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Ecosystems



Relationship of salt marsh vegetation zonation to spatial patterns in soil moisture, salinity and topography

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Ecosystems

Salt marsh spatial patterns and zonation

Moffett, Robinson, Gorelick

Relationship of salt marsh vegetation zonation to

spatial patterns in soil moisture, salinity and topography

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Moffett, Robinson, Gorelick

6217 ABSTRACT.

An intertidal San Francisco Bay salt marsh was used to study the spatial relationships between biotic and abiotic ecosystem components: specifically, between vegetation patterns and hydrologic and edaphic variables. Multiple abiotic variables were represented by six metrics: elevation, distance to major tidal channels and to the nearest channel of any size, edaphic conditions during dry and wet circumstances, and the magnitude of tidally-induced changes in soil saturation and salinity. A new approach, quantitative differential electromagnetic induction (O-DEMI), was developed to obtain the last metric. The approach takes the difference in soil induction from dry to wet conditions and converts that information to quantitative maps of tidally-induced changes in root zone soil water content and salinity. The result is a spatially exhaustive map of edaphic changes throughout the ecosystem. Spatially-distributed data on the six metrics were used to explore two hypotheses. 1) Multiple abiotic variables relevant to vegetation zonation each exhibit different, uncorrelated, spatial patterns throughout an intertidal salt marsh ecosystem. 2) Vegetation zones and habitats of individual plant species are uniquely characterized by different combinations of key metrics. The first hypothesis was supported by observed, uncorrelated spatial variability in the metrics. The second hypothesis was supported by binary logistic regression models that identified key vegetation zone and species habitat characteristics from among the six metrics. Based on results from 108 models, the Q-DEMI map of saturation and salinity change was the most useful metric for distinguishing different vegetation zones and species habitats in the salt marsh.

22 KEYWORDS

23 pattern, salt marsh, vegetation, zonation, edaphic, wetland, geophysics, ECa, Q-DEMI

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Ecosystems

Salt marsh spatial patterns and zonation

INTRODUCTION

The segregation of a few dominant plant species into distinctive zones is characteristic of intertidal salt marshes. Each zone comprises a distinctive macrophyte assemblage and may also uniquely sustain other species of concern. For example, stands of native Spartina foliosa densely dissected by tidal channels San Francisco Bay support endangered Rallus longirostris obsoletus (California Clapper Rails), but endangered Reithrodontomys raviventris (Salt Marsh Harvest Mice) favor largely monospecific and undissected Salicornia virginica flats (USFW 2008). The nature and causes of this ecologically important vegetation zonation have been studied for decades with gradient analyses and paired plot, mesocsosm, or transplant studies. Such studies have determined that the causes of salt marsh vegetation zonation are both physical, determined in part by variability in soil (edaphic) and tidal conditions (Pennings and others 2005), and biological, the result of interspecific resource competition and biological response to periodic disturbance (Bertness and others 1992; Emery and others 2001; Pennings and Callaway 1992), even as the specific patterns and species vary regionally (Peterson and others 2008). At the ecosystem scale, it remains a challenge to explain salt marsh vegetation patterning despite knowledge of specific zonation mechanisms at the plant scale. Characterization of the spatial variability of vegetation within salt marsh ecosystems has thus far relied heavily on metrics of relative landscape position such as elevation and distance to tidal channels; however, these geographic metrics, alone, have been insufficient predictors of salt-marsh vegetation zones (Zedler and others 1999, Silvestri and others 2005). Although remote sensing has been used to map spatial patterns of tidal channels (e.g., Marani and others 2006) and marsh surface

22 elevations (e.g., Sadro and others 2007) in relation to salt marsh vegetation, such maps are highly

nonspecific, failing to distinguish unique and consistent salt marsh vegetation habitat

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Ecosystems

Salt marsh spatial patterns and zonation

Moffett, Robinson, Gorelick

6247 Fhgurect2nist2ies. Probabilistic models based on geographic metrics (e.g., Sanderson and others,
25 2001) fare somewhat better, but the fraction of marsh vegetation cover predicted correctly is
26 greatly skewed by very high or very low coverage by a given species. Part of the difficulty in
27 such analyses is that geographic metrics are only rough proxies for the combined effects of many
28 physical, chemical, and biological variables that contribute to salt marsh zonation.

In this study we explored two hypotheses about the spatial nature of multiple zonation-relevant variables and their relationship to salt marsh vegetation distribution. First, we hypothesized that different variables, such as tidal flood duration and direction, root zone soil water content, and soil salinity, may each exhibit different spatial patterns in a salt marsh. The patterns of such variables may have different characteristic spatial scales and gradients oriented in opposing directions. Second, each plant species or zone may correlate with different combinations of variables. For example, one species might grow among dry soil conditions or high soil salinity, but not both; another species might not grow among dry or salty edaphic conditions. Also, due to interspecific interactions, a zone dominated by one species may not be characterized by the same variables as the total habitat range of the species. Prior to this study, such concepts had not been tested in a spatially-distributed manner throughout a salt marsh; we investigated these hypotheses on the basis of extensive data sets spanning the full range of conditions within an intertidal salt marsh ecosystem.

