Spatial gradients in the characteristics of soil-carbon fractions are associated with abiotic features, but not microbial communities

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Abstract

Coastal terrestrial-aquatic interfaces (TAIs) are dynamic zones of biogeochemical cycling influenced by salinity gradients. However, there is significant heterogeneity in salinity influences on TAI soil biogeochemical function. This heterogeneity is perhaps related to unrecognized mechanisms associated with carbon (C) chemistry and microbial communities. To investigate this potential, we evaluated hypotheses associated with salinity-associated shifts in organic C thermodynamics, biochemical transformations, and nitrogen-, phosphorus-, and sulfur-containing heteroatom organic compounds in a first-order coastal watershed in the Olympic Peninsula of Washington state, USA. In contrast to our hypotheses, thermodynamic favorability of water soluble organic compounds in shallow soils decreased with increasing salinity (43-867 μS cm^{-1}), as did the number of inferred biochemical transformations and total heteroatom content. These patterns indicate lower microbial activity at higher salinity that is potentially constrained by accumulation of less favorable organic C. Furthermore, organic compounds appeared to be primarily marine/algal-derived in forested floodplain soils with more lipid-like and protein-like compounds, relative to upland soils that had more lignin-, tannin-, and carbohydrate-like compounds. Based on a recent simulation-based study, we further hypothesized a relationship between C chemistry and ecological assembly processes governing microbial community composition. Null modelling revealed that differences in microbial community composition—assayed using 16S rRNA gene sequencing—were primarily the result of limited exchange of organisms among communities (i.e., dispersal limitation). This results in unstructured demographic events that cause community composition to diverge stochastically, as opposed to divergence in community composition being due to deterministic selection-based processes associated with differences in environmental conditions. The strong influence of stochastic processes was further reflected in there being no statistical relationship between community

assembly processes (e.g., the relative influence of stochastic assembly processes) and C chemistry (e.g., heteroatom content). This suggests that microbial community composition does not have a mechanistic or causal linkage to C chemistry. The salinity-associated gradient in C chemistry was, therefore, likely influenced by a combination of spatially-structured inputs and salinity-associated metabolic responses of microbial communities that were independent of community composition. We propose that impacts of salinity on coastal soil biogeochemistry need to be understood in the context of C chemistry, hydrologic/depositional dynamics, and microbial physiology, while microbial composition may have less influence.

1. Introduction

The interface between terrestrial and aquatic ecosystems represent a dynamic and poorly understood component of the global carbon (C) cycle, particularly along the tidally-influenced reaches of coastal watersheds where terrestrial and marine biospheres intersect (Krauss et al., 2018; Neubauer et al., 2013a; Tank et al., 2018; Ward et al., 2017b). Moreover, the nutrient cycles occurring at these terrestrial-aquatic interfaces (TAIs) influence locally important ecosystem services like contaminant fate and transport and water quality (Conrads and Darby, 2017; Vidon et al., 2010). While coastal soil C stocks are being increasingly quantified (Hinson et al., 2017; Holmquist et al., 2018; Krauss et al., 2018), the impact of tidally-driven salinity gradients on molecular level features of the soil-C pool and the processes driving soil organic matter (OM) cycling are poorly studied (Barry et al., 2018; Hoitink et al., 2009; Sawakuchi et al., 2017; Ward et al., 2017b). This is particularly true in settings with low freshwater inputs that allows for significant seawater intrusion compared to large river systems (Hoitink and Jay, 2016).

While multiple processes impact TAI carbon pools (e.g., tidal-inputs, *in situ* root exudates and litter inputs, decomposition processes), there is some indication that microbial diversity and composition impact soil C storage and mineralization (Mau et al., 2015; Trivedi et al., 2016). This points to the intriguing possibility that processes governing microbial community assembly may be associated with OM chemistry, but evaluations of such associations are lacking. This lack of mechanistic knowledge combined with significant ecosystem heterogeneity in biogeochemical function across salinity gradients (more below), highlights a need to understand how molecular-level processes vary with seawater exposure along coastal TAIs. Doing so will help enhance predictive models of TAI biogeochemistry that can be potentially included in ecosystem models to more accurately represent the role of TAIs in the broader Earth system (U.S. DOE., 2017).

Modeling of coastal TAIs is currently impeded by poor knowledge of the mechanisms underlying salinity-driven variation in biogeochemical function of associated soils. Previous studies have evaluated function primarily as carbon dioxide (CO_2) and methane (CH_4) flux measurements from soil, and/or soil OM concentrations measured as bulk soil C, percent OM and porewater dissolved organic C (DOC) concentrations in large scale coastal plain river systems. Results from field-based natural salinity gradient studies, long-term field manipulations of salinity exposure, and lab-based incubation studies subjecting soils to varying levels of salinity broadly show the following trends: increases in CO_2 and decreases in CH_4 emissions in freshwater soils exposed to increasing salinity (Chambers et al., 2011, 2013, 2014; Liu et al., 2017; Marton et al., 2012; Neubauer et al., 2013a; Steinmuller and Chambers, 2018; Weston et al., 2006, 2011), and negative relationships between salinity and emissions of both CO_2 and CH_4 from soils with a history of exposure to high salinity (Chambers et al., 2013; Herbert et al., 2018;

96 Neubauer et al., 2005, 2013a; Weston et al., 2014) (also see Table S1). Exceptions have been
97 observed where CO_2 emissions decreased in historically freshwater coastal wetland soils exposed
98 to seawater (Ardón et al., 2018; Herbert et al., 2018), with Ardón et al. (2018) also reporting an
99 increase in CH_4 flux with salinity and Steinmuller and Chambers (2018) reporting no change in
100 CH_4 flux with increasing salinity. These observations suggest that microbial activity usually
101 increases with salinity in soils that were not previously exposed to saline conditions, while
102 simultaneously indicating reduced microbial activity with increasing salinity in soils that have a
103 historical exposure to high salinity. In contrast to relatively consistent responses of gas fluxes to
104 changes in salinity, there are strong inconsistencies in DOC responses, including no change
105 (Weston et al., 2006, 2011, 2014), increased DOC (Chambers et al., 2014; Tzortziou et al.,
106 2011), and decreased DOC (Ardón et al., 2016, 2018; Liu et al., 2017; Yang et al., 2018) with
107 increasing salinity.
108
109 Relatively consistent gas flux responses to changes in salinity combined with inconsistent DOC
110 responses to elevated salinity suggests decoupling between biogeochemical rates and the
111 concentration of DOC. This apparent decoupling between the size of the C pool (in this case the
112 concentration of DOC) and microbial activity suggests that C biogeochemistry is influenced by
113 salinity-exposure history, which in turn influences nutrient resources available to soil microbial
114 communities. Specifically, any systematic shifts in soil organic carbon (SOC) chemistry (along
115 salinity gradients) that cannot be observed with bulk C measurements (e.g., changes in chemistry
116 that reduce C bioavailability; Neubauer et al. 2013) may result in unpredictable carbon fluxes.
117 Moreover, bulk C content can show no change across gradients of salinity (Neubauer et al.,
118 2013a) and may fail to capture an integrated view of microbially driven C cycling dynamics at
119 TAIs. In contrast, detailed molecular-level evaluation of SOC composition can provide a more

120 mechanistic view of OC transformations, relative to bulk measures of C content or gas flux
121 measurements.
122
123 Despite its potential importance, a detailed understanding of the characteristics of soil organic
124 compounds (Zark and Dittmar, 2018) and their association with microbial communities in
125 coastal TAIs is currently not available. Nonetheless, we derived a series of expectations by first
126 recognizing that (1) our study system is a historically freshwater system, only recently being
127 exposed to salt water due to removal of a culvert in 2014 (see Methods), and (2) microbial
128 activity increases with increasing salinity in historically freshwater systems (Nyman and
129 Delaune, 1991; Smith et al., 1983; Tzortziou et al., 2011). In addition, it is generally expected
130 that microbes preferentially degrade compounds with higher nominal oxidation states (NOSC) or
131 lower Gibbs Free Energy (ΔG^0_{Cox}) due to greater thermodynamic favorability (Boye et al., 2017;
132 Graham et al., 2017, 2018, Stegen et al., 2018, Ward et al., 2017a). An important caveat is that
133 factors such as redox state, physical protection, mineral associations, and microbial community
134 composition can alter this pure chemistry-based expectation (Schmidt et al., 2011). As a simple
135 point of departure, however, we assume that OM reactivity follows NOSC, thereby leading to
136 our first expectation/hypothesis: the average ΔG^0_{Cox} of OM will increase with increasing salinity
137 as organic compounds with greater thermodynamic favorability are preferentially depleted
138 (LaRowe and Van Cappellen, 2011) due to microbial activity increasing with salinity. To
139 develop our second hypothesis we note that actively growing microbial communities are known
140 to enhance biochemical transformations and generate heteroatom containing organic molecules
141 [sulfur (S), nitrogen (N) and phosphorus (P)] (Guillemette et al., 2018; Koch et al., 2014;
142 Ksionzek et al., 2016); therefore greater heteroatom content and more biochemical
143 transformations are expected with increasing salinity. Our third hypothesis is based on

