

# Measuring Estuarine Total Exchange Flow from Discrete Observations

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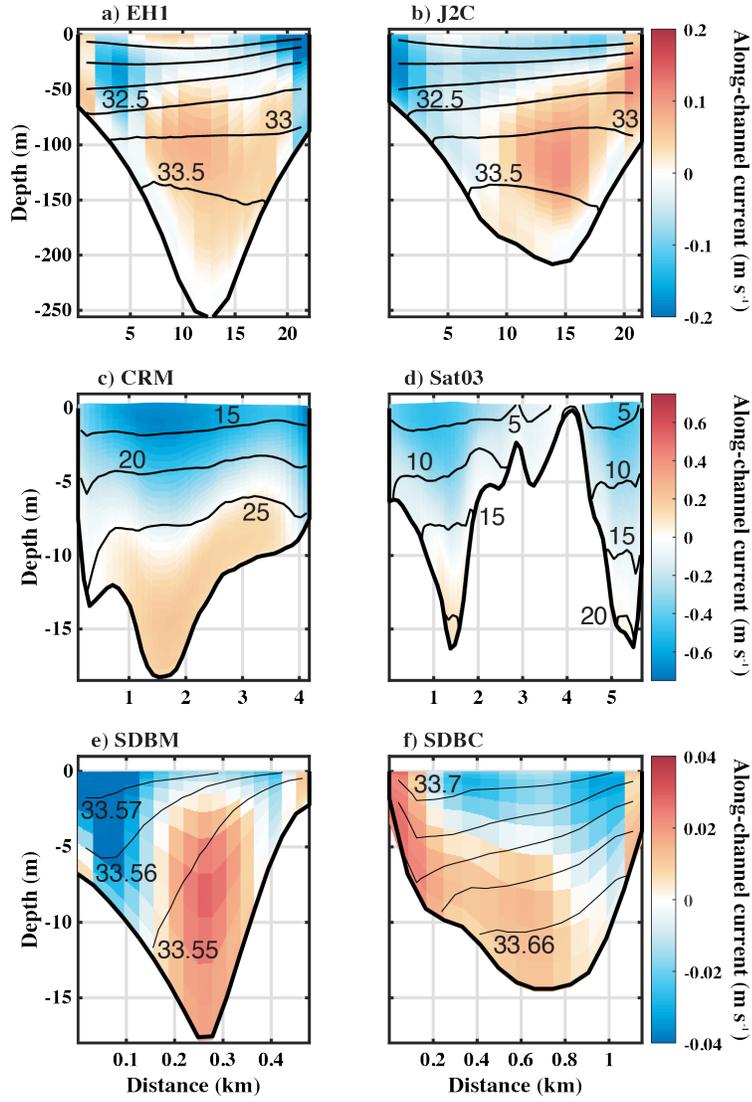
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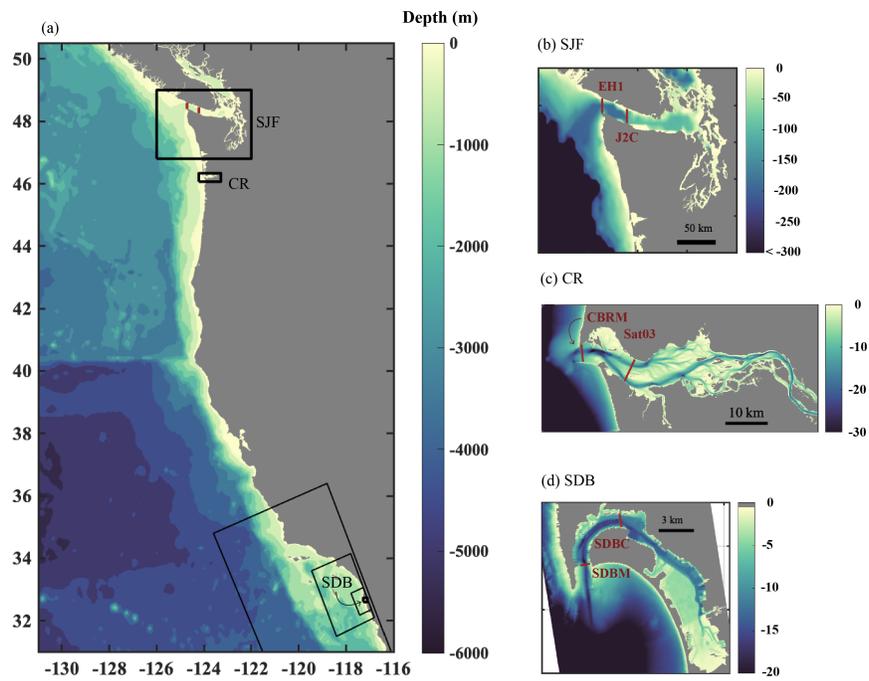
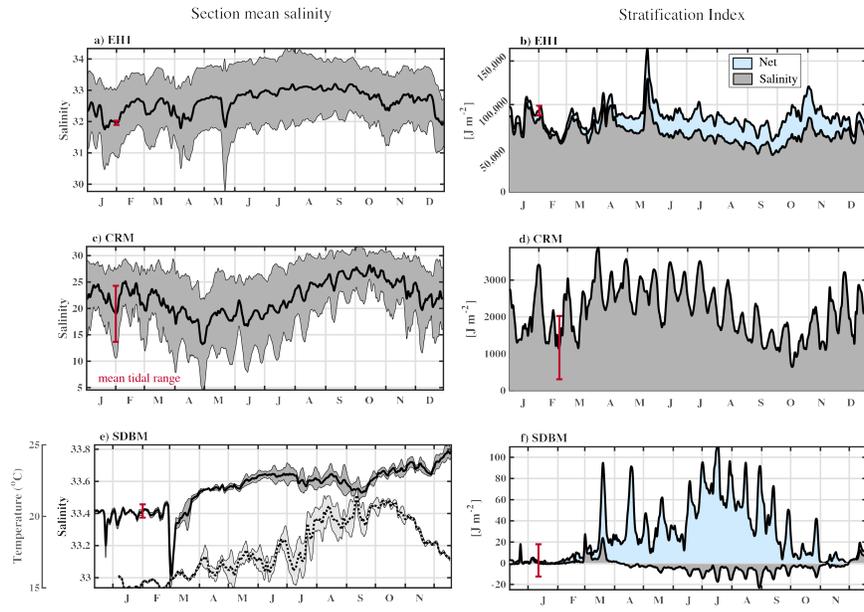
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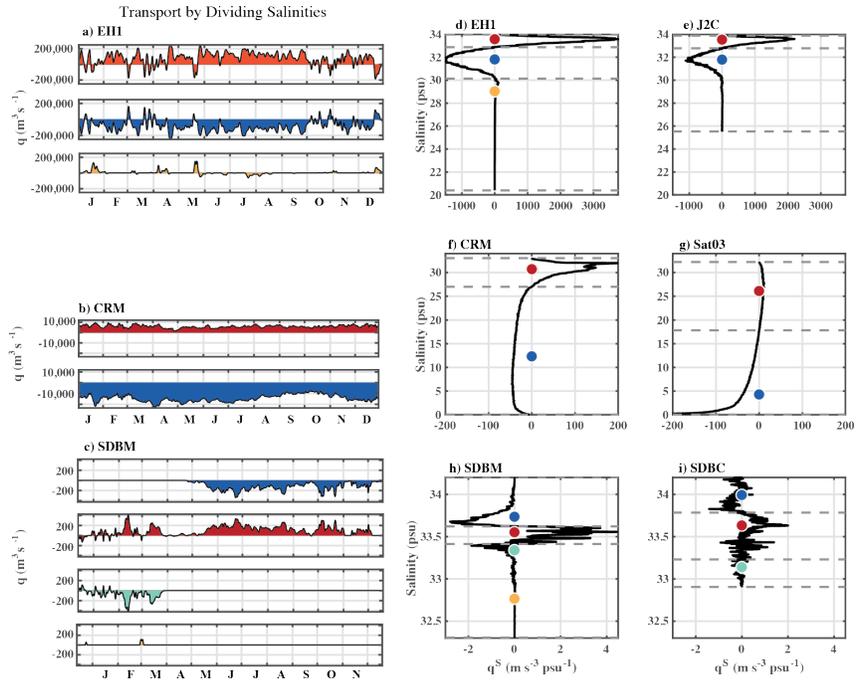
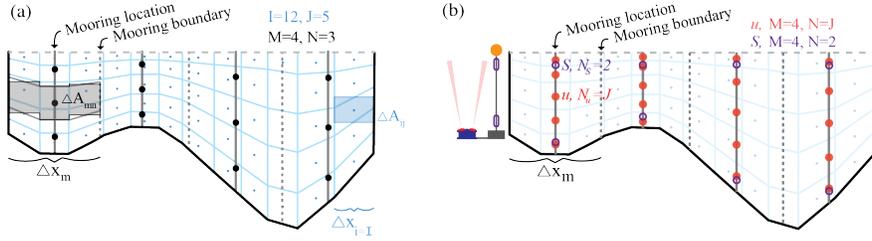
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## Abstract

The exchange between estuaries and the coastal ocean is a key dynamical driver impacting nutrient and phytoplankton concentrations and regulating estuarine residence time, hypoxia, and acidification. Estuarine exchange flows can be particularly challenging to monitor because many systems have strong vertical and lateral velocity shear and sharp gradients in water properties that vary over space and time, requiring high-resolution measurements in order to accurately constrain the flux. The Total Exchange Flow (TEF) method provides detailed information about the salinity structure of the exchange, but requires observations (or model resolution) that resolve the time and spatial co-variability of salinity and currents. The goal of this analysis is to provide recommendations for measuring TEF with the most efficient spatial sampling resolution. Results from three realistic hydrodynamic models were investigated. These model domains included three estuary types: a bay (San Diego Bay), a salt-wedge (Columbia River), and a fjord (Salish Sea). Model fields were sampled using three different mooring strategies, varying the number of mooring locations (lateral resolution) and sample depths (vertical resolution) with each method. The exchange volume transport was more sensitive than salinity to the sampling resolution. Most (>90%) of the exchange flow magnitude was captured by three to four moorings evenly distributed across the estuarine channel with a minimum threshold of 1-5 sample depths, which varied depending on the vertical stratification. These results can improve our ability to observe and monitor the exchange and transport of water masses efficiently with limited resources.







1                    **Measuring Estuarine Total Exchange Flow from**  
2                    **Discrete Observations**

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10                   **Key Points:**

- 11                   • A threshold of 1-5 sample depths was needed to measure the exchange flow.  
12                   • Most (> 90%) of the exchange flow magnitude was captured by 3-4 moorings spread  
13                   across the channel.  
14                   • The exchange volume transport was more sensitive than salinity to the sampling  
15                   resolution.

## Abstract

The exchange between estuaries and the coastal ocean is a key dynamical driver impacting nutrient and phytoplankton concentrations and regulating estuarine residence time, hypoxia, and acidification. Estuarine exchange flows can be particularly challenging to monitor because many systems have strong vertical and lateral velocity shear and sharp gradients in water properties that vary over space and time, requiring high-resolution measurements in order to accurately constrain the flux. The Total Exchange Flow (*TEF*) method provides detailed information about the salinity structure of the exchange, but requires observations (or model resolution) that resolve the time and spatial co-variability of salinity and currents. The goal of this analysis is to provide recommendations for measuring *TEF* with the most efficient spatial sampling resolution. Results from three realistic hydrodynamic models were investigated. These model domains included three estuary types: a bay (San Diego Bay), a salt-wedge (Columbia River), and a fjord (Salish Sea). Model fields were sampled using three different mooring strategies, varying the number of mooring locations (lateral resolution) and sample depths (vertical resolution) with each method. The exchange volume transport was more sensitive than salinity to the sampling resolution. Most (>90%) of the exchange flow magnitude was captured by three to four moorings evenly distributed across the estuarine channel with a minimum threshold of 1-5 sample depths, which varied depending on the vertical stratification. These results can improve our ability to observe and monitor the exchange and transport of water masses efficiently with limited resources.