We examined the first hypothesis by comparing the spatial patterns of six zonationrelevant metrics: elevation, distance to major tidal channels and to the nearest channel of any size, the soil saturation/salinity state during dry and wet marsh conditions, and the difference in this edaphic state between conditions. The first three metrics are geographic measures of landscape position and proxies for hydrologic processes relevant to salt marsh vegetation

Salt marsh spatial patterns and zonation

Ecosystems

Moffett, Robinson, Gorelick

47	zonation. Elevation is commonly employed to represent the effects of flood/exposure duration
48	and surface water ponding. A location's distance to the nearest tidal channel represents the likely
49	direction of tidal flooding and groundwater drainage and directional tidal energy effects (e.g.,
50	sediment deposition). This study uniquely considered both distances to primary tidal channels,
51	typically identified from aerial imagery, and distances to small (~ $0.1 - 0.5$ cm wide, rarely
52	mapped) surface drainage pathways hidden beneath the vegetation that we term microtributaries.
53	The remaining three metrics reflected soil properties under different hydrologic
54	conditions (dry and wet marsh soils) and the magnitude of change between conditions. The soil
55	properties considered, soil saturation, salinity, and texture, are known to contribute to salt marsh
56	zonation (Silvestri and others 2005) but previously could only be measured at points, prohibiting
57	extensive repeat sampling and marsh-wide analysis. Geophysical electromagnetic induction
58	(EMI) imaging of bulk apparent soil electrical conductivity (ECa) captures the combined state of
59	soil saturation, salinity, and texture in one ECa number (Friedman 2005, Rhoades and others
60	1999) and can be surveyed quickly over a large area. EMI has been used to investigate patterns in
61	soil properties (e.g., Lesch and others 2005, Robinson and others 2009) but its potential to
62	provide new insight into ecosystem patterning is only beginning to be explored (Stroh and others
63	2001, Robinson and others 2008). Prior to this study the method had not been tested in an
64	environment with as extremely high soil water, salt, and clay contents as in salt marshes. To
65	further the applicability of EMI to salt marsh vegetation analysis, we developed a method to
66	filter out the effects of the soil clay content on the ECa data and leverage the information on
67	changes in soil saturation and salinity from sequential EMI surveys. Our approach was to
68	subtract the data from two EMI surveys (differential or time-lapse EMI; Lesch and others 2005,
69	Robinson and others 2009) and then convert the ECa difference values (Δ ECa) to quantitative

Moffett, Robinson, Gorelick

Estimates of soil water content and salinity change using Archie's Law (Quantitative Differential EMI, or O-DEMI). Our O-DEMI methodology quantified tidally-induced saturation and salinity changes in the salt marsh root zone and enabled assessment of their spatial relationship to vegetation zonation throughout a marsh in unprecedented detail. To explore the second hypothesis, that each salt marsh plant species might bear a different relationship to a suite of relevant variables, we sought to isolate distinguishing characteristics of each of the major vegetation zones and individual species habitats composing the salt marsh ecosystem. We used logistic regression modeling to assess the correlation between vegetation patterns and the six geographic and edaphic metrics. The geophysical data on salt marsh edaphic conditions provided greater insight into the underlying abiotic characteristics of the vegetation patterns than was gained from the geographic metrics alone. In particular, spatial variability in tidally-induced changes in soil water content and salinity, reflected in the Q-DEMI ΔECa metric, were the most effective means of distinguishing vegetation zones and habitats. Multiple variables combine to support ecosystem structures, functions, habitat heterogeneity, integrity, and supply of ecosystem services of salt marshes (Turner and Chapin 2005, Peterson and others 2008), but these variables are seldom analyzed in a spatially-distributed manner. With this study we aimed to better understand how the effects of multiple abiotic variables combine into something more than the sum of the parts, a spatially-variable abiotic template upon which salt marsh vegetation patterns and biotic interactions are expressed. A system-level perspective that integrates both abiotic and biotic variables may help inform the maintenance and restoration of coastal wetlands, a matter of increasing interest worldwide amid concerns of sea level rise, increased storm activity, and coastal development pressure (Peterson and others 2008).

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Ecosystems

MATERIALS AND METHODS

Field Site and Hydrology

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95 The study site was a 0.9 ha intertidal salt marsh in southern San Francisco Bay, within the 96 Palo Alto Baylands Nature Preserve (37°27'54" N, 122°6'58" W). The geological and botanical 97 history of the surrounding Santa Clara Valley were described by Cooper (1926) and the geology 98 underlying the Palo Alto Baylands by Hamlin (1983). The history and character of the marsh 99 were similar to that described by Hinde (1954) for the adjacent marsh to the south. The 100 underlying site stratigraphy consisted of 3-5 meters of fine estuarine mud, predominantly

101 montmorillonite clay, overlying a saline aquifer system (Hamlin 1983).

102 Vegetation Mapping

103 Plant species at the site were: Spartina foliosa, Salicornia virginica (S. depressa), 104 Distichlis spicata, Jaumea carnosa, Grindelia stricta, Frankenia salina, Salsola soda, and 105 Atriplex prostrata (see USDA (2009) for synonymous species names). The habitat occupied by 106 each species was mapped by: marking the boundaries of assemblages distinguished by the 107 presence/absence of each of the species, digitizing these polygonal boundaries using streaming 108 GPS, and identifying the relative abundance of each species within each polygon. This method 109 was similar to that of Zedler and others (1999) for San Diego Bay marshes, but with greater 110 emphasis on identifying the locations of assemblage boundaries. Surveys of species' percent cover within $1-m^2$ quadrats verified assemblage composition at 69 locations. The 57/134 111 112 assemblage polygons verified by one or more quadrats accounted for 81% of the marsh area. 113 Vegetation zones were classified by the species of greatest (dominant) cover fraction in 114 each assemblage polygon. The quadrat surveys confirmed that this was a sufficient means of 115 identifying vegetation zones since assemblage composition within each zone defined in this

