleaf Cluster at the University of North  
344 Carolina at Chapel Hill. Results such as reservoir releases, canal flows, water contract deliveries,  
345 and groundwater banking balances for each simulation are stored using the hdf5 file format.  
346 Each simulation is evaluated by comparing its performance to the performance of the baseline  
347 no-infrastructure case under the same hydrologic scenario.

### 348 **3 Results**

#### 349 3.1 Evaluating new infrastructure investments along the Friant-Kern Canal

350 The candidate partnerships explored in this study display a wide range of partnership-  
351 level performance (Figures 3a-b). Five percent deliver at least 62 GL/year (50 kAF/year) of  
352 captured water gains, with a maximum of 118 GL/year (96 kAF/year). For context, Lake  
353 Millerton has a capacity of 642 GL (521 kAF), and 1 GL (1 kAF) is enough irrigate roughly  
354 0.82-2.2 km<sup>2</sup> (250-667 acres) of crops in the region (University of California Agriculture and  
355 Natural Resources, 2021). Five percent of candidate partnerships cost less than \$45/ML  
356 (\$56/AF), with a minimum of \$30/ML (\$36/AF). These results would be competitive with other  
357 water supply projects under consideration throughout the Central Valley (Jezdimirovic et al.,  
358 2020). However, most candidate partnerships perform more modestly, with a median gain of 21  
359 GL/year (17 kAF/year) and a median cost of \$120/ML (\$147/AF). This cost of gains, which  
360 includes only the investment’s debt payments and not the additional costs of procuring and  
361 transporting the water itself, would represent a significant increase above typical rates of \$32-  
362 154/ML (\$40-190/AF) charged by water districts for irrigation deliveries in the region. Worse  
363 still, 9% of candidate partnerships yield negative captured water gains, representing investments  
364 that actually reduce partners’ water deliveries on average. This occurs when new infrastructure  
365 triggers unpredictable dynamics within the water system that allow non-partners to benefit over  
366 partners. The cost of gains for these partnerships is effectively infinite, and more broadly 13% of  
367 partnerships have very high costs over \$1,000/ML (\$1,233/AF). These capital investments  
368 represent a serious financial risk if the future water gains are insufficient to allow partners to sell  
369 enough water to pay off their debt, even under our explicitly optimistic assumptions in this study  
370 that neglect the broader extremes possible under climate change. Thus, near term capital  
371 investments can have very complex and potentially severe downside risks for partners.

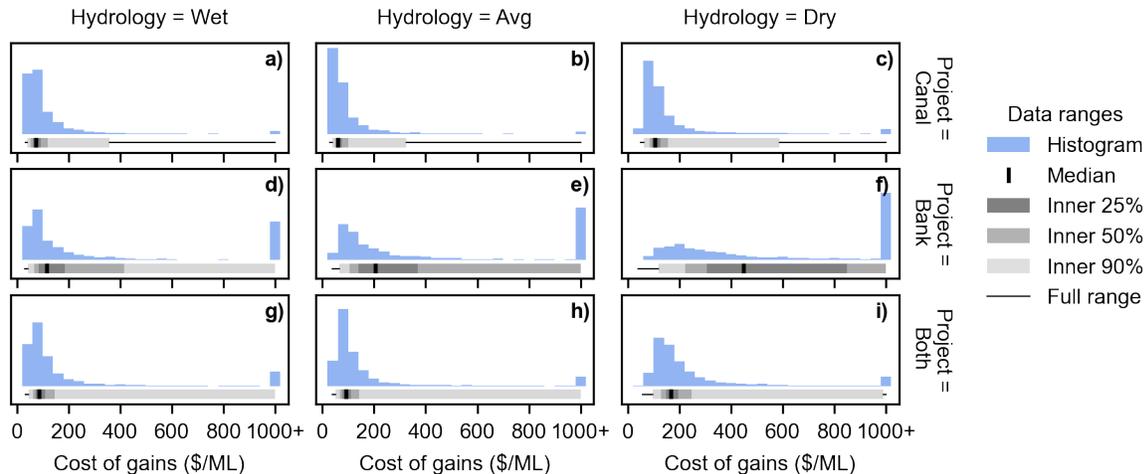


372  
 373 **Figure 3.** Distribution of performance for thousands of candidate infrastructure investment  
 374 partnership structures. (a) Distribution of partnership-level captured water gains. (b) Distribution of  
 375 partnership-level cost of gains. (c) Distribution of captured water gains for individual water  
 376 districts. (d) Distribution of cost of gains for individual water districts. Distributions show the  
 377 variability of results across all candidate partnership structures, capital projects, and hydrologic  
 378 scenarios. For Panels (b) and (d), all costs over \$1,000/ML (\$1,233/AF) are consolidated into  
 379 “1000+”. Water districts are grouped into three tiers based on the results of the experiment,  
 380 Tier 1 districts having the highest potential to benefit from new infrastructure and Tier 3 districts  
 381 having the lowest potential.

382 Decomposing these results by hydrologic scenario, the costs of gains are generally  
 383 highest in the dry scenario (Figure 4) because the size of debt payments is independent of how  
 384 much water ends up being available. However, costs in the wet scenario are similar to the  
 385 average scenario, suggesting an important asymmetry: the downside risk in a drier future is  
 386 larger than the upside risk in a wetter future. Decomposing variability by capital project, we find  
 387 similar cost distributions for canal rehabilitation projects vs. joint canal-groundwater banking  
 388 projects. However, the latter risks underperformance in dry scenarios if insufficient water is  
 389 available to warrant the capital cost of both projects. Groundwater bank-only partnerships  
 390 generally perform well in the wet scenario but much poorer in average and dry scenarios.

