

1 **Title:** Flow directions and ages of subsurface water in a salt marsh system constrained by isotope  
2 tracing

3 **Running title: Isotope tracing and water ages in a salt marsh**

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26  
27 **Abstract:**

28 Salt marshes are dynamic hydrologic systems where terrestrial groundwater, terrestrial  
29 surface water, and seawater mix due to bi-directional flows and pressure gradients. Due to the  
30 counteracting terrestrial and marine forcings that control these environments, we do not  
31 comprehensively understand water fluxes in these complex coastal systems. To understand the water  
32 sources, flow directions, and velocities in salt marsh porewater, we employed a combination of  
33 geochemical tracers and analytical models across a hillslope-to-salt marsh continuum in a salt marsh  
34 experiencing daily inundation of estuarine surface water (SW) from tides and mixing of fresh  
35 seasonal groundwater.

36 We used tritium (<sup>3</sup>H) as a hydrologic tracer to assess porewater ages and stable water isotope  
37 ( $\delta^2\text{H}$  and  $\delta^{18}\text{O}$ ) analyses to separate isotopically distinct estuarine and terrestrial groundwater across  
38 different depths and landscape positions in the study transect. We employed electrical conductivity  
39 to constrain the role of source mixing and evapotranspiration in salt marsh hydrology. Salinity and  
40 stable isotopes revealed that transpiration, rather than evaporation, increased subsurface water  
41 salinity to concentrations above estuarine SW during summer. Elevated salinity at depth indicated  
42 that salt marsh subsurface water is recharged during the dry growing season. Seasonal recharge  
43 patterns drive long-term deep subsurface water dynamics across the salt marsh, with <sup>3</sup>H ages of 3-7  
44 years, and daily tidal cycles drive short-term shallow porewater dynamics with <sup>3</sup>H ages of  $0 \pm 3.6$   
45 years. Our conceptual understanding of the spatiotemporal changes in SW-subsurface water

46 interactions at the terrestrial-marine interface quantifies the hydrological constraints we are missing  
47 to improve our understanding of biogeochemical cycles within the salt marsh.

48

49 **Key words**

50 Salt marsh hydrology, coastal hydrology, porewater exchange, isotopic tracer, tritium age

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52

## 53 1- Introduction

54 Salt marshes are dynamic hydrologic systems where terrestrial groundwater (GW), terrestrial  
55 surface water, and seawater mix. These systems play an essential role in global biogeochemical  
56 cycles, promoting carbon storage and nitrogen removal due to the saturated conditions resulting  
57 from frequent inundation (Robinson *et al.*, 2018). Further, salt marsh systems can act as buffers  
58 between the terrestrial-marine interface whereby terrestrially derived nutrients may be retained or  
59 processed, potentially lessening their effects on coastal environments (Kumar *et al.*, 2019). These  
60 processes are paramount, as excess nutrients released to coastal waters can enhance eutrophication  
61 and hypoxia (Peterson *et al.*, 2016), which may worsen with expected shifts in climatic patterns  
62 (Sinha *et al.*, 2017), and projected population growth in coastal areas (Neumann *et al.*, 2015). Thus,  
63 sustainable management of coastal waters is a critical environmental challenge due to these climatic  
64 and anthropogenic pressures on coastal zones (Ferguson and Gleeson, 2012). These challenges  
65 highlight the need for an improved understanding of terrestrial-marine interactions (Day *et al.*, 2008;  
66 Michael *et al.*, 2013), including water exchanges across the terrestrial-marine interface, particularly in  
67 salt marsh systems (Borja, 2005).

68 The hydrology of salt marshes is influenced by ecological and biogeochemical processes  
69 (Wilson *et al.*, 2015b), and it is complex due to the combined effect of diurnal, tidally driven water  
70 level oscillations (Grande *et al.*, 2022a), subsurface heterogeneity (Moffett *et al.*, 2012), and variable  
71 elevation gradients (e.g., microtopography; (Wang *et al.*, 2021b). For example, in lower marsh  
72 positions, along tidal channels, tidally driven water level fluctuations in tidal creeks induce hydraulic  
73 gradients that result in water circulation into and out of creek banks (Xin *et al.*, 2011). Further,  
74 secondary porosity due to animals that burrow in these low marsh areas (e.g., crab bioturbation;  
75 (Guimond *et al.*, 2020) impacts near-creek hydrology and can increase surface water-subsurface  
76 water exchanges (Xiao *et al.*, 2019). At higher elevations, middle and upper marsh positions are  
77 influenced by vertical hydraulic gradients caused by tidal inundation, evapotranspiration,  
78 precipitation, terrestrial runoff from contributing hillslopes, and terrestrial GW inputs (Xin *et al.*,  
79 2013).

80 Several methods have been used to study surface water-GW interactions in coastal areas  
81 (Burnett *et al.*, 2006), which span various spatiotemporal scales (Guimond and Tamborski, 2021).  
82 Further, various methods tend to explain different driving forces. For local-to-regional spatial scales  
83 and over tidal cycles, seepage meters are an effective method for measuring surface water-GW  
84 exchanges directly (Rosenberry *et al.*, 2020). Hydraulic heads measured in groundwater wells at  
85 various depths and spatial extents are commonly used with Darcy's law to calculate surface water-  
86 GW exchanges (Wilson *et al.*, 2015a). The spatiotemporal resolution of groundwater wells is  
87 controlled by the spatial (and temporal) distribution of the wells. Numerical models have also been  
88 used across a broad spatiotemporal range (Reeves *et al.*, 2000). Naturally occurring tracers were  
89 helpful in studying surface water-GW exchanges in salt marshes (Xin *et al.*, 2022). For example,  
90 salinity mass balances provide quantifiable information across various spatiotemporal scales (Michael  
91 *et al.*, 2013). Isotope tracers, such as radon and radium isotopes, are widely used over short  
92 (Tamborski *et al.*, 2017; Coluccio *et al.*, 2021; Chen *et al.*, 2022) and long timescales (McKenzie *et al.*,  
93 2021) to understand rates of subsurface-surface water exchanges. The stable isotopes of hydrogen  
94 and oxygen in water ( $\delta^2\text{H}$  and  $\delta^{18}\text{O}$  values) have also been used to study surface water-GW  
95 exchanges in coastal areas (Schmidt *et al.*, 2011; Wang *et al.*, 2021a).

96 Water stable isotopes ( $\delta^2\text{H}$  and  $\delta^{18}\text{O}$  values) are conservative tracers that help to explain  
97 mixing processes at the terrestrial-marine interface, aided by the significant isotopic contrast  
98 between terrestrial waters and seawater (Povinec *et al.*, 2008). Further,  $\delta^2\text{H}$  and  $\delta^{18}\text{O}$  values in water  
99 can trace local (Grande *et al.*, 2020), regional (Bowen *et al.*, 2022), and global hydrologic flowpaths

100 (Jasechko *et al.*, 2014). Thus, the relationship between  $\delta^2\text{H}$  and  $\delta^{18}\text{O}$  values is a practical tool for  
101 understanding water mixing at the terrestrial-marine interface (Debnath *et al.*, 2019). In addition,  
102 unlike salinity, which is affected by both evaporation and transpiration,  $\delta^2\text{H}$  and  $\delta^{18}\text{O}$  values are  
103 only altered by evaporation (Zhang *et al.*, 2010; Barbeta and Peñuelas, 2017). Therefore,  $\delta^2\text{H}$  and  
104  $\delta^{18}\text{O}$  values have the potential to constrain evapotranspiration in salt marsh subsurface water and, in  
105 combination with salinity, understand the partitioning between evaporation and transpiration.

106 Tritium ( $^3\text{H}$ ) in water is an additional isotope tracer that offers potential for studying surface  
107 water-GW exchanges at the terrestrial-marine interface.  $^3\text{H}$  is the radioactive isotope of hydrogen  
108 (half-life = 12.3 years), and its radioactive decay results in a predicted activity concentration with  
109 time in GW from the moment water recharges the terrestrial aquifer, providing direct information  
110 on water residence time (Price *et al.*, 2003; Visser *et al.*, 2013). For example,  $^3\text{H}$  has been used in  
111 coastal studies in combination with its decay product  $^3\text{He}$  to calculate the apparent age distribution  
112 of the discharging GW offshore of south-eastern Sicily, Italy (Povinec *et al.*, 2006). The age of GW is  
113 essential for understanding subsurface flowpaths (Visser *et al.*, 2007), which can aid in understanding  
114 biogeochemical transformation processes in groundwater (Visser *et al.*, 2009). However, it is  
115 challenging to measure GW age distributions directly (Ekwurzel *et al.*, 1994). Another methodology  
116 to use  $^3\text{H}$  to calculate subsurface water ages is to measure  $^3\text{H}$  in local precipitation and calculate the  
117 decay time (Harms *et al.*, 2016). This technique is widely used in watershed hydrology to calculate  
118 stream water age distribution (Visser *et al.*, 2019; Grande *et al.*, 2020; Campbell *et al.*, 2021), and it  
119 offers potential for studying shallow subsurface water ages in salt marsh systems.

120 To advance our understanding of surface water-GW exchanges at the terrestrial-marine  
121 interface, we used a combination of isotopic and geochemical tracers (e.g.,  $\delta^2\text{H}$ ,  $\delta^{18}\text{O}$ ,  $^3\text{H}$ , and  
122 electrical conductivity) with analytical models that can help us understand the residence time  
123 distribution and the water sources across a salt marsh platform in the Elkhorn Slough National  
124 Estuarine Research Reserve in central California, USA. This Mediterranean system experiences  
125 semidiurnal tidal inundation of estuarine surface water in addition to fresh GW inputs, direct  
126 precipitation, and surface water inputs (e.g., agricultural drainage), which makes characterizing  
127 subsurface flowpaths challenging.

128 Our overarching objective was to develop an empirically informed conceptual model of  
129 water inputs/outputs and residence times in a salt marsh system by identifying what we hypothesize  
130 are the three dominant processes/drivers of water movement. First, we tested whether marked  
131 seasonal patterns in precipitation, common to the site's Mediterranean climate, affected terrestrial  
132 GW contributions to the salt marsh. Specifically, we studied the subsurface water flow directions  
133 and water sources across a hillslope-to-low marsh continuum. Secondly, to improve our  
134 understanding of subsurface flow velocities of salt marsh porewater, we calculated subsurface water  
135 ages and vertical recharge rates. Thirdly, we evaluated the relative role of evaporation and plant  
136 transpiration in increasing porewater salinity in the salt marsh.

## 137 2- Study area and measurements

138 This study was conducted at Elkhorn Slough (ES) in Monterey Bay, California (Figure 1A).  
139 The principal sources of surface freshwater to ES is Carneros Creek, an ephemeral stream that only  
140 flows during the wet winter months, and the Old Salinas River, a perennial stream, which discharges  
141 at the mouth of the slough (Caffrey and Broenkow, 2002). Other sources of freshwater are direct  
142 precipitation, and surface runoff via intermittent streams and channels that drain into the Slough  
143 (Figure S1).

144 Tides in the estuary are mixed semidiurnal with a mean range of 1.7 m, a spring tidal range  
145 of 2.5 m, and a neap tidal range of 0.9 m. The principal transport mechanism for water and nutrients

146 in ES occurs via tidal exchange (Caffrey and Broenkow, 2002). Monterey Bay seawater reaches up to  
147 6 km inland during high tides, and over 50% of the total water volume of the slough is flushed  
148 during each tidal cycle (Malzone, 1999).

149 The average precipitation in ES is 627 mm/year (based on 2001-2022 record), with ~ 90%  
150 of the precipitation falling between November and April as rain (Chapin *et al.*, 2004). Air  
151 temperature averages 11.1 °C in the winter and 15.4 °C in the summer (Caffrey, 2002). The  
152 Mediterranean climate of the study site results in marked wet/dry seasonal dynamics (Figure 2),  
153 which provide the conditions to resolve seasonal variations in climatic forcing that impact  
154 subsurface saturation and mixing of fresh and saline waters. In this area, the wet periods occur  
155 during the dormant winter season, while the dry periods occur during the summer growing season.  
156 Pickleweed, *Salicornia pacifica*, is the dominant marsh plant (Van Dyke and Wasson, 2005), and the  
157 dominant grazer and bioturbator is the lined shore crab, *Pachygrapsus crassipes* (Beheshti *et al.*, 2022).

## 158 **2.1 Experimental Transect and hydrologic parameters measured**

159 For this study, we focused on a 60 m experimental transect across a hillslope-to-salt-marsh  
160 continuum located in the upper northwest section of ES (Figure 1). The salt marsh section extends  
161 over the lower ~24 m of the transect with an elevation range of 0.24 m (all elevations relative to  
162 NAVD88). The salt marsh section of the transect is tidally influenced, with a local tidal range of -  
163 0.54 to 2.21 m. We delineated the salt marsh by elevation into upper (1.71 m), middle (1.65 m), and  
164 lower (1.59 m) positions based on differences in average tidal inundation duration: 5.4%, 7.2%, and  
165 9.6%, respectively (Figure 1 B-C). Further, we demarcated an upland position ~6 m inland from the  
166 upper marsh within the terrestrial-marine transitional zone at an elevation of 2.78 m. We also  
167 designed a terrestrial position in the contributing hillslope, hereafter referred to as the hillslope  
168 position, ~30 m from the upland position, at an elevation of ~6 m (Figure 1 B-C).

169 To study the spatiotemporal variations of the salt marsh hydrology, we established a network  
170 of observation and sampling piezometers. At the three salt marsh positions, we installed a network  
171 of 70 cm below-the-ground-surface (bgs) observation piezometers where we measured water level at  
172 five-minute intervals with Solinst pressure transducer loggers (Levellogger 5, Ontario, Canada). The  
173 observation piezometers had a 15 cm screen at the bottom. At each marsh position, we installed  
174 shallow sampling piezometers at 10, 30, and 50 cm-bgs (Figure 1C). Further, at each transect  
175 position (hillslope to lower marsh), we established deeper piezometers of various depths (1-3 m-bgs)  
176 used for water sampling (Figure 1C, Table S1). Except for the hillslope piezometer, we measured  
177 water level at five-minute intervals with Solinst pressure transducer loggers in the deep piezometers  
178 (Table S1). We also used a Solinst pressure transducer to measure air pressure at the transect at five-  
179 minute intervals to barometrically correct the water pressure measurements to calculate the water  
180 level. Data gaps in the water level records of the deep piezometers caused by pumping during  
181 sampling varied between hours to weeks. However, because we did not use time series analysis in  
182 this study, we did not fill the data gaps (e.g., Figure 2B).

## 183 **2.2 Ancillary data**

184 To account for multiscale climate forcing (seasonal and sub-daily) effects on the salt marsh  
185 hydrology, we used hourly meteorological data from the Elkhorn Slough Meteorological Station,  
186 located ~4.5 km from the study site (36°48'55", -121°44'17"). The station is managed and  
187 maintained by the National Estuarine Research Reserve System (NERR, 2022). Our analyses used  
188 relative humidity, barometric pressure, precipitation, wind speed, total photosynthetically active  
189 radiation (PAR), and air temperature (Figure S2).

## 190 3- Methodology

### 191 3.1 Isotopic and electrical conductivity measurements

192 To improve our understanding of the surface water-GW exchanges at the terrestrial-marine  
193 interface, we conducted periodical subsurface, surface, and precipitation water sampling across the  
194 field site (Figure 2). Water samples were analyzed for three complementary tracers (electrical  
195 conductivity (EC), stable water isotopes, and tritium) that enabled us to understand different aspects  
196 of the salt marsh hydrology: the degree of mixing between terrestrial and marine sources, the effects  
197 of evaporation and transpiration, and water ages and flow velocities. EC/salinity is subject to mixing  
198 and concentration by evaporation and transpiration. Stable isotopes are subject to mixing and  
199 fractionation by evaporation only. With these two tracers, the degree of mixing and the partitioning  
200 between evaporation and transpiration could be established. Tritium is subject to mixing and  
201 radioactive decay. Combined with the mixing fractions derived from stable isotopes, tritium  
202 provides an estimate of water age and inversely flow velocity (Beyer *et al.*, 2014). The combined  
203 interpretation of these tracers also allowed us to evaluate the effect of temporal variability in the  
204 precipitation end-member signature.