Moffett, Robinson, Gorelick

Figure 2wast consistent. In addition to the spatial distribution of major vegetation zones, in this study we were interested in the full range of conditions among which each plant species grew. We refer to a plant species' habitat as all the areas the species occupied regardless of cover density. In our vegetation discrimination analysis we assessed the salient characteristics of zones and species habitats separately and compared the results. Mapping Edaphic Conditions A logical precursor to understanding salt marsh vegetation distribution is a three-dimensional description of root zone edaphic conditions throughout the ecosystem, but obtaining

spatially-extensive data on relevant physical and chemical soil properties has been intractable with point-sampling methods. The combination of heterogeneous soil water content, salinity, and clay fraction was captured in this study by maps of bulk soil electrical conductivity (ECa). The ECa data were obtained on two separate days by repeatedly traversing the field site carrying a streaming EMI instrument (DUALEM-1S, Dualem, Inc., Milton, ON, Canada) and GPS, logged concurrently. Sequential traverses were separated to account for the $\sim 4 \text{ m}^2$ EMI measurement support area. We estimated the vertical soil interval represented by the ECa data was 0-0.40 m depth (see online supplement), approximately the depth of the salt marsh root zone. We post-processed ~5000 ECa measurements per survey (Robinson and others 2008) and corrected for effects of soil temperature (Reedy and Scanlon 2003) to produce kriged ECa maps at 2-m resolution. Successive measurements of ECa at test locations agreed to within 0.01 dS/m, which we take to be the ECa uncertainty, though the EMI instrument accuracy was 0.001 dS/m.

The two EMI surveys were timed to capture different hydrologic conditions. The first survey occurred just prior to the neap-spring tidal transition, when the marsh had not been flooded in eight days (Nov. 19, 2007); we refer to these as data from "dry" marsh conditions.

Salt marsh spatial patterns and zonation

Ecosystems

139	The second survey was partially into the next spring tide cycle, immediately following a flood
140	tide (Dec. 7); we refer to these as data from "wet" marsh conditions. We use the terms "dry" and
141	"wet" as qualitatively convenient reminders of antecedent tidal conditions although both
142	circumstances represent very moist soils (>80% saturation). Both survey times were near mid-
143	day and no rain occurred while the marsh surface was exposed during the study period.
144	The field site experiences mixed semi-diurnal tides and a semi-arid Mediterranean
145	climate with winter precipitation (~39 cm/yr). The marsh plain is above mean high water and is
146	flooded by the higher high tide on $\frac{2}{3}$ - $\frac{3}{4}$ of days during each spring-neap cycle. To verify
147	ambient hydrologic conditions, we monitored groundwater and tidal conditions at the site by
148	logging pressure and temperature every 10 minutes at the bed of the two primary tidal channels
149	and in 43 wells and piezometers installed 0.5 to 1.0 m into the marsh substrate. We monitored
150	hydraulic heads in the root zone with 23 tensiometer pairs spanning 10-15 and 20-25 cm depths.
151	Tensiometer data were collected manually during the geophysical surveys.
152	Empirical relationships have shown ECa to increase with increasing soil clay content,
153	water content (θ), or solution electrical conductivity (ECw) (e.g., Rhoades and others, 1999),
154	though not for as high values as occur in salt marshes. We conducted laboratory analyses to
155	establish the specific relationships between ECa and salt marsh soil properties. Twenty-three soil
156	sampling locations were strategically chosen using the ECa data from the first survey and
157	response-surface directed sampling (Corwin and Lesch 2005, Lesch 2005). After collecting ECa
158	data at each location, soil cores (2.5 cm diameter) were collected manually from 0-30cm and 30-
159	60cm depth. The 0-30 cm depth interval was chosen to correspond roughly to the EMI signal
160	depth, enabling correlation with ECa survey data. The 30-60 cm deep samples were used in
161	parameterizing the Q-DEMI methodology, discussed below. The cores were immediately sealed
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Figuhastia, **bags** and promptly weighed in the laboratory. Samples were air-dried for 11-28 days, homogenized subsamples weighed, oven-dried at 105° C for at least 12 hours and re-weighed, and core water fractions and bulk densities calculated. Duplicate homogenized subsamples were analyzed for soil paste extract electrical conductivity (ECe) and soil texture (University of Idaho Pedology Laboratory standard procedures). Pore water samples extracted adjacent to the coring locations from 30 cm depth using a suction lysimeter ("sipper", ~≤5 kPa suction) were analyzed in the laboratory for pore water electrical conductivity (ECw).

169 Quantitative Differential EMI Methodology

Each geophysical survey provided a snapshot of the combination of water, salt, and clay conditions throughout the salt marsh at one point in time. We developed a method to transform the difference in ECa between dry and wet tidal conditions into spatially-distributed, quantitative estimates of changes in root zone soil water content and salinity. The premise of the Q-DEMI method was that a change in the ECa value of a location was due to changing soil water content and salinity while clay content remained constant. In our Q-DEMI analysis, we subtracted the later "wet" ECa data from the earlier "dry" ECa data, simulating a case of increasing soil moisture ($\Delta ECa = ECa_{dry} - ECa_{wet}$). We then determined the nature of the edaphic change, whether caused by changing soil water content or by changing soil salinity, from the sign of ΔECa . An observed increase in ECa between dry and wet conditions (- ΔECa) indicated an increase in soil water content: an increase in salt content could not explain the change in these areas because tidal waters were known a priori to be less saline (33.4 dS/m) than the marsh pore waters (ECw \approx 57.2 dS/m) to which they were added in order to wet-up the marsh. In contrast, an observed decrease in ECa ($+\Delta$ ECa) indicated a decrease in pore water salinity: under conditions of increasing tidal water availability, water content would remain constant or increase