144 microorganisms adapting to saline conditions through the production or sequestration of
145 osmolytes (Gouffi et al., 1999b, 1999a; Sleator and Hill, 2002), a strategy that may require
146 microbes to break down organic molecules to extract N (i.e., N mining). We therefore
147 hypothesize increases in N-containing biochemical transformation with increasing salinity. Our
148 fourth hypothesis is based on the observation that soils in saturated environments like floodplains
149 are expected to be less oxygenated and can also receive deposition of marine/algal derived OM
150 and suspended sediments during tidal flooding. These factors can result in OM having lower
151 oxygen to carbon (O/C) and higher hydrogen to carbon (H/C) ratios as compared to upland soils
152 (Seidel et al., 2016; Tfaily et al., 2014; Ward et al., 2019b). We therefore hypothesize a greater
153 relative abundance of lipid- and protein-like and less lignin- and tannin-like compounds in the
154 floodplain soils, relative to upland (i.e., drained) soil.
155 While we expect systematic shifts in C chemistry across landscape scale salinity gradients, an
156 open question is the degree to which C chemistry is associated with ecological assembly
157 processes governing composition of microbial communities. Soil microorganisms transform soil
158 C, but there is limited evidence of direct links between microbial community assembly processes
159 and molecular-level soil C chemistry (Kubartová et al., 2015; Rocca et al., 2015; Trivedi et al.,
160 2016; van der Wal et al., 2015). Assembly processes, broadly divided into deterministic
161 (systematic impacts of selection) and stochastic (unstructured demographic events) factors,
162 function over space and time to influence the composition of microbial communities, which in
163 turn mediate biogeochemical cycles (Graham et al., 2016, 2017b; Nemergut et al., 2013a; Stegen
164 et al., 2015). Deterministic processes lead to selection of microbial communities resulting from
165 different organisms having different levels of fitness for a given set of environmental conditions
166 including abiotic variables and biotic factors related to organismal interactions while stochastic
167 processes include random birth/death events and unstructured dispersal. The relative influences

of stochastic and deterministic processes can be inferred from phylogenetic distances among microbial taxa using ecological null models. This approach has been widely employed to understand community assembly processes in surface and subsurface microbial ecology (Caruso et al., 2011; Dini-Andreote et al., 2015; Graham et al., 2017a, 2018; Sengupta et al., 2019; Stegen et al., 2012). Furthermore, a recent study used ecological simulation modeling to show that communities experiencing increased rates of dispersal are linked to reduced biogeochemical functioning (Graham and Stegen, 2017). This, combined with evidence of increasing microbial activity with increasing salinity (discussed above) leads to a fifth hypothesis that the influence of deterministic selection will progressively increase with salinity due to increased microbial activity.

Analyses of specific chemical biomarkers such as lignin phenols, amino acids, and lipids have been used in soils, sediments, and water to quantitatively evaluate the provenance of terrestrial-derived OM (Hedges et al., 1997), the reactivity of OM as it travels through a soil column (Shen et al., 2015), and microbial community composition (Langer and Rinklebe, 2009), respectively. While biomarkers provide quantitative details on OC cycling, they generally represent a small fraction of the total OM pool, thus, non-targeted approaches such as analysis of thousands of organic molecules via Fourier Transform Ion Cyclotron Resonance Mass Spectrometry (FTICR-MS) have become increasingly widespread for determining molecular-level organic compound signatures (Rivas-Ubach et al., 2018) across a variety of terrestrial (Bailey et al., 2017; Simon et al., 2018), aquatic/marine (Lechtenfeld et al., 2015), and transitional settings such as hyporheic zones (Graham et al., 2017a, Stegen et al., 2016, 2018) and river-ocean gradients (Medeiros et al., 2015).

191 The objective of the current study was to test the following hypotheses in a coastal forested
192 floodplain and adjacent upland forest: (1) mean Gibbs Free Energy of organic compounds will
193 increase with increasing salinity; (2 and 3) biochemical transformations, heteroatom content, and
194 N-containing biochemical transformation will increase with increasing salinity; (4) lipid- and
195 protein-like compound classes will be more prevalent in the floodplain soils compared to upland
196 soils in which lignin- and tannin-like molecules will dominate; and (5) microbial community
197 assembly processes will be increasingly deterministic as salinity increases. The chemical forms
198 of C in these soils were characterized using FTICR-MS. We also employed ecological null
199 model analysis to evaluate the relationship between C chemistry and the influences of assembly
200 processes on microbial communities. Based on our results, we propose a conceptual model of
201 organic C processing in a coastal forested floodplain in which landscape-scale gradients in C
202 chemistry are driven by a combination of spatially-structured inputs and salinity-associated
203 metabolic responses of microbial communities that are independent of community composition.
204
205 **2. Materials and Methods**:
206 **2.1 Site Information and Soil Sampling**
207 Soils along a coastal watershed draining a small first order stream, Beaver Creek, in the
208 Washington coast were selected for this study (46.907, -123.976). Beaver Creek is a tributary of
209 Johns River and experiences a high tidal range of up to 2.5 m that extends midway up the first-
210 order stream's channel and inundates the landscape in its floodplains. The confluence of Beaver
211 Creek and Johns River is roughly 2.5 km upstream of the Grays Harbor estuary and 14.5 km
212 from the Pacific Ocean, and experiences variable exposure to saline waters at high tide (Fig. 1).
213 Surface water salinity near Beaver Creek's confluence ranges from 0 practical salinity unit (psu)
214 at low tide to 30 psu at high tide during dry periods (Ward, unpublished). The Beaver Creek

watershed is 3.8 km^2. The tidal floodplain makes up 0.5 km^2 of this total watershed area. Tidal exchange to Beaver Creek was restored after 2014 when a culvert near the creek's confluence with Johns River was removed (Washington Department of Fish and Wildlife, 2019). Due to the minimal past tidal exchange, the floodplain is dominated by gymnosperm trees (*Picea sitchensis*) that are rapidly dying since the culvert removal (Ward et al., 2019a). The headwaters (before the river channel forms) is a sparsely forested, perennially inundated freshwater wetland with tidal exchange blocked by a beaver dam, followed downstream by a densely forested setting along the river channel. Towards Beaver Creek's confluence salt tolerant grasses such as *Agrostis stolinifera* (Creeping bentgrass) become the most dominant land cover as forest cover becomes more sparse. The watershed's hillslope/uplands is dominated by *Tsuga heterophylla* (Western hemlock) trees, but *Picea sitchensis* (Sitka spruce) are also present.

Two sampling transects perpendicular to the river along the up/downstream salinity gradient were established and represent a high salt exposure site close to the culvert breach location and a moderate salt exposure site upstream of the high salt exposure site. These transects represent a coastal forested wetland with brackish (semi-salty) groundwater and consisted of three terrestrial sampling points at each transect extending from the riparian zone to the beginning of the steep upslope. An additional soil sampling point ~20m uphill from the moderate salt exposure site transect served as a purely terrestrial upland endmember. The soils are Andisols and floodplain transects represented hydric soils classified as Ocosta silty clay loam while the upland site was a well-drained Mopang silt loam. The transects experience periodic inundation episodes which result in surface pooling of tidal water, which can be up to ~1m deep. The water table varies seasonally and during tidal cycles and inundation events, ranging from about 0 to 1m below the ground surface (Ward, unpublished).

Soil samples were collected in triplicate at each of the seven locations(Fig. 1) [BC2 (2.94 m), BC3 (2.63 m), and BC4 (2.82 m) at the high-salt exposure transect, locations BC12 (2.82 m) , BC13 (2.67 m), BC14 (3.07 m) at the moderate salt exposure transect, and BC15 (13.45 m) as upland site]. The high-salt exposure transect was 230 m from the moderately saline transect (0.6 km from the confluence of Beaver Creek with Johns River), and each site at the transect was ~35 m apart from the next. Each transect is ~ 90 m. For data comparison's sake, we classify BC2, BC3, BC12, and BC13 as **floodplain** sites while BC4 and BC14 are further **inland** and ~75 m away from the creek at the base of the densely wooded hillslope. Soil samples for molecular characterization studies were collected at two depths—shallow (10 cm) and deep (19-30 cm). Samples were collected from the face of soil pits using custom mini-corers, placed into sterile amber glass vials, purged with N_2 to maintain anaerobic conditions, frozen in the field within an hour at -20 °C, and stored at -80 °C on return to the lab. Bulk samples were collected for soil physicochemical characterization including texture classification with hydrometer method after organic matter removal, dry combustion with direct measure of total C, nitrogen (N) and sulfur (S) by Elementar Macro Cube, plant-available N as ammonium-nitrogen (NH_4-N) and nitrate-nitrogen (NO_3-N) with 2M KCl quantified on Lachat QuikChem 8500 Series 2 (Hach, Loveland Colorado) as colorimetric reaction, pH measured as 1:2 soil:water slurry measured with Hanna benchtop meter, specific conductivity measured as 1:5 soil:water slurry measured on MP-6p portable specific conductivity meter, gravimetric water content measured after drying soils at 105 °C for 48 hours, and bulk density and porosity measured as per standard methods (Sumner,1999). Soil chemical characterization was performed on air-dried sieved soils.

