Impact of coastal marsh eco-geomorphologic change on the prediction of saltwater intrusion under future sea level rise

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Abstract

Coastal saltwater intrusion (SWI) is one key factor affecting the hydrology, nutrient transport, and biogeochemistry of coastal marsh landscapes. Future climate change, especially intensified sea level rise (SLR), is expected to trigger SWI to encroach coastal freshwater aquifers more intensively. Numerous studies have investigated decadal/century scale SWI under SLR by assuming a static coastal landscape topography. However, coastal marshes are highly dynamic systems in response to SLR, and the impact of coastal marsh evolution on SWI has received very little attention. Thus, this study investigated how coastal marsh evolution affects future SWI with a physically-based coastal hydro-eco-geomorphologic model, ATS (Advanced Terrestrial Simulator). Our synthetic modeling experiments showed that it is very likely that the marsh elevation increases with future SLR, and a depression zone is formed due to the different marsh accretion rates between the ocean boundary and the inland. We found that, compared to the cases without marsh evolution, the marsh accretion may significantly reduce the surface saltwater inflow at the ocean boundary, and the evolved topographic depression zone may prolong the residence time of surface ponding saltwater, which causes distinct subsurface salinity distributions. We also found that the marshland may become more sensitive to the upland groundwater table that controls the freshwater flux to the marshes, compared with the cases without marsh evolution. This study demonstrates the importance of marsh evolution to the freshwater-saltwater interaction under sea level rise and can help improve our predictive understanding of the vulnerability of the coastal freshwater system to sea level rise.

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14	Key Points:
15	• The effect of coastal marsh evolution on future saltwater intrusion is examined for the
16	first time
17	• Marsh accretion under sea level rise may significantly reduce surface seawater inflow and
18	prolong the surface seawater residence time
19	• Future saltwater intrusion on the evolved marsh landscape may become more sensitive to
20	upland groundwater inflows

21 Abstract

Coastal saltwater intrusion (SWI) is one key factor that affects the hydrology, nutrient 22 transport, and biogeochemistry of coastal marsh landscapes. Future climate change, especially 23 intensified sea level rise (SLR), is expected to trigger SWI to encroach coastal freshwater 24 aquifers more intensively. Numerous studies have investigated decadal/century scale SWI under 25 SLR by assuming a static coastal landscape topography. However, coastal marshes are highly 26 27 dynamic systems in response to SLR, and the impact of coastal marsh evolution on SWI has received very little attention. Thus, this study investigated how coastal marsh evolution affects 28 future SWI with a physically-based coastal hydro-eco-geomorphologic model, ATS (Advanced 29 30 Terrestrial Simulator). Our synthetic modeling experiments showed that it is very likely that the marsh elevation increases with future SLR, and a depression zone is formed due to the different 31 marsh accretion rates between the ocean boundary and the inland. We found that, compared to 32 the cases without marsh evolution, the marsh accretion may significantly reduce the surface 33 saltwater inflow at the ocean boundary, and the evolved topographic depression zone may 34 prolong the residence time of surface ponding saltwater, which causes distinct subsurface salinity 35 distributions. We also found that the marshland may become more sensitive to the upland 36 37 groundwater table that controls the freshwater flux to the marshes, compared with the cases 38 without marsh evolution. This study demonstrates the importance of marsh evolution to the freshwater-saltwater interaction under sea level rise and can help improve our predictive 39 understanding of the vulnerability of the coastal freshwater system to sea level rise. 40 Keywords: Saltwater intrusion, Coastal marsh evolution, Sea level rise, Freshwater-saltwater 41

42 interface, Groundwater table, Coastal aquifer vulnerability

43 **1 Introduction**

Coastal wetlands, unique landscapes that connect the terrestrial landscape and the ocean, 44 are some of the most productive ecosystems on Earth (Tiner, 2013). Climate change, especially 45 sea level rise (SLR) under a warming climate, is one of the biggest threats to the stability and 46 sustainability of coastal wetland ecosystems (Burkett & Kusler, 2000). SLR-driven impacts on 47 coastal marsh ecosystems are strongly affected by changes in coastal hydrology (Zhang et al., 48 49 2019). The rising sea level alters the balance of coastal freshwater-saltwater interaction both on coastal wetland surface and in the subsurface aquifer causing the changes in saltwater intrusion 50 51 (SWI), thereby affecting soil water salinity (Guimond & Tamborski, 2021; Sorensen et al., 1984; 52 Sousa et al., 2010), triggering the mortality of salt-intolerant vegetation (Silvestri & Marani, 2004), and eventually altering the ecosystem functions of coastal wetlands (Burkett & Kusler, 53 2000). Therefore, investigating the response of SWI to SLR is critical for our understanding of 54 the SLR impact on coastal wetland ecosystems. 55

56 Numerous studies have attempted to predict and/or assess the impact of SLR on SWI for decades. These studies have aimed to track the movement of the freshwater and saltwater 57 58 interface in coastal aquifers driven by SLR at global (e.g., Ferguson & Gleeson, 2012; Michael et 59 al., 2013), regional (e.g., Oude Essink et al., 2010; Zhang et al., 2018, 2019), and local/transect scales (e.g., Ataie-Ashtiani et al., 2013; Carretero et al., 2013; Chang et al., 2011; Chen et al., 60 2015; Giambastiani et al., 2007; Hughes et al., 2009; Ketabchi et al., 2014, 2016; Langevin & 61 62 Zygnerski, 2013; Loáiciga et al., 2012; Lu et al., 2015; Masterson & Garabedian, 2007; Mazi et al., 2013; Morgan et al., 2013; Payne, 2010; Rasmussen et al., 2013; Sefelnasr & Sherif, 2014; 63 64 Sorensen et al., 1984; Vandenbohede et al., 2008; Vu et al., 2018; Watson et al., 2010; Werner et al., 2012; Werner & Simmons, 2009; Yang et al., 2015). By using analytical or numerical 65