Salt marsh spatial patterns and zonation

Ecosystems

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3 4	185	and so could not explain the ECa change in these areas. Because saturation and salinity changes
5 6 7	186	could occur simultaneously with opposing effects, the Q-DEMI calculations represent the
, 8 9	187	conservative case in which all ECa change is ascribed to the dominant process, identified by the
10 11	188	sign of Δ ECa.
12 13 14	189	Quantifying saturation and salinity changes was accomplished using Archie's Law in our
14 15 16	190	Q-DEMI methodology. Archie's Law is a well-studied geophysical relationship between ECa
17 18	191	and: pore water conductivity (ECw), a formation factor (f) related to porosity, the soil saturation
19 20 21	192	(S), and the soil mineral surface conductivity due to adsorbed ionic charge (σ_s) (Kirsch 2006).
22 23 24	193	$ECa = ECw \cdot f \cdot {}^{2} + \sigma_{s} $ (Eqn. 1) S
25 26	194	The mineral surface conductivity (σ_s) is important in soils with large clay fractions, such as in
27 28	195	our salt marsh, but has not been tabulated for salt marsh clay soils. We estimated f and σ_s using a
29 30 31	196	simple linear regression between ECa and pore water conductivity (ECw) for saturated samples
32 33	197	(S = 1). The samples used for this regression were from 30-60 cm depth since these samples were
34 35 26	198	known to be from the saturated zone below the water table. The resulting f and σ_s parameter
30 37 38	199	estimates compared favorably with estimates from more complicated methods (see online
39 40	200	supplement). These parameters permitted Q-DEMI calculation of changes in saturation, due to
41 42 42	201	aerated pore space being filled by tidal waters, and changes in salinity, due to flushing of salt
43 44 45	202	marsh soils, using variations on Archie's Law.
46 47	203	In the saturation-change dominated (- Δ ECa) areas of the marsh we solved Archie's Law

(Eqn. 1) for the net soil water content change required to account for the observed increase in ECa between dry and wet marsh conditions. To reduce one excess degree of freedom in the calculation we assumed that initially aerated pore space in the soil was completely filled by the flood tide, leading to a minimum estimate of soil saturation change since the effect of any

3087 **Figppe** deair 26 ould have reduced the magnitude of Δ ECa. The formula we derived to calculate 209 saturation change (Δ S) is shown below (Eqn. 2). The parameters are the: formation factor (f), 210 mineral surface conductivity (σ_s), tidal flood water electrical conductivity (EC_{tide}), and ECa 211 during wet (ECa_{wet}) and dry (ECa_{dry}) conditions. 212 $\frac{-(\sigma_s - ECa_{wet} - f \cdot _{tide}) - \sqrt{(\sigma_s - _{eC} - f \cdot _{tide})^2 - 4 \cdot (f \cdot _{tide}) \cdot (ECa_{wet} - ECa_{dry})}{EC}$ $\frac{-(\sigma_s - ECa_{wet} - f \cdot _{tide}) - \sqrt{(\sigma_s - _{eC} - f \cdot _{tide})^2 - 4 \cdot (f \cdot _{tide}) \cdot (ECa_{wet} - ECa_{dry})}{2 \cdot (f \cdot _{tide})}$ (Eqn. 2)

213 The real solution to Eqn. 2 using a positive discriminant root yielded unrealistic ΔS values 214 greater-than one, so was discarded.

215 In the salinity-change dominated ($+\Delta$ ECa) areas of the marsh we solved Archie's Law 216 (Eqn. 1) for the pore water electrical conductivity (ECw) under dry and wet marsh conditions.

217 We then used the seawater equation of state to convert each ECw value to a salinity value.

Subtracting the salinity values yielded the change in salinity required to account for the observed
decrease in ECa between dry and wet marsh conditions. To reduce one excess degree of freedom

in the calculation we assumed that these areas of the marsh remained water-saturated, leading to

221 a minimum estimate of salinity change since the effect of any concurrent increase in soil water

222 content would have reduced the magnitude of ΔECa . (Field observations suggested that these

223 areas of the marsh did remain saturated throughout dry and wet marsh conditions.)

In addition to the two maps of ECa, from dry and wet marsh conditions, the map of saturation and salinity changes produced by the Q-DEMI methodology provided a third spatiallydistributed metric of salt marsh root zone characteristics against which to compare salt marsh vegetation zonation.

228 Mapping Marsh Geometry

Geometric measures of spatial context within the ecosystem have traditionally been employed as indicators of salt marsh ecosystem structure and spatially-variable intertidal

Ecosystems

Moffett, Robinson, Gorelick

hydrologic effects. The most common such geographic, or "landscape position" (Zedler and others 1999), metrics are elevation and distance-to-channel. We mapped these metrics at the same high resolution as our edaphic data sets. We represented marsh plain topography by a 2-m horizontal resolution kriged map of 742 marsh plain surface elevations surveyed using a total station, verified against LIDAR data. Major tidal channels are typically identified from aerial imagery, but we could find no precedent for mapping the small, connected surficial drainage pathways hidden under the vegetation canopy ("microtributaries"). We identified the banks of major tidal channels and microtributaries by traversing them with a streaming GPS (20-cm post-processed horizontal accuracy). Two distance-to-channel metrics were calculated as the shortest straight-line distances from the center of each elevation grid cell to: 1) the nearest of the two primary tidal channels (bounding and bisecting the study area, Figure 1); 2) the nearest channel of any size.

Statistical Vegetation Differentiation

Salt marsh spatial patterns and zonation

To contrast the utility of the six metrics described above in differentiating vegetation zones and plant species habitats, we employed binary logistic regression (BLR) models (SPSS 2009). A logistic regression is analogous to a linear regression but with a categorical, instead of continuous, dependent variable. By comparing the vegetation at each location in the marsh to the collocated values of the six metrics and repeating this for all marsh locations, the BLR method built models of those combinations of the six metrics that best distinguished the selected vegetation zone or habitat type. BLR models were assessed at the 95% confidence level.

