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# <sup>1</sup> Highlights

# Rising Temperatures Increase Risk of Soil Salinity and Land Degradation in Water-Scarce Regions

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- We develop a modeling framework for salinity and sodicity under
   changing climate combining SOTE and AWE-GEN models
- We analyze sensitivity of salinity and soil degradation to changes in
   ET and rainfall (season length and extremity)
- Results show that aridity drives elevated salinity. Increases in salinity
   are most sensitive to rising ET.
- Increased ET leads to greater risk of soil degradation as indicated by
   declining saturated hydraulic conductivity

# Rising Temperatures Increase Risk of Soil Salinity and Land Degradation in Water-Scarce Regions

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### 16 Abstract

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Climate change introduces significant uncertainty when assessing the risk of soil salinity in water-scarce regions. We combine a soil-water-salinity-sodicity model (SOTE) and a weather generator model (AWE-GEN) to develop a framework for studying salinity and sodicity dynamics under changing climate definitions. Using California's San Joaquin Valley as a case study, we perform first-order sensitivity analyses for the effect of changing ET (a proxy for changing temperature), length of the rain season, and magnitude of extreme rainfall events. Higher aridity, through increased ET, shorter rainy seasons, or decreased magnitude of extreme rainfall events, drives higher salinity – with rising ET leading to the highest salinity levels. Increased ET leads to lower levels of soil hydraulic conductivity, while the opposite effect is observed when the rainfall season length is shortened and extreme rainfall events become less intense. Higher ET leads to greater unpredictability in the soil response, with the overall risk of high salinity and soil degradation increasing with ET. While the exact nature of future climate changes remains unknown,

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the results show a serious increase in salinity hazard for climate changes within the expected range of possibilities. The presented results are relevant for many other salt-affected regions, especially those characterized by intermittent wet-dry seasons. While the San Joaquin Valley is in a comparatively strong position to adapt to heightened salinity, other regions may struggle to maintain high food production levels under hotter and drier conditions.

17 Keywords: irrigation, salt-affected, climate change, sodicity, hazard,

18 agriculture

#### 19 1. Introduction

Soil salinity and sodicity present major challenges to agricultural 20 production (Howitt et al., 2009; Wallender and Tanji, 2011; Qadir et al., 21 2014; FAO and ITPS, 2015; Daliakopoulos et al., 2016; Prăvălie et al., 2021; 22 Kramer and Mau, 2023). High soil salinity inhibits plant water uptake, 23 leading to declining yields and plant death (McGeorge, 1954; Bernstein, 24 1975; Maas and Grattan, 1999; Munns, 2002). High sodicity levels can 25 trigger the breakdown of soil aggregates, limiting the flow of water and air 26 to the root zone, thereby threatening plant growth (McGeorge, 1954; 27 Mandal et al., 2008; Levy, 2011; Bardhan et al., 2016). Critically, 28 experimental and field evidence has indicated that breakdowns in soil 29 aggregates are extremely difficult to reverse, in many cases causing 30 permanent soil degradation (Bhardwaj et al., 2008; Assouline and Narkis, 31 2011; Schacht and Marschner, 2015; Adevemo et al., 2022). 32

<sup>33</sup> The threats of salinity and sodicity are especially pronounced in

water-scarce regions (FAO, 2023). Due to limited freshwater supplies, food 34 production in these areas often depends on irrigation with high salinity 35 water, including treated wastewater and saline groundwater (Oster, 1994; 36 Bixio et al., 2006; Levy, 2011; Assouline et al., 2015). With domestic water 37 needs typically prioritized over the agricultural sector's, reliance on high 38 salinity irrigation water is expected to increase over the coming decades 39 (Oster, 1994; Bixio et al., 2006; Levy, 2011; Assouline et al., 2015; Kramer 40 et al., 2022b). 41