391 Although the project type and hydrologic scenario do impact performance in important  
 392 ways, substantial variability remains after accounting for these factors (Figure 4). This variability  
 393 is attributed to the partnership structure itself (i.e., which water districts are included as partners,  
 394 and what ownership share does each partner take?). For example, the subset of candidate  
 395 partnerships that expand the canal under the wet hydrologic scenario (Figure 4a) are identical  
 396 other than their partnership structures, yet experience widely varying partnership-level costs  
 397 ranging from \$32/ML to over \$1,000/ML (\$39-1,233/AF). For candidate partnerships with

398 groundwater banking (either alone or in combination with canal expansion), three additional  
 399 capacity factors also contribute to the variability of outcomes (e.g., soil infiltration rate).  
 400 However, these appear to play a minor role compared to partnership structure. Thus, the impacts  
 401 of water portfolio investments can depend critically on the subset of water districts in the  
 402 partnership and their relative ownership shares. This represents a major challenge for standard  
 403 infrastructure planning approaches that focus primarily on physical design parameters such as  
 404 capacity (perhaps in combination with climate-related vulnerability analyses), while neglecting  
 405 contractual design factors such as infrastructure access and ownership.



407 **Figure 4.** Distribution of performance (partnership-level cost of gains) for candidate partnerships  
 408 across different capital projects and hydrologic scenarios. Panels (a-c) involve a canal expansion,  
 409 (d-f) involve a groundwater bank, and (g-i) involve both projects. Panels (a/d/g) use the wet  
 410 hydrologic scenario, (b/e/h) use the average scenario, and (c/f/i) use the dry scenario. All costs  
 411 over \$1,000/ML (\$1,233/AF) are consolidated into “1000+”.

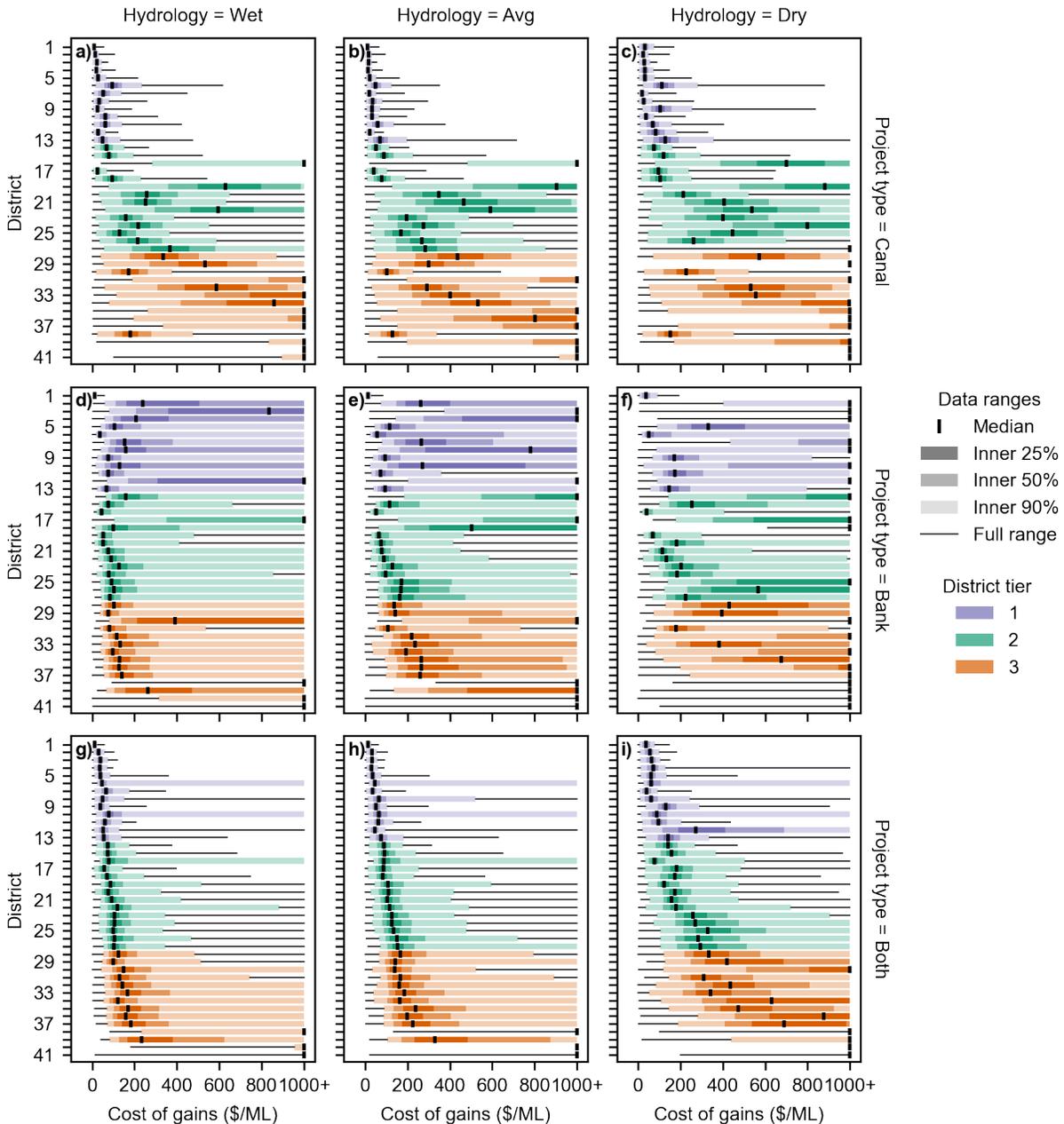
### 412 3.2 Potential benefits are highly heterogeneous across water districts