We tested 108 BLR models, including univariate and multivariate analyses for each vegetation zone and species habitat. In the univariate cases we assessed whether any of the six metrics, alone, could correctly differentiate the marsh areas inside and outside each of the six

Moffett, Robinson, Gorelick

Figior Vage28 tion zones (6 metrics x 6 zones = 36 zone models). We also tested whether any of the six metrics, alone, could correctly differentiate the marsh areas occupied or not occupied by each species, regardless of its cover density (6 metrics x 6 species = 36 habitat models). These 76 models served to test the univariate predictive capacity of each of the six metrics in relation to vegetation patterning at our site. For these models, the two-fold null hypothesis in each case was either 100% or 0% cover by the selected zone or species.

In the multivariate analyses, we built forward-conditional BLR models for each vegetation zone and species habitat. This approach tested whether a combination of multiple metrics could better identify the distinguishing characteristics of each zone and habitat than a single metric. We tested three metric combinations: 1) the three geographic metrics, 2) the three edaphic metrics, 3) all six metrics, for total of 36 multivariate models (3 combinations x (6 zones + 6 habitats) = 36 models). The forward-conditional BLR method selected only those metrics that significantly contributed to the zone or habitat prediction at the 95% confidence level. For these models, the two-fold null hypothesis in each case was either 100% or 0% cover by the selected zone or species. The results of the BLR models revealed the key characteristics distinguishing each habitat envelope and zone at our site.

270 RESULTS

271 Vegetation Patterns and Marsh Geometry

The spatial distribution of vegetation zones at the site is shown in Figure 1a, with zones labeled by the genus of the dominant species. Quadrat surveys verified that species identified as zone dominants occupied a majority (59% \pm 16%) of the zone's cover. Zones dominated by the succulent Salicornia (28% of total marsh area) and the grasses Spartina (19%) and Distichlis (47%) were most prominent at the site, with smaller areas dominated by Jaumea (4%),

Ecosystems

2 3 4	277	Frankenia (1%), and Grindelia (2%). Salsola and Atriplex individuals were present in only a few
5 6 7 8 9 10 11	278	locations. The thick black outlines in Figure 1a highlight the three major vegetation zones,
	279	dominated by Spartina, Distichlis, and Salicornia. Zone assemblage compositions are illustrated
	280	by maps of relative cover density for each species (see appendix Figure A1), which were used to
12 13 14	281	assess the total habitat occupied by each species.
15 16	282	The elevation ranges ($\mu \pm 1\sigma$ m above mean sea level) spanned by the species were not
17 18 10	283	distinct: Distichlis, 1.04 ± 0.04 ; Salicornia, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea, 1.03 ± 0.06 ; Jaumea, 1.03 ± 0.05 ; Spartina, 1.00 ± 0.06 ; Jaumea,
20 21	284	0.05; Frankenia, 1.03 ± 0.03 ; Grindelia, 1.04 ± 0.03 . Overlap between the elevation ranges of
22 23 24 25 26	285	key species is common in salt marshes despite their characteristically distinct vegetation zonation
	286	(Silvestri and others 2005, Sadro and others 2007). The average marsh plain elevation from the
27 28	287	kriged topographic data was 1.02 ± 0.06 m above mean sea level (m aMSL) and ranged from
29 30	288	0.61 to 1.32 m. The seeming visual correlation between areas of slightly lower elevation and the
31 32 33	289	southern, Spartina-dominated zone (Figure 1b) was not statistically supported because those
34 35	290	same elevations elsewhere in the marsh were dominated by different species. Employed in
36 37	291	univariate BLR models, elevation failed to justify rejecting the null hypothesis for any of the
38 39 40	292	vegetation zones or species habitats at our site.
41 42	293	Qualitative assessment of marsh locations' distance to primary tidal channels showed the
43 44 45	294	major zones dominated by Spartina, Distichlis, and Salicornia to each occur at any distance from
46 47	295	the major tidal channels that bound and bisect the marsh (Figure 2a). The Spartina-dominated
48 49	296	zone appeared to coincide with a region of dense microtributaries (Figure 2b), yet neither
50 51 52	297	distance-to-channel metric warranted rejecting the univariate BLR models' null hypothesis for

any of the vegetation zones or species habitats.

Salt marsh spatial patterns and zonation

299 Edaphic Conditions and Vegetation

Moffett, Robinson, Gorelick

Figure The spatial structure of edaphic conditions throughout the marsh, and the magnitude of ECa values reflecting these conditions, remained consistent between the dry (Figure 3a) and wet (Figure 3b) surveys. Mean ECa values for the two surveys were 13.37 and 13.71 dS/m, respectively (2.05 dS/m standard deviations; correlation coefficient r = 0.83). Tensiometer data confirmed that the root zone was drier during the first, "dry" geophysical survey than during the second, "wet" survey. Tides rapidly and uniformly covered the marsh to a depth of 0.5 m during spring tide flooding events between the surveys. The specific relationships between ECa values and edaphic conditions (soil solution and paste extract conductivities and water and clay contents) determined for this salt marsh are presented in the appendix. The configuration of vegetation zones (Figure 1a) did not resemble the spatial pattern of edaphic conditions (Figure 3). Instead, interior marsh areas that exhibited persistent high soil water content and/or salinity (high ECa) appeared coincident with major zone boundaries. A phenomenon of stressful edaphic conditions and major zone boundaries occurring in the same location was described for Spartina and Salicornia in northern San Francisco Bay salt marshes by Mahall and Park (1976a) but had not been illustrated in two dimensions; our result is consistent with this explanation of ecotone locations. Though not consistently correlated with any vegetation zone or elevation, the edaphic variability in the marsh was significantly related to the hydrologic processes represented by the distance-to-channel metrics (r = 0.36 to 0.54). Low soil saturation and/or salinity (low ECa) occurred close to tidal channels and more stressful edaphic conditions (high ECa) occurred further from the channels. Neither ECa data set provided information sufficient to reject the null hypothesis of the univariate BLR models. The spatial pattern of tidally-induced changes in edaphic conditions revealed by