Climate change introduces an additional element of uncertainty when 42 forecasting the risk of salinity-induced damage to agriculture. Rising 43 temperatures and changes to annual rainfall have the potential to further 44 aggravate water scarcity, pushing growers to even greater dependence on 45 high salinity irrigation supplies – at a time when plants are already facing 46 more intense heat stress and atmospheric demand. In areas with distinct 47 dry and wet seasons, rainfall is often critical in the natural leaching of salts 48 that accumulate from irrigation (Lado et al., 2012). Changes in rainfall 40 patterns (e.g., shorter rainfall season, reduction in rainfall amounts, or 50 increase in intermittency between storms) could disrupt these processes, 51 leading to a potential rise in average soil salinity levels, and putting the soil 52 at risk of long-term, irreversible degradation. 53

We seek to understand how the dynamics of salinity and sodicity in water-scarce regions are most likely to be affected by changing rainfall and temperature patterns. While the impact of salinity and sodicity on plants and soils has been closely studied (Minhas et al., 2020; Kramer and Mau, 2023), the effects of climate change on salinity and sodicity have received

limited attention. Most research on the intersection of agriculture and 59 salinity has focused on preventing salinity-driven damage to groundwater 60 and other natural water resources (Knapp, 1992a,b,c; Dinar et al., 1993; 61 Hansen et al., 2018; Quinn, 2020). Hassani et al. (2020, 2021) use 62 data-driven models to try and predict how primary soil salinity (i.e., 63 salinity caused by natural processes) will change over the 21st century. 64 Their models, however, do not apply to secondary salinity (salinity driven 65 by human activities), such as the irrigation-driven salinity and sodicity that 66 is common in agricultural-producing regions. Kramer and Mau (2020) 67 demonstrated that shorter rainy seasons and an increase in the magnitude 68 of extreme precipitation events have the potential to exacerbate the risk of 69 salinity- and sodicity-driven soil degradation in agricultural settings. While 70 the framework used by (Kramer and Mau, 2020) explores only one specific 71 change in rainfall patterns, without considering feedback loops between 72 salinity levels and the ability of water to move through the soil, the findings 73 underscore the fact that climate change may introduce new conditions that 74 challenge traditional salinity management strategies. Corwin (2021)75 evaluate existing research on the impact that climate change has had up to 76 now. This important review notes that remote sensing is a powerful tool for 77 monitoring salinity development and emphasises the risk that climate 78 change is already presenting in important agricultural regions. It is not, 79 however, a tool for forecasting the effect of specific climate changes on 80 salinity and sodicity dynamics. In the face of such changes, growers who 81 don't adapt may be confronted with declining yields and an increased risk of irreversible soil degradation. Given this possibility, we must develop a <sup>84</sup> core understanding of how anticipated changes in climate may affect
<sup>85</sup> salinity and sodicity trends so that policymakers and extension specialists
<sup>86</sup> can adequately prepare growers to face new challenges.

#### <sup>87</sup> 1.1. Case study: the San Joaquin Valley

As a case study for the effects of climate change on salinity and sodicity, 88 we focus on California's San Joaquin Valley (SJV). In addition to being one 89 of the most important agricultural areas in the United States, the SJV is an 90 apt case study because severely limited freshwater allocations make farmers 91 dependent on often-saline groundwater supplies for irrigation. 92 Salinity-driven environmental damage has been a concern and focus of 93 research in the SJV for more than a century (Nelson et al., 1918; Eaton, 94 1935; Tanji et al., 1972; Amundson and Smith, 1988; Fujii et al., 1988; 95 Tidball et al., 1989; Lin et al., 2000; Hanson and May, 2003; Mitchell et al., 96 2017; Hansen et al., 2018; Corwin, 2021). The focus of these studies has 97 ranged from remediation of salt-affected lands (Amundson and Lund, 98 1985), surveying the extent of existing salinity damage (Scudiero et al., 99 2014; Thellier et al., 1990), the hydrological roots of saline groundwater 100 (Schoups et al., 2005), and mapping root zone salinity using remote sensing 101 in response to climate changes (Corwin, 2021). We are unaware of any 102 studies, however, that have considered the role of future climate conditions 103 on salinity and sodicity dynamics. 104

The present SJV climate is characterized as warm-summer Mediterranean (Csb) by the Köppen-Geiger classification (Peel et al., 2007), with a rainy winter season from November to April that yields an average annual precipitation of 275 mm. Summers in the SJV are warm and dry with virtually no rainfall and a mean daily temperature of 24.6 °C.
This contrast between a wet winter season and a dry summer season is
typical of many salt-affected regions.