models, these studies examined SWI in coastal aquifers under SLR, especially under the 66 influence of different environmental settings, such as regional-scale hydrologic connectivity, 67 upland groundwater boundary condition, land surface inundation, groundwater extraction, and 68 recharge. For example, Zhang et al. (2018, 2019) investigated the groundwater flow path and 69 SWI of the coastal wetlands in North Carolina, USA, by considering regional-scale coastal 70 71 hydrologic connectivity. They found that aquifers with the largest seasonal changes of SWI are located hundreds of meters away from the shoreline, where freshwater strongly interacts with 72 saltwater. In terms of the effect of upland boundary conditions on SWI, Werner and Simmons 73 74 (2009) and Werner et al. (2012) found that SLR impact is more extensive in unconfined aquifers with a head-controlled inland boundary, compared with confined aquifers. Carretero et al. 75 (2013) found that SWI increased linearly with SLR in aquifers with a flux-controlled boundary 76 condition, but increased nonlinearly with a head-controlled boundary condition. In terms of the 77 effect of groundwater extraction and freshwater supply (recharge), Loáiciga et al. (2012) found 78 that groundwater extraction was the predominant driver of SWI in one coastal aquifer in 79 Monterey, California, compared with the effect of SLR. Using the coastal aquifer at the Western 80 Baltic Sea as an example, Rasmussen et al. (2013) found that the SWI in flux-controlled aquifers 81 82 is more sensitive to recharge than SLR. Likewise, Ataie-Ashtiani et al. (2013) found that surface inundation may induce significantly more extensive SWI than SLR. Each of these studies 83 84 provided insights into understanding SWI under SLR in the temporal scales of decades and 85 centuries, however, none of them considered the effect of coastal landscape topographic change, which may have a significant impact on SWI by altering flow paths and residence time. 86

87 The evolution of coastal marsh landscape topography under SLR has been explored 88 extensively by the coastal geomorphologic community. Many studies have predicted coastal

marsh evolution as a function of sediment erosion and deposition and found that coastal marshes 89 are not static but dynamic landscapes. Coastal marshes are very likely to keep pace with the rates 90 91 of SLR at decadal to century scales due to the net sedimentation on the marshlands (e.g., Best et al., 2018; D'Alpaos et al., 2007; Kirwan, Temmerman, et al., 2016; Kirwan, Walters, et al., 92 2016; Kirwan & Murray, 2007; Kirwan & Temmerman, 2009; Mariotti & Fagherazzi, 2010; 93 94 Zhang et al., 2020). With the changes in marsh geomorphology, the topographic and hydraulic gradient among the land, river, and ocean will change, which will alter the seawater flow path 95 and storage (Winn et al., 2006). However, so far, there is a critical knowledge gap on how the 96 97 evolution of coastal landscape may affect the SWI under SLR. This could severely limit our capability to estimate the vulnerability of coastal aquifer to SWI under SLR. To fill this 98 knowledge gap, in this study, we used synthetic coastal marsh transects to examine the effect of 99 coastal marsh evolution on coastal SWI under SLR. We simulated coastal marsh evolution and 100 SWI on the coastal marsh transects under different rates of SLR over 100 years by using a 101 102 coastal marsh evolution model and a density-dependent solute transport model in the Advanced Terrestrial Simulator (ATS) (Coon et al., 2016). We evaluated the impacts of coastal marsh 103 evolution on SWI by comparing the surface and subsurface hydrologic characteristics between 104 105 the SWI simulations with and without considering coastal marsh evolution. The surface hydrologic characteristics include seawater propagation, inflow, surface salt concentration, and 106 surface seawater penetration. The subsurface hydrologic characteristics include subsurface water 107 108 salinity distribution, seawater inflow, and the displacement of the freshwater-saltwater interaction. We hypothesized that coastal marsh evolution cannot be ignored when evaluating 109 110 coastal SWI under SLR at a decadal or century scale because marsh evolution may significantly 111 affect the surface seawater inflow rate and the surface water residence time, therefore change

- 112 SWI. The insights gained from this study can help improve our understanding of the
- 113 vulnerability of coastal freshwater systems to SWI.

114	In this paper. we first introduce the numerical model, experiment and scenario designs,
115	and evaluation metrics in Section 2, after which we present and analyze the results from the
116	numerical experiments for marsh evolution and SWI in Section 3. Lastly, Section 4 discusses the
117	implication of this study for understanding SWI under SLR from a coupled hydro-eco-
118	geomorphologic framework, its representativeness, uncertainties, and outlined future work,
119	followed by conclusions in Section 5.
120	2 Matarials and Mathads
120	

121 2.1 Mathematical models

122 This study used the Advanced Terrestrial Simulator (ATS) (Coon et al., 2016) to simulate 123 SWI and coastal marsh evolution under SLR conditions. ATS is a multi-process high-124 performance computing simulator with process kernels (PKs) for surface and subsurface 125 hydrology, energy balance, thermal dynamics, sediment transport, solute transport, coastal marsh 126 evolution, and marsh vegetation dynamics. Here we used some ATS PKs to configure a saltwater 127 intrusion model and a coastal marsh evolution model.

128 2.1.1 The configuration of coastal saltwater intrusion in ATS

129 We used the SWI configuration in ATS (hereinafter referred to as SWI model) to

130 simulate the salinity change of surface and subsurface water by coupling surface and subsurface

- 131 hydrologic processes with density-dependent solute transport processes. The integrated
- 132 hydrologic processes include a two-dimensional (2-D) diffusive-wave approximation of surface
- 133 flows (Vreugdenhil, 1994) and a variably saturated three-dimensional (3-D) Richards equation

(Richards, 1931) for subsurface flows. The governing system of PDEs for surface and subsurface
fluid mass balance is as follows,

136
$$\begin{cases} \frac{\partial d\rho_1}{\partial t} = -\nabla_1(\rho_1 d\mathbf{u}) - I_w \\ \frac{\partial \theta \sigma \rho_2}{\partial t} = -\nabla_2(\rho_2 q) + I_w \end{cases}$$
(1)

137 where $\rho_1(C_1)$ and $\rho_2(C_2)$ are the surface and subsurface fluid density (kg/m³) which are 138 functions of surface and subsurface salt concentrations (dimensionless), C_1 and C_2 , 139 respectively; *d* is the surface water depth (m); θ is the subsurface water content; σ is the soil 140 porosity; u is the depth-averaged surface flow velocity (m/s) which is, according to the

141 diffusive wave approximation, defined as follows:

142
$$u = -\frac{d^{\frac{2}{3}}}{n(|\nabla(z+d)|)^{\frac{1}{2}}}\nabla(z+d)$$
(2)

143 where z is surface elevation (m); n is the Manning's coefficient of roughness $(s/m^{1/3})$; q in 144 Eq. 1 is the subsurface water flow velocity (m/s) defined by Darcy law:

$$q = -\frac{k_r K}{\mu} \nabla h \tag{3}$$

where k_r and K are a relative (dimensionless) and absolute permeabilities (m²), respectively, μ is fluid dynamic viscosity (kg/m/s) and h is subsurface hydraulic head (m). I_w is the mass source/sink term (kg/s m²) representing the infiltration or exfiltration (positive in the downward direction)

150 In Eq. 1, the density is a linear function of water concentration (Simmons et al., 2001):

151
$$\rho = \rho_0 + a(C - C_0),$$
 (4)

- 152 where ρ is the fluid density for surface water or subsurface water (kg/m³); *C* is the fluid salt
- 153 concentration for surface and subsurface water (dimensionless); ρ_0 is the fluid density at a base

154 concentration, C_0 ; *a* is a constant coefficient of density variability.