322 subtracting the wet and dry ECa surveys (Δ ECa, Figure 4a) was more heterogeneous than the

Page 15 of 36

Salt marsh spatial patterns and zonation

Ecosystems

Moffett, Robinson, Gorelick

spatial variability in static edaphic conditions (Figure 3). The pattern of change was not altered by the Q-DEMI calculations, which converted ΔECa values to soil saturation and salinity change quantities (Figure 4b). The conversion was made using values of f = 0.223 and $\sigma_s = 2.479$ dS/m. The average estimated saturation change in the fluid-exchange dominated areas of the marsh (blue in Figure 4b) was $6.2 \pm 5.5 \%$ ($\mu \pm 1\sigma$). The average estimated salt loss from the salt-exchange dominated areas of the marsh (red in Figure 4b) was 0.77 ± 0.64 kg/m². The large standard deviations of these average results were due to highly heterogeneous soil aeration and flushing throughout the marsh. Despite the Q-DEMI methodology producing conservative estimates of the magnitude of edaphic change, we emphasize that the methodology permits mapping the magnitude of salt and water exchange in a spatially-distributed way throughout an ecosystem for the first time. Spatial patterns of saturation and salinity change did not qualitatively resemble

vegetation zonation (Figures 1a, 4b), yet BLR models based on Δ ECa were able to partially describe the zones dominated by every species except Distichlis. For the Salicornia-, Spartina-, Jaumea-, Frankenia-, and Grindelia-dominated zones, the BLR models correctly distinguished 22-44 % of the area inside each zone and 63-67% of the area outside each zone. Though short of the ideal prediction (100% correct both inside and outside each zone), these results using the Δ ECa metric were a substantial improvement over the null hypothesis returned by the models based on the other five metrics.

 Δ ECa BLR models were more successful at distinguishing between marsh areas occupied and not occupied by each of the six plant species, regardless of cover density (appendix Figure A1). Δ ECa BLR habitat models correctly identified 64% of the observed Distichlis and Salicornia occurrences and 37% and 44% of observed absences, respectively. $\Delta ECa BLR$

Moffett, Robinson, Gorelick

Figdels for Separting and Jaumea habitat correctly predicted 70% and 73% of the observed occurrences and 41% and 46% of observed absences, respectively. $\Delta ECa BLR$ models for Frankenia and Grindelia were less successful at correctly predicting occurrences of these species (28% and 23%, respectively) but more successful at correctly predicting absences (63% and 60%, respectively). For all six species, the ΔECa BLR habitat models justified rejecting the null hypotheses (95% confidence). The patterns in edaphic conditions and geographic metrics of salt marsh structure support our two hypotheses regarding the spatial nature of zonation-relevant variables and their relationship to salt marsh vegetation distribution. 1) Multiple metrics relevant to salt marsh vegetation zonation each exhibit different patterns. These patterns are characterized by different spatial scales and degrees of spatial heterogeneity. 2) Alone, only the ΔECa metric provided information useful in indentifying vegetation zones and species habitats. The relation of the

358 ΔECa metric to vegetation differed depending on the species considered and whether the species
359 was considered alone or as a zone-dominant.

360 Multivariate Vegetation Zone and Habitat Discrimination

We hypothesized that a combination of multiple metrics might better discriminate salt marsh vegetation zones and individual species habitats than univariate models. The metric combinations we tested using forward-conditional BLR models were: 1) the three geographic metrics, 2) the three edaphic metrics, and 3) all six metrics. Salient results are presented here; complete BLR model results are provided in the online supplement.

366 Except in the case of the Distichlis-dominated zone, none of the multivariate models
 367 identified vegetation zones or habitats significantly better than the univariate ΔECa BLR models.
 368 For the Distichlis-dominated zone, a BLR model including all three geographic metrics correctly

Ecosystems

Salt marsh spatial patterns and zonation

Moffett, Robinson, Gorelick predicted 45% of the marsh area within the zone and 72% of the area outside the zone, compared to the null hypothesis returned by the univariate ΔECa BLR model. This result suggests that Distichlis may be reliant on the hydrologic marsh processes implicit in the elevation and

distance-to-channel metrics to maintain a dominant cover fraction. In contrast, the dominance of

the other five major species at the site may be related to the magnitude of temporal variation in

root zone soil water content and soil salinity, represented by the ΔECa metric. For example,

 Δ ECa was the only significant predictor of the marsh areas that Jaumea occupied, whether it was

the dominant cover fraction or not, even when the other five metrics were made available to the

forward-conditional model. However, a BLR model based on AECa correctly predicted 73% of

Jaumea occurrences in the salt marsh but only 32% of Jaumea-dominated zones. The difference

between the zone and habitat models suggest that the tidally-induced changes in root zone water

and salt content represented by the ΔECa metric may affect the growth and interspecific

interactions of a species within a vegetation assemblage differently than the growth and survival

of individual plants of that species throughout the marsh.