413 Next, we disaggregate partnership-level performance into the captured water gains  
 414 (Figure 3c) and costs of gains (Figure 3d) experienced by each of the 41 water districts in the  
 415 exploratory experiment. Each district’s results include only the subset of candidate partnership  
 416 structures for which it is a participant. Most districts display a wide range of captured water  
 417 gains (e.g., from <0 to >20 GL/year), but with outcomes concentrated in a narrower band (e.g.,  
 418 >90% of partnerships between 0-10 GL/year). The range of costs experienced by each district is  
 419 even wider, with all districts except for one spanning from near-zero to over \$1,000/ML. Despite  
 420 the high variability within each district, there are clear distributional differences as well. Water  
 421 districts can be grouped into three roughly-equal sized tiers based on their median cost of water  
 422 gains: Tier 1 under \$100/ML (\$123/AF), Tier 2 under \$240/ML (\$296/AF), and Tier 3 over  
 423 \$240/ML. Tier 1 districts have the highest potential to benefit from partnering in these  
 424 investments, while Tier 3 districts have the lowest. These distributional differences provide

425 crucial information for the partnership negotiation process, but would be unavailable using  
426 traditional aggregate cost-benefit analyses.

427         Decomposing the district-level performance into separate components for each capital  
428 project and hydrologic scenario, we find costs to be highly variable across all combinations,  
429 highlighting the importance of partnership structure on district-level outcomes (Figure 5). Tier 1  
430 districts generally have much lower costs than Tier 2-3 for canal-only projects, while the  
431 distributions are more uniform across tiers for bank-only projects. Joint canal-groundwater  
432 banking projects exhibit intermediate behavior; Tier 1 districts generally have lower costs than  
433 Tier 2, which have lower costs than Tier 3, but these differences are more gradual than in the  
434 canal-only partnerships. With regards to hydrology, the costs of gains are significantly higher on  
435 average in the dry scenario than the average and wet scenarios, but this effect varies across  
436 districts, suggesting that some districts will experience more climate-related risk than others. We  
437 reiterate that our experimental design is expressly optimistic with regards to climate uncertainty,  
438 considering extreme realizations of stationary stochastic variability but not anthropogenically  
439 forced climate non-stationarity. Thus, under climate change, the downside climate-related risks  
440 are likely even greater than represented here.

441         The geographic distribution of the tiers provides useful context (Figure 1). Tier 1  
442 stretches from the Kaweah River to the Kern River along the western side of the Friant-Kern  
443 Canal, before wrapping around the other major canals to the south. These districts' proximity to  
444 the confluences of major canals and rivers allows them to receive water from multiple sources,  
445 increasing their chances of capitalizing on the investment. Additionally, many of the Tier 1  
446 districts overlie highly suitable soils for groundwater recharge (O'Geen et al., 2015) (SI Figure  
447 S2), which has allowed them to build dedicated recharge facilities within their boundaries (Alam  
448 et al., 2020; Scanlon et al., 2016; Wendt et al., 2021). These facilities increase the benefits from  
449 canal expansion by storing more surplus flows during winter/spring months when agricultural  
450 demands are limited. Districts without significant groundwater recharge capacity are more  
451 limited in how much water they can receive during these periods. This also helps to explain why  
452 the benefits from groundwater bank development are more uniformly distributed across the  
453 districts (Figure 5d-f) because these facilities level the playing field by allowing districts with  
454 unsuitable soils to store groundwater off-site. Lastly, Figure 1 shows that the water districts with  
455 the highest potential to gain from new infrastructure are not necessarily the most "obvious" ones.  
456 In this case, we find a number of Friant contractors (the group of districts currently negotiating to  
457 invest in the project (Whisnand, 2021)) in Tiers 2-3, while a number of non-contractors are in  
458 Tier 1. This demonstrates the value of the exploratory modeling approach to partnership design,  
459 which allows a much broader array of suitable alternatives to be discovered.



461 **Figure 5.** Distribution of performance (cost of gains for individual districts) for candidate  
 462 partnerships across different capital projects and hydrologic scenarios. Panels (a-c) involve a  
 463 canal expansion, (d-f) involve a groundwater bank, and (g-i) involve both projects. Panels (a/d/g)  
 464 use the wet hydrologic scenario, (b/e/h) use the average scenario, and (c/f/i) use the dry scenario.  
 465 All costs over \$1,000/ML (\$1,233/AF) are consolidated into “1000+”.