While climate models project with a high level of certainty that 112 temperatures in the SJV will rise over the remainder of the 21st century, 113 they are unclear about the precise magnitude (Pierce et al., 2013). 114 Projected changes in precipitation patterns are marked by much higher 115 levels of uncertainty, partly because inter-annual variability in rainfall 116 amounts in the region is already high (Pierce et al., 2013). Among the most 117 common probable climate projections are (i) a decrease in the overall length 118 of the winter rainfall season and (ii) intensification of extreme rainfall 119 events. The latter is primarily driven by temperature increases (Peleg 120 et al., 2020; Marra et al., 2024), and therefore is highly probable (Fowler 121 et al., 2021) even if precipitation levels remain unchanged. 122

We examine how incremental changes in each of these variables – 123 temperature, rainfall season length, and the magnitude of extreme rainfall 124 events – are likely to impact the hazard of salinity-induced crop damage 125 and sodicity-induced soil degradation in irrigated lands. While our focus on 126 the SJV reflects its central role in US food production, we would like to 127 point out that many other important agricultural regions across the US 128 Southwest and Midwest, along with other agricultural regions worldwide, 129 share similar climate profiles and are susceptible to similar pressures as a 130 result of water scarcity (Corwin, 2021). 131

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#### 132 2. Material and methods

# <sup>133</sup> 2.1. Modeling salinity, sodicity, and hydraulic conductivity dynamics

Changes in soil salinity and sodicity, and how they affect saturated 134 hydraulic conductivity, are modeled using the Salt of the Earth 2.0 (SOTE) 135 model (Kramer et al., 2022a). SOTE focuses on how irrigation (chemical 136 composition and application rates) and climate conditions (precipitation 137 and evapotranspiration fluxes) drive the dynamics of relative soil water 138 content, the electrolyte concentration of the soil water (i.e., salinity), and 139 the fraction of sodium ions in the soil's exchange complex (i.e., sodicity). 140 As the dynamics of these three state variables evolve, SOTE includes 141 feedback with saturated hydraulic conductivity,  $K_s$ . In contrast to other 142 models that track salinity and sodicity dynamics (Simunek and Suarez, 143 1994; Simůnek et al., 2013; Kroes et al., 2017; Ma et al., 2012; Russo, 1984, 144 1988; Russo et al., 2004; Russo, 2013; van der Zee et al., 2010; Shah et al., 145 2011; van der Zee et al., 2014; Kramer and Mau, 2023), SOTE includes the 146 potential for irreversible effects when modeling increases and decreases in 147 soil  $K_s$ . This is important because experimental and field evidence has 148 demonstrated that changes in  $K_s$  are marked by hysteresis (Bhardwaj 149 et al., 2008; Assouline and Narkis, 2011; Schacht and Marschner, 2015; 150 Adeyemo et al., 2022). The exclusion of hysteresis in  $K_s$  has been 151 demonstrated to significantly lower the forecasted probability of long-term 152 soil degradation, making its inclusion critical for understanding the actual 153 risks to soil health (Kramer and Mau, 2020; Kramer et al., 2022a). 154 Therefore, declines in saturated hydraulic conductivity are often used as a 155 metric for soil degradation. SOTE can be used to investigate the effect of 156

climate change on salinity and sodicity dynamics by modifying rainfall and actual evapotranspiration of the crop under non-standard conditions ( $\mathrm{ET}_{\mathrm{c} \ \mathrm{act}}$ , mm d<sup>-1</sup>) (Fernández, 2023). In this setup,  $\mathrm{ET}_{\mathrm{c} \ \mathrm{act}}$  is a proxy for the effects of temperature. All references to ET in the remainder of this article refer to  $\mathrm{ET}_{\mathrm{c} \ \mathrm{act}}$ .

The salinity, sodicity, and water dynamics in the SOTE model were successfully validated against results from a multiyear lysimeter experiment involving different irrigation water qualities and precipitation (Gonçalves et al., 2006). The hydraulic conductivity module used in SOTE has been validated through laboratory experiments (Adeyemo et al., 2022; Kramer et al., 2021). The SOTE model has also been used to examine plant responses to salinity and sodicity (Yin et al., 2021, 2023).