## Plain Language Summary

The two-way exchange of water and properties such as heat and salinity as well as other suspended material between estuaries and the coastal ocean is important to regulating these marine habitats. This exchange can be challenging to measure. The Total Exchange Flow (*TEF*) method provides a way to organize the complexity of this exchange into distinct layers based on a given water property. This method has primarily been applied in numerical models that provide high resolution output in space and time. The goal here is to identify the minimum horizontal and vertical sampling resolutions needed to measure *TEF* depending on estuary type. Results from three realistic hydrodynamics models were investigated. These models included three estuary types: bay (San Diego Bay), salt-wedge (Columbia River), and fjord (Salish Sea). The models were sampled using three different mooring strategies, varying the number of mooring locations and sample depths with each method. Most of the exchange magnitude was captured by three to four moorings evenly distributed across the estuarine channel with a minimum threshold of 1-5 sample depths, which varied by estuary. These results can improve our ability to observe and monitor the exchange and transport of water masses efficiently with limited resources.

## 1 Introduction

The exchange between estuaries and the coastal ocean is a key dynamical driver impacting biogeochemical patterns such as nutrient and phytoplankton concentrations within the estuary (e.g., Boyer et al., 2002; Brown & Ozretich, 2009) and in the coastal ocean (e.g., Davis et al., 2014). This exchange can regulate estuarine residence time, hypoxia, and acidification (e.g., MacCready et al., 2021; O’Callaghan et al., 2007). Estuaries deliver terrigenous material to the ocean including sediment, larvae, and pollutants. Estuaries can also impact coastal circulation by delivering relatively freshwater into the coastal margins (e.g., Banas et al., 2009; Giddings et al., 2014; Mazzini et al., 2014). Our ability to accurately observe the exchange at the estuary-ocean interface is therefore important to understanding the physics, biology, chemistry, and coupling of estuarine and coastal ecosystems. While a common example of exchange between estuaries and the coastal

ocean is buoyancy-driven flow (MacCready & Geyer, 2010), other mechanisms such as tidal asymmetry can drive the exchange (Burchard & Hetland, 2010). Exchange flows are also important mechanisms in the transport and mixing of water masses through inland seas (e.g., Burchard & Badewien, 2015; Becherer et al., 2016) and through straits connecting marginal seas and the coastal ocean (e.g., Reissmann et al., 2009).

Estuary exchange is particularly challenging to monitor because many estuaries have strong vertical and lateral velocity shear and salinity gradients that vary over space and time, requiring high-resolution measurements and strategic extrapolation in order to accurately constrain the flux. Efforts have been made to understand the relative importance of bathymetric variability, such as between deep channels and wide shallow shoals (Geyer et al., 2020), tidal mixing (Griffin & LeBlond, 1990; Cheng et al., 2011), and of Earth’s rotation (Valle-Levinson et al., 2003) on the exchange. Valle-Levinson (2008) developed a semi-analytical model to determine whether the along-channel velocities were horizontally or vertically sheared using the non-dimensional Kelvin ( $K_e$ ) and Ekman ( $E_k$ ) numbers. The Kelvin number  $K_e = Wf(g'H)^{-\frac{1}{2}}$  for estuary width  $W$ , depth  $H$ , reduced gravity  $g'$ , and coriolis parameter  $f$  estimates the importance of Earth’s rotation on the flow. Wide basins ( $K_e > 2$ ) are more likely to have strong horizontal shear (Garvine, 1995). The exchange flow may also depend on  $K_e$ ; for example, greater exchange due to shelf forcing has been observed in relatively narrower fjord-type estuaries (Jackson et al., 2018). The Ekman number  $E_k = A_z(fH^2)^{-1}$ , where  $A_z$  is the flow’s eddy viscosity, estimates the importance of vertical mixing (Kasai et al., 2000; Winant, 2004). Large Ekman number ( $E_k > 1$ ) basins are likely to have strong horizontal shear regardless of their width (Valle-Levinson, 2008). Other parameters have also been used to describe estuaries based on characteristics such as the tidal and freshwater forcing, but even within a given system conditions can vary substantially on sub-tidal to seasonal and interannual timescales (Geyer & MacCready, 2014).  $K_e$  and  $E_k$  provide an estimate of the degree of horizontal and vertical variability in the currents and may help predict how many moorings and vertical sample depths are needed to accurately constrain the exchange.

Regardless of the estuary size, depth, transport, degree of stratification, and of the dominant forcing mechanisms, exchange flow is governed by the Knudsen relations which use mass and salt conservation to show that the exchange flow can be many times larger than the river flow (Knudsen, 1900; Burchard et al., 2018). The Knudsen (1900) theorem calculates the inflow ( $Q_{in}$ ) and outflow ( $Q_{out}$ ) and representative salinities ( $S_{in,out}$ ) assuming the exchange flows occur in layers of constant salinity. More recently the Total Exchange Flow ( $TEF$ ) method for computing the subtidal exchange parameters  $Q_{in}$ ,  $Q_{out}$ ,  $S_{in}$ ,  $S_{out}$  was proposed by MacCready (2011) and was updated to be more numerically accurate by MacCready et al. (2018) and Lorenz et al. (2019).  $TEF$  uses isohaline coordinates (Walin, 1977) to track the exchange flow, thus extending the Knudsen (1900) theorem to conditions with time-variable stratification and flow, incorporating both subtidal and tidal fluxes (Chen et al., 2012).  $TEF$  provides detailed information about the salinity structure of the exchange flow, can identify multiple layers of exchange, can be applied in inverse estuarine conditions (Lorenz et al., 2019, 2020), and can be directly related to mixing (MacCready et al., 2018).

$TEF$  has been widely applied in estuarine research and exchange flows more generally. The  $TEF$  framework has been used to determine freshwater fluxes from a small groundwater-driven estuary (Ganju et al., 2012), to estimate estuarine residence times (Sutherland et al., 2011; Lemagie & Lerczak, 2014), to examine the relationship between exchange flow and mixing (Wang et al., 2017), and study seasonal variability (Giddings & MacCready, 2017; Conroy et al., 2020), among others. Most of the aforementioned examples are modeling studies, with the exception of Ganju et al. (2012). There are also analyses of salinity flux from observations that do not use the  $TEF$  framework (e.g., Lerczak et al., 2006; MacDonald & Horner-Devine, 2008), but calculations of salt flux from

119 observations are limited due to the large data requirement to resolve the temporal and  
 120 spatial co-variability in salinity and currents as well as a lack of knowledge regarding flux  
 121 errors when undersampling occurs.

122 The goal here is to provide recommendations for applying *TEF* to *in-situ* obser-  
 123 vations, specifically to understand the most efficient spatial sampling resolution and the  
 124 percent of the exchange flow captured under various strategies. This paper examines *TEF*  
 125 calculated from sub-sampling realistic numerical models, representative of moorings in  
 126 a channel, compared to *TEF* calculated from the full model resolution in order to com-  
 127 pare how quickly the two estimates converge as the number of moorings increase. Three  
 128 estuaries were studied in order to span much of the parameter space of estuarine char-  
 129 acteristics. The objectives of this study were (a) to test how *TEF* converged for differ-  
 130 ent sampling resolutions; (b) to examine how this varied between estuaries and sampling  
 131 strategies; (c) to attempt to outline best practices for how many moorings and instru-  
 132 ments would be required to quantify *TEF* from observations; and (d) to understand flux  
 133 errors (magnitude and potential bias) when a cross-section is under-sampled. The three  
 134 realistic estuary models, details of the *TEF* calculation, and the sampling methods are  
 135 described in the methods section 2. The current and salinity patterns characteristic of  
 136 each estuary and individual cross-section are included in section 3. The rest of the re-  
 137 sults are organized into sections based on the various sampling approaches: evenly dis-  
 138 tributed moorings (section 4), strategically distributed moorings (section 5), and a case  
 139 study designed to approximate a simple observational approach (section 6).

## 140 2 Methods

### 141 2.1 Realistic hydrodynamic models

142 Realistic hydrodynamic numerical models of three estuaries and their adjacent coastal  
 143 regions were used (Figure 1). These span different estuary types and geometries and in-  
 144 clude a small bay (San Diego Bay), salt-wedge (Columbia River), and large fjord (Sal-  
 145 ish Sea) (e.g., Geyer & MacCready, 2014). Results were extracted hourly from two across-  
 146 channel sections in each model over a full year of simulation time. Further details about  
 147 each model are outlined in the following paragraphs.

148 The Salish Sea model including the Strait of Juan de Fuca employed the Regional  
 149 Ocean Modeling System (ROMS, Shchepetkin & McWilliams, 2005). Simulations from  
 150 2004 to 2007 were developed by the University of Washington Coastal Modeling Group  
 151 (MacCready et al., 2009; Sutherland et al., 2011; Giddings et al., 2014). The model was  
 152 forced with realistic river flow, tides, wind stress, surface heat flux, and open boundary  
 153 conditions (e.g., Giddings et al., 2014) with initial and open boundary values for trac-  
 154 ers, subtidal velocity, and subtidal surface height from the Navy Coastal Ocean Model  
 155 (NCOM) (Barron et al., 2006; Kara et al., 2006). The domain spans the inland waters  
 156 of the Salish Sea (including Puget Sound, the Strait of Georgia, and the Strait of Juan  
 157 de Fuca) and coastal ocean from 43N to 50N and 200 km offshore with a horizontal res-  
 158 olution of 1.5 km at the coast to 4.5 km far offshore. There were 40 sigma layers with  
 159 enhanced vertical resolution near the surface and bottom. This analysis focuses on data  
 160 extracted from 2005 at two cross-sections spanning the Strait of Juan de Fuca: near the  
 161 ocean entrance (EH1), and 16 km upstream (J2C; Figure 1). These sections correspond  
 162 with previous analyses and validation of the model results (Giddings et al., 2014). At  
 163 EH1 and J2C the channel is 22.1 and 21.5 km wide, respectively, and 73 and 60 m deep  
 164 (Table 1). Model skill was high ( $\geq 0.92$ ) relative to observed currents, tidal sea surface  
 165 elevation, salinity and temperature, although overall slightly too salty (by  $\sim 1.5$  psu and  
 166  $\sim 0.5^\circ\text{C}$ ) within the Salish Sea (Giddings et al., 2014). Most pertinent to this study, the  
 167 exchange flow through the Strait of Juan de Fuca compared well with observations and  
 168 was insensitive to model resolution (Giddings & MacCready, 2017).

169 The Columbia River simulation used the unstructured grid, finite element model  
 170 SELFE (Zhang & Baptista, 2008) version 4.0.1. Results for these analyses were accessed  
 171 through the Columbia River Inter-Tribal Fish Commission Coastal Margin Observation  
 172 and Prediction Program (CRITFC-CMOP; [stccmop.org](http://stccmop.org)). Temperature, salinity, and  
 173 water elevations were imposed at the oceanic boundary from the Navy Coastal Ocean  
 174 Model (NCOM) for years 1999-2012 (analysis here focuses on the year 2012; Barron et  
 175 al., 2006). The domain extended from 39N to 50N and  $\sim 300$  km in the offshore direc-  
 176 tion with horizontal resolution from tens of meters in the estuary and river to 3 km in  
 177 the ocean (Karna et al., 2015; Karna & Baptista, 2016). The vertical grid consisted of  
 178 37 sigma levels between sea level and 100 m datum and an equipotential z-grid below  
 179 100 m. Data were extracted from two cross-sections: the river mouth (CRM) and 14 km  
 180 upstream at site Saturn-03 (Sat03), to match previous studies where model validation  
 181 was performed (Figure 1; e.g. Karna et al., 2015). At CRM and Sat03 the channel is 4.2  
 182 and 5.7 km wide and 18 and 16 m deep, respectively (Table 1). The model demonstrated  
 183 high skill compared to long term observations, particularly outside of high discharge and  
 184 neap tide conditions which are estimated to occur only 16% of the time (Karna & Bap-  
 185 tista, 2016).