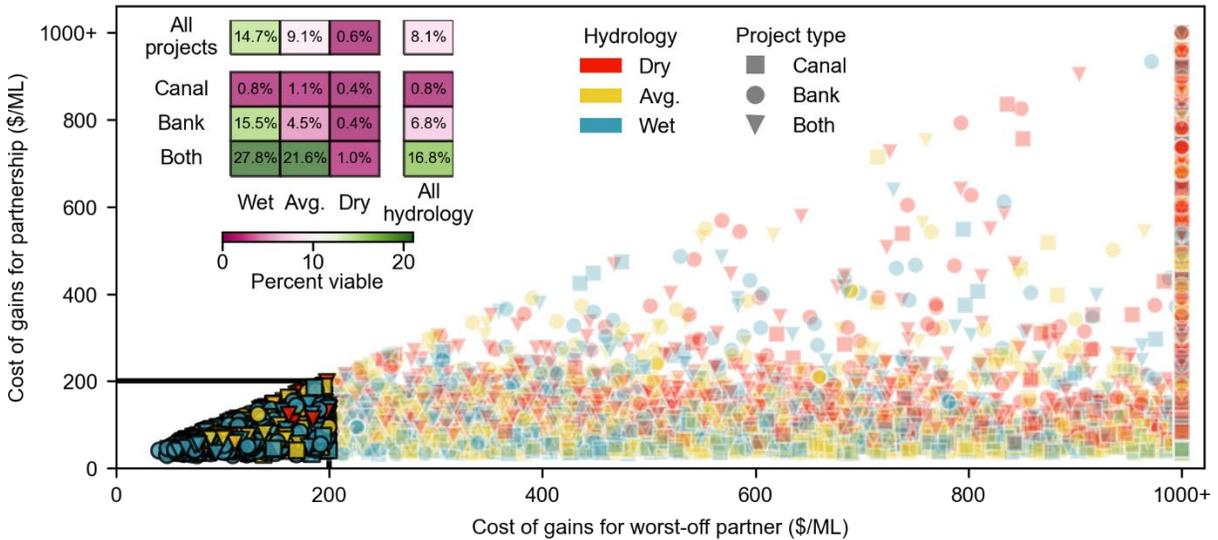
466 3.3 Heterogeneity limits the viability of partnerships

467 Results thus far have shown how the benefits of water portfolio partnerships can vary  
 468 widely based on the type of infrastructure, the hydrologic scenario, and the partnership structure

469 itself. We now demonstrate how this range of district-level outcomes can threaten investment  
470 viability. The partnerships in this work are assumed to be voluntary, meaning all partners have  
471 agreed to the terms of cooperation. This implies that each partner believes they will be better off  
472 with the partnership than without it (Giglio & Wrightington, 1972). In practice, coalition-  
473 building is complex due to power and information asymmetries, incentive misalignments, and  
474 relationships between participants (BenDor & Scheffran, 2019; Hansen et al., 2020; Lubell,  
475 Blomquist, & Beutler, 2020; Madani, 2010; Read, Madani, & Inanloo, 2014). However, we  
476 adopt the following simple definition: an infrastructure partnership is considered viable under a  
477 particular hydrologic scenario if all partners pay less than \$200/ML (\$247/AF) for their share of  
478 captured water gains. This is an optimistically conservative definition, in the sense of not  
479 erroneously labelling viable partnerships as non-viable, because water districts in the region  
480 typically charge their customers \$32-154/ML (\$40-190/AF) for irrigation water (University of  
481 California Agriculture and Natural Resources, 2021). Therefore, \$200/ML in new project debt,  
482 on top of additional water procurement costs, would represent a significant and likely  
483 unacceptable increase.

484 Despite our optimistically conservative viability definition, only 8% of candidate  
485 partnerships we have explored are viable (Figure 6). In the preponderance of cases, the worst-off  
486 partner pays significantly more than the partnership-level cost. In 61% of cases, performance is  
487 considered viable at the partnership level (i.e., overall cost is <\$200/ML), but non-viable when  
488 impacts are disaggregated (i.e., at least one partner pays >\$200/ML). This suggests that in a  
489 majority of cases, the limiting factor for project viability is not the overall volume of captured  
490 water gains, but rather the uneven distribution of these gains across the partnership in light of  
491 each partner's share of project debt. Traditional planning approaches based on aggregate impacts  
492 are incapable of uncovering these distributional issues, and thus risk leading to cooperative  
493 capital investments that are harmful to a subset of partners.

494 Decomposing viability by hydrologic scenario also delivers valuable insights (Figure 6  
495 inset). In the wet hydrologic scenario, 15% of candidate partnerships are found to be viable,  
496 compared to 9% in the average scenario and 1% in the dry scenario. This suggests that if the  
497 future is drier than the past, investment partnerships will be more likely to fail for at least one  
498 partner. California has already begun getting hotter and drier under anthropogenic climate  
499 change (Gonzalez et al., 2018), which presents a major obstacle to meeting the state's  
500 collaborative portfolio investment goals. Decomposing the results by capital project shows that  
501 the canal expansion alone is very unlikely to be viable (1%). Groundwater bank development  
502 improves the chances of viability, either in isolation (7%) or in combination with canal  
503 expansion (17%). It is instructive to compare these results to Figure 4, which shows that canal  
504 expansion projects and joint canal-groundwater banking projects have roughly similar odds of  
505 achieving viability at the partnership level (i.e., aggregate costs <\$200/ML). Thus, although  
506 canal expansion appears promising in aggregate, it tends to distribute captured water gains more  
507 unevenly and thus is less likely to satisfy all of the participating districts (see also Figure 5).