# 169 2.2. Generating stochastic weather

The present and future rainfall and evapotranspiration time series were 170 generated using the 1-dimension version of the AWE-GEN (Advanced 171 Weather Generator) model (Fatichi et al., 2011; Ivanov et al., 2007). This 172 hourly weather generator is capable of reproducing the key climatic 173 variables required for agro-hydrological applications, such as precipitation, 174 cloud cover, temperature, radiation, and humidity, while preserving their 175 temporal correlations. The low- and high-order statistics of the generated 176 time series are realistically emulated by employing physically-based and 177 stochastic approaches. For example, the precipitation module is based on a 178 Poisson-cluster process, while the near-surface air temperature module 179 includes a stochastic component to generate the hourly time series 180 according to the diurnal cycle and seasonality, physically constrained with 181

the hourly cloud cover and radiation budget. Readers are referred to (Fatichi et al., 2011) for more information regarding AWE-GEN; Fatichi et al. (2013) provides an overview of model parameterization for future climate conditions. AWE-GEN is a robust model that has been used to generate long and non-stationary time series of climatic variables for multiple applications (e.g., (Fatichi et al., 2021; Cache et al., 2023; Ramirez et al., 2023)).

The model was calibrated to generate ET and rainfall hourly time series using 40 years of ERA5 climate data (1980 to 2020) for Fresno County in the SJV. The model was validated against measured values over the same period (Supplemental Materials 1). To account for the natural climate variability (inter- and intra-annual variations), we generated 500 unique years of baseline rainfall and ET data (Fig. 1).

#### 195 2.3. Simulations framework

The objective of our study is to understand how long-term trends in 196 salinity and sodicity dynamics will be affected by potential changes in 197 climate. To facilitate this goal, we use the one-at-a-time technique where 198 the effect of one parameter (evapotranspiration, rainfall season length, 199 rainfall intensity) is analyzed while keeping the others fixed. In this local 200 sensitivity analysis approach (Razavi and Gupta, 2015), variations in 201 output are then a measure of how susceptible the system is to changes in 202 that particular input variable. Such a framework enables the 203 straightforward identification of potential trends, e.g., the effect of 204 increasing ET on overall salinity or saturated hydraulic conductivity levels, 205 while avoiding the intense computational demands of a global sensitivity 206



Figure 1: The 500 years of stochastic simulations of rainfall (a; grey lines) and ET (b; grey points) generated using the AWE-GEN model. Orange lines highlight the 50 years in which total annual rainfall was between the 45th and 55th percentiles of the ensemble. Orange points are elevated ET by 1.5 mm  $d^{-1}$ .

analysis. It also allows us to probe for the existence of "cutoff thresholds" –
points beyond which irreversible soil degradation might occur.

To account for natural variations in climate, the simulations are divided into scenarios, each composed of a unique set of input conditions. Each scenario is made up of a stochastic ensemble of 50 climatic realizations, each realization 15 years long, sampled from the pool of 500 unique data years (a similar conceptual framework as suggested by (Fatichi et al., 2016)). For each stochastic realization in the ensemble, the results consider only the average conditions over the final three years of the 15-year simulation period. This approach minimizes the impact of any one extreme year or set of climate conditions, while also highlighting the role of natural climate variations on the set of final results. Focusing on average conditions at the end of the simulation period is important because changes in salinity and sodicity levels sometimes take several years to manifest and stabilize.

In the results that follow, we focus on the following groups of scenarios, describing changes in evapotranspiration, rainfall season length, and extreme rainfall events intensity. All three groups share the same baseline scenario (Sec. 2.2), against which each treatment is compared.

Evapotranspiration. We describe nine scenarios, corresponding to additive changes between -0.5 and +1.5 mm d<sup>-1</sup> with respect to the baseline ET, with increments of 0.25 mm d<sup>-1</sup>. To minimize variation due to annual rainfall, these simulations use only the 50 colored trajectories in Fig. 1a. The annual precipitation for each of these trajectories was within 10 percent of the median annual total.