186 The San Diego Bay and adjacent coastal dynamics were simulated using the Cou-  
 187 pled Ocean-Atmosphere-Wave-Sediment-Transport (COAWST) model system (Warner  
 188 et al., 2010; Kumar et al., 2012) to represent the surfzone and shelf circulation (Wu et  
 189 al., 2020, 2021). This model grid sits within three one-way nested parent models using  
 190 ROMS (Shchepetkin & McWilliams, 2005) and is coupled with the Simulating Waves  
 191 Nearshore (SWAN) model to include surface gravity waves (Booij et al., 1999). Bound-  
 192 ary and initial conditions for the outermost domain were from the California State Es-  
 193 timate (CASE) solution (Marshall et al., 1997) with tides from the ADCIRC tidal database  
 194 (Westernik et al., 1993) and surface forcing from the North American Mesoscale Fore-  
 195 cast (NAM) and the Coupled Ocean-Atmosphere Mesoscale System (COAMPS). The  
 196 largest grid extends between 29N to 36N and over 500 km offshore with 2 km horizon-  
 197 tal resolution, which is downscaled to the finest grid with horizontal resolution from 8 m  
 198 near the coast to 110 m at the western boundary, and has 10 stretched vertical levels (Wu  
 199 et al., 2020, 2021). This study focuses on 2016 results at the estuary mouth (SDBM) and  
 200 4 km upstream (SDBC; Figure 1). At SDBM and SDBC, the channel is 0.5 and 1.2 km  
 201 wide and up to 19 and 14 m deep, respectively (Table 1). This model has not been rig-  
 202 orously validated against observations of San Diego Bay, but exhibits circulation sim-  
 203 ilar to prior observations (Largier et al., 1996).

## 204 2.2 Stratification and *TEF* calculations

The vertical stratification index was used to characterize the water column at each  
 cross section by

$$\phi = -H^{-1} \int_{-H}^0 (\rho - \bar{\rho})gzdz, \quad (1)$$

where the overbar denotes a vertical mean. Vertical mean density  $\bar{\rho}$  was computed

$$\bar{\rho} = H^{-1} \int_{-H}^0 \rho dz \quad (2)$$

205 following Simpson et al. (1981).  $\phi$  gives an estimate of the potential energy of the wa-  
 206 ter column relative to the mixed state such that in a vertically-well mixed water column  
 207  $\phi = 0$ . The influence of salinity and temperature on the density structure were com-  
 208 puted by substituting  $\rho = \rho(S(z), \bar{T})$  and  $\rho = \rho(\bar{S}, T(z))$ , respectively. While avail-  
 209 able potential energy is another useful framework for understanding estuarine systems  
 210 (e.g., MacCready & Giddings, 2016),  $\phi$  is useful in this context because as  $\phi$  approaches  
 211 0  $S_{in}$  and  $S_{out}$  converge.  $\phi$  was computed for each lateral column separately before cal-  
 212 culating an area-weighted cross-sectional mean value.

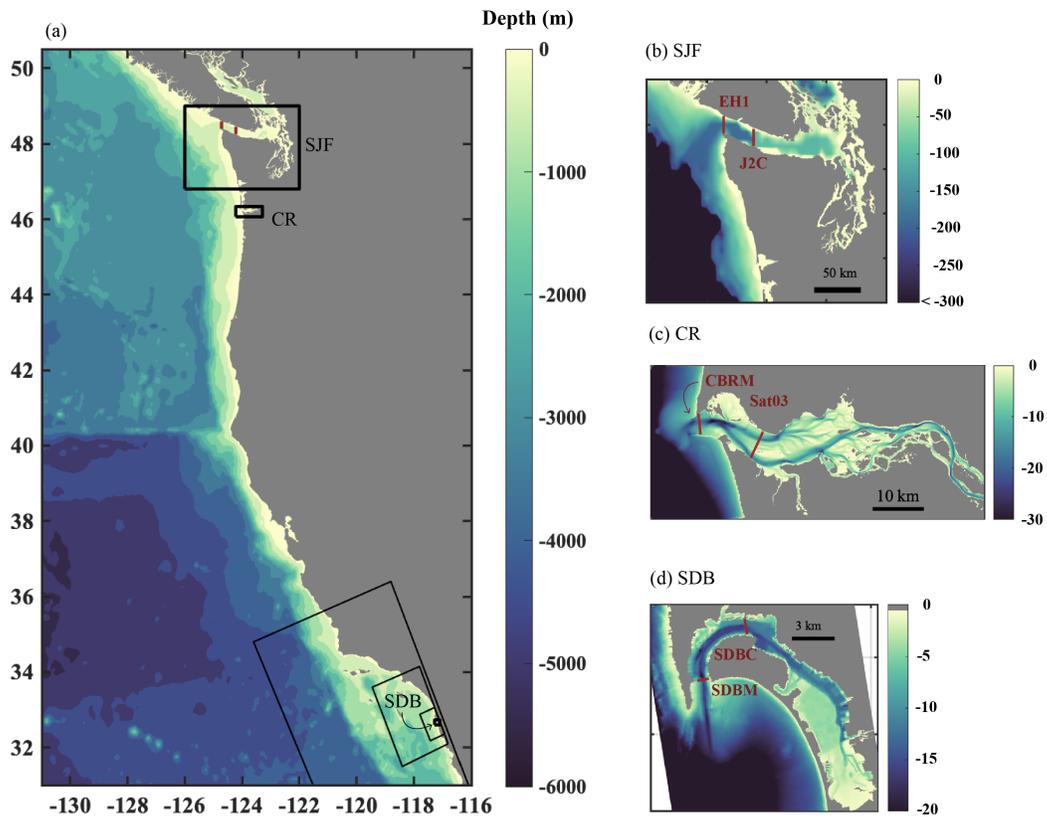
**Table 1.** Description of each model simulation and cross-section, including the latitude and longitude, simulation year, maximum cross-section width  $W$ , number of grid points across the section  $I$  and mean grid width  $\Delta x$ , maximum cross-section depth  $H$ , number of vertical layers  $J$  and the mean depth of each layer  $\Delta z$ .

Run	lat,lon	Year	$W$ (km)	$I$	$\Delta x$ (m)	$H$ (m)	$J$	$\Delta z$ (m)
SJF <sup>a</sup> (EH1)	(-124.71,48.49)	2005	22.1	14	1578±24	258	40	4.1±2.6
SJF (J2C)	(-124.21,48.35)	2005	21.5	14	1537±12	209	40	3.8±2.2
CR <sup>b</sup> (CRM)	(-124.04,46.23)	2012	4.2	41	102	18	37	0.3±0.1
CR (Sat03)	(-123.94,46.20)	2012	5.7	58	98±0.5	16	37	0.2±0.1
SDB <sup>c</sup> (SDBM)	(-117.23,32.69)	2016	0.5	16	29±1 m	19	10	1.1±0.8
SDB (SDBC)	(-117.20,32.72)	2016	1.2	14	82±0.7 m	14	10	1.1±0.8

<sup>a</sup>SJF = Strait of Jaun de Fuca.

<sup>b</sup>CR = Columbia River.

<sup>c</sup>SDB = San Diego Bay.



**Figure 1.** (a) The US west coast with the region around each realistic numerical model outlined with black boxes. Corresponding with the boxes in (a) more detail is shown around the estuarine cross-sections in (b) the Strait of Juan de Fuca, SJF, (c) the Columbia River, CR, and (d) San Diego Bay, SDB. Red lines mark each of the cross sections examined as part of this study. Colors denote bathymetric depth. Note the lateral and color scales vary between maps.

Subtidal exchange flow is calculated using isohaline coordinates following the *TEF* dividing salinity method (Lorenz et al., 2019). Following the *TEF* framework (MacCready, 2011), the net transport of a tracer  $c$  through a cross-sectional area  $A(S > S')$  determined by salinity  $S'$  is defined as:

$$Q^c(S, t) = \left\langle \int_{A(S > S')} cudA \right\rangle. \quad (3)$$

where  $u$  is the velocity normal to the cross-section (positive values are into the estuary),  $t$  is time, and  $\langle \rangle$  denotes a subtidal filter (here the Godin low pass filter, Thomson & Emery, 2014). A profile of the tracer exchange can also be determined by differentiation:

$$q^c(S, t) = -\frac{\partial Q^c(S)}{\partial S}. \quad (4)$$

The transport profile was separated into distinct inflow and outflow layers by finding the extrema in the  $Q^c$  profiles, ignoring extrema below a certain noise threshold  $Q_{thresh}$  (Lorenz et al., 2019).  $Q_{thresh}$  was defined here as a fixed percentage of the maximum transport magnitude,  $Q_{thresh} = Q_{percent} * \max(|Q(S)|)$ , where  $Q_{percent} = 0.01$ . The salinity values associated with the  $Q^c$  extrema—along with the salinity endpoints  $S_{min}$ ,  $S_{max}$ —made up the dividing salinities  $S_{div}$  and the transport in each layer was

$$\Delta Q_l^c(t) = \int_{S_{div,l}}^{S_{div,l+1}} q^c dS. \quad (5)$$

213 Inflow was positive and layers were defined by the sign of the net transport:

$$\Delta Q_{in,a}^c(t) \equiv \Delta Q_l^c(t) > 0, \quad \Delta Q_{out,b}^c(t) \equiv \Delta Q_l^c(t) < 0. \quad (6)$$

214 Subscripts  $a$  and  $b$  are used to enumerate inflow and outflow layers, following Lorenz et  
215 al. (2019). The net exchanges were defined by:

$$Q_{in}^c(t) \equiv \sum_a \Delta Q_{in,a}^c(t), \quad Q_{out}^c(t) \equiv \sum_b \Delta Q_{out,b}^c(t). \quad (7)$$

This summation transforms the results from  $l = 1 : L$  individual layers into two layers, which does not greatly impact the result if the flow is predominantly two-layered. The mean inflow and outflow salinities can be calculated by:

$$S_{in}(t) = \frac{Q_{in}^S(t)}{Q_{in}(t)}, \quad S_{out}(t) = \frac{Q_{out}^S(t)}{Q_{out}(t)}. \quad (8)$$

216 where the tracer,  $c$ , is the salinity,  $S$  and no superscript implies the volume flux only,  $Q(S, t) =$   
217  $\left\langle \int_{A(S > S')} udA \right\rangle$ . The above exchange flows and corresponding salinities are referred to  
218 as the *TEF* bulk values (e.g., Lorenz et al., 2019).

219 Currents and salinities were extracted hourly from each cross-section. Since the SELFE  
220 model uses an unstructured grid, CR output were interpolated onto horizontally fixed  
221 straight cross-sections at Sat03 and CRM, roughly matching the spatial resolution of the  
222 model grid. ROMS variables were extracted at the grid resolution. Velocities were ro-  
223 tated onto along- and across-channel coordinates, defined by the angle of the cross-section.  
224 The principle axis of the area-averaged currents over the year were closely aligned with  
225 each cross-section (e.g., Table 2). In order to avoid tidal aliasing, the start and end times  
226 were estimated by the timing of the spring tidal sea level maximum closest to each cal-  
227 endar end point.

## 2.3 Sampling strategies

Four methods of sampling the cross-sectional fields and calculating  $TEF$  were compared: (1) using horizontal and vertical resolution from the  $I \times J$  model grid to calculate  $TEF_{IJ}$ ; (2) using an  $M \times N$  array of evenly spaced samples to calculate  $TEF_{MN}$  from an increasing integer number of “moorings”  $M$  evenly distributed across the channel width with  $N$  sample depths each, which were evenly distributed across the time-averaged channel depth at each location  $x = x_m$  (e.g. Figure 2a); (3) using  $\mu$  moorings with strategic placement of each mooring  $\mu = 1, \dots, I$  determined by maximizing the correlation between  $TEF_{\mu J}$  and  $TEF_{IJ}$ ; and (4) a case study designed to imitate observations with a single bottom-mounted Acoustic Doppler Current Profiler (ADCP) and salinity measurements near the surface and bottom of the water column at  $M$  evenly distributed mooring locations (e.g. Figure 2b). Each of these methods is described in more detail in the following paragraphs.