155 The generic form of saltwater concentration is calculated based on the salt mass balance 156 equation following *Herbert et al.* (1988) and *Simmons et al.* (2001)

157
$$\begin{cases} \frac{\partial (d\rho_1 C_1)}{\partial t} = -\nabla_1 (d\rho_1 u C_1) + \nabla_1 [d\rho_1 (D_0 + D) \cdot \nabla_1 C_1] - I_{salt} \\ \frac{\partial (\theta \sigma \rho_2 C_2)}{\partial t} = -\nabla_2 (\theta \sigma \rho_2 q C_2) + \nabla_2 [\theta \sigma \rho_2 (D_0 + D) \cdot \nabla_2 C_2] + I_{salt} \end{cases}$$
(5)

where C_1 and C_2 are the salt concentration of surface and subsurface water (dimensionless), 158 respectively. D_0 is the molecular diffusivity (m²/s). **D** is the transverse and longitudinal 159 dispersivities; Likewise, I_{salt} is the mass source/sink term (kg/s m²) representing the infiltration 160 or exfiltration (positive in the downward direction). The capability of the SWI model in 161 capturing saltwater intrusion through surface and subsurface flow under tidal conditions was well 162 validated by comparing the model simulation with a lab experiment of SWI with tidal variation 163 by *Kuan et al.* (2019) (see the validation results in the supplementary information Text S1 and 164 Figs. S1 and S2). 165

166 2.1.2 The configuration of coastal marsh evolution in ATS

For the marsh evolution modeling, we used the coastal marsh evolution model (hereinafter referred to as Sed model) configured by the 2-D surface flow, sediment transport and marsh evolution PKs in ATS. The Sed model tracks the change of marsh surface elevation as a function of sediment erosion, sediment settling, sediment trapping by vegetation, and vegetation organic matter production. Namely,

$$\frac{dz}{dt} = \frac{1}{1-\sigma} \left(D_s + D_t + D_{org} - E \right) \tag{6}$$

where z is the surface elevation (m); t is the time (s); σ is the porosity of bed sediment; D_s is the inorganic sediment settling rate (m/s); D_t is the inorganic sediment trapping rate due to the effect of vegetation canopy (m/s); D_{org} is the organic matter production rate (m/s); *E* represents local sediment erosion rate (m/s).

The Sed model follows the forms of sediment erosion and deposition in D'Alpaos et al. 177 (2007), but improved the representation of surface hydrodynamics, instead of using an 178 equilibrium assumption for surface hydrodynamics. Specifically, in the Sed model, the surface 179 180 hydrodynamics due to tide and SLR is represented by a depth-averaged diffusive-wave scheme (Eq. 2) considering the spatial and temporal variations of water propagation landward. Sediment 181 erosion (E in Eq. 6) is estimated as a linear function of dynamic bed shear stress depending on 182 surface water flow velocity. Erosion occurs when the bed shear stress (τ_0) due to water flow is 183 184 greater than the critical shear stress (τ_e) for erosion. Namely,

185
$$E = \begin{cases} \alpha \left(\frac{\tau_0}{\tau_e} - 1\right) & \text{if } \tau_0 > \tau_e \\ 0 & \text{if } \tau_0 < \tau_e \end{cases}$$
(7)

186 where α is the erosion coefficient. Likewise, sediment settling (D_s in Eq. 6) describes the process 187 that particulates settle to the bottom of a liquid and form sediment due to gravity, which is also 188 assumed as a linear function of dynamic bed shear stress. Sediment settling occurs when the bed 189 shear stress (τ_0) is smaller than the critical shear stress (τ_d) for settling, viz

190
$$D_{s} = \begin{cases} w_{s}C\left(1 - \frac{\tau_{0}}{\tau_{d}}\right) & \text{if } \tau_{0} < \tau_{d} \\ 0 & \text{if } \tau_{0} > \tau_{d} \end{cases}$$
(8)

191 where w_s is the settling velocity (m/s); C is the suspended sediment concentration

192 (dimensionless). Sediment trapping by vegetation (D_t in Eq. 6) is given by

$$D_t = CU\epsilon d_s n_s \min[h_s, h_w]$$
(9)

where D_t is a function of water flow velocity (*U*), a capture efficiency of vegetation stems (ϵ), water depth (h_w), and several vegetation characteristics, such as plant stem diameter (d_s), stem density (n_s), and vegetation height (h_s). The vegetation properties are determined by vegetation biomass, which is assumed as a linear function of marsh surface elevation relative to the mean highest tide level. Also, the vegetation organic matter production (D_{org} in Eq. 6) is a linear

199 function of vegetation biomass, viz

200

$$D_o = K_b \frac{B}{B_{max}},\tag{10}$$

where K_b is the maximum production rate of belowground organic material [m/s]; *B* is the aboveground plant dry biomass at the current time [g/m²]; and B_{max} is the maximum vegetation biomass [g/m²]. In this study, we assume that the aboveground biomass of the salt-tolerant marsh vegetation linearly increases with the inundation level (Morris et al., 2002). The details of the questions are referred to D'Alpaos et al. (2007) and Zhang et al. (2020).

An offline coupling approach was used in this study to integrate the Sed model and the SWI model (an online coupling scheme is under development). Specifically, we 1) first simulate the coastal marsh evolution by using the Sed model and 2) use the simulated future surface elevation from step (1) as the initial topographic and morphologic condition to simulate future SWI until equilibrium by using the SWI model. 211 2.2 Numerical experiment design

To investigate the effect of marsh evolution on SWI, we designed two 2-D synthetic 212 coastal marsh transects that include a 1 m wide coastal marshland (2000 m long) and upland 213 region (1000 m long) with different slopes to represent real coastal marshes at the Atlantic coast, 214 particularly the Delaware Bay and Chesapeake Bay areas (see Fig. 1). We assumed that the 215 coastal marshland is covered by salt-tolerant marsh species, such as Spartina (Morris et al., 216 217 2002). The slope of the coastal marsh (1:2500) represents an averaged slope of the coastal marsh transects in Delaware Bay, USA, measured by this study. The upland slopes of 1:1000 and 1:250 218 in this area are used to represent different upland controls on surface-subsurface water 219 220 propagation and sediment transport (Fagherazzi et al., 2019). We assumed a homogeneous landscape with homogeneous surface roughness and homogeneous soil properties, such as 221 surface Manning's coefficient, soil porosity, soil hydraulic conductivity, and van Genuchten 222 water retention parameters (see the details of parameter values in Table 1). The soil thickness at 223 the upland and the ocean boundaries are 20 m and 18.2 m, respectively, based on the 224 measurements in Sanford et al. (2012). 225