**Rainfall season length.** The baseline length of 190 days was multiplied 231 by a factor between 0.6 to 1.2, with 0.1 increments, totaling seven scenarios. 232 **Extreme rainfall events** The highest 20% of rainfall events for each year 233 were multiplied by a factor ranging from 0.5 to 2.0, with 0.25 increments. 234 The smallest 20% of rainfall events were multiplied by the inverse of the 235 factor. Within each group of scenarios, the simulations start using the same 236 random seed, such that the hourly ET and rain inputs are identical across 237 the groups, with the only differences due to the applied rainfall/ET factors. 238 In discussing the results, we introduce a modified aridity index. Because 239 ET and precipitation can both vary across the simulation sets, the aridity 240

index is useful as a single metric for changes in water stress. Here, the aridity index is defined as the ratio of total evapotranspiration to the sum of all precipitation and irrigation inputs, i.e., higher values correspond to more arid conditions. The other input parameters used to run the simulations, including soil physical and chemical properties and the chemical composition of the irrigation water, are presented in Supplementary Materials 2.

### 247 **3. Results**

# 248 3.1. Effects of changing ET on soil system

The simulations reveal a multi-faceted relationship between changing ET and the health of the soil system (Fig. 2). While salinity increases linearly with ET, the effects of ET on soil degradation are more varied, such that rising ET leads to higher unpredictability in relative  $K_s$ . Likewise, the relationship between relative  $K_s$  and salinity evolves as ET changes, eluding simple classification.

Fig. 2a shows the non-linear relationship between salinity and relative 255 saturated hydraulic conductivity. As salinity increases, relative  $K_s$  values 256 initially decline. When salinity levels exceed 200  $\text{mmol}_{c} L^{-1}$ , however, this 257 trend reverses: relative  $K_s$  begins to increase and eventually surpass the  $K_s$ 258 values observed when salinity was lowest. We can also see that the 259 relationship between salinity and relative  $K_s$  changes as aridity increases. 260 Because variations in total rainfall in this set of simulations were limited, 261 aridity index values are primarily a function of the input ET. We observe 262 that the least desired results — high salinity and decreases in relative  $K_s$ 263 (at around 200  $\rm mmol_c\,L^{-1})$  – occur as aridity increases. As aridity 264

increases, we also note that there is higher variability in the scatter of 265 salinity and relative  $K_s$ ; the lowest aridity values (purple) are grouped 266 closely together, while the high aridity (yellow) points are more spread out. 267 The sensitivity of salinity and sodicity to aridity is further explored in 268 Fig. 2b-c. There is a significant linear relationship between increasing aridity 269 and salinity in Fig. 2b ( $\mathbb{R}^2$ : 0.95, p < 0.05), with distinct clouds of points 270 corresponding to the incremental jumps in input ET used in the simulations. 271 Fig. 2c presents a significant negative, but less intense, trend in relative  $K_s$ 272 as aridity increases (R<sup>2</sup>: 0.52, p < 0.05), and emphasizes how relative  $K_s$  is 273 prone to greater unpredictability as aridity increases. 274

# 275 3.2. Effects of changing rainfall season length on soil system

The simulation results show that longer rainfall seasons (lower aridity) 276 lead to a noticeable decline in overall salinity and slight drops in relative  $K_s$ 277 (Fig. 3), and vice versa. These relationships are weak, however, in comparison 278 to those observed when analyzing the effects of ET (note the scale differences 279 between Fig. 2 and Fig. 3). Changes in rainfall season length lead to smaller 280 ranges in aridity index, effectively leading to a less extreme set of climate 281 conditions. Yet even within this limited range of aridity, the relationship 282 between aridity and salinity and relative  $K_s$ , respectively, is less intense. 283 Fig. 3b-c show that there is a wide scatter around the regression line for 284 both salinity and relative  $K_s$ , indicating a wide range of potential salinity 285 and relative  $K_s$  values for each aridity index value. This is further reflected in 286 the relatively low  $\mathbb{R}^2$  values for the relationship between salinity and aridity, 287 and between relative  $K_s$  and aridity (0.39 and 0.12, respectively). 288



Figure 2: The effects of rising ET on the soil system. (a) The non-linear relationship between salinity and relative  $K_s$ . (b) the positive effect of rising ET on salinity (R<sup>2</sup>: 0.95, p < 0.05). (c) The negative relationship between ET and relative  $K_s$  (R<sup>2</sup>: 0.52, p < 0.05). Black lines are linear regression.