### 2.3.1 Full $TEF$ , $TEF_{IJ}$

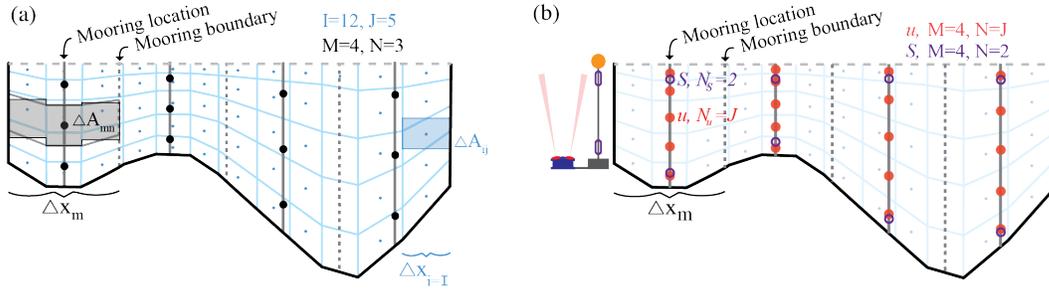
$TEF_{IJ}$  used the full vertical and horizontal resolution from the model grid.  $TEF_{IJ}$  represented the expected value which other estimates of  $TEF$  were hypothesized to converge towards at high sampling resolutions. Velocity and salinity fields were sampled at points  $(x_i, z_{ij}(t))$  and the area represented by each sample was computed by  $\Delta A_{i,j}(t) = \Delta x_i \Delta z_{ij}(t)$ . Subscripts  $i$  and  $j$  indicate the indices on the model grid in the across-channel and vertical direction, respectively (e.g., Figure 2).

### 2.3.2 Evenly distributed subsamples, $TEF_{MN}$

$TEF_{MN}$  used an  $M \times N$  array of samples evenly distributed across the channel and throughout the water column (e.g. Figure 2a). This method was chosen to test the convergence of  $TEF_{MN}$  towards  $TEF_{IJ}$  as the number of moorings ( $M$ ) or sample depths ( $N$ ) were increased. This method is simple enough to be consistently applied in every case. However, when there is sharp bathymetric variability within width  $\Delta x_m$ , it is not obvious how to estimate area  $\Delta A_{m,n}(t)$ . For example, on a steep slope a gridded area could either over-estimate or under-estimate the flux. To address this, two approaches were compared: the first method assumed that  $u$ ,  $S$  were constant with depth over distance  $\Delta x_m$  and the second assumed the profiles of  $u$ ,  $S$  had a consistent shape over distance  $\Delta x_m$  and were thus constant across  $\sigma$ -levels as in Lerczak et al. (2006). The difference in the results between approaches was negligible. The results reported herein assume that  $u$  and  $S$  were constant along  $\sigma$ -levels to estimate  $\Delta A_{MN}$  (Equation 3) as illustrated on Figure 2a.

### 2.3.3 Strategically located subsamples, $TEF_{\mu J}$

$TEF_{\mu J}$  was calculated by incrementally adding moorings in order based on identifying the mooring which contributed the largest improvement in the correlation coefficient between  $TEF_{\mu J}$  and  $TEF_{IJ}$ , similar to the approach of using maximum explained variance used by Wei et al. (2020). Lateral mooring placement was sampled at  $I$  grid locations while  $J$  sampling depths were included at each mooring location. Since  $TEF$  is a derived flux quantity, it was necessary to estimate the cross-sectional area represented by each sample of  $u$  and  $S$ . This computation used linear interpolation assuming  $u$  and  $S$  were constant across  $\sigma$ -levels, analogous to  $TEF_{MN}$ . The lateral edges of the regions represented by each mooring  $\mu$  were defined by the channel edges and the mid-point between each mooring pair in the across-channel direction. The sample area  $\Delta A_{\mu J}$  was calculated as the total model grid area at a given  $\sigma$ -level between the lateral edges bounding each mooring  $\mu$ .



**Figure 2.** Illustration of the grid points used for the three  $TEF$  calculations. In (a) thin blue lines indicate the full model resolution and blue dots mark cell centers where model fields were extracted,  $TEF_{IJ} = TEF_{12,5}$ . Blue shading is a sample model grid cell area, which varies over time with sea level. Dark gray lines indicate example mooring locations for  $M = 4$  with black dots indicating example mooring vertical sampling for  $N = 3$ . Dark gray dashed lines indicate the boundaries centered between simulated mooring locations for  $TEF_{MN} = TEF_{4,3}$ . Gray shading demonstrates the mooring sample area for  $TEF_{MN}$  assuming that model fields  $u$  and  $S$  are constant across  $\sigma$ -levels over distance  $\Delta x_m$ . In (b) the filled red dots and open purple circles indicate sample locations for currents and salinity, respectively for  $TEF_{cs}$ .

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### 2.3.4 Case study, $TEF_{cs}$

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A case study was designed to imitate a sampling plan where  $S$  and  $u$  observations are not co-located and are constrained by common instrument and deployment logistics (Figure 2b).  $M$  moorings were evenly distributed laterally over the cross-section. At each mooring  $u$  was sampled at the full model resolution, to approximate having a bottom-mounted ADCP and salinity was sampled 1 m off of the bottom and 1 m below sea level, to approximate having a bottom-mounted sensor as well as one mounted from a surface float. From this sampling distribution two variations of vertical salinity interpolation were compared. In case A,  $S$  was linearly interpolated to the velocity sample depths (i.e. the model grid). In case B, a two-layer system was assumed having well-mixed surface and bottom layers each with constant  $S$ . The depth of the boundary between the well-mixed layers was approximated by the mean depth of the 0-crossing between inflow and outflow in the deepest part of the channel (50, 8, and 5 m at sections EH1, CRM, and SDBM, respectively). With observations this interpolation could be calculated during the analysis stage using observed currents, therefore this estimate of the mixed layer depth does not rely on *a priori* knowledge. In reality, vertical patterns of currents and salinities may be decoupled, or may vary over time, however this simplified approach is applied for the case study to be most relevant to an observational study with limited to no *a priori* knowledge about the system. While not always dynamically appropriate, it is a reasonable simplified approach for many estuarine systems (e.g. Lerczak et al., 2006; Aristizábal & Chant, 2015). Tests with extrapolated currents in the top and bottom 10% of the water column, and 2-5 m from the bottom—to simulate ADCP limitations—had negligible impact on the results.

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### 2.3.5 Discrete calculations

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For the discrete calculation of Equation 3 applied to each  $TEF$  method the spatial coordinates were first converted to isohaline coordinates. The salinity range was defined by the minimum and maximum salinity sampled at each cross-section over the year. This range was divided into  $N_{bins} = 500$  evenly spaced salinity bins. The currents  $u$ , salinity  $S$ , and area  $A$  were interpolated onto the spatial grid defined for each method

304 (e.g. Figure 2) and then mapped into these discrete salinity bins at each time prior to  
 305 the calculation of  $TEF$  (Equation 3).

### 306 **3 Exchange Flow**

307 The estuaries and individual cross-sections chosen for this study differ in the de-  
 308 gree of stratification and shear, the range of seasonal and tidal variability, as well as in  
 309 the channel width and bathymetric complexity. These features contribute to differences  
 310 in  $TEF$ . Before presenting a comparison of the  $TEF$  calculated by different methods  
 311 and resolutions, the characteristics of the salinity, along-channel currents, and  $TEF$  at  
 312 the full model resolution is discussed.

#### 313 **3.1 Strait of Juan de Fuca**

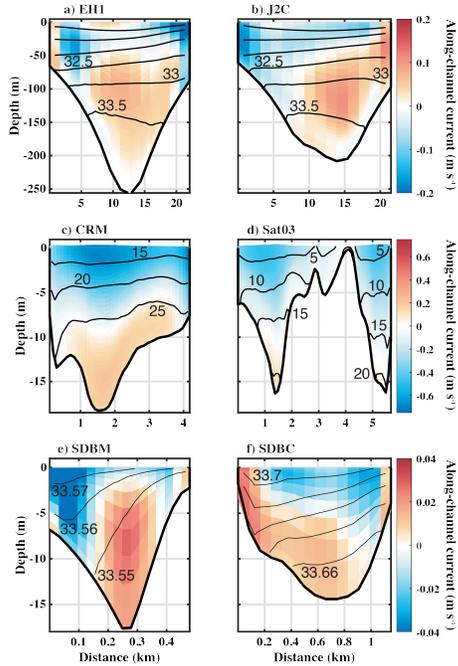
314 Annual mean currents and salinity in the Strait of Juan de Fuca generally exhibit  
 315 a classical pattern of estuarine circulation with outflow and relatively fresher water near  
 316 the surface as well as inflow and saltier water at depth (Figure 3a). Annual mean  $TEF$   
 317 was also predominantly two-layered, with outflow of fresher water and inflow of relatively  
 318 saltier water (Figure 5d,e). Occasional intrusions of the Columbia River plume during  
 319 prolonged downwelling-favorable winds (Hickey et al., 2009; Giddings & MacCready, 2017)  
 320 were apparent in the mean currents as an upstream flow near the surface at EH1 (Fig-  
 321 ure 3a); there was also a small, intermittent intrusion of fresher coastal water (<30 psu),  
 322 associated with increased horizontal shear (Giddings & MacCready, 2017). This small  
 323 incoming layer is distinguished as a third inflow layer in the  $TEF$  analysis at EH1 (Fig-  
 324 ure 5a,d). This fresh surface inflow was not evident in the mean currents at J2C (Fig-  
 325 ure 3b, Figure 5e), which is further from the mouth. Neither the salinity (Figure 4a) nor  
 326 stratification (Figure 4b) had strong subtidal variability, although temperature had a greater  
 327 contribution to the vertical stratification in the latter months of the year than in the early  
 328 spring.

#### 329 **3.2 Columbia River**

330 At CRM near the Columbia River mouth there was also a mean outflow of fresher  
 331 water near the surface and saltier near-bottom inflow (Figure 3c).  $TEF$  reflects this two-  
 332 layered exchange with outflow of fresher water and inflow of relatively saltier water (Fig-  
 333 ure 5f,g). At CRM there was a seasonal cycle in the salinity with a half-amplitude of 5 psu,  
 334 similar to the mean tidal range (the salinity time series was low-pass filtered using a Godin  
 335 filter; the scale of tidal variability is indicated in red). The stratification at CRM had  
 336 large variability (similar to the mean) at seasonal, spring-neap, and higher tidal frequen-  
 337 cies (Figure 4d). Note that stratification is calculated from salinity only. The bathymetry  
 338 is more complicated at Sat03 where a shallow sill bisects the channel. At Sat03 the mean  
 339 inflow is weak and the mean currents are mostly out of the estuary (Figure 3d). The salin-  
 340 ity range of the inflow  $S_{in}$  was greater at Sat03 than at CRM, with more evenly distributed  
 341 transport across the salinity range, while the outflow across Sat03 was predominantly  
 342 at salinities <10 psu. At both sections the outflow  $Q_{out}$  had greater seasonal variabil-  
 343 ity than the inflow  $Q_{in}$  (Figure 5b).