2.2 FTICR-MS solvent extraction and data acquisition

Soil organic compounds were extracted using a sequential extraction protocol with polar {water (H_2O)} and non-polar {chloroform ($CHCl_3$) (representing mineral-bound fraction)} solvents per standardized protocols (Graham et al., 2017a; Tfaily et al., 2015, 2017), which extract about 2-15% of total organic carbon and represent both polar and non-polar soil organic carbon fractions. Importantly, our analyses do not depend on extracting a large portion of the C found within a given soil sample. Instead, we assume that the extracted fraction is a representative sub-sample. This is a standard approach and assumption made in any study examining metabolites or other types of organic molecules in soil. Briefly, extracts were prepared by adding 5 ml of MilliQ H_2O to 5 g of each of the replicate samples in sterile polypropylene centrifuge tubes (Genesee Scientific, San Diego, USA) suitable for organic solvent extractions and shaking for 2 h on a Thermo Scientific LP Vortex Mixer. Samples were removed from the shaker and centrifuged for 5 minutes at 6000 rpm, and the supernatant was removed into a fresh centrifuge tube. This step was repeated two more times, with the 15 ml supernatant pooled for each sample and stored at -80 °C until further processing. Next, Folch extraction with $CHCl_3$ and CH_3OH was performed for each soil pellet left over from the water extraction. Folch extraction entailed adding 2 ml CH_3OH, vortexing for 5 seconds, adding 4 ml $CHCl_3$, vortexing for 5 seconds, followed by 0.25 ml of MilliQ H_2O. The samples were shaken for 1 hr and another 1.25 ml MilliQ H_2O was added and left overnight at 4 °C to obtain bi-layer separation of upper (polar) layer and the lower (non-polar) layer. The extracts were stored in glass vials at -20 °C until ready to be used. The water soluble organic carbon (WSOC) fraction was further purified using a sequential phase extraction (SPE) protocol to remove salts as per Dittmar et al., 2008. Briefly, samples were acidified to pH 2 with 85% phosphoric acid. The samples were passed through Bond Elut PPL cartridges (©Agilent Technologies) that were preactivated with CH_3OH. The cartridges were washed 5x with 10mM HCl followed by nitrogen-gas drying. Next, 1.5 ml CH3OH, a solvent that is

compatible with direct analysis on the FTICR-MS, was used to elute the samples from the cartridge thus avoiding an additional evaporation step and reducing the chance of losing volatile organic compounds and saving time during sample preparation. While SPE by PPL have shown not to be very effective in extracting several major classes of DOM compounds that had high ESI efficiencies, such as carboxylic acids and organo-sulfur compounds, and that out-competed other less functionalized compounds (e.g., carbohydrates) for charge in the ESI source (Tfaily et al., 2012), it is highly efficient for marine and estuary DOM samples as it provides complete desalination of the sample. Loss of small molecules such as simple sugars is known to happen during SPE however this is not a concern for the current study as FTICR-MS is sensitive for compounds above 200 Da. In this study, SPE by PPL isolates a major DOM fraction, that is salt-free, allowing for DOM characterization by FTICR-MS(Dittmar et al., 2008b) . While we didn't measure SPE extraction efficiency for this study, it usually ranges between 40 and 62 % depending on the sample (Dittmar et al. 2008). Samples that are collected from the same ecosystem have shown to have similar extraction efficiency. The extracts were injected into a 12 Tesla Bruker SolariX FTICR-MS located at Environmental Molecular Sciences Laboratory (EMSL) in Richland, WA, USA. Detailed methods for instrument calibration, experimental conditions, and data acquisition are provided in Graham et al., 2017a and Tfaily et al., 2017.

2.3 FTICR-MS Data Processing

One hundred forty-four individual scans were averaged for each sample and internally calibrated using an organic matter homologous series separated by 14 Da ($–CH_2$ groups). The mass measurement accuracy was less than 1 ppm for singly charged ions across a broad m/z range (100 - 900 m/z). Data Analysis software (Bruker Daltonik version 4.2) was used to convert raw spectra to a list of m/z values applying FTMS peak picker module with a signal-to-noise ratio

(S/N) threshold set to 7 and absolute intensity threshold to the default value of 100. Chemical formulae were then assigned using in-house software following the Compound Identification Algorithm, proposed by Kujawinski and Behn (2006), modified by Minor et al. (2012), and described in Tolić et al. (2017). Peaks below 200 and above 900 were dropped to select only for calibrated and assigned peaks. Chemical formulae were assigned based on the following criteria: S/N >7, and mass measurement error < 0.5 ppm, taking into consideration the presence of C, H, O, N, S, P, and excluding other elements. Detected peaks and associated molecular formula were uploaded to the in-house pipeline FTICR R Exploratory Data Analysis (FREDA) to obtain: (i) NOSC values from elemental composition of the organic compounds(Koch and Dittmar, 2006, 2016), (ii) thermodynamic favorability of the compounds calculated as Gibbs Free Energy for the oxidation half reactions of the organic compounds (ΔG^0_{Cox}) based on the equation ΔG^0_{Cox} = 60.3-28.5*NOSC (LaRowe and Van Cappellen, 2011), where a higher ΔG^0_{Cox} indicates a less thermodynamically favorable species than a lower value (LaRowe and Van Cappellen, 2011), (iii) abundance of compounds grouped into elemental groups (CHO, CHOS, CHOP, CHNOS, CHNO, CHNOP, CHOSP, and CHNOSP), and (iv) abundance of compound classes (carbohydrate-, lipid-, protein-, amino sugar-, lignin-, tannin-, condensed hydrocarbon-, and unsaturated hydrocarbon-like) based on molar H:C and O:C ratios of the compounds (Bailey et al., 2017).

Biochemical transformations potentially occurring in each sample were inferred from the FTICR-MS data by comparing mass differences in peaks within each sample to precise mass differences for commonly observed biochemical transformations (Breitling et al., 2006; Stegen et al., 2018b). The ultra-high mass accuracy of FTICR-MS allows precise mass differences to be

counted for the number of times each transformation was observed within each sample. Briefly, the mass difference between m/z peaks extracted from each spectrum were compared to commonly observed mass differences associated with 92 common biochemical transformations provided in previous publications (Graham et al., 2017a; Stegen et al., 2018c). All possible pairwise mass differences were calculated within each extraction type for each sample, as shown in Fig. 2 where a mass difference of 2.01586 indicates a hydrogenation reaction. It is important to note that direct injection electrospray ionization (ESI) FTICR-MS approach cannot distinguish between isomers such as in the case of a mass difference corresponding to a loss of grain of either glucose, fructose and galactose.

2.4 Ecological Modeling

Null modeling was used to estimate influences of ecological processes on microbial community composition from rarefied (10000) 16S rRNA amplicon data, providing estimates of microbial community composition and phylogenetic relatedness. The extraction, purification, and sequencing of soil microbial DNA were performed according to published protocol (Bottos et al., 2018). Briefly, microbial DNA was extracted from 0.25 g of each sample using the MoBio Power Soil DNA Isolation Kit and cleaned-up using Zymo ZR-96 Genomic DNA Clean and Concentrator-5 kit (Zymo Research Corporation, Irvine, CA) as per manufacturer instructions. The V4 region of the 16S rRNA gene was amplified by polymerase chain reaction and amplicons sequenced on Illumina MiSeq using the 500 Miseq Reagent Kit v2 (Illumina Inc., San Diego, CA) according to manufacturer's instructions. Sequence pre-processing, operational taxonomic unit (OTU) table construction and phylogenetic tree building were performed using an in-house pipeline, HUNDO (Brown et al., 2018). Briefly, sequence demultiplexing was done using EA-Utils (Aronesty, 2013), reads quality filtered with BBDuk2 (BBMap, 2014), and merged using

USEARCH (Edgar, 2010). Sequence de-replication and clustering was performed, taxonomy was assigned to operational taxonomic unit (OTU), and chimeras were removed using USEARCH. Raw sequences are archived at NCBI (BioProject PRJNA541992) at the following website: https://dataview.ncbi.nlm.nih.gov/object/PRJNA541992?reviewer=b55qu29emsinvk3udb2rmuffqh.

Null modeling was performed as described previously (Stegen et al., 2013, 2015) with a total of 35 samples to estimate relative influences of deterministic and stochastic selection processes. Briefly, samples that passed quality control and rarefaction threshold were evaluated for pairwise phylogenetic turnover between communities, calculated as the difference between the observed values of the β-mean-nearest-taxon-distance (βMNTD) and mean of the null βMNTD distribution, in units of standard deviation (see Stegen et al. 2012 for details). The difference between observed βMNTD and the null distribution is known as the β-nearest taxon index (βNTI). Deterministic assembly process are inferred to be dominant when βNTI > 2 or <-2. When βNTI is > 2 it indicates that deterministic processes have driven community composition apart, which is referred to as 'variable selection' (Dini-Andreote et al. 2015; Stegen et al. 2015). When βNTI is < -2 it indicates that deterministic processes have caused community composition to be similar between a given pair of communities, which is referred to as 'homogeneous selection' (Dini-Andreote et al. 2015; Stegen et al. 2015).