# 289 3.3. Effects of extreme rainfall on soil system

We found that an increase in the magnitude of the extreme rainfall events leads to lower salinity and lower values of relative  $K_s$  (Fig. 4). The heavy



Figure 3: The effect of changes in rainfall season length on the soil system. (a) The relationship between salinity and relative  $K_s$ . (b) The effect of rainfall season length on salinity (R<sup>2</sup>: 0.39). (c) The relationship between aridity and relative  $K_s$  (R<sup>2</sup>: 0.12). Black lines are linear regression.

<sup>292</sup> concentration of purple points in the bottom left of Fig. 4a corresponds to <sup>293</sup> the simulations with the lowest aridity values, which in this case are the simulations with the highest magnitude of extreme rainfall events. Likewise, Figs. 4b-c indicate positive linear relationships between aridity and salinity and aridity and relative  $K_s$  (R<sup>2</sup> values of 0.39 and 0.46, respectively).



Figure 4: The effect of changes in extreme rainfall on soil system. (a) The relationship between salinity and relative  $K_s$ . (b) and (c) present the effect of extreme rainfall on salinity (R<sup>2</sup>: 0.39) and relative  $K_s$  (R<sup>2</sup>: 0.46), respectively. Black lines are linear regression.

#### <sup>297</sup> 4. Discussion

# 298 4.1. Shifting dynamics as a result of changes in ET

This shifting response of relative  $K_s$  dynamics observed in Sec. 3.1 is 299 not entirely surprising given previous work on the effects of salinity and 300 sodicity on  $K_s$ . Several experimental and modeling studies have 301 demonstrated that seasonal fluctuations in salinity — typically as a result 302 of high salinity irrigation water applied during dry months being leached by 303 winter rainfall — have the potential to increase the risk of soil degradation 304 (Shainberg and Shalhevet, 1984; van der Zee et al., 2014; Mau and 305 Porporato, 2015; Kramer and Mau, 2020). This occurs because the fraction 306 of sodium in the soil exchange complex changes at a slower rate than 307 overall salinity, and degradation is most likely to occur when salinity is 308 moderately low and the sodicity fraction relatively high (Shainberg and 309 Shalhevet, 1984; van der Zee et al., 2014; Mau and Porporato, 2015; 310 Kramer and Mau, 2020). These same studies, however, have demonstrated 311 that extremely high levels of salinity are likely to insulate the soil system 312 against degradation hazards, no matter how high the sodicity fraction. In 313 these cases, extreme salinity levels mask the relatively weak ionic bonding 314 strength of the sodium cations. 315

The similar distribution of the points within each of the clouds in the ET simulations is a feature of the modeling setup. The same random seed was used before each simulation set to restrict variation in the final results to the effect of initial ET (Sec. 2.3). While differences in annual rainfall in this set of simulations were intentionally restricted, most of the variation in results at the selected ET increments can be explained by rainfall (Supplemental 322 Materials 3).

# 323 4.2. Rainfall vs. ET simulations

The results from the rainfall season length (Sec. 3.2) and extreme 324 rainfall (Sec. 3.3) exhibit several differences in comparison to the ET 325 simulations (Sec. 3.1). In the ET simulations the relationship between 326 salinity and relative  $K_s$  switches from a negative correlation to positive as 327 salinity increases. The rainfall simulations, by contrast, feature a 328 consistently negative relationship between the two variables. This is partly 329 explained by the fact that the rainfall and ET simulations showcase a 330 difference in the relationship between aridity and relative  $K_s$ . In the ET 331 simulations, aridity and relative  $K_s$  have a moderately negative correlation, 332 while in the rainfall simulations, the correlation between the two variables 333 is slightly positive. 334