#### 344 **3.3 San Diego Bay**

345 In San Diego Bay, the annual mean exchange was out of the estuary near the sur-  
 346 face, but the surface waters are saltier than at the bottom (Figure 3e,f). In this shal-  
 347 low system with relatively little rainfall, vertical stratification at the mouth is thermally  
 348 controlled much of the year (Chadwick et al., 1996) and varies at seasonal and spring-  
 349 neap tidal frequencies (e.g., Figure 4f). Also, the spatial standard deviation of salinity  
 350 across the section SDBM was small relative to the fluctuation of the mean. Variability

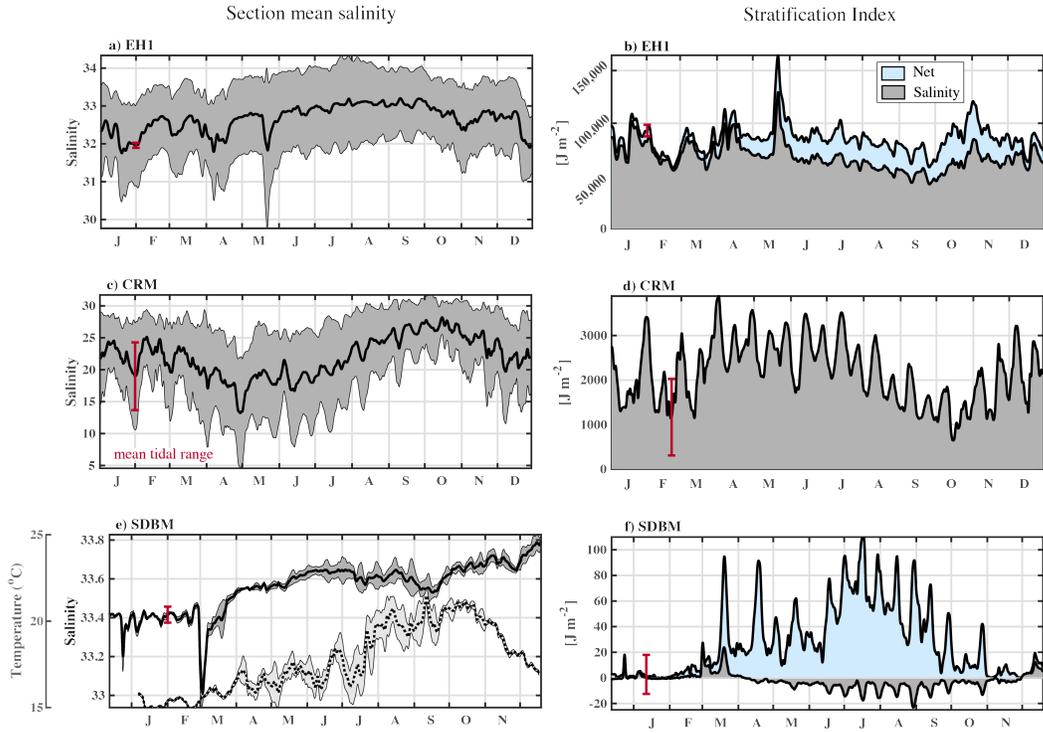


**Figure 3.** Cross-sections of annual-mean fields, taken between the first and last spring high-tide of the year to avoid tidal aliasing. Shading is along-channel currents, with positive values (warm colors) indicating flow into the estuary. Contours are isohalines. Note the axes, isohalines, and color scales on each subplot are different.

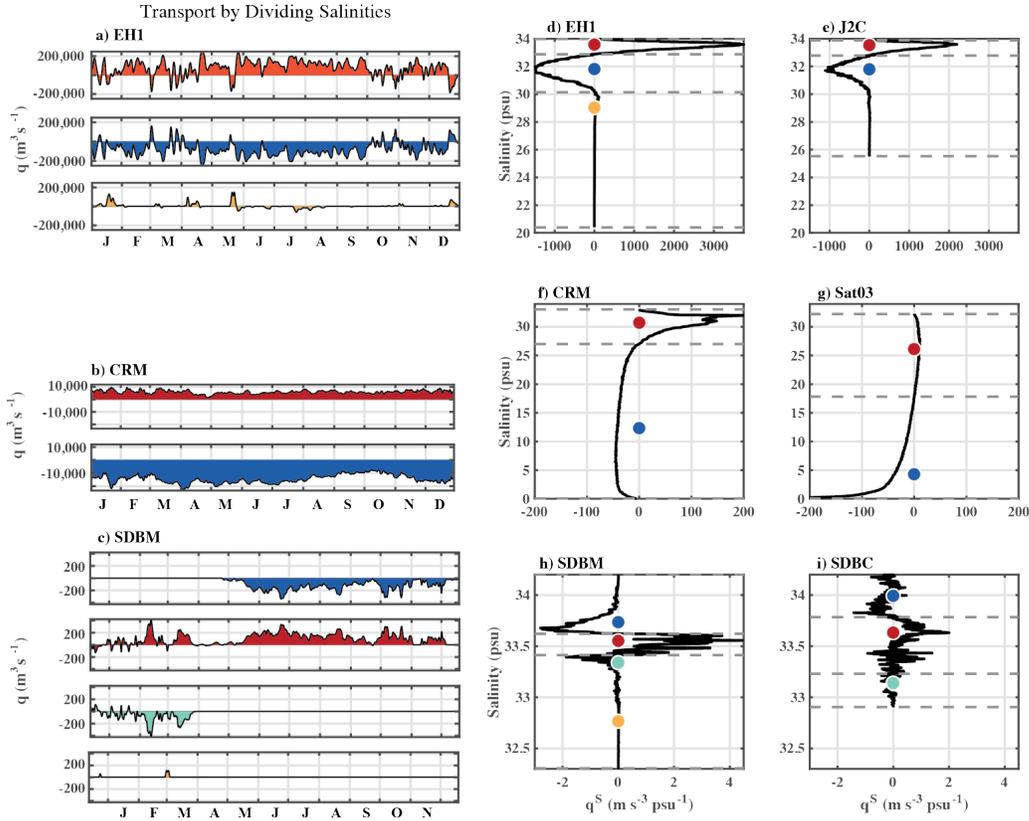
351 in this system is driven by surface heating and evaporation as well as tidal advection and  
 352 river discharge (Largier, 1995; Largier et al., 1997).  $TEF$  was only examined in isoha-  
 353 line coordinates for this study although an analysis in temperature-salinity space is also  
 354 possible (Lorenz et al., 2020). The seasonal variability in the exchange at SDBM near  
 355 the mouth was reflected in the results. In winter there was inflow of relatively saltier wa-  
 356 ter and outflow of fresher water, with a difference between  $S_{in}$  and  $S_{out}$  of  $<0.3$  psu (Fig-  
 357 ure 5c). When the stratification was thermally dominated, the isohaline coordinate  $TEF$   
 358 reversed and the inflow was fresher than the outflow, again by a small margin. This pat-  
 359 tern was similar further upstream at SDBC, but at SDBC the magnitude of the exchange,  
 360 was weaker and to identify three distinct layers of the exchange flow as shown in Fig-  
 361 ure 5i, the threshold used to calculate the dividing salinities was adjusted to  $Q_{percent}=0.1$ ,  
 362 instead of 0.01. For consistency in the following sections, the same value of  $Q_{percent}=0.01$   
 363 was used for all of the simulations. The time series of  $TEF$  parameters  $Q_{in}$  and  $S_{in}$  ( $Q_{out}$   
 364 and  $S_{out}$ ) were summed (and averaged, following Equation 8) across all inflow (outflow)  
 365 layers and were not sensitive to  $Q_{thresh}$ .

### 366 3.4 Horizontal and vertical shear

367 The degree of horizontal shear relative to vertical shear is likely important for es-  
 368 timating how many moorings and sample depths would be needed to capture the exchange  
 369 flow. Following the model developed by Valle-Levinson (2008),  $K_e$  and  $E_k$  can be used  
 370 to evaluate the degree of horizontal and vertical shear at a channel section. At EH1 and  
 371 J2C the channel is wide enough that the Coriolis force can be important ( $K_e \sim 2$ ) as  
 372 can be seen in the tilted isopycnals at J2C (Figure 3b). A detailed analysis of the mech-  
 373 anisms driving  $TEF$  at EH1 and J2C found that the flow was in geostrophic balance,  
 374 but also that temporal changes were driven by the baroclinic pressure gradient and ad-



**Figure 4.** Subtidal salinity variability spatially averaged across sections (a) EH1, (c) CRM, and (e) SDBM. Gray shading is the standard deviation of the spatial mean. Temperature variability is also shown for (e) SDBM, where temperature can dominate the stratification. Right panels show the stratification index  $\phi$  at (b) EH1, (d) CRM, and (f) SDBM. Gray shading is the stratification due to salinity, while blue shading is the net stratification index. The red vertical bar on each plot shows the mean range of these values over a tidal cycle.



**Figure 5.** The profile of annual-mean salinity exchange across each model section. Dashed lines mark the dividing salinities between inflow and outflow layers and colored dots indicate  $S_{in}$  (warm colors) and  $S_{out}$  (cool colors) values. Note that the exchange threshold used to calculate the transport in each layer following Equation ( 3) is  $Q_{thresh} = 0.01$ , except as shown for cross-section SDBC (i), where for the same threshold there were 18 dividing salinities. For SDBC shown here  $Q_{thresh}$  is 0.1.

**Table 2.** Oceanographic characteristics, including the time and area-averaged mean salinity, and along-channel current magnitude, the principle axis of the currents, and the Kelvin and Ekman numbers calculated at each cross section from unfiltered time series. Along-channel flow was defined as positive flowing into the estuary normal to the cross-section and the principle axis is reported here as degrees counter-clockwise from section normal (with the normal vector directed into the estuary).

Run	mean $S$ psu	range $S$ psu	mean $ u $ ( $\text{m s}^{-1}$ )	Princ.Ax. degrees	Ke	Ek
SJF (EH1)	32.7	[20.4, 34.0]	$0.34 \pm 0.26$	3	1.68	$6.03 \times 10^{-4}$
SJF (J2C)	32.6	[25.5, 33.9]	$0.36 \pm 0.26$	-22	1.81	$1.07 \times 10^{-3}$
CR (CRM)	21.8	[0.0, 33.1]	$0.73 \pm 0.54$	26	0.46	
CR (Sat03)	8.7	[0.0, 32.3]	$0.58 \pm 0.71$	4	0.85	
SDB (SDBM)	33.6	[32.2, 34.2]	$0.26 \pm 0.18$	1	0.57	0.17
SDB (SDBC)	33.7	[32.9, 34.4]	$0.16 \pm 0.11$	-14	1.36	0.22

vection (Giddings & MacCready, 2017). The eddy viscosity was not available from the Columbia River SELFE model, and  $E_k$  could not be explicitly calculated. The low  $Ke < 2$  suggests that horizontal shear is likely to be small relative to vertical shear. However, at Sat03 the steep bathymetry that divides the flow between two channels contributes to across-estuary shear. While the role of friction is expected to be stronger in the shallow San Diego Bay, neither  $E_k$  or  $Ke$  values indicate strong horizontal shear; however  $E_k$  approaches the moderate frictional regime where both horizontal and vertical shear can be found (Valle-Levinson, 2008). Channel curvature (Figure 1) may also contribute to the observed across-channel shear and salinity gradients (Figure 3e,f). Overall, all of the cross-sections presented here exhibit both horizontal and vertical shear.

#### 4 Evenly Distributed Moorings, $TEF_{MN}$

One goal of this analysis was to understand how quickly various sampling approaches will converge to  $TEF_{IJ}$ . A first approach,  $TEF_{MN}$ , used evenly distributed samples (e.g., Figure 2a). Due to variability in the exchange flow parameters with the number of moorings as well as sample depths (Figure 6), it takes some care to identify the minimum number of moorings and samples required to capture  $TEF_{IJ}$  across each section. An appropriate definition of convergence between these methods could depend on the specific application or research question. One comparison that demonstrated utility was to identify the threshold for which the magnitude of each  $TEF_{MN}$  parameter consistently remained within  $\leq 10\%$  of the  $TEF_{IJ}$  value (Table 3). However, 3 psu was large relative to the observed salinity range (Figure 5). In order to normalize salinity values for this comparison,  $S_{in,out}$  were converted to their equivalent freshwater fraction using the tidal maximum salinity over time at each cross-section,  $S_{max}$  by  $FWF_{in,out} = (S_{max} - S_{in,out})(S_{max})^{-1}$ .