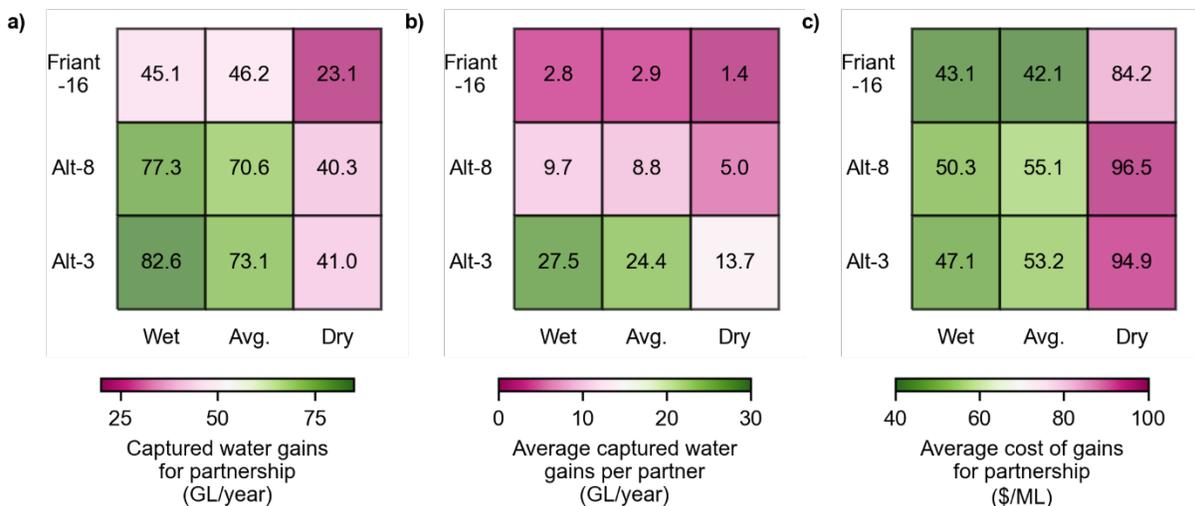
509 **Figure 6.** Viability of candidate infrastructure investment partnerships. Each simulated  
 510 partnership is plotted according to the cost of gains for the partnership as a whole vs. the worst-  
 511 off partner. The project type and hydrologic scenario used for each simulation are represented by  
 512 marker type and color, respectively. Viable partnerships (those with costs <\$200/ML (\$247/AF)  
 513 for the worst-off partner) are represented with black outlines and higher opacity. All costs over  
 514 \$1,000/ML (\$1,233/AF) are consolidated into “1000+”. Inset shows the viability of candidate  
 515 partnership structures under each combination of capital project and hydrologic scenario,  
 516 represented by color as well as the percentage printed in each square.

### 517 3.4 Comparing alternative partnerships

518 We now explore these issues in more detail through a comparison of three alternative  
 519 partnership structures. First, we consider a partnership between 16 Friant contractors (“Friant-  
 520 16”) for canal expansion only. In contrast to the randomly sampled partnerships from the  
 521 exploratory study, Friant-16 is deliberately constructed to represent the baseline performance  
 522 under business-as-usual planning because the Friant contractors are currently negotiating to  
 523 establish such a partnership (Whisnand, 2021). The ownership shares for Friant-16 are assumed  
 524 to be proportional to historical deliveries of CVP-Friant contract water (Congressional Research  
 525 Service, 2007).

526 In addition, we consider two high-performing partnerships from the exploratory study:  
 527 one with eight partners (“Alt-8”) and one with three (“Alt-3”). Both are joint canal-groundwater  
 528 banking investments. The selection procedure for choosing Alt-8 and Alt-3 is as follows. First,  
 529 the set of 27,000 simulations is filtered down to include only those that meet each of the  
 530 following criteria: (1) the partnership is viable, i.e., the cost of gains for the worst-off partner is  
 531 less than \$200/ML; (2) the partnership-wide captured water gain is greater than 55 GL/year (45  
 532 kAF/year); and (3) the total captured water gain for all non-partners in the region is greater than  
 533 zero. These thresholds are set *a posteriori* by iteratively increasing the constraints until an elite  
 534 subset of solutions remain. In practice, the thresholds could be defined by decision-makers based

535 on their particular context and preferences. Note that Friant-16 fails on all three criteria under the  
 536 average hydrologic scenario: its worst-off partner cost is over \$1000/ML, its total captured water  
 537 gain for the partnership is 46 GL/year, and its captured water gain for non-partners is -2 GL/year  
 538 (i.e., the new infrastructure reduces deliveries to regional non-partner districts). Each of these  
 539 criteria eliminates a significant subset of simulations (SI Figures S3-S4). After filtering by the  
 540 three criteria, only 103 simulations remain: 51 from the wet scenario and 52 from the average  
 541 scenario, while no partnerships remain from the dry scenario. SI Figure S5 shows how these  
 542 simulations vary along multiple performance metrics. From the 52 candidate partnerships that  
 543 perform satisfactorily in the average scenario, Alt-3 and Alt-8 are selected manually as examples  
 544 that perform well across all metrics. Alt-3 is chosen as a representative small successful  
 545 partnership, while Alt-8 is chosen to from among the larger partnerships, which may be  
 546 preferable due to political and financial concerns.



548 **Figure 7.** Comparison of three alternative infrastructure investment partnerships at aggregate  
 549 scale. For each partnership and each hydrologic scenario, Panel (a) shows the total captured  
 550 water gain for the partnership, Panel (b) shows the average captured water gain per partner, and  
 551 Panel (c) shows the average cost of gains for the partnership. Results are represented by color as  
 552 well as the number printed in each square.