205 To study the spatiotemporal distribution of different water sources, we collected and  
206 analyzed samples for stable water isotopes and EC from various depths ( $n = 86$ , Table S2). We  
207 collected precipitation water at the site using a precipitation funnel, the volume of collected water  
208 for each sample varied as a result of variability in precipitation intensity, amount and collection  
209 duration (Table 1). In addition, we collected a sample from the tidal creek in winter, during the rainy  
210 season (Figure 2), and two irrigation water samples from the agricultural field above the study  
211 site (Figure 1A) which is pumped from local GW (~400 m bgs), hereafter referred to as deep GW.  
212 We analyzed  $^3\text{H}$  in 22 piezometer samples collected from 11 positions and depths and calculated the  
213 mean  $^3\text{H}$  in precipitation from 21 samples.

214 We analyzed the stable water isotope samples by cavity ring down spectroscopy in a Picarro  
215 L2130-i at the University of California Santa Cruz. We measured electric conductivity of the samples  
216 with a Orion Star™ A329 multiparameter meter (Thermo Fisher Scientific, Massachusetts, USA).  
217 We analyzed  $^3\text{H}$  samples at Lawrence Livermore National Laboratory by helium-3 accumulation  
218 (Clarke *et al.*, 1976; Surano *et al.*, 1992).

219

### 220 3.2 Local meteoric water line and line conditioning excess

221 The local meteoric water line (LMWL) is the site-specific long-term covariation of  $\delta^2\text{H}$  and  
222  $\delta^{18}\text{O}$  (Rozanski *et al.*, 1993). Thus, conceptually, the LMWL of an area is a simplified representation  
223 of the average  $\delta^2\text{H}$  and  $\delta^{18}\text{O}$  relationship in precipitation (Putman *et al.*, 2019). For this study, we  
224 calculated the LMWL using monthly  $\delta^2\text{H}$  and  $\delta^{18}\text{O}$  values from the Online Isotopes in Precipitation  
225 Calculator (Bowen and Revenaugh, 2003; Bowen, 2022) and calculated the unweighted, orthogonal-  
226 least-squares fit of the relationship between  $\delta^2\text{H}$  and  $\delta^{18}\text{O}$  at the site. This LMWL is a practical tool  
227 that provides a hydrologic framework for evaluating hydroclimatic processes such as evaporation.

228 We evaluated the spatiotemporal variations between  $\delta^2\text{H}$  and  $\delta^{18}\text{O}$  by calculating the  
229 deviations from the LMWL. We described these variations in terms of the line conditioning excess  
230 (lc-excess), where  $\text{lc-excess} = \delta^2\text{H} - m * \delta^{18}\text{O} - b$  (Landwehr and Coplen, 2006), where  $m$  and  $b$  are  
231 the slope and the intercept of the LMWL, respectively. The lc-excess mathematically expresses the  
232 offset between the LMWL and the studied samples in dual-isotope space. Most precipitation water  
233 will lie close to the LMWL and have a nearly zero lc-excess. When water evaporates, it becomes  
234 enriched in heavy isotopes such that they deviate from LMWL, and the lc-excess value becomes

235 progressively negative during evaporation. Therefore, lc-excess is a valuable indicator of evaporative  
 236 fractionation. It is worth noticing that precipitation samples can naturally deviate from the LMWL  
 237 and consequently have a nonzero lc-excess. These deviations are accredited to variations in moisture  
 238 sources, air mass trajectories, and cloud processes (Dansgaard, 1964).

### 239 3.3 Mixing models, tritium ages and uncertainty estimations

240 We explained subsurface water flow directions by studying the different water sources in the  
 241 salt marsh subsurface (e.g., terrestrial and marine/estuarine). For this, we used stable water isotopes  
 242 as tracers. In the analysis, we identified two clear end members, Elkhorn Slough surface water,  
 243 representing marine/estuarine surface water, and terrestrial water, represented by precipitation and  
 244 deep GW sampled at the site. We calculated mixing ratios between terrestrial and marine (estuarine)  
 245 water using end-member-mixing analysis, where

$$246 f_{ES} = \frac{C_S - C_P}{C_{ES} - C_P} \quad (1)$$

247 and

$$248 f_{ES} + f_P = 1 \quad (2)$$

249 Where  $f$  [%] is the calculated fraction of each end member (estuarine or  
 250 precipitation/terrestrial,  $ES$  or  $P$ , respectively), and  $C$  is the tracer concentration ( $\delta^{18}O$  value) in the  
 251 end members  $ES$  and  $P$  and the sample ( $s$ ). We quantify the uncertainty of the mixing using a  
 252 Gaussian error propagation (Genereux, 1998) as:

$$253 W_{f_{ES}} = W_{f_P} = \sqrt{\left[\frac{f_P}{C_{ES} - C_P} W_{C_P}\right]^2 + \left[\frac{f_{ES}}{C_{ES} - C_P} W_{C_{ES}}\right]^2 + \left[\frac{-1}{C_{ES} - C_P} W_{C_S}\right]^2} \quad (3)$$

254 Where  $W$  is the error (analytical error/uncertainty of the tracer measurement).

255 To understand the flow velocities of subsurface water in the salt marsh, we measured  $^3H$   
 256 activity in subsurface water across the experimental transect and compared that with  $^3H$  activity in  
 257 Elkhorn Slough surface water (SW) and precipitation. The  $^3H$  activity in precipitation at Elkhorn  
 258 Slough was based on a regional estimate of  $^3H$  in precipitation for central coastal California (Harms  
 259 *et al.*, 2016; Visser *et al.*, 2018) and  $^3H$  in precipitation in the city of Oakland, 115 km north of the  
 260 field site (Grande *et al.*, 2020). For samples that showed limited mixing with ES water (i.e. sourced  
 261 from precipitation infiltration), we calculated the apparent tritium age ( $\tau$ [year]) as:

$$262 \tau = \frac{\ln\left(\frac{C_0}{C}\right)}{\lambda} \quad (4)$$

263 Where  $\lambda$  [ $\text{year}^{-1}$ ] is the  $^3H$  decay constant (0.0563/year),  $C_0$  [pCi/L] is the initial  $^3H$  activity  
 264 concentration in precipitation, and  $C$  [pCi/L] is  $^3H$  activity concentration in the sample. The  
 265 apparent tritium age calculation is based on the assumptions that (1) the sample represents a single  
 266 age (i.e. piston flow distribution model) and (2) that the input of tritium in precipitation has  
 267 effectively been constant (i.e. was the same at the time of recharge as today). A detailed analysis  
 268 (Supporting Information Text S1) of several age distribution models (e.g. exponential model,  
 269 dispersion model) considering the historical peak of tritium in precipitation in the 1960s and 1970s  
 270 showed that the apparent tritium age adequately reflects the mean residence time of most likely age  
 271 models.

272 We calculated the error of the age  $\sigma_\tau$  [yrs] resulting from the analytical uncertainty of the  
 273 tritium measurement as:

$$274 \sigma_\tau = \lambda^{-1} \sqrt{\left(\frac{\sigma_C}{C}\right)^2 + \left(\frac{\sigma_{C_0}}{C_0}\right)^2} \quad (6)$$

275 Where  $\sigma_C$  [pCi/L] is the uncertainty of the  $^3H$  activity concentration measured in the  
 276 subsurface water sample,  $\sigma_{C_0}$  [pCi/L] is the uncertainty in initial  $^3H$  concentration. For mixed

277 samples with contributions of both terrestrial flow paths and ES surface water, the age uncertainty  
278 was propagated numerically, by adding random noise (n=1000, normal distribution, standard  
279 deviation equal to the measurement uncertainty) to the  $\delta^{18}\text{O}$  and  $^3\text{H}$  measurements. After  
280 performing the end-member mixing and age calculations on the ensemble, we report the mean and  
281 standard deviation of the ensemble.

282 For terrestrial sourced samples, we estimated recharge flow velocities ( $v_V$ [cm/yrs]) as:

$$283 \quad v_V = \frac{\tau}{\Delta Z} \quad (7)$$

284 where  $\Delta Z$  [cm] is the sample depth. We calculated recharge rates ( $q_V$ [cm/yrs]) as:

$$285 \quad q_V = v_V \phi \quad (8)$$

286 Where  $\phi$  is the porosity. We measured porosity and bulk density for each position from soil  
287 samples extracted during the installation of the piezometers (Grande *et al.*, 2022a).

288 Because of the study's various sampling depths and the differences in piezometer depths  
289 (less than 0.5 m and larger than 1.7 m), we grouped the sampling piezometers into shallow and deep  
290 samples (unless otherwise specified). We refer to samples from piezometers with depths  $\leq 50$  cm-  
291 bgs as shallow samples and samples from piezometers  $> 50$  cm-bgs as deep samples.

### 292 **3.4 Statistical analysis**

293 We analyzed the normality of all of the data distributions using histograms and Shapiro-Wilk  
294 tests (Shapiro and Wilk, 1965) using the “shapiro.test” function in base R (R Core Team, 2021). We  
295 tested if the variations in porewater electrical conductivity (EC),  $^3\text{H}$ , and stable water isotopes  
296 differed significantly among positions and depths using Levene’s test (Schultz, 1985). We used a  
297 significance level ( $\alpha$ ) of 0.05 for statistical significance. We used the “leveneTest” function from the  
298 “car” package in R (Fox and Weisberg, 2019). We used the Kruskal-Wallis test (Breslow, 1970) to  
299 analyze if EC,  $^3\text{H}$ , and stable water isotopes are identical for the wet and dry seasons and to evaluate  
300 differences in their porewater signature/activity across depths and positions. We used a level ( $\alpha$ ) of  
301 0.05 for statistical significance. We used the “kruskal.test” function in base R (R Core Team, 2021)  
302 for this analysis. We further analyzed all the significant results with pairwise Mann-Whitney U test  
303 (Rosner and Grove, 1999) to correct the significance level for multiple comparisons.

## 304 **4- Results**

### 305 **4.1 Salt marsh hydrology and regional climate**

306 During the 2020 and 2021 water year study period, we observed precipitation totals of 395.1  
307 mm and 280 mm, respectively (Figure 2A). These are both well below the twenty year mean annual  
308 precipitation for the area (627 mm). We also observed marked wet/dry seasonality in the  
309 fluctuations of the upland water level record (Figure 2B). Over the study period, we observed a  
310 difference of 1.34 m between the peak terrestrial GW level (2.75 m relative to NAVD88) in the wet  
311 periods and the lowest level in dry periods (1.39 m).

312 The terrestrial water level increases during precipitation events/periods (Figure 2B). Marsh  
313 subsurface water levels are subject to daily, biweekly and seasonal tidal cycle inundation dynamics  
314 (that vary due to the position of the moon/earth), resulting in multiple water level fluctuation  
315 frequencies (Grande *et al.*, 2022a). Subsurface water levels at the lower marsh position are  
316 consistently lower than at the upper and middle marsh positions during low tides, draining below the  
317 marsh land surface elevation in each tidal cycle (Figure 2 C-E). However, the subsurface water level  
318 in the upper and middle marshes did not drop below the marsh land surface elevation during the  
319 wet and dormant winter season. This indicates that the marsh does not drain substantially between  
320 daily tidal cycles during wet periods. As the system transitions into the dry seasons, we observed that



321 the terrestrial GW levels decreased, and the salt marsh subsurface water levels in the middle and  
322 upper positions also dropped below the salt marsh surface between tidal inundation periods (Figure  
323 2 B-E).

#### 324 **4.2 Electrical conductivity of salt marsh subsurface water**

325 EC varied across sampling sites (Table 2), generally decreasing from Elkhorn Slough surface  
326 water (SW) and the salt marsh positions to the hillslope position (Figure 3A-D), although not all the  
327 positions had significantly different EC values (Table S3). Notably, the tidal creek EC was not  
328 significantly different from any other position. The median SW EC was 55.3 mS/cm and varied  
329 between 46.6 and 60.7 mS/cm during the dry and wet seasons, respectively (Figure 3A-D).  
330 However, wet/dry seasonality did not cause significant differences in the EC values of SW ( $p = 0.8$ ).  
331 Upland, hillslope, and precipitation were all significantly lower than SW. EC did not differ  
332 significantly between SW and the three salt marsh positions ( $p = 0.3$ ) because of significant seasonal  
333 variation in the variability of salt marsh EC. Additionally, upland, hillslope, and GW were all  
334 significantly lower than the EC in each of the salt marsh positions (Table S3).

335 During the dry season, porewater EC in all marsh positions was higher than EC in SW due  
336 to evapotranspiration (Figure 3A-D). We used this hyper-salinity (EC values elevated above SW) as a  
337 valuable geochemical tracer to understand deeper marsh water flow directions and sources. In  
338 contrast to the dry season, wet season EC in all three marsh positions varied from above SW EC to  
339 below SW EC due to mixing with either precipitation or terrestrial water sources (upland, GW). The  
340 variance of salt marsh porewater EC was more pronounced during the wet season than during the  
341 dry season. The varying degrees of mixing resulted in significant differences between the lower, the  
342 middle, and upper marsh positions during the wet season, but not during the dry season. The upper  
343 marsh position showed the largest range in EC values, indicating the largest degree of mixing with  
344 terrestrial sources. However, evaporation did not result in significant variations in hyper-salinity in  
345 salt marsh porewater EC during the dry season ( $p = 0.6$ , Figure 3 A-D) suggesting that a biological  
346 control (i.e., plant transpiration) places an upper limit to the process of salinity concentration.

347 The deep salt marsh samples had overall significantly higher EC than shallow salt marsh  
348 samples. This significant difference is caused by lower EC values in shallow piezometers as a result  
349 of mixing with precipitation or terrestrial water during the wet season (Figure 4 A). In contrast, there  
350 were no significant differences in salt marsh EC with depth for the dry season ( $p = 0.2$ , Figure 4D).

#### 351 **4.3 Stable water isotopes as tracers of subsurface water exchange in salt marsh** 352 **systems**

353 Stable water isotope values varied across the transect (Table 3). The stable water isotope  
354 analysis revealed two end members: SW (mean  $\delta^{18}\text{O} = 0.22\text{‰}$ ), representing the marine end-  
355 member, and deep GW from the area (mean  $\delta^{18}\text{O} = -5.93\text{‰}$ ), likely representing the long-term  
356 signature of local precipitation (Figure 5). Precipitation water collected at the site had a mean  $\delta^{18}\text{O}$   
357 and a standard deviation  $\delta^{18}\text{O}$  of  $-4.99\text{‰}$  and  $1.38\text{‰}$ , respectively (Table 3). Terrestrial water  
358 samples (upland and hillslope) and wet season precipitation closely resemble the low  $\delta^{18}\text{O}$  value of  
359 GW, while dry season precipitation shows a larger range in  $\delta^{18}\text{O}$ . The lower and middle marsh  
360 positions were significantly higher, and more similar to SW (i.e. not significantly different from SW),  
361 than the upper marsh and the other terrestrial samples (Figure 3 B-E, Table S4).

362 The effect of wet/dry seasonality in the three salt marsh positions resulted in significantly  
363 lower  $\delta^{18}\text{O}$  values in the wet season than in the dry season (Figure 3 B-E) as a result of terrestrial  
364 inputs into the salt marsh. However, wet/dry seasonality did not affect the terrestrial positions  
365 (Figure 3 B-E).

366 During the wet season, the shallow samples were isotopically lighter (i.e. lower  $\delta^{18}\text{O}$  values)  
367 than the deep samples as a result of terrestrial or precipitation inputs (Figure 4 B-E). However, in  
368 the dry season, shallow samples were isotopically heavier (i.e. higher  $\delta^{18}\text{O}$  values) than the deep  
369 samples, possibly as a result of evaporative fractionation ( $p = 0.02$ , Figure 4 B-E). Moreover, for  
370 individual marsh positions, the differences in isotopic composition and depth were only significant  
371 for the lower and upper marsh positions in the wet season and the upper marsh in the dry season  
372 (Figure 4 B-E). The tidal creek sample collected after a precipitation event in the wet winter season  
373 (12/28/2021, Figure 2A) had an unusually light isotopic signature with  $\delta^{18}\text{O} = -4 \text{ ‰}$ , suggesting a  
374 large component of terrestrial water, and was similar to shallow samples collected in the lower marsh  
375 during the wet season (Figure 6).