$TEF_{MN}$  parameters  $Q_{in,out}$  converged towards  $TEF_{IJ}$  parameters  $Q_{in,out}$  as the number of moorings  $M$  increased as long as there was a minimum number of sample depths  $N$  at each mooring (Figure 6). The minimum threshold of sample depths for convergence between  $TEF_{MN}$  and  $TEF_{IJ}$  was  $N \leq 4$  in most cases (Table 3). Sampling the model fields at fewer depths resulted in smaller  $Q_{in,out}$  magnitudes. For a small number of sample depths, e.g.  $N < 4$ ,  $TEF_{MN}$  diverged from  $TEF_{IJ}$  and approached 0 as the number of moorings  $M$  increased. At sections EH1 and SDBM  $Q_{in,out}$  depended on whether the channel center was sampled; the magnitudes were overestimated if only a single centered mooring was sampled and were underestimated if only two moorings were sampled

**Table 3.** For each parameter ( $M, N$ ) are the minimum number of evenly distributed moorings ( $M$ ) and the minimum number of depth samples ( $N$ ) for which the magnitude of  $TEF_{MN}$  parameters converge towards the  $TEF_{IJ}$  values calculated at the full model resolution. Convergence is defined here by  $Q_{in,out}$  and the freshwater fraction equivalent of  $S_{in,out}$  consistently reaching within 10% of the full model value. These estimates of  $M$  are conservative and in some cases  $M$  can be lower such as with a greater number of sample depths  $N$  (as indicated in each footnote).

Run	$Q_{in}$ $M$	$Q_{out}$ $M$	$FWF_{in}$ $M$	$FWF_{out}$ $M$	$Q_{in}$ $N$	$Q_{out}$ $N$	$FWF_{in}$ $N$	$FWF_{out}$ $N$
SJF (EH1)	3	4	3 <sup>a</sup>	2 <sup>b</sup>	4	4	5	2
SJF (J2C)	4 <sup>c</sup>	6 <sup>d</sup>	3	1	4	4	8	3
CR (CRM)	3	1	1	1	2	2	2	2
CR (Sat03)	7 <sup>e</sup>	3	6 <sup>f</sup>	1	3	1	1	1
SDB (SDBM)	3	4	1	1	4	4	1	1
SDB (SDBC)	3	7	1	3 <sup>g</sup>	4	4	1	1

<sup>a</sup>For  $N \geq 5$ ,  $S_{in}$  also converges for  $M = 1$ , but not  $M = 2$ .

<sup>b</sup>For  $N \geq 6$ ,  $S_{out}$  also converges for  $M = 1$ .

<sup>c</sup>For  $N \geq 4$ ,  $Q_{in}$  converges for  $M = 3$ , but not  $M = 4$ .

<sup>d</sup> $Q_{out}$  also converges with as few as  $M = 4$  moorings for  $N \geq 7$ .

<sup>e</sup>For  $N \geq 3$ ,  $Q_{in}$  also converges for  $M = 5$ , but not  $M = 6$ .

<sup>f</sup> $S_{in}$  also converges for  $M = 3$ .

<sup>g</sup> $S_{out}$  also converges for  $M = 1$ , but not  $M = 2$ .

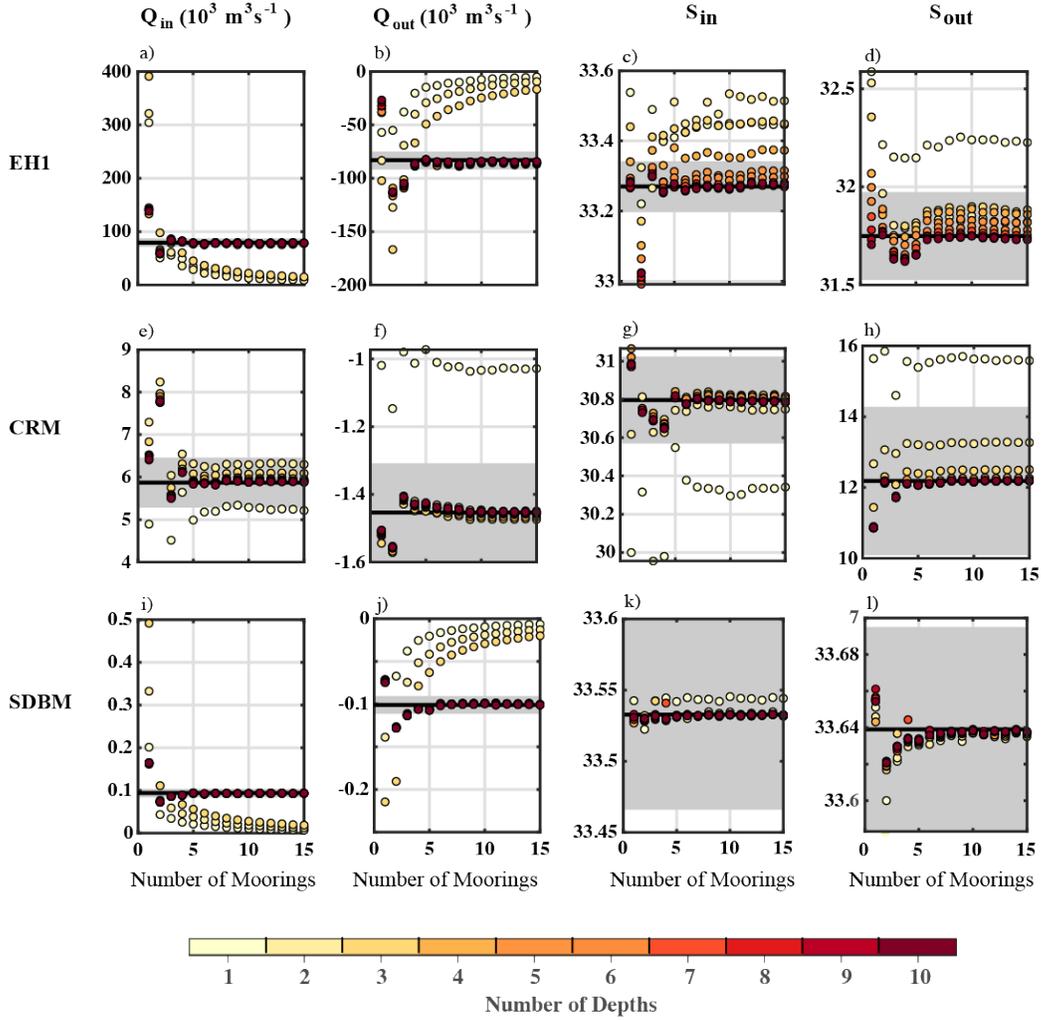
407 (i.e., not sampling the channel center; Figure 6). Similarly, the variance of  $Q_{in,out}$  that  
 408 was captured when two moorings span the channel center was less than with a single cen-  
 409 tral mooring (Figure 7).

410 The deviations of  $S_{in,out}$  from  $TEF_{IJ}$  values were relatively small (Figure 6)—within  
 411 10% of the freshwater fraction even with only a single mooring, in most cases (Table 3).  
 412 The salinity range associated with  $\pm 10\%$  of the freshwater fraction is smaller for  $S_{out}$   
 413 because the freshwater fraction itself is smaller; this may contribute to the higher value  
 414 of  $N$  at EH1 and J2C.

415 If too few depths were sampled the annual mean magnitude of  $Q_{in,out}$  and  $S_{in,out}$   
 416 calculated from  $TEF_{MN}$  did not converge to  $TEF_{IJ}$ , but the number of sample depths  
 417 required for convergence did not change as  $M$  increased (e.g. Figure 6a). In most cases  
 418 the  $TEF_{MN}$  parameters  $S_{in,out}$  converged towards  $TEF_{IJ}$  parameters at values of  $M$   
 419 and  $N$  that were similar to or smaller than the number of moorings and sample depths  
 420 over which  $Q_{in,out}$  converged. Thus,  $Q_{in,out}$  were the limiting parameters with the  $TEF_{MN}$   
 421 method.

## 422 5 Strategically Distributed Moorings, $TEF_{\mu J}$

423 In order to assess the sensitivity of  $TEF$  on sampling method, a strategic approach  
 424 to lateral mooring placement rather than a geometric distribution was tested. The strate-  
 425 gic approach,  $TEF_{\mu J}$ , incrementally added moorings that contributed the maximum cor-  
 426 relation improvement between the sampled parameters,  $TEF_{\mu J}$  and  $TEF_{IJ}$ . Due to the  
 427 iterative nature of this method, only variability in the lateral mooring placement was as-  
 428 sessed, and the water column was sampled at the full vertical model resolution,  $N =$   
 429  $J$ . While the maximum correlation for a given number of moorings was slightly less with  
 430 fewer vertical samples ( $N < J$ ), the patterns were similar to those presented here for  
 431  $N = J$ , particularly for  $N \geq 4$ .



**Figure 6.**  $TEF_{MN}$  parameters calculated for  $m=1-15$  evenly distributed moorings across the channel and  $n=1-10$  depths evenly distributed in the vertical direction along each mooring at sections (a-d) EH1, (e-h) CRM, and (i-l) SDBM. Colors indicate the number of depths at each mooring. Black lines and gray shading indicate the magnitude of each  $TEF_{IJ}$  parameter and the  $\pm 10\%$  range (calculated using the freshwater fraction for salinities), respectively.

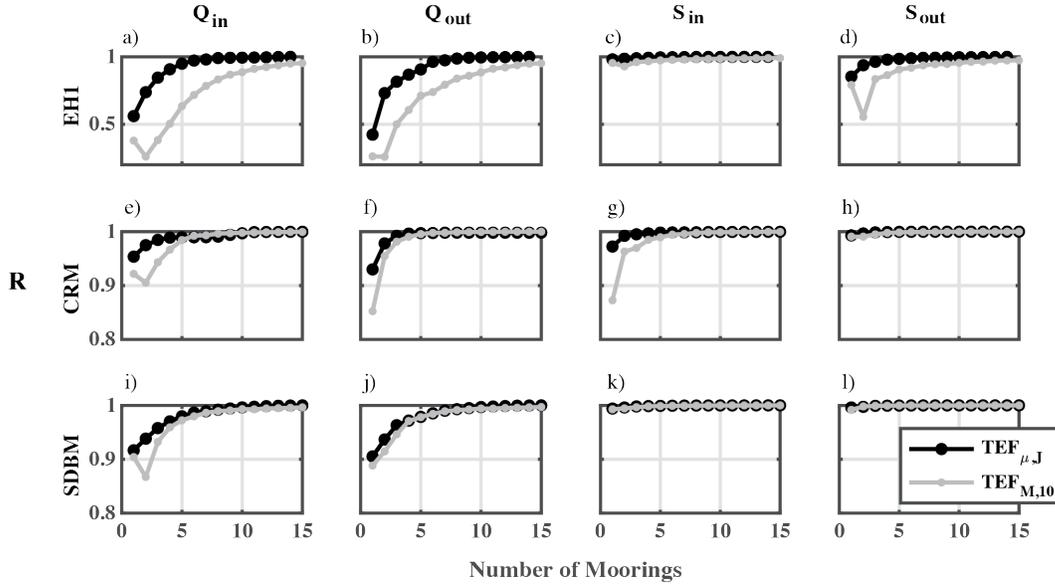
Using the time series correlation to strategically select lateral mooring locations, the correlation between the sub-sampled  $TEF_{\mu J}$  and  $TEF_{IJ}$  parameters converged to 1 for fewer moorings than  $TEF_{MN}$  parameters (Figure 7). However, this distinction is minimal for the cross-sections in the Columbia River and San Diego Bay (e.g. Figure 7e-l) where the correlation is high ( $> 0.8$ ) even when only one or two moorings were used to sample the exchange flow. The high correlation between  $TEF_{IJ}$  and that from a single mooring is likely due to the temporal variability in the exchange flow having a wide spatial signal, i.e. similar temporal variability over the full cross-section. This is opposed to the Strait of Juan de Fuca where temporal variations in exchange have a strong spatial signature, such as those associated with intermittent downwelling-favorable winds (e.g., Figure 3a and Giddings & MacCready, 2017).

The mooring order varied based on which  $TEF$  parameter they were designed to capture (i.e.,  $Q_{in}$ ,  $Q_{out}$ ,  $S_{in}$  or  $S_{out}$ ; Figure 8). However, in general the strategic mooring placement spanned the section. If each cross-section was geometrically divided into thirds, the first triad of moorings (i.e. the three tallest bars of each color) was roughly distributed across those three sections, resulting in a sampling distribution that was similar between  $TEF_{\mu=3,J}$  and  $TEF_{M=3,N}$ . Despite this similarity, and despite the correlation converging more quickly for strategic moorings (Figure 7), the minimum number of moorings  $\mu$  before  $Q_{in,out}$  converged to within 10% of  $TEF_{IJ}$  values was greater for the strategic sampling approach (Table 4) than when using the evenly spaced sampling approach (Table 3). This implies that the areas of high  $TEF$  variance do not fully correspond with the maximum  $TEF$  magnitude.