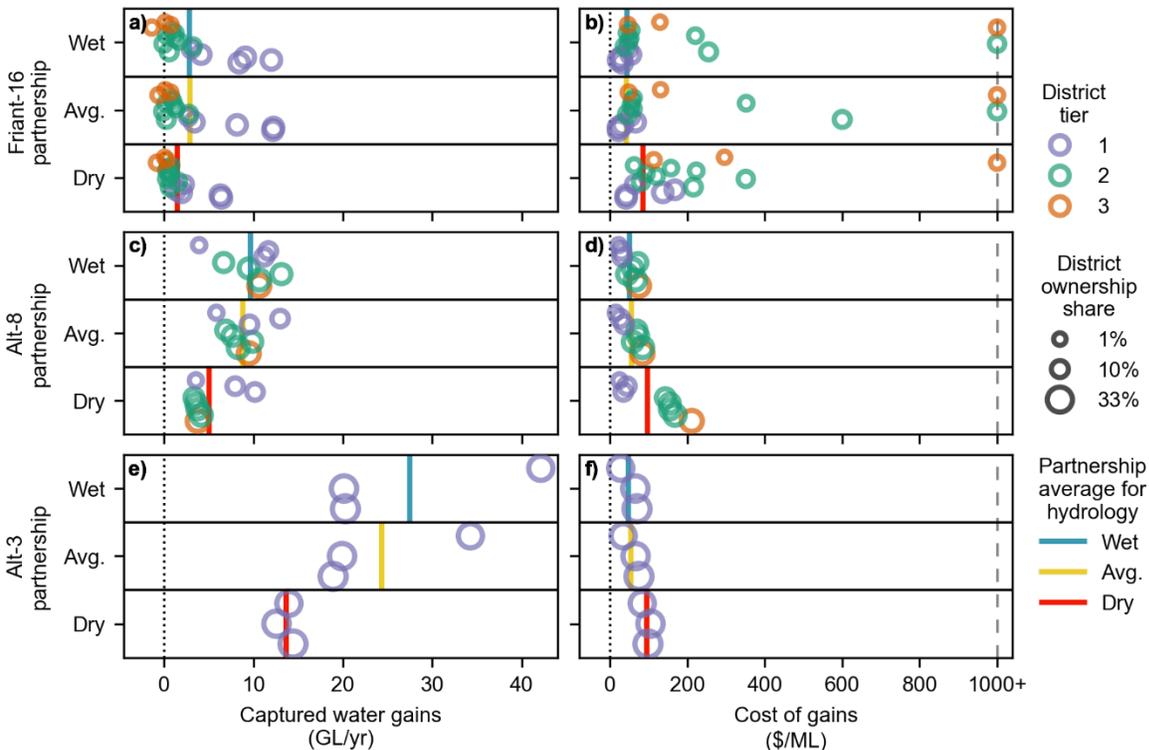
553 At the partnership level, Alt-3 is found to deliver the highest total captured water gains, at  
 554 73 GL/year (59 kAF/year) in the average scenario (Figure 7a). Alt-8 delivers 71 GL/year (57  
 555 kAF/year) of gains, while Friant-16 delivers significantly less at 46 GL/year (38 kAF/year).  
 556 Gains in the wet scenario are slightly lower for Friant-16 and slightly higher for the other two  
 557 partnerships, while dry scenario gains are roughly 50% lower across all partnerships. Because  
 558 Alt-3 only has three partners, it has significantly higher gains on a per-partner basis (Figure 7b).  
 559 However, after accounting for the larger debt shares borne by each district in the smaller  
 560 partnerships and the additional capital expense required for Alt-3 and Alt-8 to develop a  
 561 groundwater bank, the three partnerships are found to have similar costs of gains (Figure 7c):

562 \$43-50/ML (\$53-62/AF) in the wet scenario, \$42-53/ML (\$52-68/AF) in the average scenario,  
563 and \$84-97/ML (\$104-119/AF) in the dry scenario.

564 When performance is disaggregated, however, significant differences emerge (Figure 8).  
565 District-level captured water gains for each partnership are quite heterogeneous, with some  
566 districts receiving less than average and others receiving more. This is not necessarily a problem;  
567 as long as each district's captured water gain is proportional to its ownership share and thus its  
568 share of project debt, then the costs of gains across the partnership will be uniform. However,  
569 this is found to be very uncommon. The costs of gains for Friant-16 are especially  
570 heterogeneous, spanning from under \$23/ML to over \$1,000/ML. This means that some districts  
571 receive more than their "fair share" of gains based on their ownership share, and others receive  
572 far less than their fair share (in fact, some districts' gains can be negative). The other two  
573 partnerships experience smaller but meaningful levels of cost heterogeneity. Interestingly, the  
574 costs for Alt-8 are much more heterogeneous under the dry scenario than the wet and average  
575 scenarios. This suggests that climate-related risk is unevenly held across the partnership, which  
576 is vital information for partners to have when planning major investments under uncertainty. If  
577 the future turns out to be drier than the historical record, then planning studies based on historical  
578 records are likely to underestimate not only the overall cost of gains from the capital project but  
579 also the level of inequality in how these costs are distributed.