#### 376 4.3.1 Local meteoric water line and line conditioned excess

377 Based on interpolated  $\delta^2\text{H}$  and  $\delta^{18}\text{O}$  estimates of precipitation values in the area downloaded  
378 from the Online Isotopes in Precipitation Calculator database, the LMWL has a slope of 7.1 and an  
379 intercept of 1.6 (Figure 6). The different precipitation events we sampled at the site during the 2022  
380 water year do not align with the LMWL or the volume-weighted mean of these local precipitation  
381 samples (Figure S3, Table 3). However, the deep GW samples fall on the LMWL, suggesting that  
382 this water is a good indicator of the long-term precipitation in the area (Figure 6).

383 Lc-excess varied among transect positions (Figure 3 C-D). Negative and near-zero lc-excess  
384 values in SW contrast with positive lc-excess in the upland, hillslope, deep GW, and precipitation  
385 samples (Table 3). However, we found that not all the relationships were significantly different  
386 (Table S5). Notably, the lc-excess differed significantly among the three salt marsh positions and the  
387 terrestrial positions (Figure S4). Further, we found that wet/dry seasonality did not result in  
388 significant differences in lc-excess in any position ( $p = 0.6$ , Figure 3 C-F). lc-excess differed among  
389 shallow and deep samples in the salt marsh during the wet season, but not during the dry season ( $p$   
390  $= 0.7$ , Figure 4).

391 Typically, evaporative fractionation results in progressively lower lc-excess values as  $\delta^2\text{H}$  and  
392  $\delta^{18}\text{O}$  shift along a line with a slope of 5. We find no clear indication of evaporative fractionation in  
393 our samples. While the lc-excess of marsh water varies between 7 ‰ and -2 ‰, the range of values  
394 can largely be explained by the mixing between precipitation (6-12 ‰), terrestrial sources (3-6 ‰),  
395 and SW (-1 - -3 ‰; Fig S4). In the dry season, several samples plot below the mixing line, with more  
396 negative lc-excess values than mixing could explain, suggesting limited effect of evaporative  
397 fractionation.

#### 398 4.3.2 Estuarine surface water and terrestrial water mixing

399 The salt marsh positions had significantly higher fractions of SW than the upland and  
400 hillslope positions (Figure 7). However, not all salt marsh positions were significantly different from  
401 another (Table S6). Further, the upper marsh position differed significantly from the lower and  
402 middle marsh, and from the upland and hillslope positions (Table S6).

403 The effect of wet/dry seasonality was evident in the fractional mix of SW and terrestrial  
404 water. The fraction of SW water in the salt marsh was significantly larger in the dry season than in  
405 the wet season (Table 4). Notably, the median fraction of estuarine surface water varied between  
406 0.64 and 0.83 in the upper marsh position for the wet and dry seasons, respectively. During the dry  
407 season, six samples had higher  $\delta^{18}\text{O}$  than SW, and during the wet season, one sample had higher  
408  $\delta^{18}\text{O}$  than SW, resulting in fractional mixing ratios above 100% (Figure 7). However, we did not find  
409 significant differences in wet/dry seasonality for the upland and hillslope positions.

410 During the wet season, shallow samples had smaller fractions of SW than deep samples  
411 (Figure 8). These results contrast with the dry season, when shallow samples had higher fractions of

412 SW than the deep samples. This variation in shallow mixing ratios is caused by precipitation and  
413 terrestrial sources mixing in the shallow subsurface, while deep marsh water appears to have a more  
414 consistent origin.

#### 415 **4.4 Tritium as a salt marsh porewater age tracer**

416 The activity of  $^3\text{H}$  in precipitation was based on 21 samples collected between 2014 and  
417 2022 (Table 5).  $^3\text{H}$  activities for the available samples varied between 4.0 pCi/L and 14.5 pCi/L.  
418 Each sample represented between 0.2% and 7% of the total precipitation in the water year it was  
419 collected. Because we found no correlation with sample date ( $R^2 < 0.01$ ), daily precipitation amount  
420 ( $R^2 < 0.01$ ), mean daily temperature ( $R^2 = 0.13$ ), or wind direction ( $R^2 = 0.09$ , Figure S5), we  
421 considered these values to be randomly sampled from the natural distribution of  $^3\text{H}$  in precipitation  
422 and calculated the mean (8.6 pCi/L), standard deviation (3.3 pCi/L) and used the uncertainty around  
423 the mean (0.7 pCi/L) for uncertainty propagation.

424  $^3\text{H}$  in subsurface water varied between 5.4 pCi/L and 1.9 pCi/L in the upland and lower  
425 marsh, respectively (Figure 9A). The deep GW sample contained less than 1 pCi/L  $^3\text{H}$  (Figure 9B)  
426 suggesting it recharged entirely before the 1950s. Shallow samples had higher  $^3\text{H}$  activity  
427 concentrations than the deep samples for these marsh positions (Figure 9A).

428 Along the transect,  $^3\text{H}$  is subject to both mixing of different sources, and radioactive decay  
429 reflecting the time since infiltration. We used  $\delta^{18}$  to evaluate mixing and reconstruct the initial  $^3\text{H}$   
430 concentration from which we calculated water age (Figure 9B). The upper, upland, and hillslope  
431 deep piezometer samples are mixtures between the terrestrial and ES water. The degree of mixing in  
432 the hillslope sample is negligible ( $f_{\text{ES}} = 0.03 \pm 0.03$ ) and the apparent tritium age of  $12.7 \pm 1.8$   
433 yrs. The degree of mixing in the upland sample is limited ( $f_{\text{ES}} = 0.20 \pm 0.02$ ) and the apparent  
434 tritium age in the shallower piezometer (250 cm) is younger ( $5.8 \pm 1.8$  yrs) than in the deeper  
435 hillslope piezometer (430 cm). Due to the mixing in the Upper Marsh position ( $f_{\text{ES}} = 0.17 \pm$   
436  $0.03$ ), the uncertainty associated with the apparent tritium age is larger ( $11.8 \pm 5.6$  yrs) but the age is  
437 within the range of the upland and hillslope piezometers, suggesting converging terrestrial flow  
438 paths mix with marsh subsurface water.  $^3\text{H}$  activities of lower and middle positions cluster closely  
439 around the ES SW value, suggesting the activities are not affected by significant decay. Assuming the  
440 ES SW value as the initial concentration, the age in the middle marsh deep piezometer is  $2.3 \pm 4$  yrs,  
441 providing an upper limit to the range of ages in the marsh.

442 The analysis of these ages results in varying values of recharge velocity and recharge rates  
443 along the transect. Recharge velocities of 34 and 43 cm/yr in the upland and hillslope positions  
444 (respectively) translate to a recharge rate of 13 cm/yr (considering differences in porosity). For the  
445 lower marsh positions, vertical flow velocities of more than 1 m/yr, based on the young ages in the  
446 deep piezometers, are consistent with the seasonal variability observed in the shallower piezometers.  
447 These differences in recharge velocities, and their corresponding recharge rates (Table 6), reflect  
448 different positions along flow paths with different origins and salinity, as demonstrated by the  
449 mixing ratios discussed previously.

#### 450 **5- Discussion**

451 The results of this study provide important conceptual insights into surface water-subsurface  
452 water exchanges at the terrestrial-marine interface, especially for salt marsh systems. This study  
453 demonstrates that a combination of stable water isotopes, tritium, and electrical conductivity in  
454 subsurface water of a hillslope to salt marsh continuum is a valuable tool to understand the effect of  
455 wet/dry seasonality typical to Mediterranean climates in the subsurface flow direction and porewater  
456 exchange. Our general results on the hydrological processes of the salt marsh subsurface agree with  
457 previous studies based on numerical models informed by piezometers and seepage meters (Michael

458 *et al.*, 2005), porewater salinity (Shen *et al.*, 2015), and radon isotopes (Smith *et al.*, 2008; Wang *et al.*,  
459 2021c). However, our study provides additional insights on porewater flow velocities, subsurface  
460 saltmarsh water recharge, and discretizes evaporation and transpiration processes in the salt marsh  
461 subsurface.

### 462 **5.1 Salt marsh subsurface water evapotranspiration signatures differ by season**

463 Salinity concentration in salt marsh subsurface water at Elkhorn Slough is primarily  
464 controlled by plant transpiration as stable water isotopes revealed no major evaporative fractionation  
465 (Figure 6). Subsurface water salinity in the salt marsh, and consequently EC, can increase due to  
466 evaporation and plant transpiration, and decrease by dilution from lower salinity waters such as  
467 precipitation, SW, and fresh GW (Nuttle and Hemond, 1988; Miklesh and Meile, 2018). In  
468 Mediterranean climates, the growing season coincides with the dry season (Feng *et al.*, 2019), creating  
469 “ideal” conditions for high evapotranspiration that can result in elevated subsurface water EC  
470 during the dry summer season. Salinity at the surface varied between sub-SW in wet periods (due to  
471 precipitation and terrestrial, freshwater inputs, Figure 3A-D), similar to SW (during the transition  
472 between wet-dry seasonality), and elevated-SW (during the dry growing season, Figure 5). The  
473 absence of clearly evaporated samples, based on both  $\delta^{18}O$ -excess and stable water isotopes (Figure 3  
474 C-F, Figure 6), suggests plant uptake of water as the dominant mechanism for excess EC in salt  
475 marsh subsurface water.

476 A special case of observation bias could also result in the lack of evidence for evaporative  
477 fractionation in the salt marsh subsurface while salinity is elevated above the SW source end-  
478 member: evaporated (fractionated) water is produced at the land surface, where it is frequently  
479 inundated and thus flushed out (Grande *et al.*, 2022a). During the study, we saw changes in the  
480 saturation stage in the salt marsh subsurface with different hydrologic dynamics during wet and dry  
481 periods (Figure 2 C-E). In the dry season, the salt marsh subsurface drains below the salt marsh  
482 elevation at intra-tidal scales, resulting in vertical hydraulic gradients that favors SW  
483 infiltration/recharge during spring tides. Plant uptake happens below the surface. Thus, inundation  
484 could not flush away high EC water resulting from transpiration. Periodic evaporation and recharge  
485 from lake beds during the Pleistocene at Gold Flat Playa (Nevada, USA) also resulted in elevated  
486 salinity without isotopic evidence of evaporative fractionation in the Pahute Mesa groundwater  
487 system (Kwicklis *et al.*, 2021).

### 488 **5.2 Seasonality drives dry season recharge of hyper-salinity water to the salt marsh** 489 **subsurface**

490 Hyper-salinity is only found during the dry growing season in shallow salt marsh water, but  
491 is observed throughout the year in the deep salt marsh samples. As hyper-salinity is a result of  
492 evapotranspiration at the surface or at rooting depth, we believe the dry season shallow water is the  
493 source of the deep hyper-salinity water. This is evidence that recharge of the deep salt marsh system  
494 must occur during the dry growing season (Figure 10). This unusual timing of recharge is explained  
495 by the relatively lower deep groundwater levels during the dry season, which allow for downward  
496 vertical flow. Higher water tables during the wet season limit the downward vertical flow thus  
497 recharge cannot take place in winter. Salt marsh recharge has been studied in the context of the  
498 primary mechanisms by which the salt marsh subsurface is recharged. For example, freshwater can  
499 recharge the salt marsh subsurface vertically due to direct precipitation on the salt marsh (Hemond *et*  
500 *al.*, 1984), or from underlying aquifers (Nuttle and Harvey, 1995), laterally from upland terrestrial  
501 GW inflow (Wilson *et al.*, 2015b), or streams and drainage channels connected to the salt marsh  
502 (Zhao *et al.*, 2021). Further, salt marshes can also be recharged by saline water that occurs vertically  
503 during spring tides that flood the salt marsh or laterally through the drainage creek banks during

504 rising tides (Harvey *et al.*, 1987). However, less is known about the temporal changes in salt marsh  
505 subsurface water recharge. Here, we showed that deep salt marsh samples are recharged during the  
506 dry growing season (Figure 10).

### 507 **5.3 Subsurface flow direction across the hillslope to salt marsh continuum**

508 Wet season precipitation increases terrestrial GW levels, driving observed salt marsh  
509 hydrologic patterns (Figure 2). In the wet season, precipitation recharges GW in the hillslope and  
510 upland positions, resulting in a hydraulic gradient from the upland towards the salt marsh. Stable  
511 water isotopes revealed a decrease in terrestrial source contribution from the upland to the lower salt  
512 marsh position (i.e., an increase in the fraction of SW coastward), confirming the water level  
513 measurements (Figure 7). The stable isotope data corroborated the trends in salinity that suggest that  
514 lower marsh positions are more often inundated. Our results align with previous estimations in other  
515 salt marsh systems using numerical models and salinity gradients that have suggested that upper  
516 marsh positions are more influenced by terrestrial waters than lower marsh positions (Wilson and  
517 Gardner, 2006; Wilson *et al.*, 2015b).

518 Identifying the terrestrial end-member is critical for understanding the interactions between  
519 terrestrial and marine waters, as it permits us to account for freshwater contributions to the coastal  
520 ocean. Most of the research at the terrestrial-marine interface focuses on identifying submarine GW  
521 discharge or porewater exchange fluxes. However, only recently some researchers have focused on  
522 identifying the fraction of terrestrial freshwater from marine water using, for example, airborne  
523 thermal infrared (Tamborski *et al.*, 2015), a combination of remote sensing and radon-222 (Cheng *et al.*  
524 *et al.*, 2020), water balance of the discharging aquifer (Zhou *et al.*, 2019), salinity (Ibáñez *et al.*, 2021),  
525 and stable water isotopes (Rocha *et al.*, 2016; Wang *et al.*, 2021a). We observed that the isotopic  
526 signature of the hillslope and the upland positions was similar to the deep GW sample from the  
527 regional aquifer (Figure 7). It is worth noting that the deep GW sample might be fossil water and  
528 reflect paleoclimatic conditions, but its similarity with the water sampled from the hillslope and  
529 upland positions suggests this is not the case (Figure 6). Furthermore, the terrestrial end-member is  
530 not necessarily the same as the weighted precipitation average measured during the study due to land  
531 surface processes that can affect the isotopic composition of the sample during recharge (i.e.,  
532 evaporation).

533 The lower and middle marsh positions have a dominant SW source, while the upper marsh  
534 position is always a mix of terrestrial and SW (Figure 7). Previous studies have successfully used  
535 salinity/EC to study marine intrusion/recharge in coastal areas (Michael *et al.*, 2003; Cardenas *et al.*,  
536 2010; Santos *et al.*, 2021). However, EC was not a good tracer in this site because it is affected by  
537 evaporation and transpiration, resulting in a significant excess in subsurface water EC higher than  
538 the marine end-member for several samples (Figure 3). Stable water isotopes proved to be a more  
539 conservative tracer than EC and provided more robust estimations of the fractional mix of SW  
540 across the study transect (Figure 10).

541 Wet/dry seasonality drives the direction of subsurface water flow in the salt marsh system  
542 (Figure 7). These findings are also in line with previous work based on hydraulic gradients in coastal  
543 aquifers showing that a shift in the freshwater–saltwater interface, influenced by seasonal variations  
544 in terrestrial GW levels, can justify saline discharges that lag inland recharge cycles (Michael *et al.*,  
545 2005). Our results show that SW is drawn into upper marsh positions as the terrestrial–marine water  
546 interface moves landward during the dry season and discharges back into estuarine waters as the  
547 interface moves coastward during wet periods (Figure 7A). We found a connection between this  
548 Mediterranean climate's marked wet/dry seasonality, controlling terrestrial hydrological processes,  
549 and the salt marsh subsurface water, influenced by estuarine water contributions. These hydrological  
550 fluxes are important drivers of biochemical processes and chemical loadings in coastal waters that

551 have shown shifts in salt marsh biogeochemical behavior as a function of wet/dry seasonality  
552 (Grande *et al.*, 2022b). The wet/dry seasonality results in significant shifts of the terrestrial-marine  
553 water interface across depths. However, the seasonality is limited to the shallow marsh, with more  
554 variable ranges of estuarine water fraction during wet periods. In contrast, the deep salt marsh  
555 sample did not experience significant shifts in the water sources across wet and dry periods.