This investigation of  $TEF_{\mu J}$  provided insight into the sensitivity of the results to the specific sample locations by comparing the variation in results using the mooring order from each  $TEF_{\mu J}$  parameter (Figure 9). The results presented in Figures 7 and 8 and throughout the text primarily focus on the outcomes of each parameter of  $TEF_{\mu J}$  with strategic moorings selected based on that same parameter. This means that the single mooring ( $\mu = 1$ ) with the highest correlation to  $TEF_{IJ}$  for the parameter  $Q_{in}$  is not necessarily the same mooring location with the highest correlation to  $TEF_{IJ}$  for the parameters  $Q_{out}$ ,  $S_{in}$ , or  $S_{out}$ . Similarly, the second mooring ( $\mu = 2$ ) is not necessarily the same across parameters  $Q_{in}$ ,  $Q_{out}$ ,  $S_{in}$ , or  $S_{out}$  and so on. For example, there was some variation in the pattern of  $Q_{in}$  for increasing  $\mu$ , between the mooring order strategically determined using  $Q_{in}$  compared to the mooring order strategically determined using  $Q_{out}$ ,  $S_{in}$ , or  $S_{out}$  (Figure 9a). As the number of moorings increased, the magnitude of  $Q_{in,out}$  and  $S_{in,out}$  converged towards the  $TEF_{IJ}$  results. In most cases the rate of convergence was qualitatively similar regardless of which  $TEF$  parameter was used to determine the mooring order, with some variation in the point at which convergence within 10% of  $TEF_{IJ}$  was reached (Figure 9). One exception that stood out was that the number of moorings ( $\mu$ ) at CRM before  $Q_{in}$  converged to within 10% of the  $TEF_{IJ}$  value was more than double the minimum number of moorings ( $\mu$ ) to reach the same threshold when the mooring placement was optimized for  $Q_{out}$ ,  $S_{in}$ , and  $S_{out}$  (Figure 9e). It is unclear why the mooring placement for  $Q_{in}$  was particularly inefficient for capturing the magnitude of  $Q_{in}$ , but individual components of the exchange flow and salt flux can exhibit different spatial patterns and timescales of variability (e.g. Lerczak et al., 2006), which likely contributed to a misalignment between the locations with the most variance and those with the greatest magnitude in the exchange flow.

## 6 Case Studies: Hypothetical mooring deployments, $TEF_{cs}$

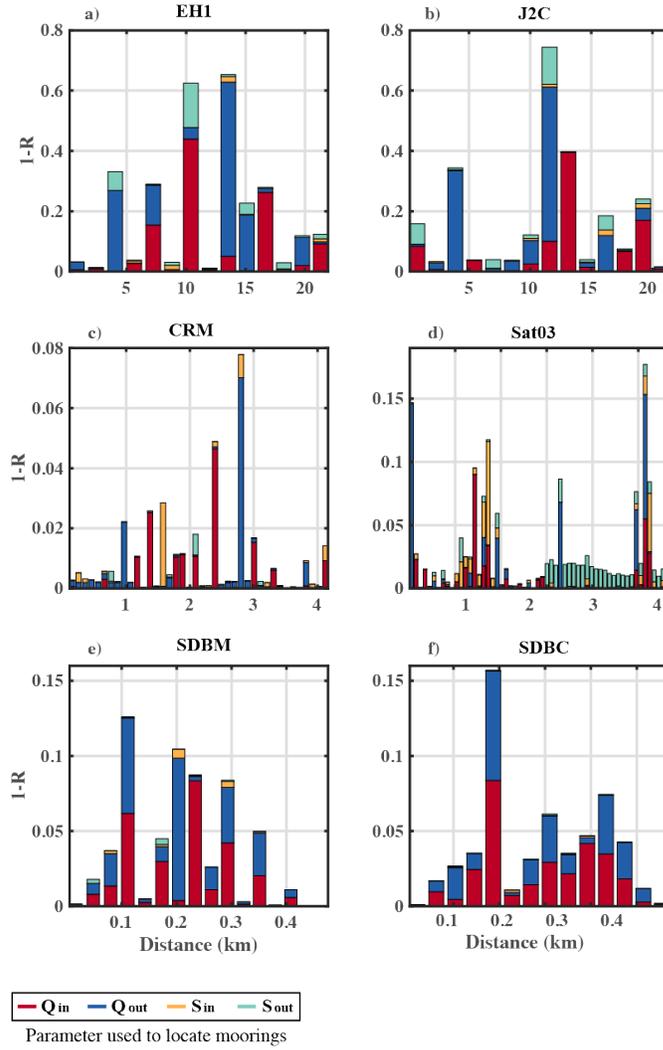
Both evenly and strategically distributed mooring approaches tested here are skewed towards numerical modeling applications because of the sampling resolution, which would require many sensors and mooring lines that extend across nearly the full water column. In particular, strategic mooring placement requires extensive *a priori* knowledge of the



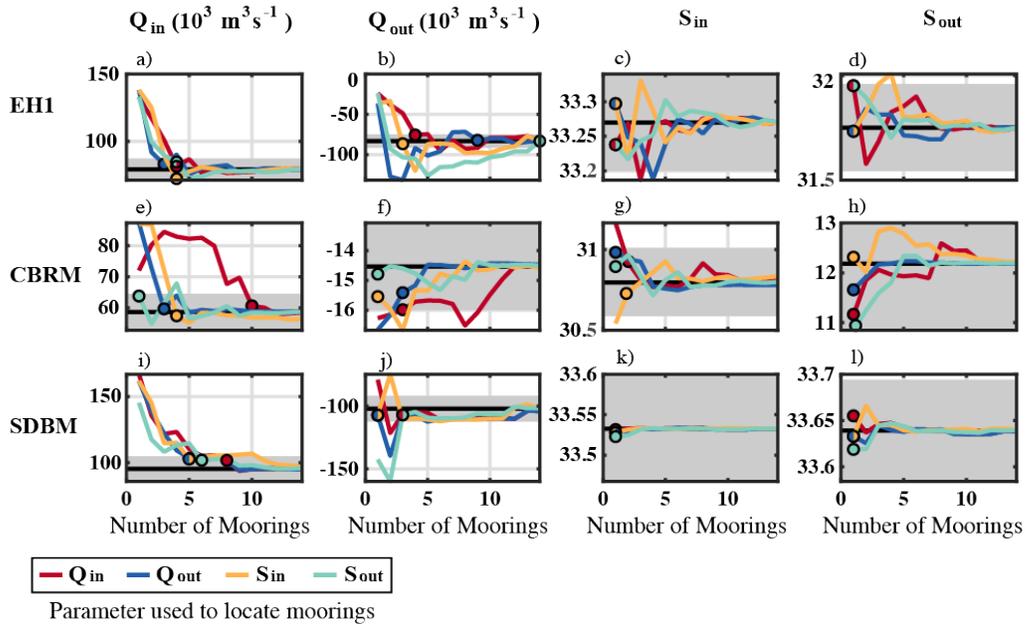
**Figure 7.** The correlation coefficient between time series of parameters calculated from the full model resolution  $TEF_{IJ}$  compared to strategically placed moorings  $TEF_{\mu J}$  (black) and evenly distributed moorings  $TEF_{MN}$  (gray;  $N = 10$ ) as the number of moorings is increased.

**Table 4.** The minimum number of strategically distributed moorings for which the magnitude of  $TEF_{\mu J}$  parameters  $Q_{in,out}$  and  $S_{in,out}$  converge towards the  $TEF_{IJ}$  values calculated at the full model resolution. Convergence is defined here by  $Q_{in,out}$  and the freshwater fraction equivalent of  $S_{in,out}$  reaching within 10% of the full model value.

Run	$Q_{in}$ $\mu$	$Q_{out}$ $\mu$	$FWF_{in}$ $\mu$	$FWF_{out}$ $\mu$
SJF (EH1)	4	9	1	1
SJF (J2C)	6	2	3	1
CR (CRM)	10	3	2	1
CR (Sat03)	16	2	1	1
SDB (SDBM)	8	1	1	1
SDB (SDBC)	7	13	1	1



**Figure 8.** The improvement in the time series correlation between  $TEF_{\mu J}$  and  $TEF_{I,J}$  realized from adding each additional mooring  $\mu$ . The horizontal axis marks the location of each mooring along the cross-section (bathymetry shown in Figure 1). Colors indicate different parameters  $Q_{in}$ ,  $Q_{out}$ ,  $S_{in}$  and  $S_{out}$ .



**Figure 9.**  $TEF_{\mu,J}$  parameters calculated for  $\mu=1-15$  moorings strategically placed across the channel, using the full vertical grid resolution at each mooring. Black lines and gray shading indicate the magnitude of each  $TEF_{IJ}$  parameter and the  $\pm 10\%$  range (calculated using the freshwater fraction for salinities), respectively. Colored lines indicate the parameter used to determine the order of mooring placement by maximizing the correlation between  $TEF_{\mu,J}$  and  $TEF_{IJ}$ . Colored dots mark the first point where each  $TEF_{\mu,J}$  parameter converges to within 10% of the corresponding  $TEF_{IJ}$  parameter (e.g. Table 4)

483 system. In an effort to connect this analysis more closely to observations, a specific case  
 484 study,  $TEF_{cs}$ , was also examined as described in section 2.3.4.

485 The salinity interpolation impacted both the calculated exchange volume transport  
 486 and salinities (Figure 10). Similar to the other sampling methods presented, as the num-  
 487 ber of moorings  $M$  increased,  $TEF_{cs}$  converged to a similar value for  $M \geq 4$ , although  
 488 not necessarily towards  $TEF_{IJ}$ . At SDBM, where salinity stratification is weak,  $TEF$   
 489 for both case studies converged within  $\approx 10\%$  of  $TEF_{IJ}$  with  $\geq 4$  moorings. In other  
 490 words, the vertical salinity interpolation method did not matter. At sections EH1 and  
 491 CRM, however, the results were mixed with the two-layer approximation generally per-  
 492 forming better than the linear interpolation with varying sensitivity.