226 The hydrodynamics are driven by tides and SLR at the ocean boundary. Initially, the 227 mean sea level (MSL) is equal to 0 m, the elevation of the current coastal marsh near the ocean boundary (see Fig. 1). We adopted two widely-used future global mean SLR scenarios based on 228 the Representative Concentration Pathways (RCP) 4.5 and RCP 8.5 scenarios in Phase 5 of the 229 230 Coupled Model Intercomparison Project (CMIP5) (Spencer et al., 2016) for our numerical experiments, including (1) a relatively low SLR rate of 0.5 m/100-yr (Da Lio et al., 2013; Ganju 231 et al., 2020; Kirwan & Temmerman, 2009; Sefelnasr & Sherif, 2014; Spencer et al., 2016) and 232 (2) a relatively high SLR rate of 1 m/100-yr (Carretero et al., 2013; Langevin & Zygnerski, 233

2013; Lu et al., 2015; Michael et al., 2013; Sefelnasr & Sherif, 2014; Watson et al., 2010; Yang 234 et al., 2015). We applied a sinusoidal semi-diurnal tide with a tidal range of 1.6 m based on the 235 National Oceanic and Atmospheric Administration (NOAA) tide and current observation at the 236 Delaware Bay, USA (Cape May Station, station ID: 8536110). 237 Freshwater supply in the coastal aquifer is controlled by the hydrostatic groundwater 238 table (GWT) at the upland boundary (the left boundary of the domain in Fig. 1). This study used 239 240 three upland GWT scenarios to represent (1) a present-day upland GWT condition (1.3 m above the initial MSL), (2) a future GWT condition with increased groundwater extraction and/or a 241 drier climate (0.9 m above the initial MSL), and (3) a future GWT condition with a wetter 242 243 climate (1.8 m above the initial MSL), respectively. To simplify the control factors and present a more focused study on the effect of marsh evolution on SWI, we did not include the effect of 244 rainfall recharge in this study, which will be explored in future work. We used a head-controlled 245 upland GWT boundary condition, which may predict more SWI than the flux-controlled upland 246 boundary condition for unconfined aquifers (Werner & Simmons, 2009). Therefore, our 247 scenarios may represent a relatively more aggressive prediction of SWI under SLR. All scenarios 248 are summarized in Table 2. 249

The 2-D model domain in Fig. 1 was decomposed into logically structured mesh with a horizontal resolution of ~5 m and vertical resolution of ~1 m (20 layers). We simulated 100-yr marsh evolution, and the SWI experiments were simulated until the freshwater-saltwater interface reached an equilibrium state.





Figure 1. Sketches of the 2-D model domain. The surface elevation at the ocean boundary (right side) is 0 m. The elevation at the upland boundary is 1.8 m for the milder upslope domain and 4.8 m for the steeper upslope domain. MHTL stands for the mean highest tide level and MSL is mean sea level. The present-day MSL is at 0 m level. h at the left side of the domain indicates the upland groundwater level. Three upland groundwater level scenarios are used: h = 1.3 m (present-day level), h = 1.8 (reflecting future wetter climate), and h = 0.9 m (reflecting future drier climate or more groundwater extraction)

260

261

Table 1. Key parameter values used in the numerical experiments

	Parameters	Values	References
	Porosity	0.46	(Kuan et al., 2012)
	Hydraulic conductivity (m/day)	10	(Kuan et al., 2012)
	Manning's n (s m ^{-1/3})	0.06	(Arcement & Schneider, 1989)
	Diffusion coefficient (m ² s ⁻¹)	1.00E-09	(Kuan et al., 2012)
	Longitudinal dispersivity coefficient (m)	0.002	(Kuan et al., 2019)
SWI-related parameters	Transversal dispersivity coefficient (m)	0.0004	(Kuan et al., 2019)
	van Genuchten α for water retention (m ⁻¹)	0.48	(Benson et al., 2014)
	van Genuchten <i>n</i> for water retention	2.54	(Benson et al., 2014)
	Residual saturation	0.1	(Kuan et al., 2012)
	Saltwater concentration in the ocean (kg salt per kg seawater)	0.0357	(Michael et al., 2013)
Marsh	Sediment concentration in the	50	(Kirwan, Walters, et al., 2016)

evolution-	ocean (mg/I	L)			
parameters	Erosion coefficient	$t\left(\frac{kg}{m^2sP_a}\right)$	1.12E-04	(D'Alpaos et	al., 2007)
	Critical shear stress for erosion (P_a)		0.4	(Thompson et al., 2004)	
	Critical shear stre deposition (F	ess for P_a)	0.1	(Parchure Trimbak M. & Mehta Ashish J., 1985)	
	Sediment settling ve	elocity $(\frac{m}{s})$	1.00E-04	(Riazi & Türker, 2019)	
	Belowground or production $(\frac{1}{2})$	rganic m yr)	0.003	(Morris et a	1., 2016)
	Table 2. The numerical experin	nent cases with dif Milder u	ferent upland slope, s	SLR, and GWT scena	rios
	Table 2. The numerical experin	nent cases with dif Milder u Higher SLR	ferent upland slope, s pland slope Lower SLR	SLR, and GWT scena Steeper up Higher SLR	orios Dland slope Lower SLR
Medium uj (prese	Table 2. The numerical experim pland GWT (h=1.3 m) ent-day scenario)	nent cases with di <u>f</u> Milder u Higher SLR Case 1	ferent upland slope, s pland slope Lower SLR Case 2	SLR, and GWT scena Steeper up Higher SLR Case 3	oland slope Lower SLR Case 4
Medium up (prese High upl (wetter	Table 2. The numerical experim pland GWT (h=1.3 m) ent-day scenario) and GWT (h=1.8 m) c climate scenario)	nent cases with dif Milder u Higher SLR Case 1 Case 5	ferent upland slope, s pland slope Lower SLR Case 2 Case 6	SLR, and GWT scena Steeper up Higher SLR Case 3 Case 7	oland slope Lower SLR Case 4 Case 8
Medium up (press High upl (wetter Low upla (drier c	pland GWT (h=1.3 m) ent-day scenario) and GWT (h=1.8 m) c climate scenario) and GWT (h=0.9 m) limate and/or more	ment cases with dif Milder u Higher SLR Case 1 Case 5 Case 9	ferent upland slope, 2 pland slope Lower SLR Case 2 Case 6 Case 10	SLR, and GWT scena Steeper up Higher SLR Case 3 Case 7 Case 11	oland slope Lower SLI Case 4 Case 8 Case 12

We evaluated the effect of marsh evolution on SWI by comparing the SWI simulation with and without marsh evolution. Specifically, we created two groups of experiments. Each

268 group consisted of all 12 experimental cases listed in Table. 2. The first group of experiments

was based on the current marsh topography as illustrated in Fig. 1. SWI was simulated under

270 different rates of SLR and different levels of the upland GWT in the future 100 years without

271 considering coastal marsh evolution. In contrast, the second group conducted the same SWI

simulations as the first group experiments but used the marsh topography from the marsh
evolution simulations in the future 100 years as the topographic condition. We examined the
surface and subsurface freshwater and saltwater changes in the two groups of numerical
experiments. Specifically, we analyzed the changes in surface seawater propagation, inflow,
concentration, and infiltration and subsurface water salinity, seawater inflow, and the
displacement of the freshwater-seawater interface.