These differences point to important distinctions in how the selected 335 climate variables affect the soil system. The ET simulations experience a 336 wider range of salinity levels than observed in the rainfall simulations (the 337 reader's attention is drawn to the different axis limits in Figs. 2–4). 338 Specifically, the results show that for the scenarios examined, increasing ET 339 drives higher salinity levels than in any of the rainfall simulations. It is also 340 worth making clear that the extreme salinity levels recorded in the ET 341 simulations are beyond the tolerance levels of even the most salt-resistant 342 crops. 343

The model results suggest that farmers under such conditions would have no choice but to (a) spend more water by increasing the leaching fraction to stimulate the leaching of salts from the root zone, (b) search for irrigation water with a less saline chemical composition or, (c) abandon agricultural production altogether. Given that such regions are already facing water scarcity, solutions (a) and (b) will be difficult to apply, while option (c) would endanger food security and economic output.

At the same aridity index values, the rainfall simulations exhibit lower 351 salinity levels compared to the ET simulation. For example, when the 352 aridity index value is 1, the ET simulations show an average salinity of 353 approximately 100 mmol<sub>c</sub>  $L^{-1}$  (Fig. 2b), while the average salinity levels are 354 less than 80 mmol<sub>c</sub>  $L^{-1}$  in the two rainfall simulations when the aridity 355 index is 1 (Figs. 3b and 4b). One possible explanation for this difference is 356 that the increased rainfall drives additional leaching of salts from the root 357 zone. While leaching can certainly contribute to lower salinity values, 358 Sec. 4.4 discusses some potential limitations concerning the model's ability 359 to fully forecast the effects of extreme rainfall. 360

#### <sup>361</sup> 4.3. Impact on soil health hazards

One of the clearest contrasts between the three sets of simulations is 362 how the changing climatic variables affect the overall hazard of dangerous 363 salinity and relative  $K_s$  levels. This point is emphasized in Fig. 5, which 364 presents probability density functions (PDFs) for each of the sets of 365 The PDFs for the ET simulations show the highest levels of scenarios. 366 variation, with rising ET strongly contributing to increased salinity hazards 367 and soil degradation, affecting soil health and agriculture production. In 368 Fig. 5a, the PDFs shift from right to left as ET increases, indicating lower 369 averages for relative  $K_s$ , while in Fig. 5b the PDFs shift from left to right 370 as ET increases, corresponding to elevated salinity levels. In both cases, not 371

only do the PDFs shift to the less desirable range of values, but the PDFs
themselves become flatter, indicating a wider range of potential values –
i.e., that the final results are characterized by higher levels of uncertainty.

These dynamics are present to a less significant degree in the other sets 375 of simulations. As rainfall season length becomes shorter, the PDFs move 376 rightward (Fig. 5c-d), consistent with the higher salinity values observed in 377 Sec. 3.2. The PDFs also become narrower, indicating not only that the model 378 forecasts increased salinity as ET rises, but also a high level of certainty in this 379 outcome. The effect of rainfall season length on relative  $K_s$  has minimal effect 380 on the PDFs, again consistent with the lower correlation observed between 381 aridity and relative  $K_s$  in Fig. 3c. The PDFs for the rainfall extremity 382 simulations exhibit a gradual shift to the left for the relative  $K_s$  output 383 (Fig. 5e-f), while increased rainfall extremity actually causes the salinity 384 PDFs to shift to the left. 385

# 386 4.4. Modeling limitations

The simulations presented here help understand how salinity and 387 sodicity dynamics might be affected by changes in climate, but inherent 388 modeling limitations should be considered when assessing the results. The 389 simulations were intentionally narrow in scope, focusing on sensitivity to a 390 single climate feature at a time. While this approach is important for 391 building initial understandings, it is more likely that future climate 392 conditions will involve parallel changes to rainfall duration and intensity, 393 ET, and possibly other variables. Future research should explore how 394 interactions between climate variables will affect the system as a whole. 395 While such an investigation is within the capabilities of the combined 396



Figure 5: The probability distribution functions (PDFs) for relative  $K_s$  results for (a) ET, (c) rainfall season length, and (e) rainfall extremity simulations. (b), (d), and (f) present the PDFs for salinity values for the same sets of simulations.

<sup>397</sup> SOTE-AWE-GEN framework, it is beyond the scope of this study.