556 On shorter time scales, hydrological fluxes are driven by precipitation, as precipitation water  
557 exchanges with shallow salt marsh subsurface water, leading to distinct isotopic signatures (similar to  
558 terrestrial water sources) in the tidal creek and shallow subsurface water samples collected during  
559 and after episodic precipitation events (Figure 6). Notably, many of these samples were collected  
560 during spring tides, thus our findings show that terrestrial water inputs to the salt marsh are not  
561 significantly diluted by frequent tidal inundation. During precipitation events, terrestrial water can be  
562 delivered to the salt marsh as overland flow through the intermittent streams and drainage network  
563 at the study transect (Figure S1), by infiltration of direct precipitation on the salt marsh, or by  
564 subsurface flow as the hillslope and upland positions become saturated and contribute freshwater to  
565 the shallow salt marsh subsurface. Recent work conducted at this site using high-frequency  
566 measurements of subsurface water nitrate (Grande *et al.*, 2022b) and wavelet and information theory  
567 on continuous redox potential (Grande *et al.*, 2022a) have proposed that precipitation water  
568 disturbed the subsurface water chemistry at shallow depths (down to 50 cm-bgs) over short periods  
569 from the onset of precipitation. Our results using stable water isotopes confirm these previous  
570 findings.

#### 571 **5.4 Tritium ages constrain subsurface flow velocities and recharge rates**

572 While the shallow (10-50 cm) depths of the salt marsh are flushed regularly by tidal  
573 inundation (Grande *et al.*, 2022a), tritium ages show water moves through the deeper (~3 m) salt  
574 marsh on a time scale of 0-6 years. These ages constrain the water recharge component of  
575 subsurface flow velocities to 34-43 cm per year (Table 6). The substantial uncertainty (1.7-5.6 yrs) of  
576 the age estimates, and consequently the velocities, stems from the reconstruction of the initial  $^3\text{H}$   
577 activity concentrations of mixed samples (based on  $\delta^{18}\text{O}$ ) as well as input variability. These initial  
578 tritium-based velocity estimates provide valuable constraints to numerical models aimed at  
579 understanding the complex three-dimensional transient subsurface flow field.

580 Shallow salt marsh porewater has been proposed to have short residence times throughout  
581 the literature. Researchers have found that shallow subsurface waters in marsh systems are affected  
582 by near-surface and lateral flow paths, resulting from tidal inundations and tidal pumping, with short  
583 residence times (Tamborski *et al.*, 2021). Other researchers have used radon mass balances to study  
584 lateral fluxes of carbon and methane from a low salt marsh position to a tidal creek, finding high  
585 fluxes from shallow samples that suggest short residence times (Chen *et al.*, 2022). These findings  
586 have been mostly described in lower marsh positions, where hydraulic conductivity is usually higher  
587 (Xiao *et al.*, 2019). However, using a tracer specific for subsurface water dating, we show that short  
588 residence times are expected in shallow subsurface waters across higher marsh positions as well.  
589 Tritium ages and recharge flow velocities, in combination with radiocarbon analyses, are powerful  
590 constraints for the interpretation and quantification of carbon turnover times, such as salt marshes  
591 or peatland ecosystems (Wilson *et al.*, 2021).

592 Subsurface water ages in the deep samples varied along the transect. In the lower salt marsh,  
593 subsurface waters are younger and recharge flow velocities are higher, indicating a short water  
594 residence time in this position and a more frequent surface water-subsurface water exchange (Table  
595 6). This observation aligns with the conceptual understanding of lower marsh positions, which are  
596 more affected by secondary porosity, such as animal burrows, and thus have higher hydraulic  
597 conductivity that favors water circulation (Guimond and Tamborski, 2021). The deep middle and

598 upper marsh positions presented water ages from a few years to a decade (Table 6). The isotopic  
599 composition of the deep samples did not vary significantly during the study period (Figure 4B-E),  
600 confirming that these positions have a longer residence time and that this water is not affected by  
601 episodic events, such as king tides or precipitation events.

### 602 **5.5 Implications for carbon storage and water quality**

603 The short residence time of shallow subsurface water in the salt marsh has important  
604 implications for our conceptual understanding of carbon storage in salt marsh systems. Given their  
605 high rates of primary productivity and slow decomposition in anoxic soils, salt marshes can store  
606 carbon for extended periods (Kathilankal *et al.*, 2008). However, soil oxidation, due to the circulation  
607 of oxygen-rich surface water recharging the shallow salt marsh subsurface, with a short residence  
608 time, has the potential to increase carbon oxidation and release back into the atmosphere (Lee,  
609 2008). Further, surface water-subsurface water exchanges with short residence times have been  
610 observed to increase the lateral flux of inorganic carbon from salt marshes to the coastal ocean  
611 (Chen *et al.*, 2022).

612 The bimodal time-scales of subsurface water cycles in the salt marsh system can influence  
613 our understanding of nutrient budgets, such as nitrate, and the overall water quality of coastal zones.  
614 In systems such as Elkhorn Slough, where high nitrate concentrations are a water quality concern  
615 that has resulted in eutrophication events (Hicks *et al.*, 2019), understanding the water flow velocity  
616 and residence time of subsurface salt marsh water is valuable in evaluating total nutrient loads or  
617 removal potential. Our analysis can improve the understanding of nitrate retention in the salt marsh  
618 system, which had not been considered in this area (Broenkow and Breaker, 2019; Van Dop *et al.*,  
619 2019). On the one hand, a long residence time of deep subsurface water, recharged in the summer,  
620 can potentially retain nitrate by dissimilatory nitrate reduction to ammonium due to the  
621 accumulation of recalcitrant carbon and sulfides in deep marsh sediments (Koop-Jakobsen and  
622 Giblin, 2009). In contrast, short residence time and fast exchange of shallow subsurface water with  
623 SW are influenced by oxygen and nutrient-rich surface water to labile carbon-rich shallow sediments  
624 that can result in nitrogen removal (denitrification) dominating over retention (Almaraz *et al.*, 2020).

### 625 **6- Conclusion**

626 The results of this study provide an empirically-informed conceptual model into salt marsh  
627 water recharge and subsurface flow paths. We found that subsurface water cycling has two dominant  
628 frequencies, seasonal patterns of terrestrial inputs that interact with daily tidal cycles driving short-  
629 term porewater dynamics in the shallow salt marsh. In summer, the water sources in the shallow  
630 marsh are controlled by tidal inundation and pore water uptake by vegetation. During the wet winter  
631 season, precipitation and terrestrial sources mix with daily inputs of SW. This results in fast water  
632 cycling in the shallow and lower marsh positions, but slower downward flow paths to deeper parts  
633 of the subsurface in the upper marsh, upland and hillslope positions.

634 Plant transpiration is the principal driver of increased salinity, because no evidence of  
635 evaporative fractionation is present. Our conceptual model suggests that deeper salt marsh  
636 subsurface water is preferentially recharged with this super-saline water that was formed during  
637 summer by plant transpiration, while local winter precipitation recharges the terrestrial positions.  
638 Understanding the subsurface hydrology and the spatiotemporal variability of interacting hydro-  
639 geochemical processes in these dynamic hydrologic systems where terrestrial groundwater, terrestrial  
640 surface water, and seawater mix is essential for understanding how salt marshes might respond to  
641 disturbance or climate change.

642

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651

652 **Data availability:** Upon acceptance for publication, the data supporting this manuscript will be  
653 uploaded to an open source repository (DOE-managed ESS DIVE). The codes used in the  
654 manuscript will be uploaded to the corresponding author's GitHub page.

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950 the Near-Global Coast. *Geophysical Research Letters* **46** (11): 5855–5863 DOI:  
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954 **List of Tables**

955 *Table 1. Summary of stable water isotopes of precipitation water collected in Elkhorn Slough. The Date*  
 956 *Time start/end refers to the date and time at which the storm event sampled started and ended. We obtained the*  
 957 *precipitation amounts from the National Estuarine Research Reserve weather station in Elkhorn Slough (~4.5 km*  
 958 *from the study site).*

Date Time start	Date Time end	$\delta^{18}\text{O}$ [‰]	$\delta^{18}\text{O}$ SD [‰]	$\delta^2\text{H}$ [‰]	$\delta^2\text{H}$ SD [‰]	Precip amount [mm]
10/21/21 00:00	10/22/21 08:00	-3.08	0.09	-12.02	0.19	0.2
10/25/21 08:00	10/26/21 13:00	-5.86	0.11	-33.74	0.20	0.3
11/23/21 00:00	12/28/21 20:00	-6.14	0.12	-30.2	0.23	63.3
12/28/21 20:00	04/12/22 00:00	-4.87	0.05	-27.73	0.03	208.6

959 *Table 2. Summary of electrical conductivity for precipitation, estuarine (SW), tidal creek, deep groundwater, lower*  
 960 *marsh (Low), middle marsh (Mid), upper marsh (Upp), upland (Upl), and hillslope positions (Hill).*

Sample	EC [mS/cm]				Wet Season EC [mS/cm]			Dry Season EC [mS/cm]		
	Min	Max	Mean	n	Min	Max	Mean	Min	Max	Mean
Precip	0.01	0.2	0.06	4	0.02	0.2	0.11	0.01	0.02	0.02
SW	46.59	60.7	55.3	4	58.3	60.7	59.5	46.6	52.3	49.4
Tidal Creek			24	1			24			
Deep GW	0.8	0.8	0.8	2			0.8	0.8	0.8	0.8
Low	27.7	67.4	58.3	21	27.7	66.7	51.9	53.1	67.4	59.1
Mid	42.9	69.4	61.8	17	42.9	69.5	56.6	56.1	67.0	64.1
Upp	10.4	75.2	54.2	19	10.4	75.3	47.6	51.9	65.7	61.5
Upl	0.02	23.7	5.7	14	2.5	10.2	5.5	0.02	23.7	6.3
Hill	1.7	2.7	2	4	1.7	1.8	1.7	1.9	2.7	2.3

962 *Table 3. Summary of stable water isotopes and lc-excess for precipitation, estuarine (SW), tidal creek, deep*  
 963 *groundwater, lower marsh (Low), middle marsh (Mid), upper marsh (Upp), upland (Upl), and hillslope positions*  
 964 *(Hill). \*  $\delta^2\text{H}$  volume weighted mean = -28.3 ‰, #  $\delta^{18}\text{O}$  volume weighted mean = -5.2 ‰.*

Sample	$\delta^{18}\text{O}$ [‰]			$\delta^2\text{H}$ [‰]			LC_excess [‰]			n
	Min	Max	Mean	Min	Max	Mean	Min	Max	Mean	
Precip	-6.14	-3.08	-4.99 <sup>#</sup>	-33.74	-12.02	-25.92 <sup>*</sup>	5.5	6.5	8.1	4
SW	0.04	0.96	0.37	-1.27	5.34	0.32	-3.08	-1.10	-2.24	4

<b>Tidal Creek</b>			-4.94			-26.01			7.69	1
<b>Deep GW</b>	-5.93	-5.93	-5.93	-37.75	-37.75	-37.75	3.02	3.02	3.02	2
<b>Low</b>	-3.88	1.14	0.03	-19.77	5.69	-0.03	-0.94	-4.02	6.44	21
<b>Mid</b>	-1.91	0.39	-0.24	-12.61	1.76	-2.20	-1.95	-3.57	0.85	17
<b>Upp</b>	-4.47	-0.01	-1.54	-27.94	-1.61	-10.74	-1.12	-3.97	3.30	19
<b>Upl</b>	-5.61	-4.04	-5.14	-34.23	-24.64	-32.22	3.80	0.18	5.96	14
<b>Hill</b>	-5.72	-4.51	-5.25	-37.40	-27.75	-32.38	3.06	1.87	6.09	4

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Table 4. Summary statistics of Kruskal-Wallis test showing the effect of wet/dry seasonality in the fractional mix of estuarine surface water for the three salt marsh positions. All the relationships had one degree of freedom and a p-value < 0.01.

<b>Marsh Position</b>	<b>Median fraction wet season estuarine surface water [%]</b>	<b>Median fraction dry season estuarine surface water [%]</b>	<b>Kruskal-Wallis: H</b>
Lower	0.89	1.0	7.8
Middle	0.9	0.98	6.3
Upper	0.64	0.83	8.4

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Table 5. Summary of tritium results in precipitation. \*Grande et al. (2022), #Harms et al. (2016), & unpublished. The error of the weighted mean is the standard error of the mean.

<b>Date</b>	<b><sup>3</sup>H [pCi/L]</b>	<b><sup>3</sup>H error [pCi/L]</b>	<b>Precip [mm]</b>	<b>% of WY Precip</b>	<b>Location</b>
03/05/13	6.0	0.4	20	5%	Oakland#
02/07/14	6.9	0.6	16	4%	Oakland#
03/26/14	4.0	0.4	25	6%	Oakland#
03/29/14	14.8	0.8	24	6%	Oakland#
02/06/15	4.4	0.3	35	6%	Oakland#
07/09/15	11.7	0.5	3	0%	Oakland#
09/16/15	5.7	0.2	4	1%	Oakland#
11/02/15	7.3	0.4	12	2%	Oakland#
11/09/15	8.8	0.6	10	2%	Oakland#
11/24/15	12.2	0.6	6	1%	Oakland#



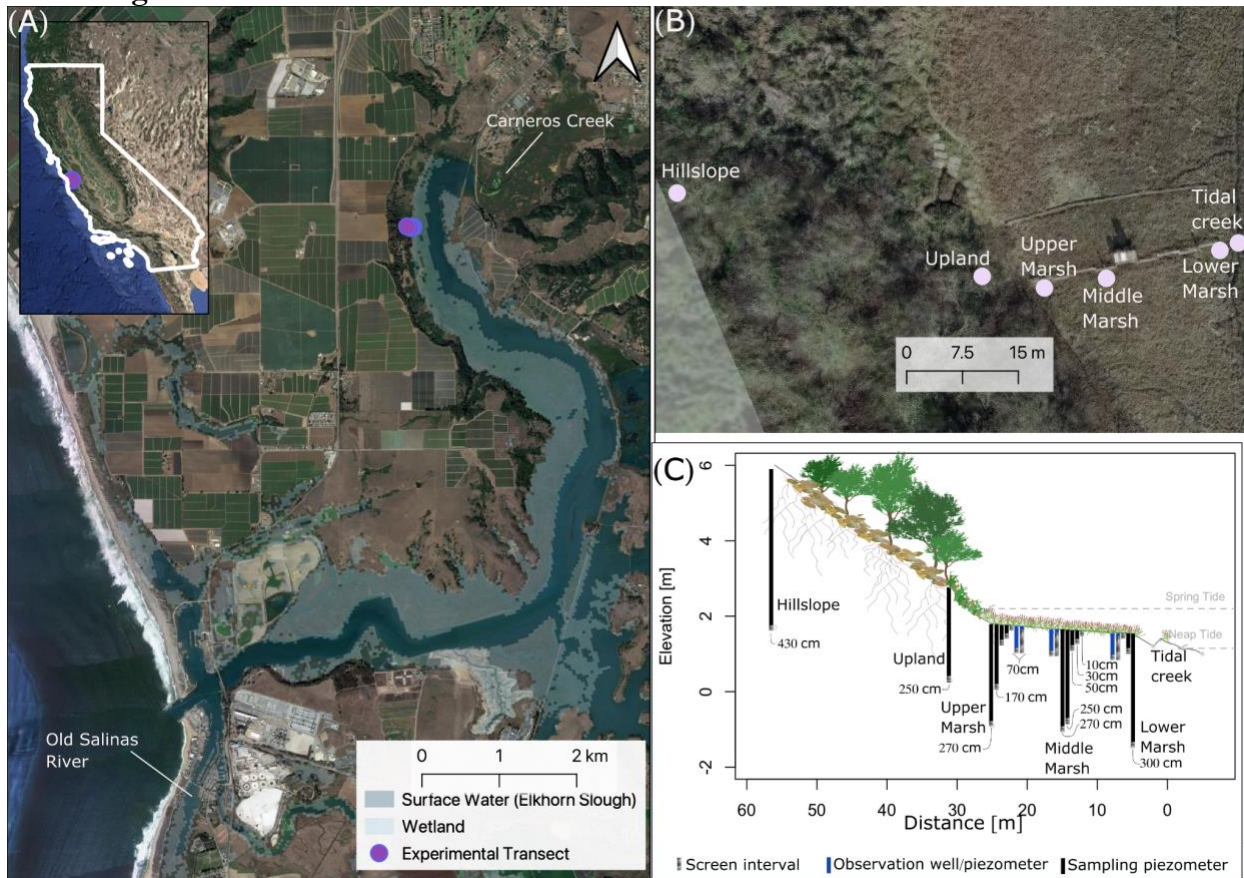
10/14/16	10.0	0.5	33	4%	Oakland*
03/03/18	13.1	0.7	1	0%	Oakland*
04/07/18	10.0	0.5	27	5%	Oakland*
03/05/21	4.9	0.2	6	2%	Oakland <sup>&amp;</sup>
03/06/21	4.9	0.3	2	1%	Oakland <sup>&amp;</sup>
03/09/21	9.1	0.9	9	3%	Oakland <sup>&amp;</sup>
03/10/21	10.0	0.4	12	4%	Oakland <sup>&amp;</sup>
03/11/21	11.2	0.5	2	1%	Oakland <sup>&amp;</sup>
03/14/21	13.3	0.6	10	3%	Oakland <sup>&amp;</sup>
10/22/21	6.1	0.4	22	7%	Elkhorn Slough
12/28/21	5.6	0.2	0	0%	Elkhorn Slough
<b>Mean</b>	<b>8.6</b>				
<b>St.Dev.</b>	<b>3.3</b>				
<b>N</b>	<b>21</b>				
<b>Unc.Mean</b>	<b>0.7</b>				