278 **3 Results**

279 3.1 Coastal marsh evolution driven by tidal and SLR forcing

With the SLR rates of 0. 5 m/100-yr and 1 m/100-yr, the future MSL increases to 0.5 m 280 and 1 m in 100 years, respectively. Accordingly, the mean highest tide levels (MHTL, equal to 281 MSL + tidal amplitude) rise from 0.8 m to 1.3 m and 1.8 m, respectively. Driven by the future 282 283 SLR, our modeling results showed that the future marsh elevation rises substantially with SLR and with a larger increase near the ocean boundary and a smaller increase for the inland marsh 284 (Figs. 2a and b) due to a gradient in sedimentation rates (Fig. 3). Therefore, a topographic 285 depression forms in the middle of the marshland. The future elevations near the ocean boundary 286 287 are close to the future MHTLs under both the higher and lower SLR rates. Correspondingly, the vegetation co-evolves with the topographic change and future inundation condition. The salt-288 tolerant vegetation biomass is modeled to increases with inundation level, which results in higher 289 vegetation biomass in the middle of the domain and lower vegetation biomass at the ocean and 290 291 upland sides (Figs. 2c and d). Relative to the initial marshland (the gray dashed lines in Figs. 2c and d from 1000m to 3000m from the upland boundary), there was a landward expansion of 292 293 marsh vegetation that varied with SLR and upland slopes. For example, the marsh vegetation

covers the entire upland region in the case with the milder slope and higher SLR rate (the light 294 green line in Fig. 2c) because the future MHTL (1.8 m) is the same as the elevation of the upland 295 boundary leading to an inundation condition favorable for vegetation growth. For the cases with 296 the steeper upland slope, a large portion of the upland areas is still higher than the MHTL, 297 therefore no vegetation presents on the upland areas (Fig. 2d). 298 Among sedimentation rates across the domain, the sediment settling rate is the greatest 299 300 near the ocean boundary (the light blue lines in Fig. 3) because of a higher sediment concentration near the ocean boundary. Moving landward, the sediment settling rate decreases 301 due to a lower sediment concentration, but the vegetation organic matter production rate 302 303 increases because of higher vegetation biomass, which contributes more to the sedimentation of

the inland marsh (the purple lines in Fig. 3). Vegetation sediment trapping is more significant 304 near the ocean boundary where the sediment concentration and flow velocity are higher. Moving 305 landward, due to the decrease of sediment concentration and flow velocity, the trapping effect 306 gradually vanishes (the orange lines in Fig. 3). Due to the vegetation effect on reducing water 307 flow velocity, sediment erosion is only observed at the beginning of the simulation when the 308 future SLR is applied to the domain boundary (not shown). Later in the simulation, the surface 309 310 elevation increases due to a higher sedimentation rate. Therefore, the seawater inflow velocity 311 decreases because of a decrease of the hydraulic gradient between the land and the ocean, and 312 the erosion rate drops to zero.



Figure 2. The distribution of future elevation and vegetation biomass of the domain with a milder upland slope (a and c) and a
 steeper upland slope (b and d). The gray dashed lines indicate the initial surface elevation and vegetation biomass distribution,
 respectively. MHTL stands for the mean highest tide level. RSLR is the rate of sea level rise.



318 Figure 3. The spatial distribution of the sediment fluxes at the end of the 100-year simulation. The plots with various colors 319 represent different fluxes in different scenarios.

320 3.2 SWI with and without marsh evolution

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321 3.2.1 The influence of marsh evolution on coastal hydrodynamics

After predicting the topographic change of the coastal marsh landscapes 100 years in the

future, we compared the SWI under SLR with and without considering the marsh evolution. For

324 both SLR rates, the upland GWT is held fixed at 1.3 m above the initial MSL (0 m), but the future MSL increases (see details of the MSL rise in Subsection 3.1). Without considering marsh 325 topographic change in the future, the rising sea level increases the hydraulic gradient between the 326 ocean and the marshland, thus more saltwater was predicted to flow onto the marshland with a 327 relatively larger maximum inflow rate ($\sim 0.25 \text{ m}^3/\text{s}$) under the higher SLR scenario and a 328 relatively lower rate ($\sim 0.16 \text{ m}^3/\text{s}$) under the lower SLR scenario (the gray bars in Fig. 4). 329 However, considering marsh evolution, the marsh surface elevation increases with SLR, 330 especially for marshland near the ocean boundary (e.g., Fig. 2a and b), which decreases the 331 332 hydraulic gradient between the ocean and the marshland. Thus, the saltwater maximum inflows are predicted to be ~0.009 m³/s for the higher SLR rate scenario and ~0.0048 m³/s for the lower 333 SLR rate scenario (the black bars in Fig. 4). These rates are two orders of magnitude smaller than 334 the maximum inflow rates without considering marsh evolution. 335



Figure 4. The maximum seawater inflows under future sea level for the simulations with and without considering coastal marsh
 evolution. The gray and black bars indicate the simulated seawater inflow without and with considering marsh evolution,
 respectively.