Pairwise community comparisons that do not deviate significantly from the null distribution (i.e., 2> βNTI>-2) indicate the dominance of stochastic processes (including homogenizing dispersal and dispersal limitation), or a scenario in which neither deterministic or stochastic processes

dominate (referred to as undominated). Homogenizing dispersal occurs when rate of dispersal between two communities result in community composition becoming relatively similar between the two communities, and potentially overwhelming other assembly processes (e.g., variable selection). Dispersal limitation is the result of very low rates of organismal exchange between communities, which can result in the stochastic divergence of community composition through the accumulated outcomes of random birth/death events (i.e., ecological drift).

For pairwise comparisons that were not associated with deterministic processes (i.e., when 2> βNTI>-2), we use a second null model to test for influences of homogenizing dispersal or dispersal limitation. This second null model is referred to as RC_{bray}, and like βNTI accounts for variation in OTU relative abundances (Stegen et al., 2013, 2015). Homogenizing dispersal was inferred to be the dominant process for pairwise comparisons characterized by 2>βNTI>-2 and RC_{bray}<-0.95. Dispersal limitation was inferred to be the dominant process for pairwise comparisons characterized by 2>βNTI>-2 and RC_{bray}>0.95. The relative influences of variable selection, homogeneous selection, dispersal limitation, and homogenizing dispersal were quantified by the fraction of pairwise comparisons that were dominated by each ecological process (Stegen et al. 2013). The relative contribution of scenarios in which the system was undominated was estimated as the fraction of pairwise comparisons characterized by 2>βNTI >-2 and 0.95>RC_{bray}>-0.95 (Stegen et al. 2015).

2.5 Statistical Methods

Samples were separately analyzed for WSOC and $CHCl_3$ fractions. Within each solvent fraction, samples were grouped into shallow or deep depths. FTICR-MS dependent metrics including ΔG^0_{Cox}, and relative abundance of compound classes, total transformations, nitrogen-containing

transformations, and organic nitrogen containing compounds were regressed against specific conductivity. Regressions were considered significant if $R^2 \geq 0.50$ and $p \leq 0.05$. The transformation profiles were also regressed with the community assembly processes to determine the relationship between deterministic/stochastic processes and organic compound transformations. Mantel tests were used to evaluate similarity between BNTI matrix and Sorensen matrix of peak presence/absence. The Sorensen distance matrices of WSOC and $CHCl_3$ peaks were regressed against measured variables (soil physicochemical properties) and community assembly process-variables to determine correlations. Finally, a redundancy analysis –based stepwise model building with forward model choice was performed to determine variation in the Hellinger-transformed water-fraction peaks and $CHCl_3$ fraction peaks as explained by explanatory variables (which included measured soil variables, modeled community assembly variables, and categorical variables depth and location). All statistical analyses were performed in the statistical computing language R version 3.5.3 (R Development Core Team, 2019).

3. Results

3.1 Soil characterization. The percent of total soil C (%C) in the shallow soils (26.3 $\pm$ 8.3%) was higher than the deeper soils (4.0 $\pm$ 1.3%) for the lowland soils (i.e. "floodplain" and "inland" sites), while the upland site had an average %C of 7.4 $\pm$ 0.27% at 10 cm and 2.13 $\pm$ 0.06% at 30 cm (Table S2). No significant relation was observed between %C in the shallow inland and floodplain soils along the salinity gradient. The percent of total soil N (%N) of the shallow soils were higher (1.5 $\pm$ 0.40%) than the deeper soils (0.4 $\pm$ 0.08%) for the lowland soils and co-varied with %C (r^2=0.95). The pH of all soils were acidic (5.64 $\pm$ 0.70). The concentrations of NH_4-N and NO_3-N showed a consistent trend where NH_4-N was 1-2 orders of magnitude higher than

NO_3-N in all samples. The specific conductivity (used as a measurement of salinity in this study) of the shallow soils ranged from 206-866 ($\pm$12) $\mu S\ cm^{-1}$ in the lowland soils to 43$\pm$5 $\mu S\ cm^{-1}$ in the terrestrial end-member site. The deep soils exhibited specific conductivity ranging from to 148-524 ($\pm$11) $\mu S\ cm^{-1}$ in the lowland soils to 29.2 $\pm$8 $\mu S\ cm^{-1}$ in the terrestrial end-member site. Texture analysis revealed a broad range of sand (4.1 – 40 %), silt (21.4 – 57.9%), and clay (28.6 – 64.8%) fractions.

3.2 Thermodynamics, compound classes, and elemental composition. The calculated ΔG^0_{Cox} WSOC (Table S3)in shallow soils was consistent with our hypothesis of decreasing thermodynamic favorability with increasing conductivity. Average ΔG^0_{Cox} ranged from 53-71 kJ mol C^{-1} (R^2= 0.78, p <0.00001), while no significant relationship between ΔG^0_{Cox} and specific conductivity was observed for WSOC fraction in the deeper soils (averaging 51-54 kJ mol C^{-1}) for the floodplain and inland samples (Fig. 3). The upland site had significantly higher average ΔG^0_{Cox} (67-70 kJ mol C^{-1}) than the soils near the lowland. The ΔG^0_{Cox} ($CHCl_3$) at both depths (Table S4) was higher than the water fractions (ranging between 96-105 kJ mol C^{-1}) but did not show significant relationship with respect to specific conductivity.

Peak profiles for each solvent extraction showed distinct compound classes in the van Krevelen space, with peaks assigned to specific compound classes according to rules outlined in Kim et al., 2003 and modified by Bailey et al., 2017. The WSOC fraction was dominated by compounds classified as protein-, amino sugar-, lignin-, condensed hydrocarbon-, carbohydrate-, and tannin-like compounds (Table 1), while the $CHCl_3$ fraction had relative high abundances (75% and higher) of lipid-like compounds (data not shown). A modest percentage of peaks (11-17%) did not have classes assigned. Unique and common peaks extracted in the WSOC fraction in samples

grouped according to their landscape position and depth [four sites in the floodplain (BC2, BC3, BC12, and BC13), two sites inland (BC4 and BC14), and one upland site (BC15)] are represented as H/C to O/C ratio of the compounds ($p = 0.05$) in Fig. S1.

The shallow WSOC in the floodplain had greater relative abundance of unique lipid (28%)- and protein (25%)-like compounds with relatively high H:C and low O:C ratios as compared to the upland site (BC15), which had an 31%, 30%, and 19% unique peaks representing lignin-, tannin-, and carbohydrate-like compounds respectively. About 23% of peaks were common in both groups, including lignin- and condensed hydrocarbon-like compounds (Fig. S1a). Between the floodplain and the inland samples, high H:C and low O:C ratios representing 47% lipid-, 38% protein-, and 22% amino sugar-like peaks were uniquely present in the floodplain samples (Fig. S1b). The inland shallow soils had 19% unique higher H:C peaks representing condensed hydrocarbon-like compounds compared to 1.2% in the upland soil , though most of the compound classes were observed at both locations (Fig. S1c). Linear regression with specific conductivity profiles showed significant positive correlation with amino sugar-, protein-, lipid-, and unsaturated hydrocarbon-like compounds, while condensed hydrocarbon-like compounds were significantly negatively correlated (Table S5)

For the deep soils, the upland site had 32% unique peaks with relatively high H:C ratios and low O:C ratios commonly associated with unsaturated hydrocarbon-like compounds, as compared to the 0.7% in the floodplain which had higher prevalence of unique peaks representing condensed hydrocarbon (36%)-, and tannin-like (35%) compounds (Table 1, Fig. S1d). The floodplain vs inland samples had thrice as many unique peaks with high H:C and low O:C ratios representing lipid-like compounds in the floodplain samples. Comparisons between inland and upland end-member samples revealed 43% and 37% unique peaks representing low H:C and high O: C ratio

hydrocarbon- and tannin-like compounds respectively in inland samples, while 32%, 14% 9%, and 12% of unique peaks were matched to unsaturated hydrocarbon-, lipid-, protein-, and amino sugar-like compounds respectively in the latter (Table 1, Fig. S1e, f). No significant relationship between compound-class abundances and specific conductivity was observed (Table S5). For the $CHCl_3$ fraction, peaks of lipid-like and unsaturated hydrocarbon-like compounds were observed to be common in all samples (data not shown) and regressions against specific conductivity were not significant for the compound classes.

Compositional differences of the organic compounds showed variable heteroatom abundances, with cumulative heteroatom abundance decreasing with increasing salinity (R^2=0.43, p = 0.009) for shallow fraction of the WSOC. For the WSOC fraction, heteroatom abundance of CHOP (R^2 = 0.61) and CHNOP (R^2 = 0.50) containing compounds was consistent with our hypothesis and significantly ($p < 0.01$) increased, while CHNOS (R^2 = 0.66), and CHNOSP (R^2 = 0.62) abundances were inconsistent with our hypothesis and significantly decreased with increasing specific conductivity. The elemental composition of the WSOC compounds for deep soils did not show any significant trend with respect to conductivity. For the $CHCl_3$ fraction, relative abundance of CHNOP in the shallow soils significantly decreased with specific conductivity (R^2 = 0.57, $p < 0.01$).