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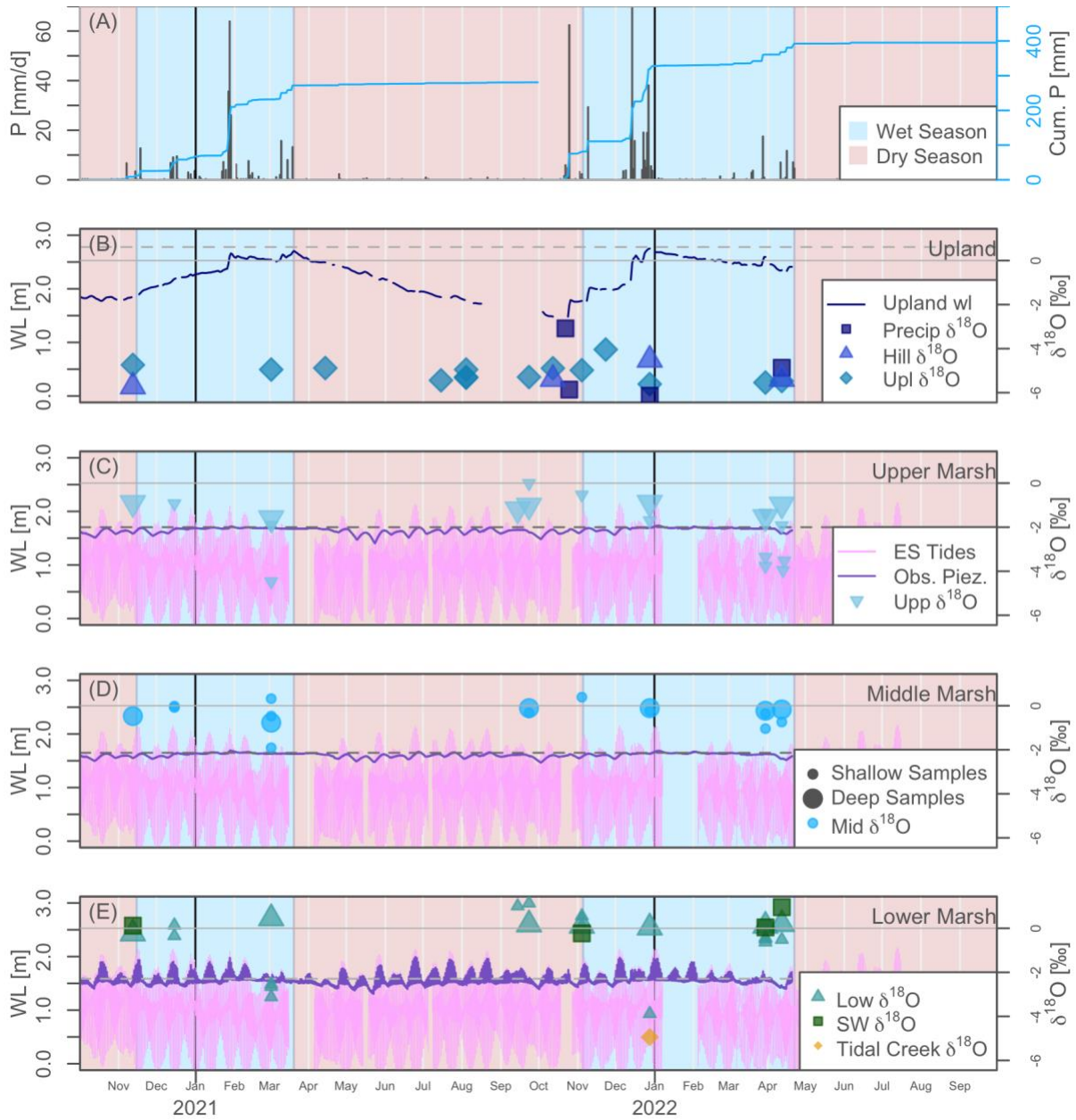
*Table 6. Summary of tritium ages, recharge flow velocities and recharge rates of subsurface water across the study transect.*

<b>Date</b>	<b>Position</b>	<b>Depth [cm]</b>	<b>Age [years]</b>	<b>Age uncertainty [years]</b>	<b>Recharge flow velocity [cm/yr]</b>	<b>Recharge rate [cm/yr]</b>
11/12/20	Upl	250	5.8	1.7	43	13
11/12/20	Hill	430	12.7	1.8	34	13

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 982 *Figure 1. (A) Map of Elkhorn Slough with the extent of wetlands outlined in light blue. The purple symbol*  
 983 *marks the location of the study transect. Notice that the experimental transect is adjacent to agricultural fields. (B)*  
 984 *Map view of the experimental transect showing the location of the hillslope and upland samling piezometers in relation*  
 985 *to the salt marsh transect. (C) Illustration of the experimental transect showing the spatial distribution of the sampling*  
 986 *(black) and observation piezometers and wells (blue). The elevation ( m relative to NAVD88) of the salt marsh*  
 987 *positions are: 1.79 m, 1.65 m, and 1.55 m for the upper, middle, and lower marsh, respectively. The elevation of the*  
 988 *hillslope and upland positions is 6 m and 2.4 m, respectively.*  
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 991 *Figure 2. Time series of precipitation (A) and subsurface water level (WL) in the upland (B) and across the*  
 992 *salt marsh (C-E). The figure shows data for two water years (2020 and 2021) and illustrates the seasonal variations*  
 993 *in the terrestrial water level as measured in the upland position (B). The water level elevations are relative to*  
 994 *NAVD88. The Mediterranean climate of the study area, with marked seasonality in precipitation (A), results in a*  
 995 *drop of 1.34 m in the terrestrial groundwater level between the rainy and the dry seasons. The horizontal dashed lines*  
 996 *in C-E represent the salt marsh elevation at each marsh position. The symbols are  $\delta^{18}\text{O}$  time series across the different*  
 997 *positions. The small symbols correspond to shallow samples ( $\leq 50$  cm-bgs) and the larger symbols correspond to deep*  
 998 *samples ( $> 50$  cm-bgs). (C-E) shows the tidal record over the study period(ES Tides). Notice that spring tides flood*  
 999 *the salt marsh platform.*

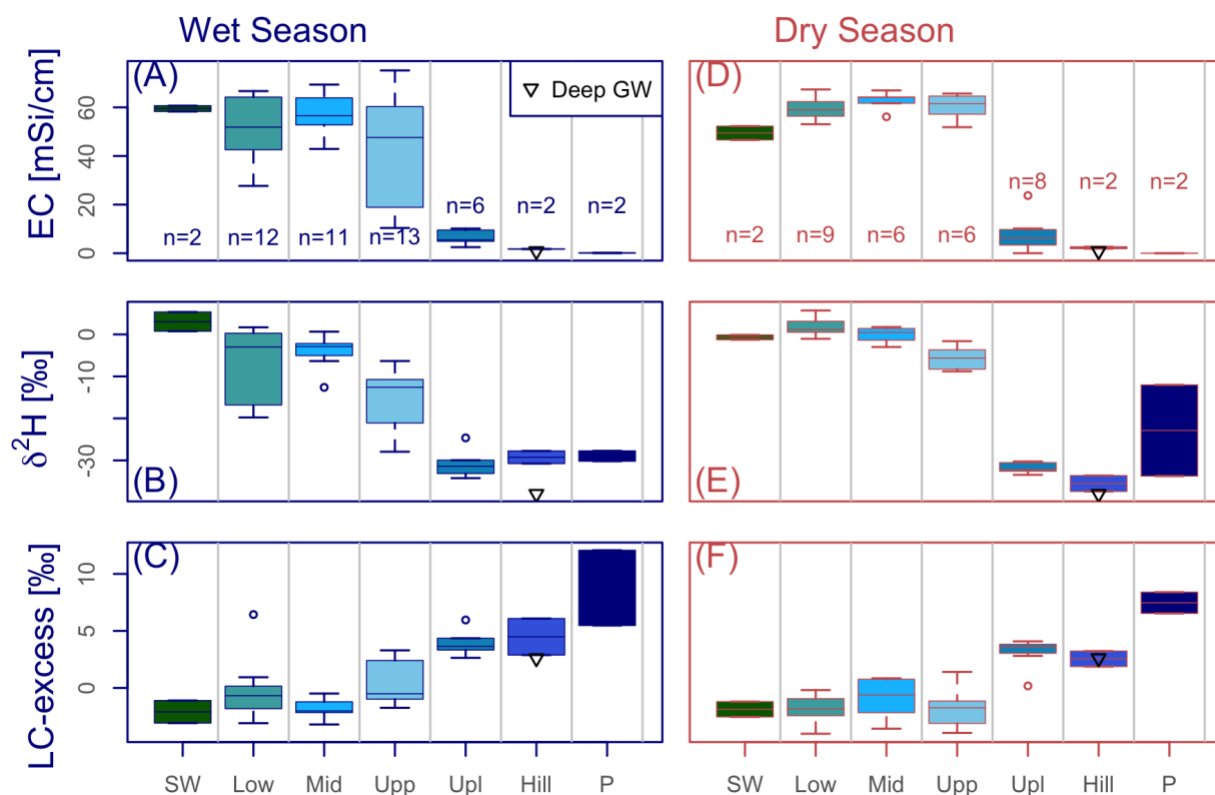
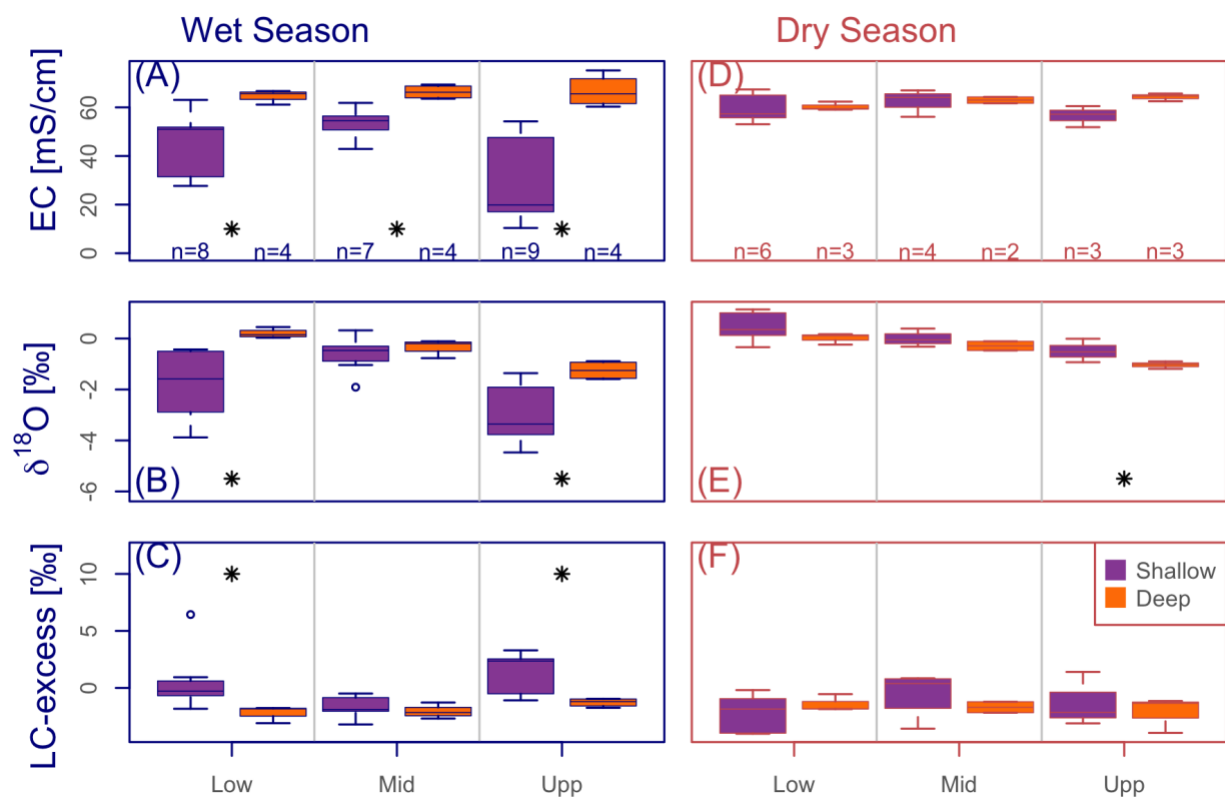
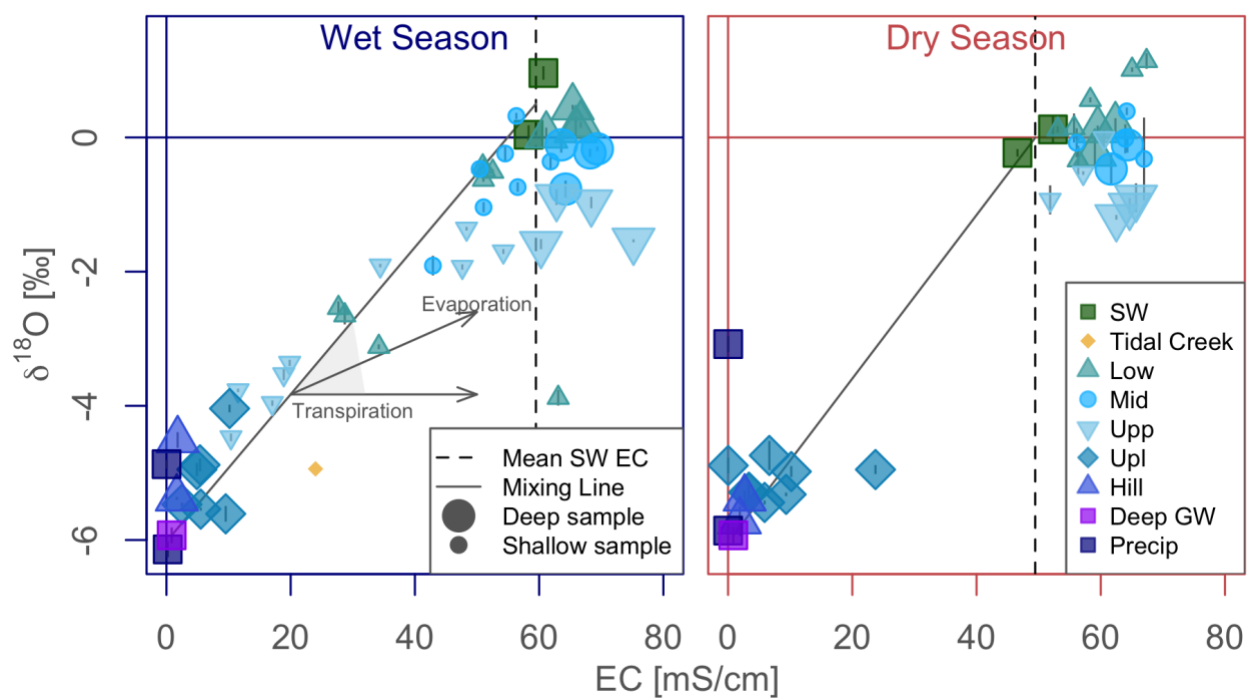


Figure 3. Boxplots of conductivity (A-D),  $\delta^{18}\text{O}$  (B-E) and lc-excess (C-F) for all estuarine/ surface water (SW), lower marsh (low), middle marsh (mid), upper marsh (Upp), upland (Upl), and hillslope positions (Hill), and precipitation (P). A-C correspond to the wet season whereas D-F correspond to the dry season. The inverted triangle symbol represents the deep groundwater sample collected from the irrigation well in the agricultural field above the field site (Figure S1).