DISCUSSION

Vegetation Zonation

Our use of geophysical EMI technology was motivated by the inherent spatial limitations of transect- and plot-based methods of investigating in situ salt marsh vegetation patterning. Logistic regression models based on extensive two-dimensional data identified major characteristics that distinguished the vegetation zones and species habitats at our site. The detail of our regression models, based on over 2000 data points, makes it striking that some zones and species habitats were uniquely identified by a combination of multiple variables (e.g., Distichlis) but others were best identified by a single variable (e.g., Jaumea). It is also significant that there

Ecosystems

Salt marsh spatial patterns and zonation

Moffett, Robinson, Gorelick

Figure large difference in the ability of the models to describe the key characteristics of the total habitat envelope of a species versus the zone for which it provided the dominant cover class. Existing theory suggests that the survival of individual salt marsh plants may be strongly affected by environmental conditions but that the growth rate and relative cover of a species may be more strongly affected by interspecific interactions with its neighbors (e.g., Emery and others, 2001). This hypothesis seems to be supported by the greater success with which our regression models, based on metrics of abiotic ecosystem variability, were able to identify characteristic individual species habitats compared to vegetation zones. Tide-induced Edaphic Change

The most striking result of the vegetation discrimination analysis was the utility of the ΔECa metric, alone, in identifying vegetation zones and species habitats. The O-DEMI method and soil core analyses showed that ΔECa represented the amount of water and salt exchanged from the root zone (Figure 4b) between two points in time. The data from this study could not definitively separate, however, whether observed changes in edaphic conditions were due entirely to intervening tidal flooding, or due to a combination of physical and biological effects. The lack of correlation between ΔECa and either elevation or distance-to-channel argues against the hydrologic processes implied by the elevation and distance-to-channel metrics as the dominant determinants of spatial patterns in edaphic change.

The phenomenon of large, broadly distributed decreases in soil salinity, identified in this
study by decreases in ECa between dry and wet marsh conditions, has not previously been
reported and the precise cause is unknown. Potential mechanisms for what was apparently rapid
flood tide-induced salt removal from the salt marsh root zone include: diffusion, leaching, or
dissolution of salt from the surface; plant salt uptake; or dilution by convective mixing in soil

Page 19 of 36

Ecosystems

Moffett, Robinson, Gorelick

macropores. On average, the 0.77 kg/m² of salt loss from the salt-exchange dominated areas of
the marsh constituted approximately 15% of the salt contained in the root zone pore water fluids.
Were dilution the only mechanism in operation, a salinity decrease of this magnitude would have
required approximately 70% of the root zone pore water to be replaced by the less salty tidal
waters, on average across the marsh. Salt uptake by vegetation would reduce the amount of pore
water turn-over required to match the geophysical observations.

421 Plant – Soil Interactions

Salt marsh spatial patterns and zonation

Qualitative comparison of tide-induced edaphic change and vegetation zonation (Figures
1a, 4b) suggests that the greatest density of areas experiencing large saturation increases due to
tidal flooding occurred in the Spartina-dominated zone, the greatest density of areas
experiencing salt loss occurred in the Salicornia-dominated zone, and changes of intermediate
magnitude prevailed in the Distichlis-dominated zone. We hypothesize that these spatial
coincidences may be related to three different ecosystem engineering effects enabled by the
different physiologies and morphologies of these three species.

First, enhanced sediment deposition within, and enhanced erosion around, Spartina clusters has been reported at low marsh elevations (Temmerman and others 2007, van Hulzen and others 2007). This mechanism may also explain the great density of microtributaries in the Spartina-dominated zone (Figure 2) at our high-elevation marsh site. Because Spartina is more productive in low-salinity conditions than Salicornia and Distichlis (Bertness and others 1992, Mahall and Park 1976b), it may remain dominant precisely where surrounding microtributaries enhance pore water drainage and flushing. This hypothesis is supported by experimental manipulations of marsh hydrology (Balling and Resh 1983, Wiegert and others 1983).

Figure Second, Salicornia is one of the most salt-tolerant halophytes, producing more biomass \$3277 than Spartina and Distichlis at high salinities and water contents, but less at low salinities (Pearcy and Ustin 1984). We hypothesize that efficient root salt exclusion and water use by Salicornia (Mahall and Park 1976b) may locally maintain a saturated and salty root zone, enhancing Salicornia competitiveness. Third, the matted morphology of Distichlis may decrease surface water velocities and so inhibit erosion of microtributaries (and so Spartina-dominance) in areas of low salinity favorable for both grasses (Bertness and others 1992). Distichlis, like Salicornia, is quite effective at excluding ions at the root membrane (Marcum and others 2007), and so may persist in areas of intermediate salinity, where it is often found nearly co-dominant with Salicornia at our site. A lack of known disturbance at our study site in at least 30 years argues against disturbance as an explanation for the distribution of Distichlis at the site; furthermore, both Distichlis and Salicornia are very effective at recovering from disturbance in northern California salt marshes, neither necessarily competitively displacing the other (Allison, 1995). CONCLUSION This study explored the spatial relationships between salt marsh vegetation patterns and six zonation-related metrics. The metrics were based on geographic and edaphic data: elevation, distance to major tidal channels and to the nearest channel of any size, the soil saturation/salinity state during dry and wet marsh conditions, and the difference in this edaphic state between conditions. The metrics, mapped at high resolution throughout a salt marsh ecosystem, exhibited very different spatial patterns. Among the six metrics, information on tide-induced edaphic change was most useful in discriminating salt marsh vegetation zones and individual species habitats. Unexpectedly poor spatial correlation between edaphic conditions and proxies for

Ecosystems

Moffett, Robinson, Gorelick

460 hydrologic forcing suggest, instead, the potential importance of plant-soil relations in spatial461 patterns of tide-induced edaphic change.

Salt marsh spatial patterns and zonation

The results of this study supported two hypotheses relating the spatial patterns of salt marsh vegetation and abiotic variables. The hypotheses extend classic concepts of niche breadth and overlap (Colwell and Futuyma 1971) into three spatial dimensions. First, the influence of each resource used by, and stressor endured by, salt marsh vegetation may be spatially variable. These spatial patterns are combined implicitly in nature by superposition and may exhibit emergent patterns and properties that are more than the sum of the contributing variables. Second, multiple contributing variables may affect the distribution of interacting species assemblages differently than the distribution of individuals. Multivariate relationships between abiotic and biotic ecosystem patterns are difficult to assess without high-resolution spatially-distributed data at the ecosystem-scale. Geophysical methods such as EMI and Q-DEMI provide means to obtain high-resolution, spatially-distributed data on root zone soil properties that have previously been prohibitively difficult to obtain. In this study, such edaphic data was more useful in characterizing salt marsh vegetation zones and habitats than traditional geographic metrics such as elevation and distance-to-channel.