3.3 Transformation profiles. In contrast to our expectations, the number of transformations decreased with increasing salinity in the water fraction of shallow soils (R^2= 0.60, $p < 0.01$) (Fig. 4a, Table S3). We also evaluated N-containing transformations and the abundance of N-containing compounds in the system. Total nitrogen-containing transformations also decreased significantly with conductivity but the correlation was not as strong (R^2= 0.40, $p < 0.01$). Total N

502 containing compounds (Fig. 4b, Table S3) as well as their relative abundance decreased
503 significantly (R^2= 0.74, $p < 0.01$), with increasing conductivity in the shallow soils for water
504 fraction.
505
506 **3.4 Ecological processes impacting community composition**
507 Null modeling revealed that microbial community assembly processes were influenced by
508 variable selection (βNTI>2), homogenous selection (βNTI<-2), dispersal limitation (2>βNTI>-2
509 and RC_{bray}>0.95), homogenizing dispersal (2>βNTI>-2 and RC_{bray}<-0.95), and undominated
510 processes (2>βNTI>-2 and 0.95> RC_{bray}>-0.95) (Fig. 5). Dispersal limitation had the greatest
511 influence, responsible for 54% of the variation in community composition. The lowest signal
512 was for homogenizing dispersal (1%), and the signal for homogenous selection (23%) was higher
513 than variable selection (9%). Together, deterministic processes (variable selection plus
514 homogeneous selection) were responsible for 32% of the variation in community composition,
515 with 55% contributed by stochastic processes (dispersal limitation plus homogenizing dispersal).
516 Variation not accounted by dispersal or selection (i.e., influenced by a mixture of processes)
517 accounted for the remaining signal (23%). Consistent with influences from both stochastic and
518 deterministic processes, βNTI relationships with environmental variables were significant ($p <$
519 0.05 by Mantel test), but relatively weak (r=0.46 for pH and r=0.31 for specific conductivity)
520 (Fig. S2).
521
522 To evaluate associations between microbial community assembly processes and chemistry,
523 process estimates were regressed against features of the organic C profiles. Deterministic
524 processes decreased (Fig S3a) while community assembly processes influenced by non-
525 deterministic processes increased with increasing number of transformations of organic

compounds (Fig. S3b), although no strong relationships were observed between assembly processes and transformations ($p = 0.027$, $R^2 = 0.11$ for deterministic/non-deterministic processes, $p = 0.475$, $R^2 = 0.015$ for variable selection, $p = 0.054$, $R^2 = 0.10$ for homogenous selection, $p = 0.514$, $R^2 = 0.013$ for dispersal limitation, and $p = 0.627$, $R^2 = 0.007$ for homogenizing dispersal). No significant relationships were observed between assembly processes and the number of N-containing transformations. Sorensen dissimilarity values based on the detected FTICR peaks for the water fraction were correlated with distance matrices of measured environmental variables and estimates of community assembly processes. Weak positive correlations were observed with NH_4-N ($r = 0.28$), pH ($r = 0.27$), specific conductivity ($r = 0.41$), NO_3-N, silt, and clay ($r = 0.30$) while for the $CHCl_3$ fraction, weak positive correlations were observed with specific conductivity and NO_3-N ($r = 0.26$) (Fig. S4). A Mantel test of FTICR Sorensen dissimilarity vs βNTI values yielded a significant relationship ($r = 0.213$, $p = 0.003$) for water fraction but not for $CHCl_3$ fraction ($r=0.076$, $p = 0.152$). The stepwise model building yielded a combination of five variables that were weakly associated with the composition of water fraction peaks ($p=0.026$, adj. $R^2 = 0.217$), including sand, dispersal limitation, NH_4-N concentration, specific conductivity, and location. The model explaining variation in the composition of $CHCl_3$ fraction peaks was non-significant ($p = 0.1$, adj. $R^2 = 0.05$).

4. DISCUSSION

Sea level rise is increasing the inland extent of tides and exacerbating storm surge, resulting in greater salinity intrusion and altered ecosystem behavior across coastal TAIs (Conrads and Darby, 2017; Ensign and Noe, 2018; Langston et al., 2017; McCarthy et al., 2018; Neubauer et al., 2013b). Site-driven variations in the responses of bulk soil biogeochemical processes (i.e.,

gas flux and DOC release) to elevated salinity suggests potentially important influences of underlying features such as C chemistry and microbial communities. To provide a foundation for understanding the role of C chemistry and microbial communities on biogeochemical cycling in coastal soils, we evaluated associations among a landscape-scale soil salinity gradient, molecular-level soil carbon chemistry, and microbial community assembly processes in order to ultimately inform future improvements for predictive models. In soils associated with a coastal first-order drainage basin, we observed salinity-associated gradients in soil organic carbon fractions that were not associated with microbial community assembly processes. Our results are consistent with C chemistry being driven by a combination of spatially-structured inputs driven by landscape structure (i.e., terrestrial inputs further inland, marine inputs further downstream) and salinity-associated metabolic responses of microbial communities that are independent of microbial community composition. An important caveat is that we did not measure microbial metabolism, but instead infer an influence of microbial metabolism due to microbial composition being independent of C chemistry. To more directly evaluate these inferences, additional work is needed that focuses on quantifying inputs (e.g., via stable isotopes) and measuring microbial metabolism (e.g., via metatranscriptomics). Future work should also use tools like Nuclear Magnetic Resonance and Gas Chromatograph-Mass Spectrometry to evaluate low molecular weight OC (like those contributed by root exudates) vary with salinity.

4.1 Molecular characterization reveals chemical gradients not seen in the bulk C pool

The systematic shifts observed in the molecular signatures compared to non-significant changes in bulk C chemistry shows that molecular-level investigations are particularly relevant to process-based resolution of C biogeochemistry. The absence of bulk C signals mimicking molecular C signals parallel studies indicating rapid change in molecular constituents of the soil

C pool with no change in gross C content (Graham et al., 2018; Reynolds et al., 2018). A faster turnover time of C has been observed in microbial biomass as compared to bulk soil organic matter (Kramer and Gleixner, 2008), which is likely to impact microbe-mediated biochemical C transformations and lead to chemically complex heterogeneous C signatures likely to be missed in bulk analysis (Tfaily et al., 2015). The systematic shifts in chemical characteristics of soil carbon fractions exhibited by samples at the shallow depth suggests that organic C compound pools in shallower soil depths are sensitive to salinity gradients while deeper depth signatures do not vary systematically across the landscape. The landscape gradient observed in the shallow soils is likely influenced by a combination of reduced litterfall due to trees suffering under recent increases in salinity, changing understory vegetation, and algae-rich particulate OM deposition during inundation events that presumably initiated after the recent culvert removal (Wang et. al, 2019). In contrast, the deeper soil depths were more similar to older organo-mineral complexed C in terrestrial soils across various ecosystems and land uses (Conant et al., 2011; Dungait et al., 2012; Jobbágy and Jackson, 2000; Kramer and Gleixner, 2006, 2008). The lack of any systematic gradients in the mineral-associated soil C provides further evidence in support of these interpretations, in addition to previous studies showing mineral-associated soil C to be less responsive to environmental forcings, relative to water soluble C (Reynolds et al., 2018).

4.2 Decreases in organic C thermodynamic favorability may restrict microbial activity

Consistent with our first hypothesis, systematic changes in chemical characteristics of soil carbon fractions were observed with thermodynamically less favorable C present at high salinity in shallow soils. This gradient was expected to emerge from increased microbial activity at higher salinity leaving behind less favorable organic C. However, decreases in the number of inferred biochemical transformations and heteroatom abundances with increasing salinity

suggests that microbial activity decreased with increasing salinity imply (but do not quantify) lower microbial activity at higher salinity. While difficult to infer direction of causality, these patterns suggest that less favorable C at higher salinities may constrain microbial activity, leading to fewer biochemical transformations of the organic C. Thermodynamic limitation of organic C transformation is likely due to anaerobic conditions (LaRowe and Van Cappellen, 2011), which are indicated by high-moisture content of soils, high NH_4-N, and low NO_3-N. Anaerobic conditions restrict oxidation of C compounds based on thermodynamic properties (i.e., NOSC and ΔG^0_{Cox}) (Boye et al., 2017), and our data suggest that this has the potential to lead to lower microbial activity in conditions with less favorable organic C.

4.3 Compound class landscape gradients suggest influences of spatially structured inputs

Similar to patterns in C thermodynamic favorability, C compound classes showed significant heterogeneity in shallow soils but had conserved characteristics in deeper soils. The lipid-like peaks observed in the shallow floodplain samples suggest marine-associated algal-derived lipid organic matter similar to results observed by Ward et al., 2019 in a coastal wetland setting. In contrast, lignin-like signatures in the upland site suggest terrestrially derived OM, as has been observed in other environments where terrestrially-derived organic molecules have a high abundance of vascular-plant derived material such as lignin (Hedges and Oades, 1997; Ward et al., 2013). These characteristics also align with reports of saturated soil environments (*e.g.*, floodplains) exhibiting greater abundance of less-oxygenated organic matter than aerobic environments (*e.g.*, upland soils) as reported by Tfaily et al., 2014 in organic matter transformation of a peat column. Our observed landscape gradients in compound class composition indicate spatially structured inputs of organic C such as particulate OM deposition (Langley et al., 2007). Combining this outcome with gradients observed in the total number of

biochemical transformations and the contribution of heteroatoms suggests that sources of C (marine vs terrestrial) and *in situ* processing combine to influence landscape-scale gradients molecular-level organic C chemistry.