The marsh topographic change also affects surface seawater propagation, ponding water 340 depth, surface water residence time, and saltwater concentration on the marsh surface. For 341 example, Figure 5 shows the surface water propagation and saltwater concentration for the cases 342 without considering marsh evolution (the top four plots with gray backgrounds) and with marsh 343 evolution (the bottom four plots with white backgrounds). Without considering marsh evolution, 344 345 seawater propagates landward during the high tides and causes a surface inundation with a maximum inundation depth higher than 1.3 m near the ocean boundary (the upper dashed lines in 346 Fig. 5a, b, c, and d). During the high tides, the saltwater concentration in the ponding water is 347 almost the same as the concentration in the ocean (3.5%; the blue dashed lines in Fig. 5a, b, c, 348 and d), except for the upland regions, where the subsurface freshwater exfiltrates to the surface 349 and dilutes the surface saltwater. During the low tides, the surface ponding water flows out from 350 the marsh domain (the lower dashed lines in Fig. 5a, b, c, and d). Therefore, the residence time 351 for surface saltwater is tightly controlled by the tidal frequency (~12 h). 352

In contrast, considering marsh evolution, the depression zone in the middle of the 353 marshland significantly changes the surface water propagation. Saltwater flows onto the 354 marshland during the high tides with a much smaller inflow rate as illustrated in Fig. 4, and then 355 356 the saltwater gradually accumulates in the depression zone without flowing out from the domain 357 during the low tides, which largely increases the residence time of saltwater (the black dashed lines in Fig. 5e, f, g, and h). A large portion of the surface ponded saltwater exits the marsh 358 surface only through infiltration. Meanwhile, freshwater flows into the depression zone through 359 360 subsurface freshwater exfiltration. In the cases with a higher SLR rate (Figs. 5e and g), the upland GWT (1.3 m) is lower than the future MHTL (1.8 m). For most of the time during a tidal 361 cycle, fresh groundwater cannot easily flow into the marsh aquifer. Thus, at equilibrium, the 362

surface ponding water consists almost of saltwater with very little freshwater, and the surface 363 saltwater concentration is almost equal to the concentration in the ocean for the entire surface 364 ponding water (the blue lines in Figs. 5e and g). However, in the cases with a lower SLR rate 365 (Figs. 5f and h), the upland GWT (1.3 m) is at the same level as the future MHTL (1.3 m). Fresh 366 groundwater can be more easily flow into the marsh aquifer and exfiltrate to the surface when the 367 instantaneous sea level is below the MHTL during a tidal cycle. Thus, a larger amount of fresh 368 groundwater accumulates in the depression zone to dilute saltwater inflow. Therefore, we only 369 observed an increase of surface saltwater concentration near the ocean boundary (the blue lines 370 371 in Figs. 5f and h).





Figure 5. The surface water propagation and associated surface water concentration in the simulations with and without the considerations of marsh topographic change. (a), (b), (c), and (d) with the gray background are the cases without considering marsh evolution. (e), (f), (g), and (h)are the cases considering marsh evolution. The black solid lines are the surface elevation.
The black dashed lines indicate the maximum and minimum surface water propagation under the high and low tides. The blue lines indicate the distribution of the saltwater concentration under the maximum surface water propagation conditions. HSLR and LSLR stand for the higher and lower SLR rate scenarios, respectively.

379 3.2.2 Subsurface salinity distribution

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Under the present-day condition (upland GWT=1.3m) without SLR, Figs. 6a and b show
the equilibrium salinity distribution in the aquifer with milder and steeper upland slopes,
respectively. The toes and heads of the freshwater-saltwater interfaces (the interface between the
blue and the cyan colors) are at ~1250 m and ~1500 m from the upland boundary, respectively.
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Starting from this present-day equilibrium condition, we simulated the SWI under SLR. For the 384 cases without considering marsh evolution (Figs. 6c, d, e, and f), we predicted an increase of 385 SWI with the increase of SLR magnitudes. Both the toes and heads of the freshwater-saltwater 386 interfaces are observed to move landward 500m to 1000m under the lower and higher SLR rates. 387 The increases in SWI are attributed to the SLR-induced surface saltwater infiltration and 388 subsurface lateral saltwater inflow. With the increased sea level, more surface seawater infiltrates 389 to the subsurface aquifer. Meanwhile, more seawater flows into the aquifer directly through the 390 subsurface lateral flow due to the increased hydraulic gradient between the sea level and the 391 392 inland water table. We analyzed the changes in surface saltwater infiltration and subsurface lateral inflows from the beginning of the simulation to the end when the subsurface salinity 393 distribution reached an equilibrium (see Fig. S3 and Text S2). In general, the surface saltwater 394 infiltration had a much larger contribution to the subsurface SWI than the subsurface lateral 395 inflow (see the blue solid lines in Fig. S3) due to the high inundation level during the high tides 396 as illustrated in Fig. 5. 397

When marsh evolution was included, the depression zone formed by the marsh evolution 398 process significantly increases surface water residence time. This occurs even though the 399 simulated surface saltwater inflow and surface inundation depth at the ocean boundary 400 401 (illustrated in Figs. 4 and 5) are much lower than the simulations without marsh evolution, Therefore, the surface ponding water accumulates seawater and prolongs the time of surface 402 saltwater infiltration in contributing to the landward movement of the freshwater-saltwater 403 404 interface. Thus, under the higher SLR rate, the equilibrium freshwater-saltwater interfaces almost reach the upland boundary (Figs. 6g and h), similar to the simulations without marsh evolution. 405 However, under the lower SLR rate, more freshwater can flow into the domain due to a lower 406

hydraulic gradient between the sea level (0.5 m) and the upland GWT (1.3 m), there is no 407 significant increase in saltwater concentration in the surface ponding water (e.g., the blue lines in 408 Figs. 5f and h). Therefore, even though the surface saltwater infiltration still dominates the SWI 409 (the blue dashed lines in Figs. 3Sb and d), its rate is smaller than the rate under the higher SLR 410 rate and is located only near the ocean boundary (Figs. 5f and h). We see a distinct subsurface 411 412 saltwater distribution at equilibrium under the lower SLR rate: the toe of the freshwater-saltwater interface moves towards the upland boundary, but the freshwater-saltwater interface near the 413 marsh surface moves toward the ocean boundary. More freshwater occupies the upper part of the 414 aquifer in the middle of the marshland (Figs. 5i and j). 415



416



420 3.3 SWI under different upland GWT level

421 We also investigated the future SWI in response to the changes in upland GWT 422 conditions caused by a wetter or drier future climate or intensified groundwater extraction (cases 423 5 to 12 in Table 2). With the same model settings and same SLR rate scenarios, a higher upland GWT (1.8 m above the initial MSL) under a wetter future climate causes a larger hydraulic 424 gradient towards the ocean, thereby more freshwater flows towards the ocean. Therefore, the 425 simulations predict a lower SWI for all cases under the high upland GWT (Fig. 7), compared 426 with the cases under the present-day upland GWT (Fig. 6). Specifically, the cases without marsh 427 428 evolution show slightly less SWI (Figs. 7c, d, e, and f), compared with the cases under the present-day GWT (Figs. 6c, d, e, and f). However, the cases with the marsh evolution and higher 429 SLR rate show a larger decrease of SWI (Figs. 7g and h), compared with the corresponding cases 430 431 under the present-day upland GWT (Figs. 6g and h). This is because the sufficient fresh groundwater supply from the upland pushes the saltwater back toward the ocean and dilutes the 432 surface ponding saltwater. 433