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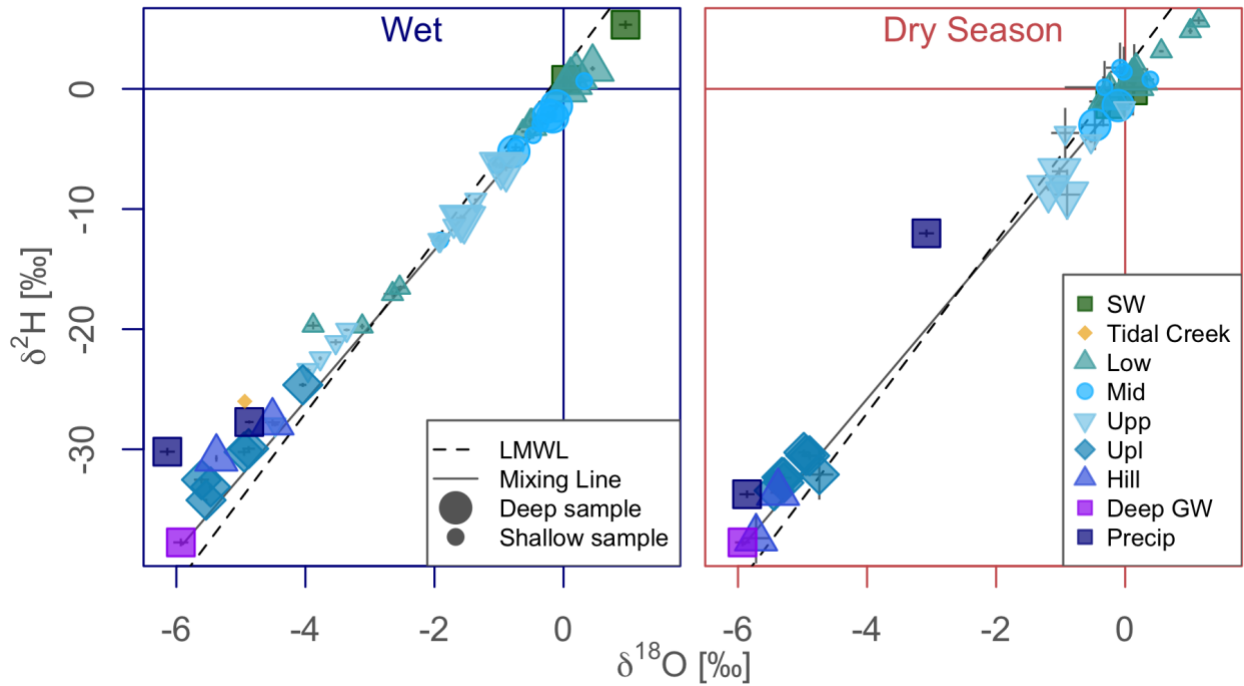


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 1010 *Figure 4. Boxplots of conductivity (A-D),  $\delta^{18}\text{O}$  (B-E) and lc-excess (C-F) for the three marsh positions,*  
 1011 *lower marsh (low), middle marsh (mid), upper marsh (Upp), separated by shallow (purple) and deep (orange)*  
 1012 *sampling depths. A-C correspond to the wet season whereas D-F correspond to the dry season. The asterisks*  
 1013 *demarcate significant differences.*  
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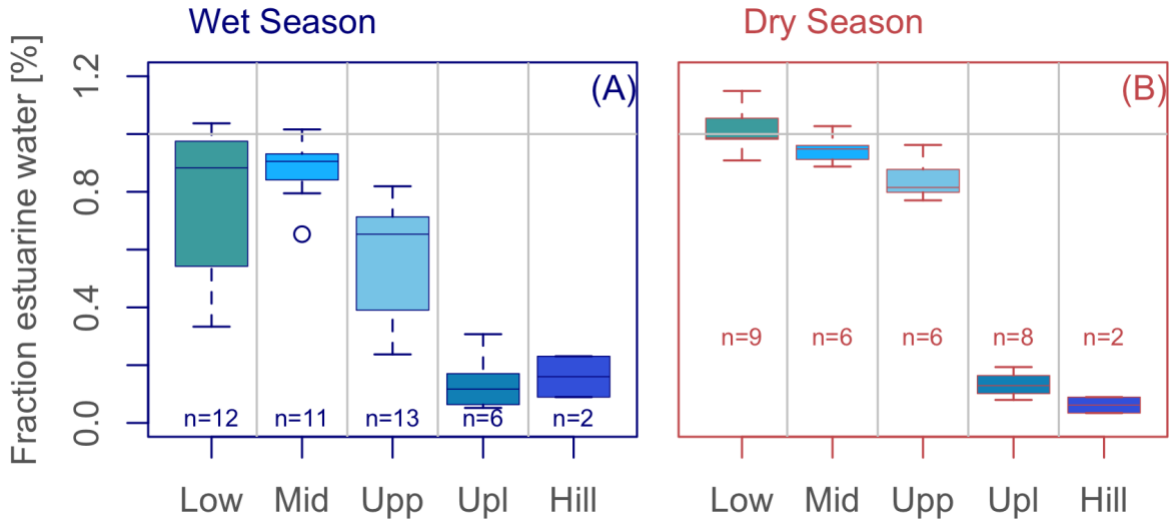
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Figure 5. Biplot of  $\delta^{18}\text{O}$  and electrical conductivity of wet and dry seasons with precipitation (precip), groundwater from the agricultural field adjacent to the study site (Deep GW), Elkborn Slough surface water (SW), the tidal creek, the lower marsh (Low), the middle marsh (Mid), the upper marsh (Upp), the uland position (Upl), and the hillslope position (Hill). The mixing line connects the deep groundwater, representing the long-term mix of local precipitation, and the mean Elkborn Slough surface water, representing marine water. All the samples are separated by shape and color (see the figure legend). The size of the points separate shallow (smaller symbols) and deep samples (larger symbols). The error bars are the standard deviation (analytical uncertainty) from individual measurements (notice that in most samples the error bars are within the symbol size).



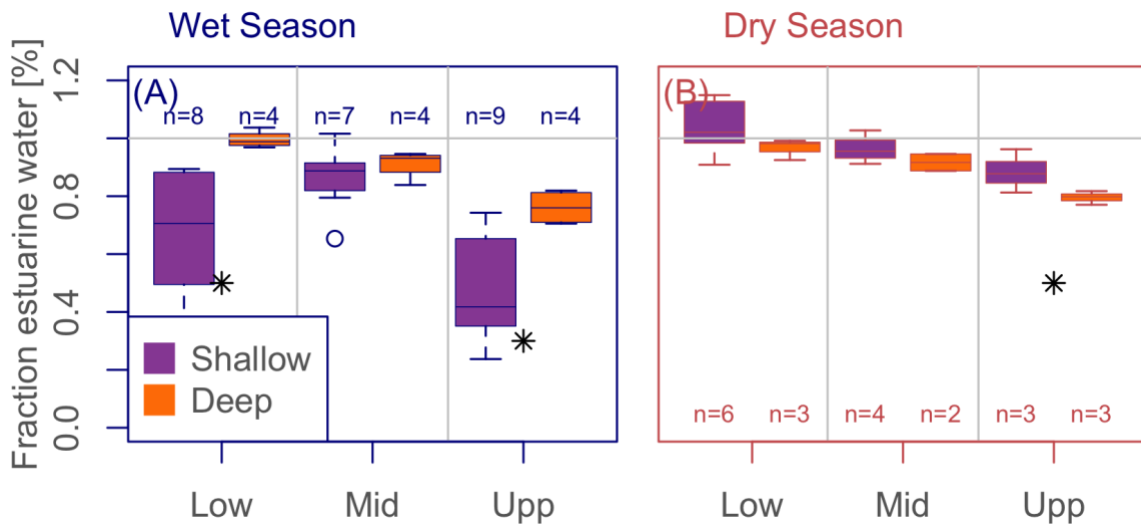
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Figure 6. Dual isotope plot of wet and dry seasons with precipitation (precip), groundwater from the agricultural field adjacent to the study site (Deep GW), Elkborn Slough surface water (SW), the tidal creek, the lower marsh (Low), the middle marsh (Mid), the upper marsh (Upp), the uland position (Upl), and the hillslope position (Hill). The local meteoric water line (LMWL) is:  $\delta^2\text{H} = 7.1 \times \delta^{18}\text{O} + 1.6$ . The mixing line connects the deep groundwater, representing the long-term mix of local precipitation, and the mean Elkborn Slough surface water, representing marine water. All the samples are separated by shape and color (see the figure legend). The size of the points separate shallow (smaller symbols) and deep samples (larger symbols). The error bars are the standard deviation (analytical uncertainty) from individual measurements (notice that in most samples the error bars are within the symbol size).



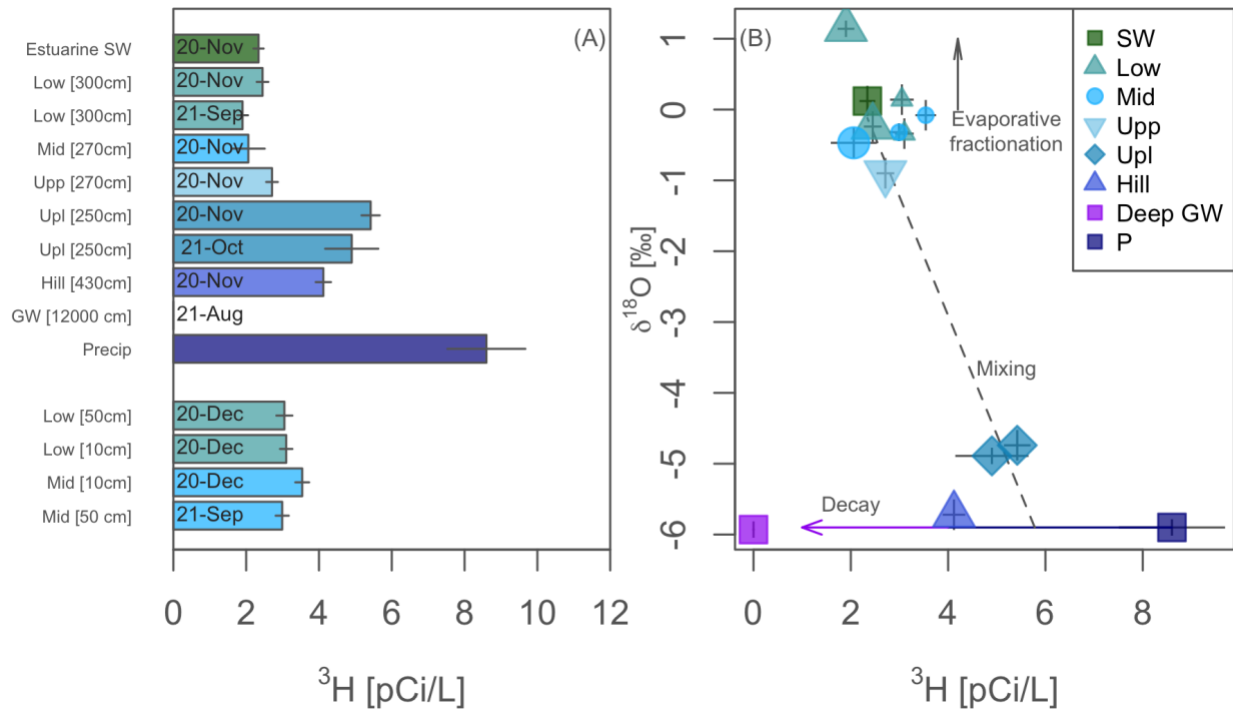
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Figure 7. Boxplots of fractions of estuarine water (Elkborn Slough) across the different positions for the wet (A) and dry seasons (B) calculated using  $\delta^{18}\text{O}$  as a tracer. Notice that during the dry season, some salt marsh samples are isotopically heavier than Elkborn Slough water.



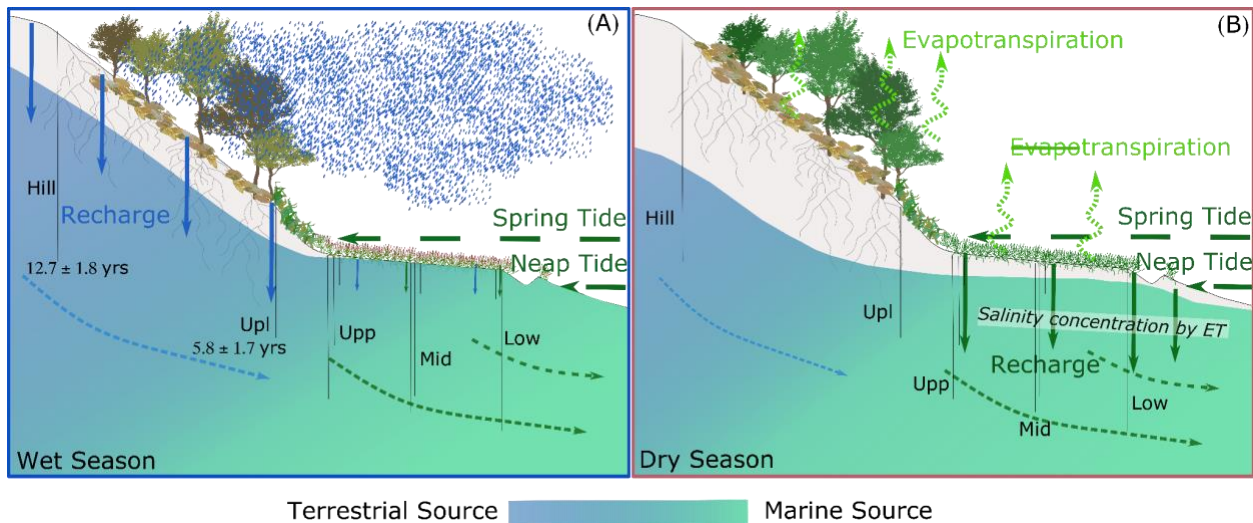
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Figure 8. Boxplot of fractional mix of Elkborn Slough water in the three salt marsh positions separated by wet (A) and dry seasons (B). The asterisks demarcate significant differences.



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Figure 9. Barplot of tritium in subsurface water (A) and  $\delta^{18}\text{O}$  and  $^3\text{H}$  biplot illustrating the mixing and decay processes for the different positions (B). The terrestrial age was calculated from the interception of the mixing line and the decay lines. The labels in the bars in (A) are the month/year when the samples were collected. The names of each sample in (A) also shows the depth of the sample in brackets. The error bars are the analytical error. The error bar of the precipitation is the standard error of the mean. GW is the deep groundwater sample.