4.4 Ecological assembly processes are weakly associated with organic C

Our results show that microbial community assembly is driven by a combination of dispersal limitation (a stochastic process) and deterministic selection most likely associated with pH, as is often observed in soils (Fierer, 2017; Fierer and Jackson, 2006; Garbeva et al., 2004). In contrast, variation in organic C character was associated primarily with specific conductivity. This suggests that the composition of microbial communities is not mechanistically related to C chemistry. Consistent with this inference, we found a very weak association between βNTI and organic C characteristics. Furthermore, and contrary to our hypothesis, we observed a weak negative association between the influence of deterministic processes and the number of organic C transformations.

Relatively fast changes of organic C chemistry compared to relatively slow changes in microbial composition may underlie the lack of association between assembly processes and C chemistry (Bramucci et al., 2013). Supporting this interpretation, a recent study evaluating microbial community composition and C biogeochemistry of soils in a mesohaline marsh following saltwater intrusion reported immediate changes in C mineralization rates with delayed shifts in microbial community composition (Dang et al., 2019). Similarly, a 17-year dryland soil transplant experiment showed large shifts in microbial activity with no change in community composition (Bond-Lamberty et al., 2016). Furthermore, studies across diverse systems show disconnect in function and composition. For example, C chemistry and not microbial community

structure or gene expression was found to significantly influence freshwater hyporheic zone organic matter processing (Graham et al., 2018); environmental conditions influenced the distribution of functional groups, but not taxonomic composition of marine bacterial and archaeal communities (Lima-Mendez et al., 2015; Louca et al., 2016); and dynamic community shifts did not impact functional stability of a methanogenic reactor (Fernández et al., 1999). Combining our study with these previous investigations provides evidence that is consistent with (but does not prove) that soil microbial community composition can be independent of C chemistry, though this certainly varies across systems (e.g., Stegen et al. 2018).

In our system, lack of an association between microbial composition and organic C chemistry is also likely due to a strong influence of stochastic community assembly. Our null modeling indicated that dispersal limitation was responsible for 54% of variation in community composition. Dispersal limitation influences composition by restricting the movement of organisms through space. Restricted movement enhances the influences of stochastic ecological drift, which arises through birth and death events that are randomly distributed across taxa (Green et al., 2004, 2008; Hubbell, 2001; Martiny et al., 2006; McClain et al., 2012; Stegen et al., 2015). Because ecological drift (enabled by dispersal limitation) can lead to the random loss of taxa within local communities, it can result in different communities containing different, but functionally redundant taxa (Loreau, 2004). Moreover, one can argue as per Louca et al., 2018 that in an open system with regular exposure to external inputs (e.g., via tides), functional redundancy is expected to occur and lead to a decoupling of microbial structure and function (Burke et al., 2011; Liebold and Chase, 2017; Nemergut et al., 2013b).

Conclusions

Our results have revealed landscape scale gradients in soil C chemistry in a coastal forested floodplain, but also show that such gradients are different across soil depths and OC fractions—occurring only in the shallow, water soluble C pool. In addition, we found little evidence of an association between C chemistry and microbial community assembly processes, likely due to a dominant influence of stochastic community assembly (as indicated by a strong influence of dispersal limitation). We propose that the disconnect between C chemistry and microbial communities is enhanced by differences in the time scales for which C chemistry and microbial community composition shift.

Our findings suggest that cross-system heterogeneity observed in coastal soil biogeochemical responses to salinity are likely associated with molecular-level C chemistry and microbial physiological responses that are contingent on historical conditions (Fig. 6) (Goldman et al., 2017; Hawkes and Keitt, 2015; Hawkes et al., 2017; Stegen et al., 2018a). We further suggest that microbial community composition may not strongly influence biogeochemical function in coastal soils. Processes associated with molecular-level C chemistry dynamics are therefore likely to be a critical component of ecosystem responses to changing salinity dynamics in coastal TAIs. A full elucidation of these processes will lay a foundation for the development of mechanistic models of coastal TAI biogeochemical dynamics, providing an opportunity for better representation of these ecosystems in local, regional, and Earth system models.

Code and data availability

Raw sequence data has been uploaded to the National Center for Biotechnology Information's (NCBI) Sequence Read Archive (SRA) under BioProject PRJNA541992. All other data are available at DataHub (a PNNL database server) upon manuscript acceptance. Original codes for

694 community assembly metric calculation are available at Stegen_etal_ISME 2013 github
695 repository https://github.com/stegen/Stegen_etal_ISME_2013.

696

697 **Author contribution**

698 AS designed the study, performed the experiments, conducted data analyses and interpretation,
699 and wrote the original draft. JI and CG collected the samples and created site maps. MTF, RKC,
700 and JT provided input on FTICR methodology, conducted the FTICR-MS instrument run, and
701 handled quality filtering and pre-processing of FTICR scans. VLB and NDW contributed to
702 funding acquisition, site selection, study design conceptualization, interpretation of results and
703 editing. JCS contributed to funding acquisition, study design conceptualization, interpretation of
704 results, reviewing and editing. All authors provided feedback on the manuscript.

705

706 **Competing interests**

707 The authors declare no conflict of interest.

708

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Table 1. Relative peak abundances (%) of compound classes in the water extracted organic carbon fraction averaged across replicates per site. Samples are ordered according to their depth profile (shallow and deep) and their relative position in the landscape: floodplain (Fp), inland (I), and upland (U). Abbreviations: Con HC (condensed hydrocarbon), UnsatHC (unsaturated hydrocarbon), Other (no classification assigned)

Site/Depth	Landscape position	Protein	Amino Sugar	Lipid	Lignin	Con HC	Tannin	Other	Carb	Unsat HC
BC2_Shallow	Fp	17.2	3.3	9.4	31.0	22.3	13.2	0.5	1.8	1.3
BC3_Shallow	Fp	21.6	3.8	11.5	27.3	23.0	9.8	0.4	1.5	1.2
BC4_Shallow	I	1.6	0.6	0.3	45.3	32.2	18.9	0.04	0.8	0.2
BC12_Shallow	Fp	7.6	1.8	4.0	38.1	31.2	15.3	0.1	1.2	0.7
BC13_Shallow	Fp	13.3	2.6	5.9	33.4	28.6	14.4	0.2	0.9	1.0
BC14_Shallow	I	6.1	1.7	1.6	37.0	35.8	16.	0.2	0.8	0.5
BC15_Shallow	U	3.7	1.5	1.3	51.8	18.5	21.0	0.2	1.5	0.5
BC2_Deep	Fp	2.3	0.5	1.5	41.2	27.2	25.7	0.2	1.1	0.3
BC3_Deep	Fp	3.2	0.3	3.1	34.1	33.4	24.4	0.3	0.9	0.2
BC4_Deep	I	2.8	0.8	0.6	50.4	27.7	16.5	0.2	0.7	0.2
BC12_Deep	Fp	2.29	0.40	1.43	43.3	27.9	22.9	0.2	1.2	0.3
BC13_Deep	Fp	3.47	0.62	2.00	39.8	33.6	19.2	0.2	0.8	0.3
BC14_Deep	I	1.71	0.76	0.57	43.7	32.5	19.34	0.2	1.0	0.2
BC15_Deep	U	9.51	2.55	4.70	63.8	5.1	9.93	0.7	1.0	2.6