580 District tier is generally indicative of performance across the three partnerships, but  
581 imperfectly so. The highest-performing partnership, Alt-3, is composed entirely of Tier 1  
582 districts. In Alt-8, Tier 1 districts have the lowest costs in general, followed by Tiers 2-3. This  
583 disparity is exacerbated in the dry scenario, where Tiers 2-3 bear almost all of the negative  
584 impacts. Friant-16 displays weaker clustering, with districts from all three tiers present across the  
585 low- to mid-ranges of the cost spectrum. However, the highest costs are paid by districts in Tiers  
586 2-3. Overall, these results are consistent with the district-level variability in Figures 3d and 5. It  
587 is better, in general, to be Tier 1 than Tier 3, but Tier 3 districts can do very well in well-  
588 designed partnerships and Tier 1 districts can do very poorly in poorly-designed partnerships.

589 Lastly, an important factor in the success of Alt-3 is simply the size of the partnership.  
590 All else equal, smaller coalitions are more likely to remain viable because they have fewer  
591 partners to satisfy. Smaller partnerships also deliver more water to each partner on average,  
592 which is beneficial as districts attempt to maximize their capture of surface water supplies to  
593 avoid falling under the Sustainable Groundwater Management Act (Hanak et al., 2020). On  
594 the other hand, many districts are incapable of accepting very large quantities of water during  
595 short-lived high-flow events, and thus may prefer a smaller ownership share within a larger  
596 partnership. Additionally, larger and more diverse partnerships will be more capable of  
597 harnessing subsidies to lower costs (Cypher & Grinnell, 2007; Hansen et al., 2020; Newsom et  
598 al., 2020).



600 **Figure 8.** Comparison of three alternative infrastructure investment partnerships. Panels (a/c/e)  
 601 show the captured water gains for each district in the Friant-16, Alt-8, and Alt-3 partnerships,  
 602 respectively, while Panels (b/d/f) show the cost of gains for each district in these partnerships.  
 603 Each panel is split into three layers showing performance on the wet, average, and dry  
 604 hydrologic scenarios. Each district is represented by a circle, with color and size representing the  
 605 district's tier and its ownership share in the project, respectively. Within each layer, districts are  
 606 arranged by ownership share from smallest (top) to largest (bottom). The vertical blue, yellow,  
 607 and red lines represent partnership-level averages under the wet, average, and dry scenarios. All  
 608 costs over \$1,000/ML (\$1,233/AF) are consolidated into "1000+".

## 609 4 Discussion

### 610 4.1 Policy implications

611 This work has a number of important policy implications for regions working to adapt to  
 612 water scarcity through collaborative infrastructure investments. First, these results caution  
 613 against the use of highly aggregated models and mean cost-benefit performance assessments that  
 614 fail to resolve specific multi-actor dynamics within complex water supply systems. Traditional  
 615 capital investment planning frameworks tacitly assume that all partners will benefit equally from  
 616 joint infrastructure investments, or that benefits will be distributed according to historical usage  
 617 patterns. Our results show that investment partnership outcomes are often highly heterogeneous  
 618 across participating water districts, highlighting the importance of disaggregation to ensure that  
 619 investments provide benefits not only to the "average" partner, but to all partners. In the case of

620 the planned Friant-Kern Canal rehabilitation project, results suggest that the business-as-usual  
621 partnership (the Friant contractors) may not be the ideal set of partners, and a wider set of  
622 regional water districts should be considered for participation. More broadly, this work  
623 highlights that contractual details regarding the operation and ownership of shared infrastructure  
624 are crucial design elements on par with physical design parameters such as conveyance capacity.  
625 Neglecting these details (a current standard practice in planning studies) can cause large errors in  
626 predicted performance.

627 Our results also emphasize the interconnectedness of the individual components within  
628 institutionally complex water supply systems. Multiple capital investments under consideration  
629 should be evaluated concurrently based on their interactive and cumulative effects rather than in  
630 isolation. Moreover, the bundling of multiple components into a joint portfolio of investments  
631 can yield synergistic benefits. For example, pairing canal expansion with storage infrastructure  
632 such as groundwater recharge facilities can improve the value of conveyance by increasing local  
633 capacity to store surplus flows from the canal. Groundwater banking can also widen the set of  
634 water districts willing to invest in conveyance by providing a mechanism for districts with poor  
635 local recharge capacity to store their water elsewhere. More broadly, these synergistic effects  
636 support California's vision of a flexible infrastructure network that encourages coordination and  
637 cooperation across scales.

638 However, this interconnectedness also amplifies the challenge of accurately evaluating  
639 capital projects within complex supply networks. For example, the local capacity for  
640 groundwater recharge (both in-district recharge and out-of-district banking) has a large impact on  
641 overall project performance, but information about groundwater recharge capacity across the  
642 region is widely dispersed and, in some cases, non-existent. Moreover, these capacities are  
643 evolving quickly as water districts adapt to new requirements under the Sustainable Groundwater  
644 Management Act (Hanak et al., 2020). This makes it difficult to accurately represent  
645 groundwater recharge within planning models and therefore to evaluate candidate capital  
646 projects. More broadly, this highlights the challenge of modeling an increasingly complex and  
647 interconnected system with data that is siloed and diffuse. The state is working to improve data  
648 availability following the Open and Transparent Water Data Act of 2016 (California Department  
649 of Water Resources, 2021), but more work is needed to inform planning efforts under the WRPI.  
650 This is a ubiquitous challenge for water resources systems globally.