The challenge of predicting the vegetation distribution of intertidal salt marsh ecosystems
persists. Despite functional similarity between different salt marsh species around the world,
regional and latitudinal differences so far prohibit development of a universally-applicable,
mechanistic, zonation model (Farina and others, 2009; Pennings and others 2003). Even if such a
model were possible, its accuracy would necessarily vary from site to site. Some of the most
pressing questions regarding salt marsh vegetation zonation, such as the expected response of a
marsh to restoration efforts or to an invasive species, must be answered on a site-by-site basis

Ecosystems

Salt marsh spatial patterns and zonation

3 4	48 37	Englirea 3/a;e2/bire probabilistic, not deterministic, answers. Three-dimensional mapping, EMI
5 6 7	484	geophysics, and the Q-DEMI methodology can provide a cost-effective, rapid, and repeatable
, 8 9	485	means to statistically characterize a salt marsh site. The resulting spatial and temporal patterns
10 11	486	can then be used as a foundation upon which to interpret or predict vegetation distributions and
12 13 14	487	biotic interactions based on existing region- and species-specific knowledge. Linking plot-scale
15 16	488	studies of plant-soil relations and interspecific interactions to marsh-scale studies of spatial
17 18	489	variability such as this one may provide the most promising means to fill the gap between the
19 20 21	490	general principles and site-specific needs of salt marsh vegetation zonation science.
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26 27 28	493	University. Any opinions, findings, and conclusions or recommendations expressed in this
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Salt marsh spatial patterns and zonation

Ecosystems

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Page 25 of 36

Salt marsh spatial patterns and zonation

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- 4 588 APPENDIX
- 4 588 APPENDIX
- 589 Vegetation Habitat Distribution Maps
- 590 (Figure A1 and caption.)
- 591 Relating ECa to Salt Marsh Soil Properties
- 592 Despite the extreme environment, correlations between our ECa and soil core data 593 showed that salt marsh ECa measurements can be interpreted in terms of three key edaphic

Ecosystems

Salt marsh spatial patterns and zonation

properties: water content, salt content, and clay content. Variability in ECa values was significantly related to variability in each of these edaphic properties (p < 0.005, Table A1). At our site the EMI signal was dominated by the total salt content of the soil (as measured by the soil paste extract conductivity. ECe) but the soil water (θ) and clay contents also contributed. The clay content throughout the marsh was remarkably uniform and so did not figure significantly in our analysis. See the online supplement for comparison of our salt marsh relationships with prior published relationships at lower water, salt, or clay contents. In brief, we conclude that the salt marsh ECa – ECe and ECa – θ relationships scale as in other environments but that the soil pore soil c... solution conductivity (ECw) and soil clay content of intertidal salt marshes have unique effects

on EMI signals.

Figure 1. a) Major vegetation zones, classified by the species of greatest cover fraction. b) Site

FIGHEREA AND TABLE CAPTIONS

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606	topography, units: meters above mean sea level.
607	Figure 2. a) Shortest distance to one of the main tidal channels, shown in light blue bounding and
608	bisecting the marsh site. b) Shortest distance to the nearest channel of any size,
609	including microtributaries shown in dark blue.
610	Figure 3. Root zone bulk soil electrical conductivity (ECa) from (a) dry and (b) wet marsh
611	conditions. Dark blue lines are channel and microtributary banks, black lines depict
612	major vegetation zone boundaries.
613	Figure 4. a) Edaphic change between dry and wet marsh conditions, represented by the change in
614	bulk soil electrical conductivity (Δ ECa, dS/m). b) Result of Q-DEMI conversion of
615	Δ ECa to changes in root zone saturation (%) or salinity (kg/m ³) between dry and wet
616	marsh conditions. Blue areas were dominated by net saturation increase between dry
617	and wet conditions, red areas were dominated by net salinity decrease. Dark blue lines
618	are channel and microtributary banks, black lines depict major vegetation zone
619	boundaries.
620	Figure A1. Relative cover fraction of each major plant species at the site: primary cover,
621	secondary cover, tertiary cover, present as minor cover. Clockwise from lower left:
622	Distichlis spicata, Salicornia virginica (S. depressa), Spartina foliosa, Grindelia

- 623 stricta, Frankenia salina, Jaumea carnosa.
- 624 Table A1. Relation of ECa data to soil properties.

Ecosystems



627 Figure 2a, 2b.



628 Figure 3a, 3b.

Salt marsh spatial patterns and zonation



629 Figure 4a, 4b.





Secondary

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Property	Sample Statistics		Correlation
	Mean	Standard Deviation	with ECa
Soil Paste Extract Electrical Conductivity ⁱ (ECe, dS/m)	68.9	15.0	0.67
Soil Pore Water Electrical Conductivity ⁱⁱ (ECw, dS/m)	57.2	7.0	0.53
Soil Volumetric Water Cor , t^{t} ()	0.83	0.15	0.43
Soil Clay Content ⁱ (%)	61.8	9.4	0.51
Soil Temperature ⁱⁱⁱ (dry conditions, °C)	13.98	0.54	
Soil Temperature ⁱⁱⁱ (wet conditions, °C)	11.56	0.41	
Tide Water Electrical Conductivity (EC _t , dS/m)	33.4		
$^{i}N = 23, \ ^{ii}N = 17, \ ^{iii}N = 14$		·	
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