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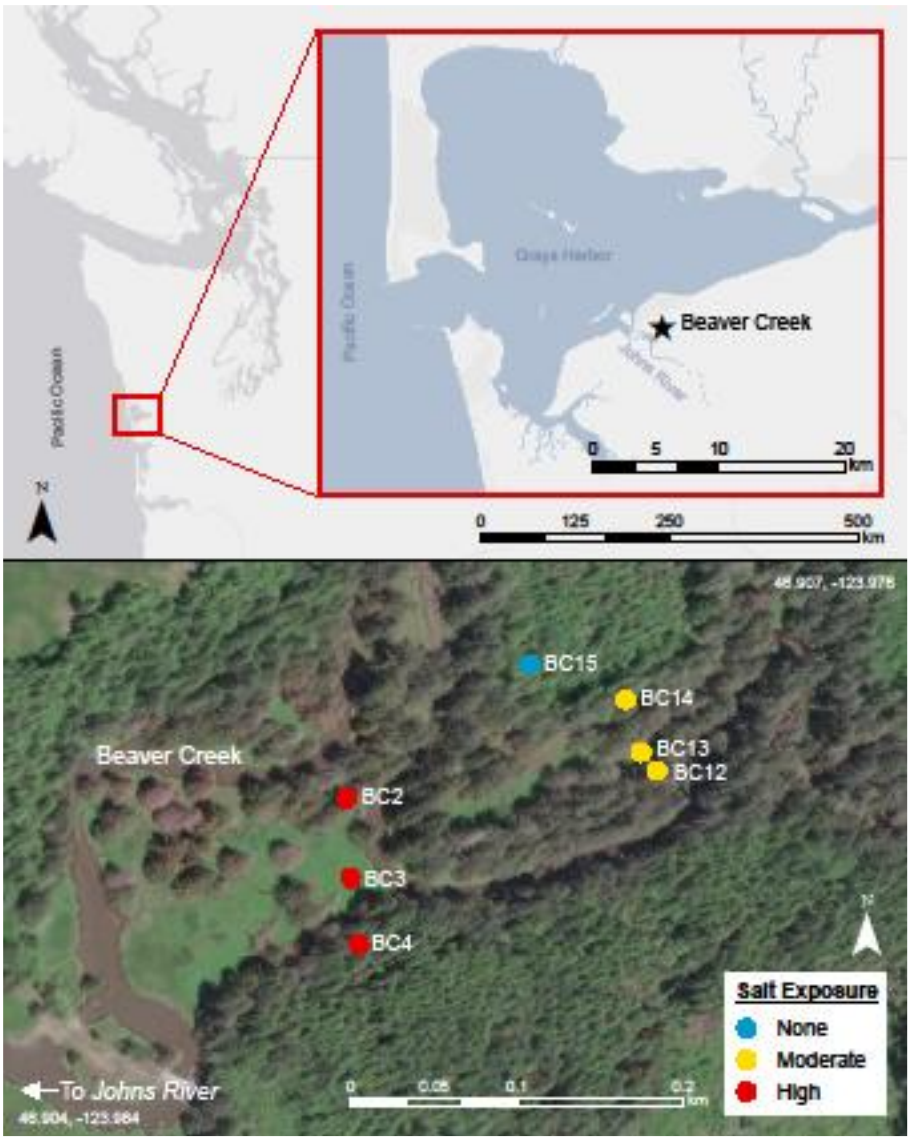


Figure 1. Study site Beaver Creek in the Olympic Peninsula in western Washington. The creek is a first order stream with tidal exchange restored in 2014. Top panel shows site location in western Washington with inset panel zoomed in to show site close to Johns River. Bottom panel shows soil sampling locations at the high salt exposure (BC2, BC3, BC4) transect, moderate salt exposure (BC12, BC13, BC14) transect, and terrestrial upland (BC15) site. The transects with six sampling sites experience periodic inundation episodes which result in surface pooling of tidal water. Map was created using ArcGIS 10.5 software (ESRI, 2017). Coordinate System: GCS WGS 1984.

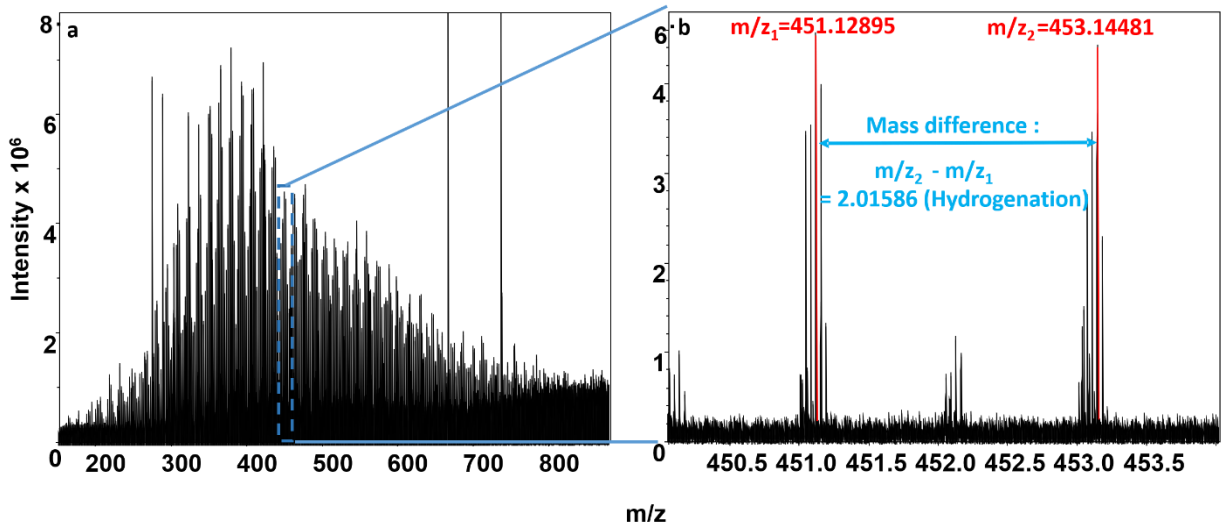


Figure 2. a) Negative mode FTICR-MS (full spectrum); b) zoom in at ~ 450 m/z showing an example of our FTICR-MS spectra overlain with peak mass assignments (red), and a biochemical transformation (mass difference between peaks, denoted in blue). Y axis denotes peak intensities, X-axis denotes mass-to-charge ratio.

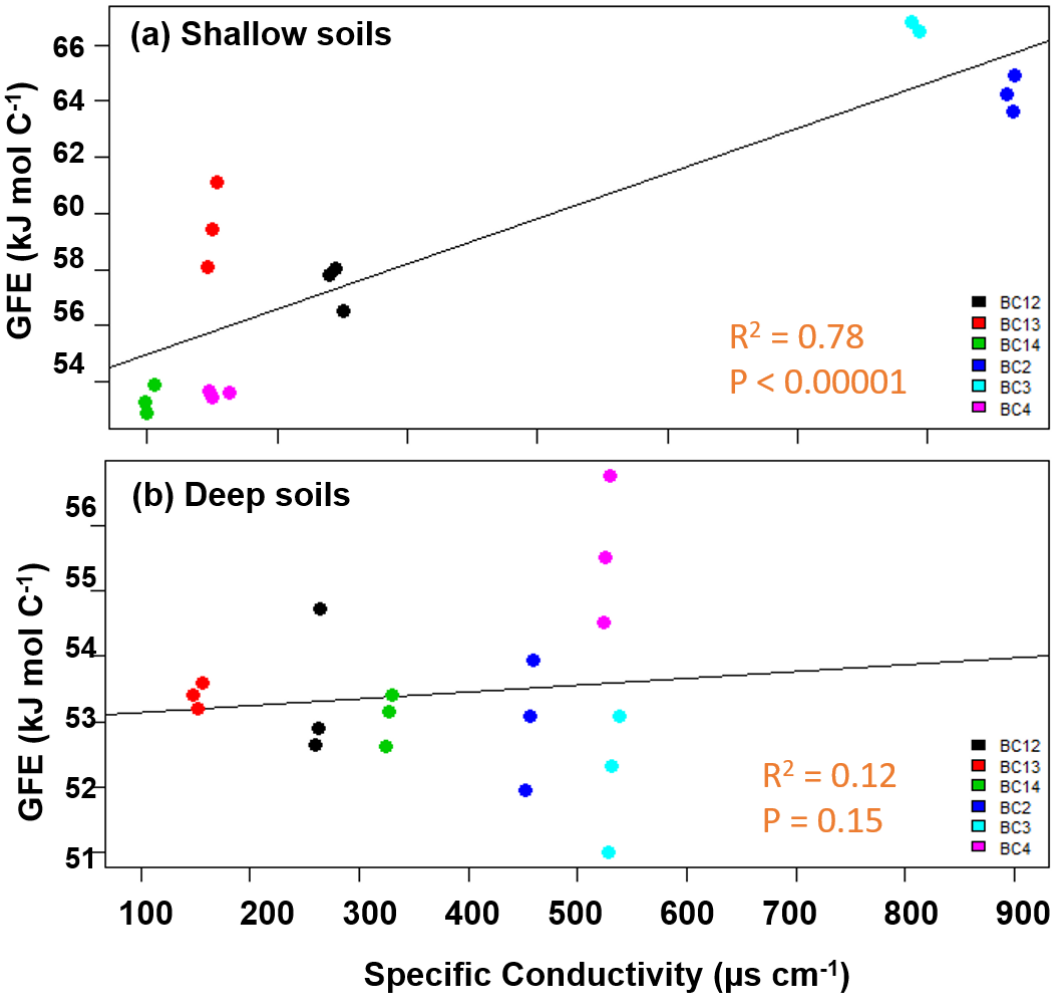
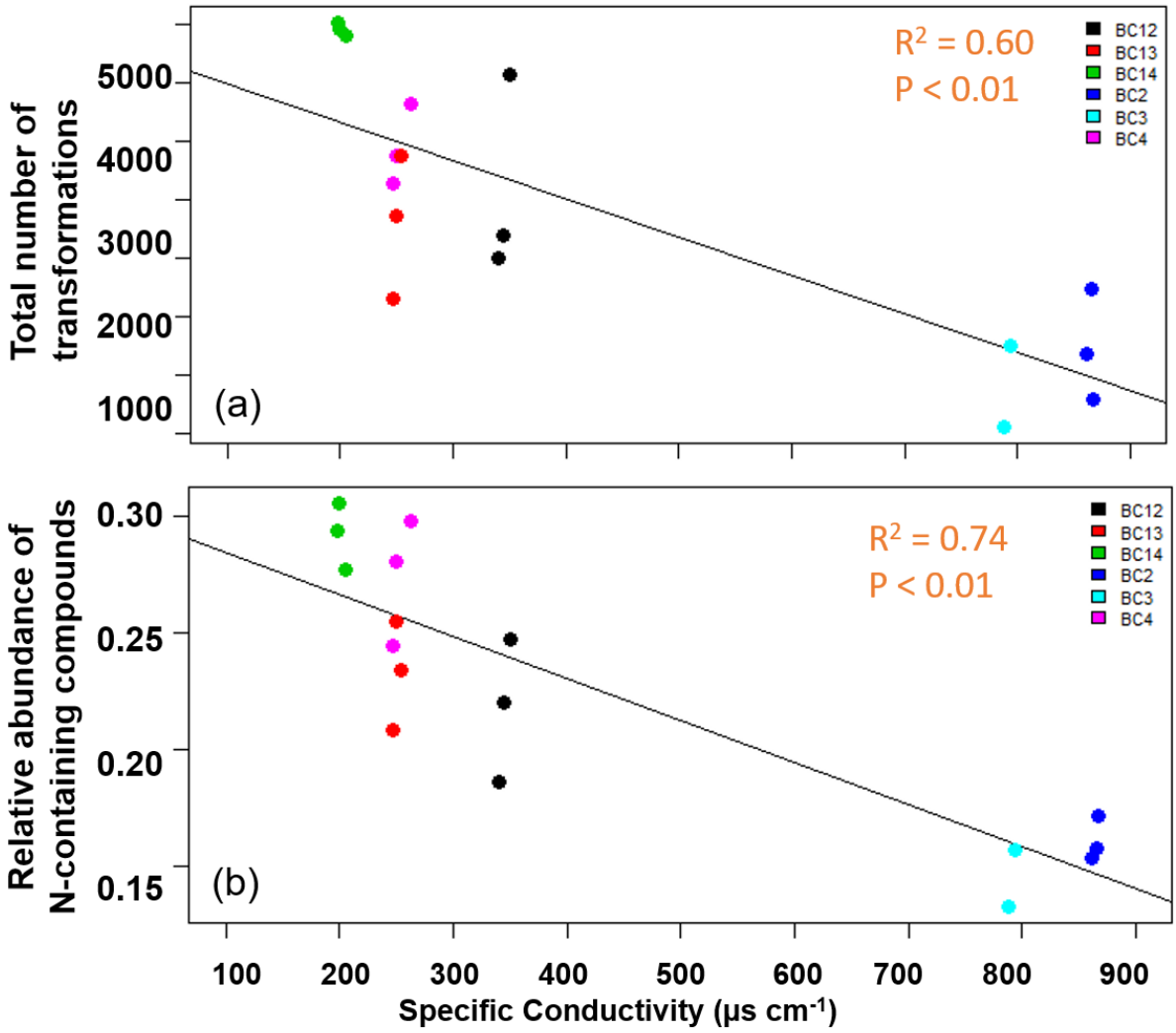