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Figure 10. Conceptual model of the study transect for the wet (A) and dry seasons (B). Semidiurnal, the salt marsh is flooded resulting in surface water-subsurface water exchanges. Frequent flushing of shallow subsurface water in the marsh results in subsurface water ages in the order of  $\sim 0$  yrs, while subsurface water ages at depth increase from the hillslope towards the lower marsh. The dashed arrows in the subsurface indicate the major flow pathways. Vertical arrows represent subsurface recharge to the terrestrial aquifer (blue) and the salt marsh (green). Although subsurface water is super-saline, we did not find any evaporative signal in our samples, indicating that plant transpiration is a first order driver increasing salinity in subsurface water. The salt marsh subsurface is recharged during the dry season.



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## Supporting information for:

### Flow directions and ages of subsurface water in a Mediterranean salt marsh system constrained by isotope tracing

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### Content of this file

- Figures S1 to S4
- Text S1
- Table S1 to S6

### Introduction

Here we provide supplementary figures and tables. Figures provide (1) an elevation model of the study area showing some of the geomorphologic features of the site that provide surface runoff onto the salt marsh (2) time series (hourly time steps) of the meteorological parameters available for this study (3) stable water isotope biplot illustrating the deviation of local precipitation collected in the 2022 water year from the long-term local meteoric water line and (4) a plot of lc-excess versus  $\delta^{18}\text{O}$  showing minimal evaporative fractionation in the salt marsh subsurface (5) a plot of tritium activity in precipitation versus several hydroclimatic parameters.

1101 The text S1 provides an explanation of the analysis of apparent tritium ages and lumped parameter  
1102 models to interpret tritium values

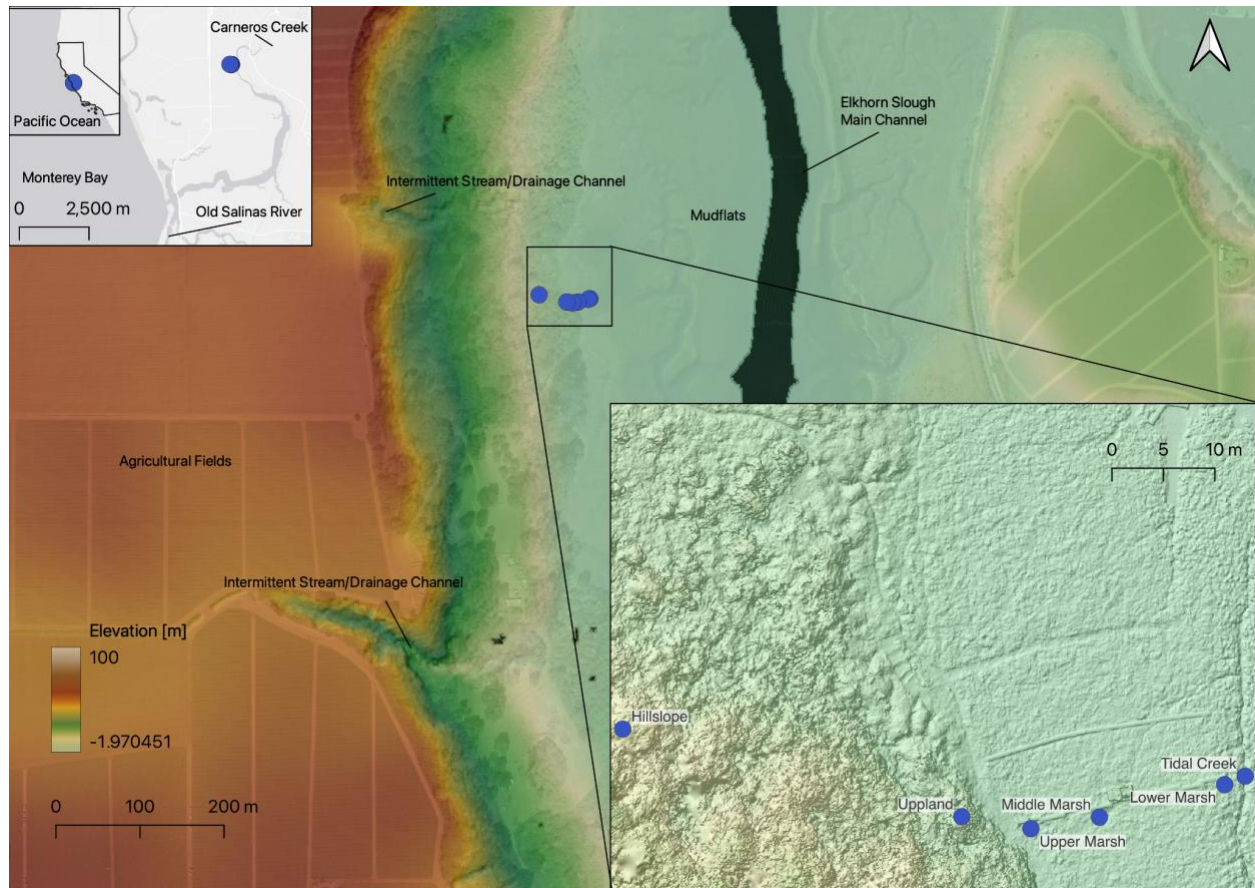
1103 The table provides (1) summary of depth and the type (observation versus sampling) piezometers  
1104 used in the study (2) summary schedule of isotope and conductivity sampling/measurement (3)  
1105 summary p-value of multiple pairwise Mann-Whitney U test for electrical conductivity across the  
1106 experimental transect (4) summary p-value of multiple pairwise Mann-Whitney U test for  $\delta^2\text{H}$  across  
1107 the experimental transect (5) summary p-value of multiple pairwise Mann-Whitney U test for lc-  
1108 excess across the experimental transect (6) summary p-value of multiple pairwise Mann-Whitney U  
1109 test for fractional mixing of Elkhorn Slough water across the experimental transect

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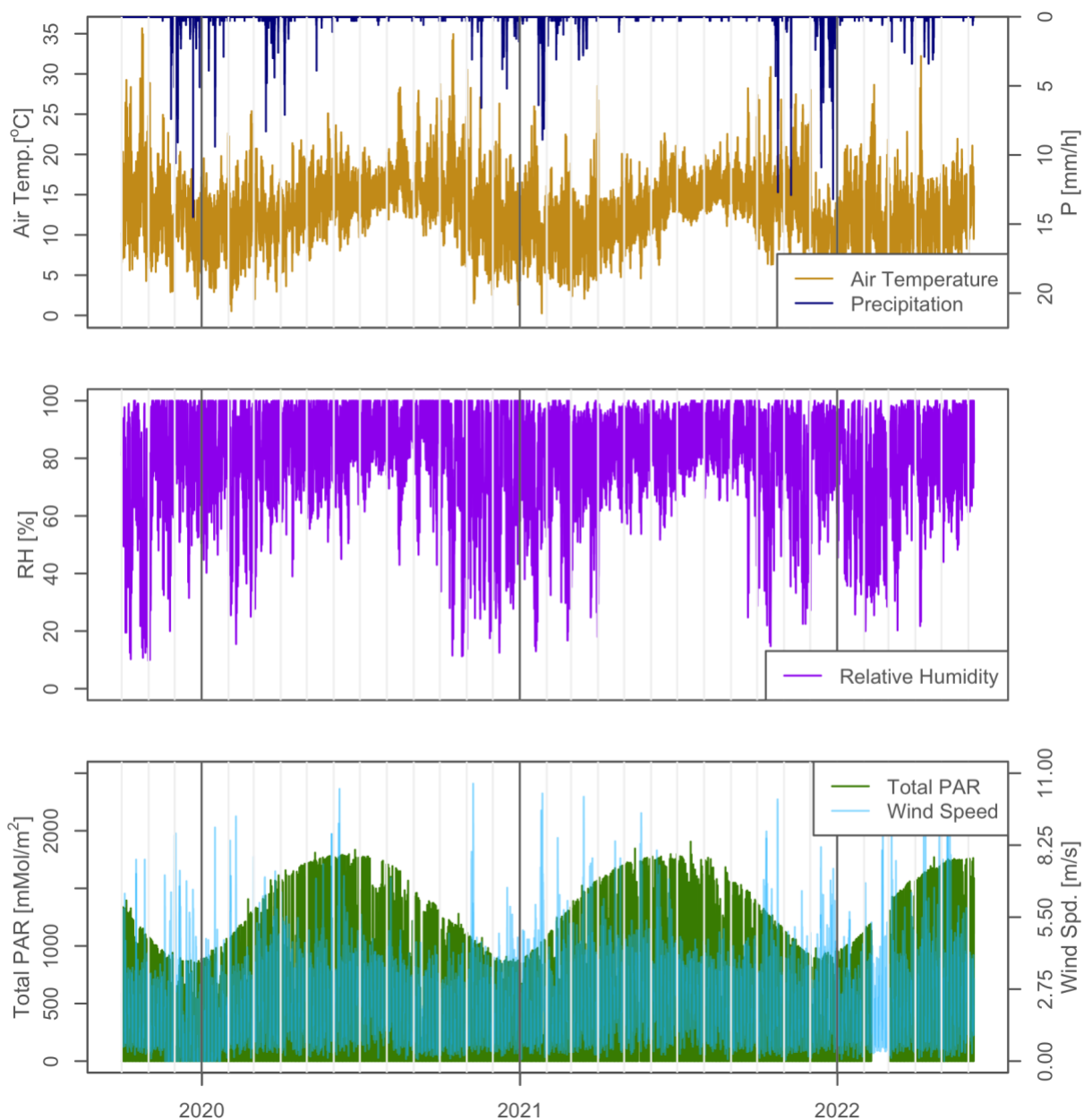
1113 **Figure S1.** Map of the study site



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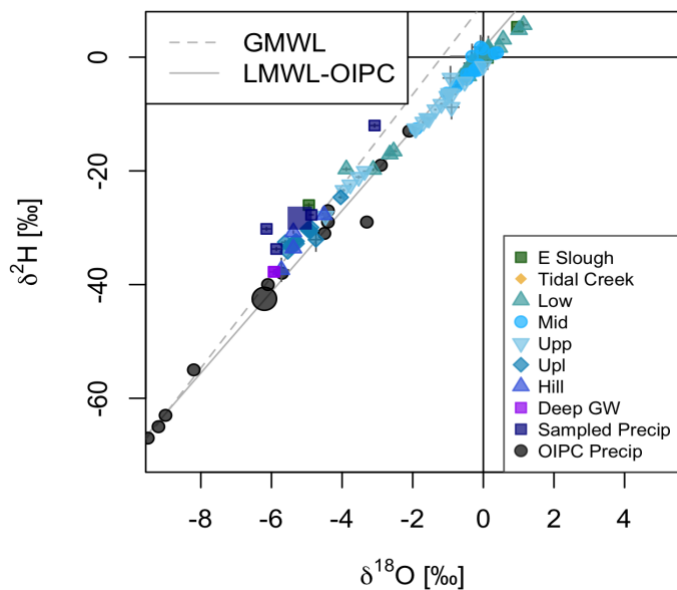
1115 *Figure S1. Map of the study site showing the elevation gradient of the area. The figure highlights the geomorphological*  
1116 *features that provide pathways for surface runoff onto the salt marsh during precipitation events. The inset shows a high*  
1117 *resolution DEM of the experimental transect, illustrating some of the salt marsh features such as the tidal creek.*

1118 **Figure S2.** Time series of meteorological data available for the study



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1120 *Figure S2. Time series of all the meteorological variables available for the study (hourly). Meteorological data is*  
1121 *measured through the Elkhorn SLough National Estuarine Research Reserve ~4.5 km from the study site. PAR is*  
1122 *photosynthetically active radiation.*

1123 **Figure S3.** Stable water isotopes biplot

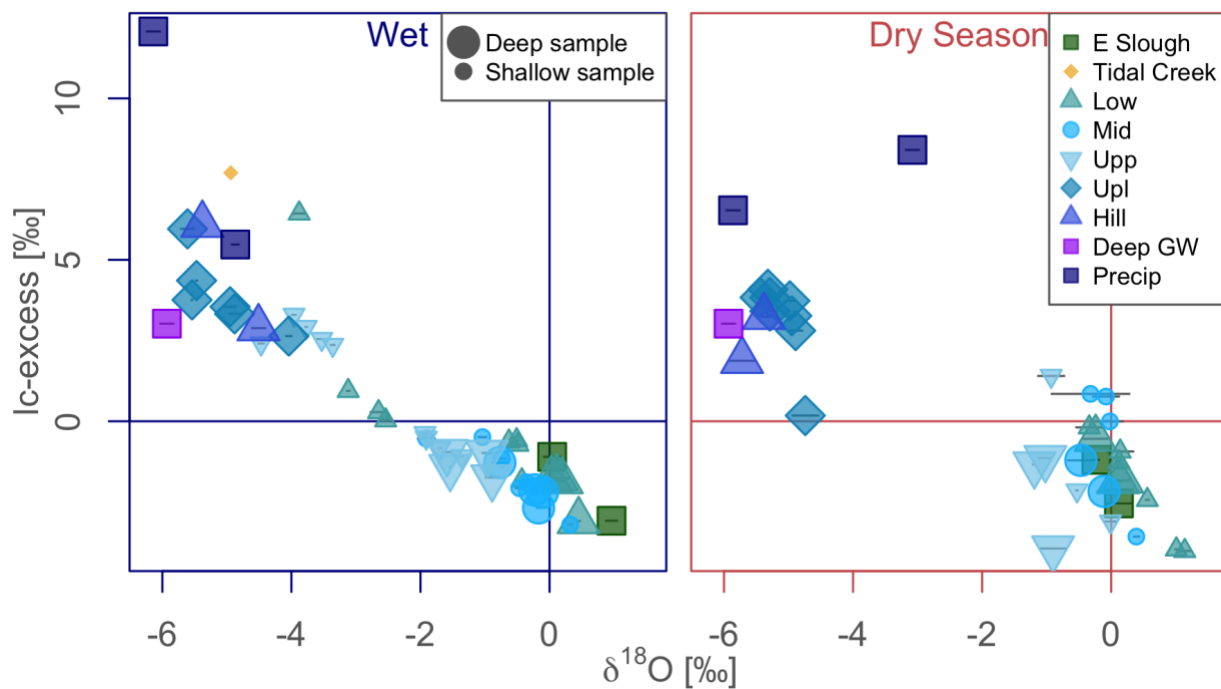


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1125 *Figure S3. Dual isotope plot illustrating the deviation of the local meteoric water line (LMWL) obtained*  
1126 *from the online isotope in precipitation calculator (OIPC). Notice that the precipitation samples, collected at the site*  
1127 *during the 2022 water year, deviate from the long-term estimates that form the LMWL. The figure also illustrates the*  
1128 *global meteoric water line (GMWL). The volume weighted sampled precipitation and the annual average isotopic*  
1129 *compositions per the OIPC at the cite are shown by the larger size symbols.*

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1131 **Figure S4.** lc-excess versus  $\delta^{18}\text{O}$



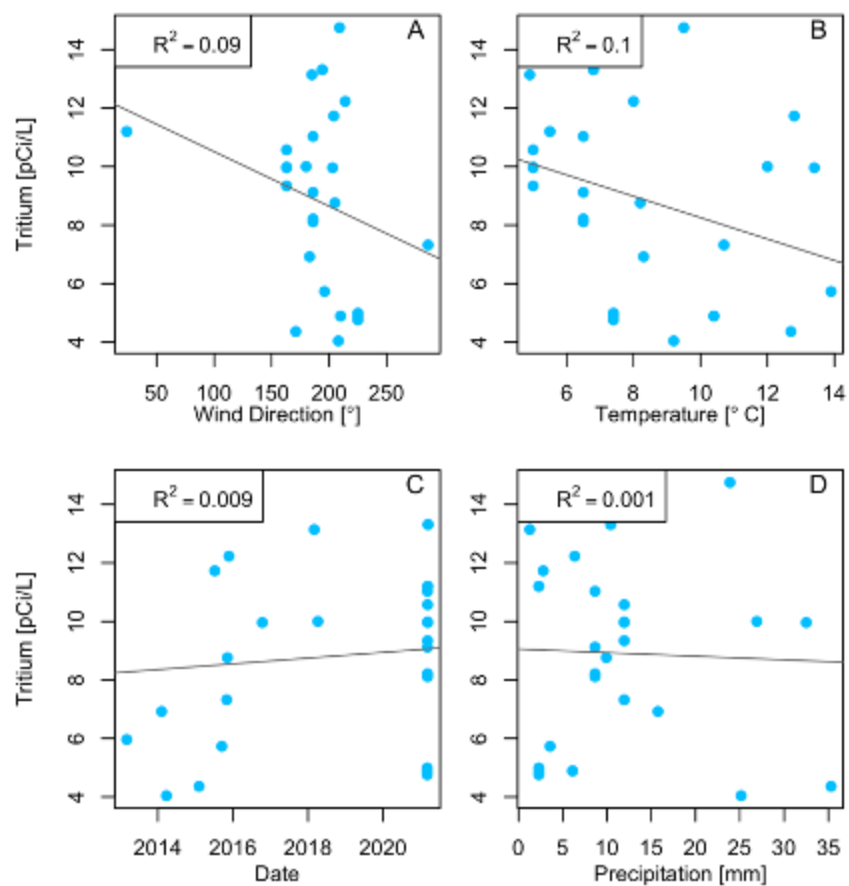
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1133 *Figure S4. Line conditioning excess versus  $\delta^{18}\text{O}$  biplot. The figure shows that there is no significant*  
1134 *evaporative fractionation in the salt marsh subsurface.*