Likewise, the present analysis focuses on changes to the soil root zone with 398 little attention to the interaction between different layers of the soil profile 399 or the potential effect of rainfall itself on a soil's physical conditions. While 400 SOTE is not by definition restricted to the analysis of specific soil depths, 401 it is a bucket model and therefore less amenable to studying interactions 402 between the upper and lower layers of the soil profile. We focused on the 403 upper layers of the soil profile since changes in salinity and infiltration rates 404 in the zone present an immediate risk to crop production. Attention to 405 lower layers of the profile, however, could be especially important in cases 406

where groundwater infiltration is of concern. Furthermore, the simulations 407 in Sec. 3.3 focused on extreme rainfall events without analyzing the potential 408 effects of impact force itself on the soil. It is well understood that extreme 409 rainfall can lead to dispersion of the particles on the soil surface, including 410 the breakdown of soil aggregates, such that infiltration rates and overall 411 hydraulic conductivity are both impacted (Assouline, 2004). To increase 412 our understanding of how extreme rainfall might affect salinity and sodicity 413 dynamics, the incorporation of these phenomena should be considered an 414 important next step. 415

# 416 5. Conclusion

We analyzed the first-order sensitivity of salinity and sodicity dynamics 417 to changes in ET, rainfall season length, and extreme rainfall. While 418 increased aridity leads to higher salinity levels in all three sets of 419 simulations, the response of relative  $K_s$  showed mixed behavior – with 420 increased aridity leading to lower relative  $K_s$  in the ET simulations, and 421 slightly higher relative  $K_s$  in the rainfall simulations. Changes in 422 temperature (ET) led to the largest variation in output levels, with higher 423 ET contributing to wider distribution in final salinity and relative  $K_s$ . 424

<sup>425</sup> Climate models have consistently pointed to a likely rise in temperature <sup>426</sup> and ET in the Fresno area, underscoring the importance of understanding <sup>427</sup> how these changes may affect soil health. The exact nature of any future <sup>428</sup> climate will of course depend on government policy, technological <sup>429</sup> developments, and potential feedback between climate variables. However, <sup>430</sup> a substantial rise in temperature and ET, on the order of that explored in this research, is well within the range of possible changes, presenting a
potentially serious threat to agricultural production.

The analyses here used the San Joaquin Valley as a case study, but the 433 results are a bellwether for other agriculturally important parts of the US 434 and beyond. Farmers throughout the rest of California, the American 435 Southwest, and large portions of the Midwest are similarly confronted by 436 the challenges of declining freshwater access and expected temperature 437 increases, while simultaneously facing pressure to improve crop yields as 438 food demand grows. Furthermore, the general climate patterns in Fresno 439 County – hot and dry summer growing seasons; seasonal rainfall during the 440 winter months – are common in other regions affected by salinity and 441 sodicity hazards, including large portions of the Middle East and North 442 Africa, the Indian sub-continent, and Australia (Kramer and Mau, 2023; 443 FAO, 2023). 444

What most separates the San Joaquin Valley from these other regions is 445 the California agricultural sector's relatively strong ability to cope with 44F climate-driven challenges. Traditionally, the most effective ways of 447 mitigating salinity hazards are irrigation with higher quality (low-salinity) 448 input water, intentional over-irrigation designed to leach salts from the root 449 zone, and transition to more salt-tolerant crops and varieties. Many 450 growers in the San Joaquin Valley focus on high-revenue specialty crops, 451 providing them with the capital necessary to invest in high-efficiency 452 irrigation systems, advanced monitoring capabilities, and automation 453 equipment – all of which can contribute to water conservation. Likewise, 454 these growers are more capable of transitioning to salt-resistant varieties. 455

Several local, state, and national funding programs provide further financial 456 aid and direct incentives to farmers interested in technological upgrades. 457 Abundant government funding can help support investment in alternatives 458 such as desalination and treated wastewater, which can provide 459 supplemental sources of irrigation water when freshwater is limited. On the 460 other hand, coping with the challenges of salinity and sodicity will be much 461 more challenging in less wealthy regions, where investment in new 462 technologies is less affordable for most food producers, and where 463 governments are less capable of funding water infrastructure projects. 464

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