Figure 3. Average Gibbs Free Energy (GFE) of samples in the water fraction of shallow soils impacted by tidal inundation increased with increasing specific conductivity (a) while no change was observed in the deeper soils (b). The salinity of the soil samples did not follow a clear spatial gradient.



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1178 **Figure 4.** The total number of inferred transformations (a) and total abundance of N-containing

1179 compounds (b) in the water fraction of shallow soils impacted by tidal inundation show

1180 significant negative correlations with increasing specific conductivity. No significant

1181 relationships were observed for water fraction of deeper soils or for the $CHCl_3$ fraction in

1182 shallow or deeper soils. The salinity of the soil samples did not follow a clear spatial gradient.

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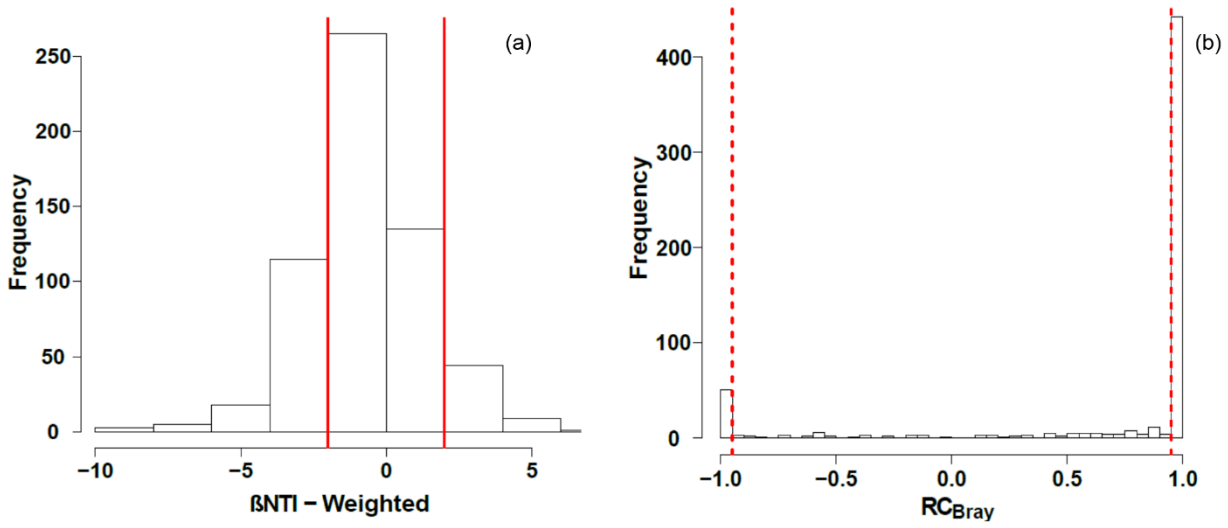

Figure 5. Histograms representing the observed distribution of comparisons based on (a) Beta-near taxon index (βNTI) and (b) Raup Crick metric (RC_{Bray}). Red lines represent the significance thresholds, whereby values outside their bounds are significantly different from the null distribution.

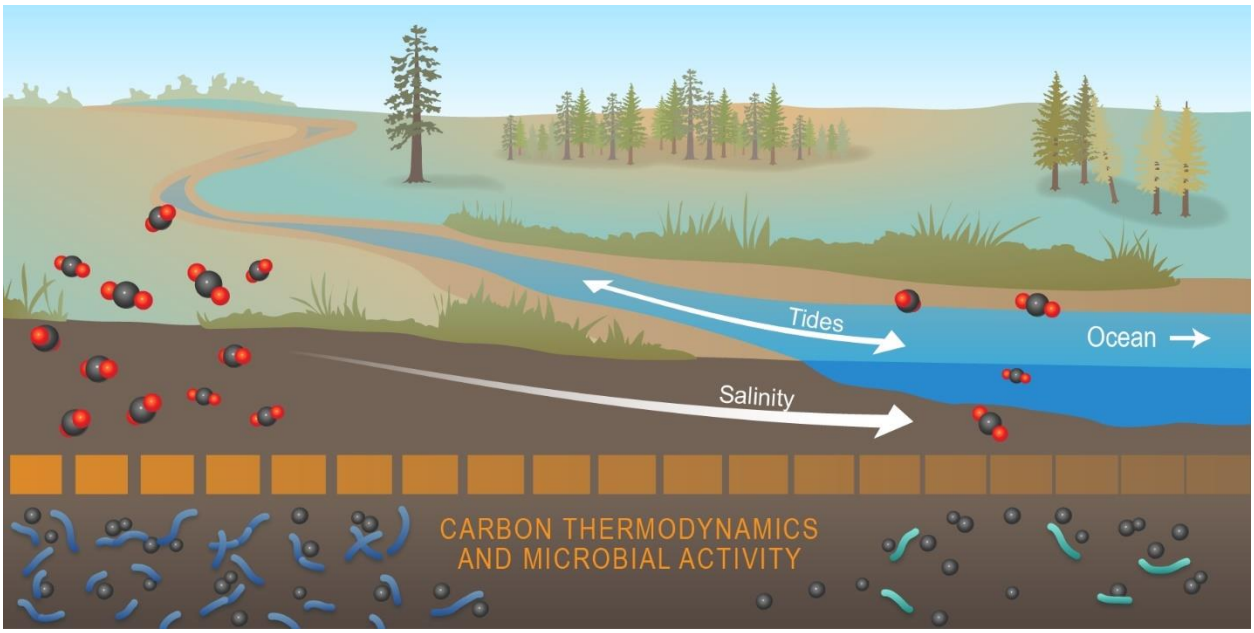

Figure 6. Conceptual model summarizing the key outcome of our study: Microbial activity at high salinity may be depressed by thermodynamically less favorable C. Collectively, our data revealed that organic C thermodynamic favorability, heteroatom content, and number of biochemical transformations all decreased with increasing salinity. This suggest that microbial activity was lower at higher salinity, and we hypothesize this was due to lower thermodynamic favorability of organic C. To evaluate generality, the salinity-associated gradients shown here need to be evaluated across coastal watersheds and mechanistically understood as they have implications for contemporary and future C cycling in coastal watersheds experiencing hydrologic disturbances (e.g., sea level rise and storm surge).