However, with a lower upland GWT (0.9 m above the initial MSL) and under a drier 434 future climate and/or increased groundwater extraction, the marshland creates a larger hydraulic 435 gradient towards the land, compared with the present-day condition, causing more saltwater 436 flows onto the marsh surface and into the aquifer. Therefore, more SWIs are observed in all 437 simulation cases with the lower future upland GWT (Fig. 8). In particular, we see a larger 438 439 increase of SWI for the cases with the marsh evolution and under the lower SLR rate (Figs. 8i 440 and j), compared with the corresponding cases under the present-day upland GWT in Figs. 6i and j. This is because the upland groundwater supply is not sufficient to counteract the saltwater 441 inflow. 442



443 444 445

Figure 7. The distribution of subsurface saltwater concentration under the present-day sea level (a and b), future sea level (c,d, e, and f), and future sea level and topographic change (g, h, i, and j). All the simulation results are under the upland GWT of 1.8 446 m. The left and right columns are the simulations with the milder and steeper upland slopes, respectively.





450

Figure 8. The distribution of subsurface saltwater concentration under the present-day sea level (a and b), future sea level (c,d, e, and f), and future sea level and topographic change (g, h, i, and j). All the simulation results are under the upland GWT of 0.9 m. The left and right columns are the simulations with the milder and steeper upland slopes, respectively.

452 **4 Discussion**

453

4.1 Effect of coastal marsh evolution on coastal SWI

This study used a representative coastal marsh landscape to investigate the effect of 454 coastal landscape evolution on coastal SWI under SLR. Our marsh evolution simulations 455 confirmed that coastal marsh landscapes are dynamic, and the evolution of coastal marsh is very 456 spatially heterogeneous in response to SLR (Fig. 2). We found that the future marsh topography 457 458 has an important influence on coastal SWI, especially on the saltwater surface inflow, surface 459 saltwater residence time, and saltwater infiltration. Firstly, the future evolved marsh topography significantly reduces the seawater inflow on the marsh surface because the increased marsh 460 elevation near the ocean boundary reduces the hydraulic gradient between the marshland and the 461 462 ocean (e.g., the saltwater inflows in Fig. 3). The predicted seawater inflow can be up to two 463 orders of magnitude smaller than the inflow without considering marsh evolution implying that the seawater inflow may be overestimated in the studies that do not consider marsh evolution. 464 Secondly, the depression zone formed during the marsh evolution processes can accumulate 465 466 seawater, which significantly increases the saltwater residence time on the marsh surface and prolongs infiltration time. Therefore, with the evolved marsh landscape, the hydrologic regime 467 may be very different from that without considering marsh evolution as we demonstrated above 468 (e.g., Figs. 4, 5, 6, and S3). In addition, our numerical experiments show that the surface 469 saltwater infiltration contributes more to the SWI than subsurface saltwater lateral inflow 470 through ocean boundary. This finding is consistent with several previous studies, such as Ataie-471 Ashtiani et al (2013) who discussed the predominant effect of surface inundation on subsurface 472 SWI. 473

474 4.2 The sensitivity of future SWI to upland groundwater supply

Our simulations under future SLR show that the upland GWT effect is more significant in 475 the cases with marsh evolution because the exfiltrated freshwater can stay longer in the surface 476 depression zone so that it has a larger influence on counteracting with surface saltwater inflow. 477 For example, with the different upland GWTs, the cases with marsh evolution show larger 478 variations in the displacements of the freshwater-saltwater interface (e.g., Figs 6, 7, and 8). 479 Quantitatively, by compiling all the cases with the different upland GWTs in Figs. 6, 7, and 8, 480 we found that the difference between the upland GWT and MHTL (GWT-MHTL) can be a good 481 metric to understand the effect of marsh evolution on SWI under future SLR. Specifically, Figure 482 483 9 shows the mean distance between the freshwater-saltwater interface and the ocean boundary as a function of GWT-MHTL based on the cases in Figs. 6, 7, and 8. When GWT-MHTL is greater 484 than zero (a hydraulic gradient towards the ocean), we found that the cases considering marsh 485 evolution predict the shortest SWI distance (the orange dots and circles in Fig. 9). However, if 486 GWT-MHTL is less than zero (a hydraulic gradient towards the inland), we observe a slight 487 difference of SWI between the cases with and without marsh evolution. Therefore, as marshes 488 evolve, the system may become more sensitive to the upland GWT variation, highlighting the 489 490 importance of protecting upland groundwater resources to prevent intensified SWI in the future.



498 4.3 Marsh evolution and its representativeness

Under the external drivers of SLR and tidal current, our marsh evolution simulations 499 show that coastal marsh elevation is very likely to increase with a higher increase near the ocean 500 boundary and a smaller increase landward due to a gradient in inorganic and organic 501 502 sedimentations. Although the future marsh topographies predicted by our study are the results of the combined effect of the specific tidal amplitude, SLR rates, sediment concentration in the 503 ocean, tidal period, and sediment diffusivity, the predicted topographic features (higher increase 504 505 at ocean boundary and lower landward) are consistent with many previous marsh evolution studies (e.g., D'Alpaos et al., 2007; Kirwan, Walters, et al., 2016; Kirwan & Temmerman, 2009; 506 Zhang et al., 2020). The formulations used in the Sed model to represent the dominant processes 507 were selected from broadly-used sedimentation, erosion, and vegetation dynamic equations. The 508 parameters used in this study were established in the literature from field measurements 509 (Fagherazzi et al., 2013; Kirwan, Temmerman, et al., 2016; Morris et al., 2002; Mudd et al., 510

511 2004) and were within the parameter range in the parametric sensitivity study by Zhang et al.

512 (2020). Thus, our simulation results of marsh evolution are representative of the future

513 marshland topographic change of some real-world coastal marshes under SLR.