651 Lastly, our results provide a stark picture of the impacts of a drier climate on water  
652 infrastructure investments. Even under expressly optimistic assumptions (e.g., a conservative  
653 viability threshold, a capital project with unusually high subsidies, and a limited range of  
654 hydrologic uncertainty that does not explicitly include climate warming), the vast majority of  
655 candidate partnerships are not viable under the dry scenario. Moreover, the downside risks are  
656 often borne unequally, with a subset of partners bearing the bulk of ill effects in unfavorable  
657 futures. This work thus provides a warning that poorly designed regional water infrastructure  
658 investment partnerships may provide marginal supply resiliency benefits if the future turns out to  
659 be drier than the past. Simultaneously, long-lived debt obligations from new investments  
660 (combined with existing obligations from past investments) can pose serious financial risks for  
661 water districts and their customers if benefits turn out to be lower than expected (Chapman &

662 Breeding, 2014; Jeffrey Hughes et al., 2014). Planning under the WRPI should consider not only  
663 water supply resilience, but also financial resilience for the organizations tasked with providing  
664 affordable water.

#### 665 4.2 Future directions

666 Future work will extend this framework through more advanced solution-generation  
667 techniques (e.g., multi-objective evolutionary algorithms (Maier et al., 2014; Nicklow et al.,  
668 2010; Reed, Hadka, Herman, Kasprzyk, & Kollat, 2013)) and a broader exploration of robustness  
669 under climatic, economic, and regulatory uncertainties (Kasprzyk, Nataraj, Reed, & Lempert,  
670 2013; Lempert, Groves, Popper, & Bankes, 2006; Moallemi, Zare, et al., 2020). Additionally,  
671 future work should consider whether supply and financial portfolios in water-scarce regions can  
672 be made more resilient using infrastructure real options and adaptive pathways (Fletcher,  
673 Lickley, & Strzepek, 2019; Haasnoot, Kwakkel, Walker, & ter Maat, 2013; Herman, Quinn,  
674 Steinschneider, Giuliani, & Fletcher, 2020), flexible partnership design (Gorelick et al., 2019), or  
675 novel financial tools such as environmental impact bonds or index insurance contracts (Brand et  
676 al., 2021; Maestro, Barnett, Coble, Garrido, & Bielza, 2016; Zeff & Characklis, 2013). Lastly,  
677 this work has focused primarily on equality at the water district level (i.e., whether costs are  
678 equally distributed across partners) as opposed to equity and justice (Jafino et al., 2021; Osman  
679 & Faust, 2021) (i.e., whether different water districts and their customers have differing access to  
680 water and differing ability to pay for infrastructure, and how these differences intersect with  
681 economic and political power, racial injustice, and responsibility for historical groundwater  
682 overdraft and subsidence) (Dobbin & Lubell, 2021; Fernandez-Bou et al., 2021; Pauloo et al.,  
683 2020). Explicit consideration of these issues in direct co-production with disadvantaged  
684 communities (Lemos et al., 2018; Minkler, Vásquez, Tajik, & Petersen, 2008) will be an  
685 important extension of this study in light of the WRPI's stated goal of alleviating the growing  
686 water affordability challenge.

#### 687 5. Conclusions

688 Population growth, groundwater overdraft, and climate change represent an  
689 unprecedented challenge to water security in California, the Western US, and other water-  
690 stressed regions around the world. The Water Resilience Portfolio Initiative provides a vision for  
691 bolstering the state's water supplies through an interconnected, collaborative, and flexible *water*  
692 *system of systems* (Newsom et al., 2020), and represents not only a roadmap for California, but  
693 also a potential template for other regions looking to develop their own resilient water supply  
694 portfolios. For such a vision to work, individual water providers within the broader system will  
695 need to collaborate in financing and building new shared infrastructure. However, traditional  
696 planning frameworks based on highly aggregated models and mean cost-benefit estimates are ill-  
697 equipped to evaluate multi-party investment partnerships due to the significant complexities and  
698 uncertainties within the distributed supply network. In this paper, we demonstrate the need to  
699 evaluate partnerships at the level of individual water providers, and the challenge of designing  
700 partnerships that can provide acceptable water supply benefits to each partner relative to its share  
701 of project debt.

702 We leverage a high-resolution water supply simulation model within a parallelized  
703 exploratory modeling framework in order to explore alternative partnership structures, capital  
704 projects, and hydrologic scenarios at an unprecedented scale. Even under explicitly optimistic  
705 assumptions regarding climate change and other uncertainties, we find that vast majority of  
706 alternative partnership structures tend to deliver water supply benefits and financial risks that are  
707 highly uneven, threatening the viability of these cooperative investments. Designing viable  
708 partnerships is especially challenging under drier hydrologic conditions, when insufficient  
709 surplus water is available to warrant the investment in additional conveyance and storage  
710 infrastructure. Additionally, our results demonstrate the synergy between conveyance and  
711 storage for capturing surplus water during peak flow events; partnerships may be more likely to  
712 succeed if they can combine multiple pieces of infrastructure that interact positively.  
713 Importantly, however, partnership design choices (i.e., which water providers are participating,  
714 and how do they distribute the project's debt obligation?) may be even more important to the  
715 success of a partnership than the future hydrology or the type of infrastructure. As a whole, this  
716 research investigates several under-studied challenges in the evaluation and planning of new  
717 infrastructure investment partnerships within complex water supply networks. Confronting these  
718 challenges will be crucial if California and other regions are to achieve their resilient water  
719 portfolio goals.

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## 727 **Open research**

728 All data and code for this project, including figure generation, will be made available on GitHub  
729 and Zenodo prior to publication.

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