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**Figure S5.** Plot of tritium activity in precipitation versus several hydroclimatic parameters



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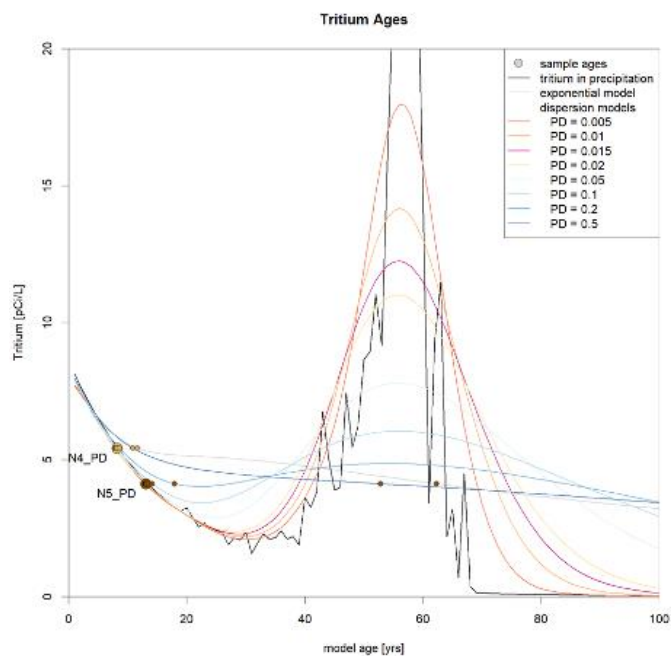
1138 *Figure S5. Tritium activity in precipitation v. wind direction (A), air temperature (B), time (C), and*  
1139 *precipitation depth (D).*

1140 **Text S1- Analysis of apparent tritium ages and lumped parameter models to interpret**  
1141 **tritium values**

1142 To examine the validity and accuracy of the apparent tritium ages, we compared the measured tritium  
1143 concentrations to the exponential and dispersion mixing models. For the dispersion model, we used a  
1144 dispersion parameter ranging from 0.005 to 0.5. The tritium input was based on Harms, multiplied by 1.06 to  
1145 obtain the present-day tritium concentration in precipitation of 8.6 pCi/L.

1146 Figure S1 shows the initial tritium in precipitation, corrected for decay up to 2020 (i.e. piston flow  
1147 model), as well as tritium concentrations predicted by various dispersion and exponential models. Dispersion  
1148 models with small dispersion parameters closely follow the tritium in precipitation while larger dispersion  
1149 parameters smooth the peak of tritium in precipitation that occurred 50-60 yrs ago. The tritium concentration  
1150 of the exponential model is continuously decreasing with age.

1151 The tritium concentrations measured in the deep terrestrial piezometers (Upland=N4, Hillslope=N5) are  
1152 plotted at their respective apparent tritium ages (larger symbols) as well as the youngest model age that results  
1153 in the same tritium value (smaller symbols). For dispersion models with a parameter of 0.1 or smaller, the  
1154 model ages are close to the apparent ages. For larger dispersion parameters and the exponential model, there  
1155 is a substantial difference between the model and apparent ages, in particular for the Hillslope (N5) deep  
1156 piezometer. For both piezometers, additional possible ages on the rising and falling edge of the bomb pulse  
1157 are possible (but not plotted here). While these older age estimates are hypothetically possible, given the  
1158 assumptions underlying these lumped parameter models, their hydrological interpretation is less probable. For  
1159 example, age interpretations around the falling edge of the bomb pulse (~40 yrs old) require very low  
1160 recharge rates (2-4 mm/yr) typically associated with desert environments.



1161 *Figure 1 Text S1: Measured tritium concentrations in Upland (N4) and Hillslope (N5) deep piezometers, together with initial*  
1162 *tritium in precipitation, corrected for decay up to 2020 (i.e. piston flow model) and tritium concentrations predicted by various*  
1163 *dispersion and exponential models.*

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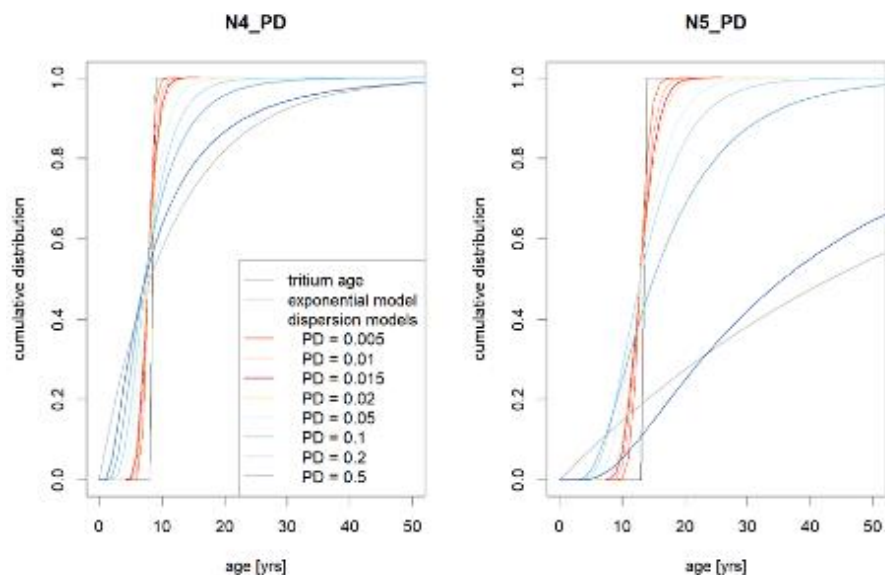
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Figure S2 shows the cumulative age distributions associated with the first intersection of the measured and modeled tritium values (as illustrated in Fig S1). Although the mean age for N4 is larger for the exponential model and the dispersion model with PD=0.5, the median age (where the cumulative age distribution reaches 0.5) for all age models is very similar. For N5, the median age of the exponential and PD=0.5 models are substantially different.



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Figure 2 Text S1: Cumulative age distributions associated with the various lumped parameter models for the Upland (N4) and Hillslope (N5) deep piezometers

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Repeated tritium measurements over the course of 16 months show stable tritium concentrations in both piezometers, suggesting that dispersion along subsurface flow paths reduce observed seasonal and interannual variability in precipitation (Table 1 Text S1). Constraining the magnitude of dispersion (PD parameter) is hypothetically possible but beyond the scope of this study.

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 1181 Table 1 Text S1. Tritium activity in the upland and hillslope across wet and dry seasons.

Position	Depth (cm)	Date	Tritium (pCi/L)	Uncertainty (pCi/L)
Upland	250	12/16/2020	5.42	0.25
Upland	250	9/23/2021	5.22	0.25
Upland	250	12/28/2021	5.47	1.50
Upland	250	4/12/2022	5.40	0.35
Hillslope	430	12/16/2020	4.12	0.21
Hillslope	430	10/12/2021	4.22	0.35
Hillslope	430	12/28/2021	4.15	0.21
Hillslope	430	4/12/2022	6.06	0.59

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**Table S1.** Summary of depth and the type of well/piezometer used in the study by depth and position

Position	well/piezometer	Depth [cm]	Pressure transducer?	sampling?
Lower Marsh	piezometer	10	no	yes
Lower Marsh	piezometer	50	no	yes
Lower Marsh	well	70	yes	no
Lower Marsh	piezometer	70	yes	no
Lower Marsh	piezometer	300	yes*	yes
Middle Marsh	piezometer	10	no	yes
Middle Marsh	piezometer	30	no	yes
Middle Marsh	piezometer	50	no	yes
Middle Marsh	well	70	yes	no
Middle Marsh	piezometer	70	yes	no
Middle Marsh	piezometer	250	no	yes
Middle Marsh	piezometer	270	no	yes
Upper Marsh	piezometer	10	no	yes
Upper Marsh	piezometer	30	no	yes
Upper Marsh	piezometer	50	no	yes
Upper Marsh	well	70	yes	no
Upper Marsh	piezometer	70	yes	no
Upper Marsh	piezometer	170	yes*	yes
Upper Marsh	piezometer	270	yes*	yes
Upland	piezometer	250	yes	yes
Hillslope	piezometer	430	no	yes

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*\*Data was not used in this study*

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**Table S2.** Summary schedule of water isotopes and conductivity samples

Sample Type	Position	Depth [cm]	Fall 2020	Winter 2021	Spring 2020	Summer 2021	n
Precipitation	-	-		x	x		4
Elkhorn Slough Surface Water	-	-	x		x		3
Tidal Creek	-	-		x			1
Deep Groundwater	-	4000		x			2
Subsurface Water	Lower Marsh	10	x				1
Subsurface Water	Lower Marsh	50	x	x	x	x	5
Subsurface Water	Lower Marsh	300	x	x	x	x	5
Subsurface Water	Middle Marsh	10	x				1
Subsurface Water	Middle Marsh	30	x				1
Subsurface Water	Middle Marsh	50	x	x	x		4
Subsurface Water	Middle Marsh	250		x	x	x	3
Subsurface Water	Middle Marsh	270	x		x		2
Subsurface Water	Upper marsh	10			x		1
Subsurface Water	Upper marsh	30			x	x	2
Subsurface Water	Upper marsh	50	x		x	x	4
Subsurface Water	Upper marsh	170		x	x	x	3
Subsurface Water	Upper marsh	270	x		x	x	4
Subsurface Water	Upland	250	x	x	x	x	8
Subsurface Water	Hillslope	430	x	x	x	x	4

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*Table S2. Summary schedule of stable water isotopes and conductivity samples*

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**Table S3.** Summary p-value of Kruskal-Wallis test of electrical conductivity amongst all marsh positions

	SW	Tidal Creek	Low	Mid	Upp	Upl	Hill	GW
Tidal Creek	0.12	-	0.11	0.11	0.25	0.12	0.16	0.16
Low	0.5	0.11	-	0.3	0.35	<b>0.0001</b>	<b>0.002</b>	<b>0.02</b>
Mid	0.2	0.11	0.3	-	<b>0.0001</b>	<b>0.0001</b>	<b>0.002</b>	<b>0.02</b>
Upp	0.9	0.25	0.35	<b>0.0001</b>	-	0.09	<b>0.002</b>	<b>0.03</b>
Upl	<b>0.003</b>	0.12	<b>0.0001</b>	<b>0.0001</b>	0.09	-	<b>0.01</b>	0.06
Hill	<b>0.02</b>	0.16	<b>0.002</b>	<b>0.002</b>	<b>0.002</b>	<b>0.01</b>	-	0.06
GW	<b>0.06</b>	0.16	<b>0.02</b>	<b>0.02</b>	<b>0.03</b>	0.06	0.06	-
P	<b>0.02</b>	0.15	<b>0.004</b>	<b>0.004</b>	<b>0.003</b>	<b>0.02</b>	<b>0.02</b>	<b>0.06</b>

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*Table S3. Summary p-value of Kruskal-Wallis test of electrical conductivity amongst all marsh positions. Boded values represent significant differences in the mean rank of EC between the two positions.*

**Table S4.** Summary p-value of Kruskal-Wallis test of  $\delta^2\text{H}$  amongst all marsh positions

	SW	Tidal Creek	Low	Mid	Upp	Upl	Hill	GW
Tidal Creek	0.2	-	0.09	0.1	0.14	0.16	0.2	0.22
Low	0.5	<b>0.09</b>	-	0.3	<b>0.0002</b>	<b>0.0001</b>	<b>0.002</b>	<b>0.002</b>
Mid	0.06	0.1	0.3	-	<b>0.0001</b>	<b>0.0001</b>	<b>0.002</b>	<b>0.02</b>
Upp	<b>0.002</b>	0.14	<b>0.0002</b>	<b>0.0001</b>	-	<b>0.0001</b>	<b>0.003</b>	<b>0.02</b>
Upl	<b>0.003</b>	0.16	<b>0.0001</b>	<b>0.0001</b>	<b>0.0001</b>	-	0.5	<b>0.02</b>
Hill	<b>0.03</b>	0.2	<b>0.002</b>	<b>0.002</b>	<b>0.003</b>	0.5	-	0.06
GW	0.08	0.22	<b>0.002</b>	<b>0.02</b>	<b>0.02</b>	<b>0.02</b>	0.06	-
P	<b>0.03</b>	0.48	<b>0.004</b>	<b>0.003</b>	<b>0.01</b>	0.2	0.9	0.06

*Table S4. Summary p-value of Kruskal-Wallis test of  $\delta^2\text{H}$  amongst all marsh positions. Boded values represent significant differences in the mean rank of  $\delta^2\text{H}$  between the two positions.*

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1199 **Table S5.** Summary p-value of Kruskal-Wallis test of lc-excess amongst all marsh positions

	ES	Hill	GW	Low	Mid	P	Tidal Creek	Upl
Hill	0.2		0.3	0.6	0.09	<b>0.003</b>	<b>0.03</b>	0.08
GW	0.3	0.3		0.4	0.2	<b>0.0001</b>	<b>0.005</b>	<b>0.04</b>
Low	0.6	0.6	0.4		0.05	<b>0.0001</b>	<b>0.002</b>	<b>0.02</b>
Mid	0.09	0.09	0.2	0.05		<b>0.0001</b>	<b>0.01</b>	<b>0.04</b>
P	<b>0.003</b>	<b>0.003</b>	<b>0.0001</b>	<b>0.0001</b>	<b>0.0001</b>		0.5	0.2
Tidal Creek	<b>0.03</b>	<b>0.03</b>	<b>0.005</b>	<b>0.002</b>	<b>0.01</b>	0.5		<b>0.04</b>
Upl	0.08	0.08	<b>0.04</b>	<b>0.02</b>	<b>0.04</b>	0.2	1	
Upp	<b>0.03</b>	1	<b>0.002</b>	<b>0.002</b>	<b>0.002</b>	<b>0.004</b>	<b>0.04</b>	0.06

1200 *Table S5. Summary p-value of Kruskal-Wallis test of lc-excess amongst all marsh positions. Boded values*  
 1201 *represent significant differences in the mean rank of lc-excess between the two positions.*

1202 **Table S6.** Summary p-value of Kruskal-Wallis test of fractional mixing of Elkhorn Slough  
 1203 water amongst all marsh positions

	Low	Mid	Upp	Upl
Mid	0.3	-	<b>0.001</b>	<b>0.0001</b>
Upp	<b>0.0001</b>	<b>0.001</b>	-	<b>0.0001</b>
Upl	<b>0.0001</b>	<b>0.0001</b>	<b>0.0001</b>	-
Hill	<b>0.002</b>	<b>0.002</b>	<b>0.003</b>	0.5

1204 *Table S5. Summary p-value of Kruskal-Wallis test of fractional mixing of Elkhorn Slough water amongst*  
 1205 *all marsh positions. Boded values represent significant differences in the mean rank of fractional mixing of Elkhorn*  
 1206 *Slough water between the two positions.*

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