514 4.4 The influence of future SWI on vegetation dynamics

Subsurface water salinity has a direct impact on vegetation growth, species richness, 515 species distribution, and migration (Antonellini & Mollema, 2010; Silvestri & Marani, 2004). In 516 517 previous studies that did not consider marsh evolution, the subsurface SWI was predicted to 518 occupy a larger area of the coastal aquifers (e.g., Kuan et al., 2012; Michael et al., 2013), which exert larger stress on vegetation growth. However, our experiments show that the subsurface 519 salinity may decrease, especially for the upper part of the aquifer, under the cases with a higher 520 521 upland GWT and lower SLR rate (e.g., Figs. 6i and j and Figs. 7g, h, i, and j). Therefore, the 522 surface ponding freshwater dilutes saltwater providing a favorable condition for salt-intolerant vegetation, which indicates that the vegetation species and distribution may be significantly 523 different from the previous studies without considering marshland evolution. 524

525

4.5 Uncertainties and future work

In this study, we chose to focus on certain elements, including sea level, tide, upland groundwater table, and topographic change, which we felt were critical to our analysis of the coastal eco-geomorphologic impact on SWI. However, several factors may affect the SWI prediction, such as the vegetation dynamic representation, types of upland groundwater boundary condition, precipitation, evaporation, waves, and the land-river-ocean interaction. Specifically, our marsh evolution simulation assumed a linear relationship between the *Spartina*-dominant vegetation biomass and the coastal inundation level based on the field observation by *Morris et*

al. (2002). However, there are also other schemes to represent the relationship between 533 vegetation biomass and inundation level for different marsh landscapes, such as the Spartina-534 nonlinear function (Mariotti & Fagherazzi, 2010) and mixed vegetation species linear function 535 (D'Alpaos et al., 2007). Zhang et al. (2020) evaluated the topographic outcomes from these three 536 vegetation schemes and found that all of the schemes predicted a higher elevation increase near 537 538 the ocean boundary and a lower increase landward. However, these schemes also showed differences in marsh elevation relief and unvegetated-vegetated ratio. Therefore, it is worth 539 540 exploring SWI under future topographic change with different vegetation dynamic 541 representations. Our ongoing development of a physically-based vegetation dynamic configuration on ATS will better link surface and subsurface water conditions, including salinity, 542 soil moisture, nutrient content, and inundation level, with vegetation growth. Therefore, the 543 newly developed model will better capture the vegetation dynamic features under different 544 subsurface salinity conditions, instead of only tracking the impact of surface inundation on 545 vegetation growth. 546

In addition, a constant GWT boundary condition is used in this study, which is suitable 547 for the coastal landscapes with a relatively small slope such as the slopes in our experiments 548 549 (Ketabchi et al., 2016). However, the freshwater supply from the upland can also be controlled 550 by continuous groundwater fluxes. Previous numerical and analytical modeling studies found that SLR-induced saltwater intrusion has a larger impact on head-controlled coastal aquifers 551 rather than flux-controlled coastal aquifer (Rasmussen et al., 2013; Werner et al., 2012; Werner 552 553 & Simmons, 2009). Therefore, our study made a more aggressive prediction of SWI under SLR 554 based on the head-controlled systems. The SWI in a flux-controlled system will be explored in future studies. 555

Precipitation and evaporation may also affect surface and subsurface saltwater transport 556 and distribution (Geng et al., 2016; Payne, 2010; Werner & Simmons, 2009). To conduct this 557 focused study, we reduced the complexity by limiting external drivers. We acknowledge that 558 precipitation and evaporation would play an important role in changing water salinity on land 559 surface and in the upper subsurface zone, yet they do not affect our conclusion on the role of 560 561 marsh evolution in controlling the saltwater inflow and surface water residence time. However, to have a more realistic prediction of SWI for a real-world coastal marshland, it is necessary to 562 take precipitation and evaporation into account in future studies. 563

Additionally, we did not consider erosion due to waves because marsh vegetation can 564 mitigate waves if the waves are caused by the regular tidal variation and wind speed (D'Alpaos 565 et al., 2007; Marani et al., 2007). However, under climatic extreme events, like hurricane and 566 storm surge, the large waves may not be effectively mitigated by the marsh vegetation, and 567 marshland erosion may occur and exceed the rate of sediment deposition. In this case, marsh 568 elevation may decrease, especially near the ocean boundary as predicted by Mariotti and 569 Fagherazzi (2010). The decrease of marsh elevation may increase surface saltwater inflow, 570 thereby stimulating SWI. Also, hurricanes may directly cause distinct surface and subsurface 571 572 water salinity distribution due to a dramatic increase in seawater level (Yu et al., 2016). 573 Although it is not the scope of this study, it is worth exploring the effect of marsh evolution on SWI under these extreme climatic events in future studies. 574

Last, this study used transects that do not allow for surface water drainage paths connecting the marshland and marsh drainage network, which may facilitate the drainage of the surface ponding water to channels. This will require a 2-D simulation that captures the complex topography of marshes with channels and drainage pathways. Also, the role of 3-D

hydrodynamics is not considered and worthy of additional future study as it would incorporate
baroclinic effects that can contribute to tidally-driven sedimentation.

581 **5 Conclusions**

582 In this study, we investigated the impact of coastal marsh evolution on SWI prediction under future SLR by using a physically-based coastal hydro-eco-geomorphologic model, ATS 583 (Advanced Terrestrial Simulator). Using a representative coastal marsh landscape, we first 584 585 predicted the marsh landscape change with different upland slopes and under two SLR scenarios. 586 We found that the coastal marsh landscape is not static but dynamic in response to SLR. The marsh elevation increases with the rising sea level due to the organic and inorganic 587 sedimentation and created a higher elevation near the ocean boundary and a depression zone in 588 589 the middle of the marshland. The marsh accretion is projected to cause a significant reduction of saltwater inflow at the ocean boundary because of the decrease of the hydraulic gradient between 590 the land and ocean. Also, the evolved topographic depression zone prolongs the residence time 591 of surface ponding water, which affects surface saltwater infiltration, therefore causes distinct 592 593 subsurface salinity distributions. With the evolved marsh landscape, we also tested the impact of different upland groundwater conditions on SWI under SLR, reflecting the impact of future 594 drier/wetter climate conditions and human groundwater extraction on fresh groundwater 595 596 dynamics. We found that with the topographic change in the future, SWI is more sensitive to the upland groundwater supply because of the more intensified freshwater-saltwater interaction in 597 the depression zone. Therefore, when predicting future SWI on coastal marsh landscape, if we do 598 not consider marsh evolution, we are very likely to overestimate SWI under future SLR if the 599 upland GWT is higher than the MHTL, whereas we may underestimate SWI if the future upland 600 GWT is lower than the MHTL. 601

This study highlighted the importance of considering coastal marsh landscape change in predicting SWI under SLR, which was not investigated in previous studies. Over the decadal and century scales, changes in coastal landscape topography can significantly affect the temporal and spatial distributions of SWI under SLR. The insights gained from this study can help improve our understanding of the vulnerability of coastal freshwater systems under SLR, marsh landscape dynamics, and changes in upland groundwater resources, where these interconnections have been previously ignored but warrant greater consideration.

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619 Dynamics Modeling System (CSDMS)

https://csdms.colorado.edu/wiki/Model_download_portal. Model parameters are listed in the
tables above.

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