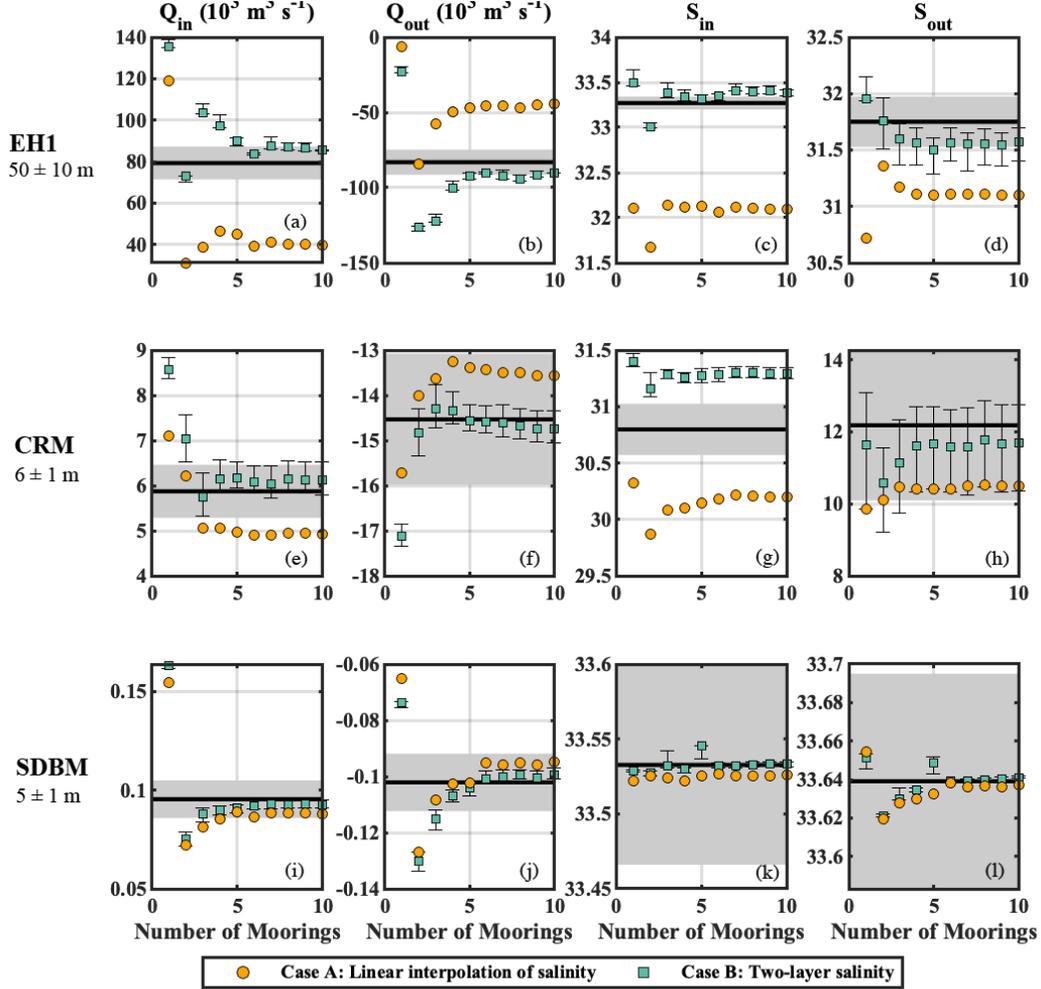
493 At EH1, linearly interpolating salinity led to underestimating the magnitude of the  
 494 exchange volume transport (Figure 10a,b). This is possible when opposing currents are  
 495 classified into the same salinity range, which reduces the net exchange magnitude within  
 496 that salinity class. In this case, a two-layer  $S$  approximation led to values of  $TEF_{cs}$  closer  
 497 to  $TEF_{IJ}$  than a linear interpolation. The results were fairly insensitive to the interface  
 498 depth between the two uniform salinity layers, except for  $S_{out}$ .

499 At CRM, the two-layer  $S$  approximation performed slightly better than the linear  
 500 interpolation (Figure 10c-h). Whether the two-layer approximation resulted in an im-  
 501 provement of  $Q_{in,out}$  over the linear approximation was sensitive to the interface depth.  
 502 In neither case did the  $TEF_{cs}$  exchange salinity  $S_{in}$  converge to within 10% of the full  
 503  $TEF_{IJ}$  value.

## 504 7 Discussion

505 The results of this study suggest that  $TEF$  via *in situ* moorings can be well ap-  
 506 proximated in many situations with  $\leq 4$  moorings across a channel. However, there are  
 507 some limitations to this analysis that may constrain the generalization of these results.  
 508 The magnitude and structure of the exchange flow for each estuary may depend on the  
 509 specific thresholds, time periods, and cross-sections. The sampling approaches explored  
 510 here were limited to a small number of relatively simple designs and were not adapted  
 511 for the particular bathymetry or oceanic conditions at each cross-section. Also the ex-  
 512 amples chosen, while spanning significant parameter space in terms of estuary type and  
 513 geometry, were not exhaustive. The practical applicability of the results presented in this  
 514 study are discussed in the following section in the context of these limitations.

515 The magnitude and structure of the exchange flow for each model may depend on  
 516 specific thresholds, time periods, and cross-sections chosen for the calculation. The  $TEF$   
 517 approach requires the number of salinity bins  $N_{bins}$  and a threshold for identifying the  
 518 cut-off between dividing salinities  $Q_{thresh}$  to be defined. While the dividing salinity ap-  
 519 proach (Lorenz et al., 2019) reduces the sensitivity to these choices (here,  $N_{bins} = 500$   
 520 and  $Q_{thresh} = 0.01$ ), in the weakly stratified San Diego Bay which experiences large  
 521 variability on tidal time scales (e.g. Figure 4f) the number and location of dividing salin-  
 522 ities varied depending on threshold choices (e.g. Figure 5i). However the results presented  
 523 here were summed over the incoming and outgoing layers and were insensitive to these  
 524 choices, even at section SDBC. At the other cross-sections, the results were insensitive  
 525 to variations in  $N_{bins}$  and  $Q_{thresh}$ . Another consideration is that the exchange flow can  
 526 vary over time (Figure 5a-c) such that the  $TEF$  magnitude may be specific to the par-  
 527 ticular periods examined (in particular annual vs. seasonal time periods). The conclu-  
 528 sions drawn from this analysis focus on comparing the relative change in the  $TEF$  across  
 529 methods for the same time period and model cross-section to reduce the dependence on  
 530 the specific time period. However, if only part of the year was examined when surface  
 531 intrusions were absent at *EH1*, for example, the optimal lateral spacing may have been  
 532 impacted.



**Figure 10.** Case study parameters calculated for  $m=1-10$  evenly distributed moorings across the channel with salinity sampled 1 m above the seafloor and below sea level and velocity sampled at the full model vertical resolution. Results are shown from sections (a-d) EH1, (e-h) CRM, and (i-l) SDBM. Two methods of interpolating salinity values to the velocity sampling points along each mooring line are shown: linear interpolation and assuming two mixed layers of uniform salinity with the interface at a fixed depth. Vertical bars mark the range of the results using the range of interface depths as listed for each section ( $50 \pm 10 \text{ m}$ ,  $6 \pm 1 \text{ m}$ , and  $5 \pm 1 \text{ m}$  at EH1, CRM, and SDBM, respectively). Black lines and gray shading indicate the magnitude of each  $TEF_{IJ}$  parameter and the  $\pm 10\%$  range (calculated using the freshwater fraction for salinities), respectively.

533 Variances in the  $TEF$  parameter time series were strongly correlated between meth-  
 534 ods ( $R > 0.85$ ) even with a single mooring, except at EH1 (Figure 7). The high cor-  
 535 relation is likely due to the strong spring-neap and seasonal variability at CRM and SDBM  
 536 (Figure 5) that are spatially coherent as opposed to the seasonal spatial variability caused  
 537 by surface intrusions at EH1. That a single mooring location captures most of the  $TEF$   
 538 variance also suggests that optimizing the correlation may not be a useful method for  
 539 strategic mooring placement to measure exchange flows. Even at EH1 where the vari-  
 540 ance in the exchange flow was improved using the  $TEF_{\mu J}$  method relative to  $TEF_{MN}$   
 541 (Figure 7a,b), the minimum number of moorings for which the  $TEF_{\mu J}$  magnitude con-  
 542 verged to within 10% of  $TEF_{IJ}$  was greater than the minimum number of geometrically  
 543 distributed moorings (Table 3 compared to 4). Also, at CRM, 10 mooring samples were  
 544 required for  $Q_{in}$  using the  $TEF_{\mu J}$  method to converge to within 10% of  $TEF_{IJ}$  result  
 545 (Table 4). These results suggest that the locations across each section with the most tem-  
 546 poral variance are not the same locations where the majority of the transport occurs.  
 547 That in most cases 4 evenly distributed moorings captured  $> 90\%$  of the exchange flow  
 548 (Table 3) and most of the flow variance (Figure 7) also suggests that a straightforward  
 549 sampling plan—as long as there are sufficient number of moorings and depths sampled—  
 550 is likely to capture both features (variance and magnitude) of the exchange. This is also  
 551 supported by the observation that there was generally little difference in the results as  
 552 mooring placement varied (Figure 9).

553 While the selected cross-sections span a bay, salt-wedge, and fjord estuary type with  
 554 differing scales and geometric complexity (Geyer & MacCready, 2014), this subset does  
 555 not comprehensively cover the full range of estuarine shapes, sizes, and dynamics. Nev-  
 556 ertheless, given the range investigated here, the similarity of the results was striking. First,  
 557 only sampling the center of each channel led to an overestimate of  $Q_{in}$  in every case. This  
 558 overestimate could be several times the magnitude of  $Q_{in}$  from the full model resolution.  
 559 In particular in SJF and SDB, with relatively simple U-shaped channels, using only two  
 560 laterally distributed moorings that did not sample the channel center led to an under-  
 561 estimate of  $Q_{in}$ . While the greatest inflow tended to be concentrated towards the cen-  
 562 tral and deeper channel, the lateral distribution of the outflow was more variable (Fig-  
 563 ure 3). Second, it was encouraging—from the perspective of capturing exchange flows  
 564 in a range of systems with limited *a priori* knowledge—that using evenly distributed moor-  
 565 ings performed as well or better than the strategic sampling strategy (Table 3 compared  
 566 to Table 4) and also that the results converged towards  $TEF_{IJ}$  even as the specific sam-  
 567 pling locations were varied (Figure 9). The mean number of evenly distributed lateral  
 568 mooring locations across each channel to resolve ( $Q_{in,out}$ ) to within 10% was  $M = 4.0 \pm 1.8$   
 569 with  $N = 3.3 \pm 1.1$  sample depths evenly distributed across the water column and  $M =$   
 570  $2.0 \pm 1.5$ ,  $N = 2.3 \pm 2.1$  to resolve  $S_{in,out}$  (Table 3).

571 Further studies of systems dominated by either horizontal or vertical shear would  
 572 be needed to assess if the sampling resolution might be reduced for flows with different  
 573  $K_e$  and  $E_k$ . However, the degree of stratification did appear to be important to under-  
 574 stand how the vertical and horizontal sampling resolution impacted  $TEF$ . At EH1 and  
 575 CRM, fewer moorings  $M$  were needed for  $TEF_{MN}$  to converge with evenly distributed  
 576 moorings (Figure 6) than for  $TEF_{cs}$  in case study B (Figure 10). This suggests that when  
 577 stratification is important more vertical resolution of  $S$  reduces the number of moorings  
 578 needed across the channel. Also, at these stratified sections the result depends on the  
 579 vertical interpolation of  $S$ . In contrast, at SDBM the number of moorings needed to ac-  
 580 curately calculate  $TEF_{MN}$  (Figure 6) was similar to  $TEF_{cs}$  (Figure 10) and was simi-  
 581 lar between cases A and B. At SDBM with  $u$  and  $S$  samples evenly distributed verti-  
 582 cally the exchange volume flux only converged for  $N \geq 4$ , while in the case studies the  
 583  $TEF_{cs}$  exchange volume fluxes converged to  $TEF_{IJ}$  when  $S$  was sampled at only two  
 584 depths, near-surface and near-bottom.

585 Given that only 3 estuaries (6 cross-sections) were examined here and that this study  
 586 utilized numerical model output, the question remains: how realistic would it be to ap-  
 587 ply the results of these experiments to observations and in other systems? One general  
 588 limitation of observational studies is the cost of moorings (anchors, line, floats, etc.) as  
 589 well as of the individual sensors. The results of this study may be useful to constrain an  
 590 estimate of the sign and possibly the magnitude of the error for sampling studies with  
 591 fewer moorings. For example, one could extrapolate that a single mooring centered in  
 592 a deep part of the channel is likely to overestimate the magnitude of the exchange vol-  
 593 ume transport. The outcome of the case studies also suggest that the salinity does not  
 594 have to be sampled at the same resolution as the currents to estimate the exchange flow  
 595 with relatively high accuracy, although as stratification increases, identifying the best  
 596 salinity interpolation remains a challenge (Aristizábal & Chant, 2015). In channels with  
 597 high ship traffic, near-surface measurements can be particularly challenging. However,  
 598 it may be encouraging that the results here demonstrated relatively little sensitivity to  
 599 the specific sample location, i.e., one could place a mooring just outside a shipping lane.

## 600 8 Conclusions

601 Exchange between estuaries and the coastal ocean or through inland seas is an im-  
 602 portant driver of the circulation, mixing, biology, and chemistry on both ends of the ex-  
 603 change. Significant progress has been made in calculating this exchange in estuarine con-  
 604 ditions with time-varying stratification and flow, strong vertical and lateral velocity shear  
 605 and salinity gradients, and complex bathymetry using the *TEF* method (Lorenz et al.,  
 606 2019, 2020). However, application of this theory has predominantly utilized numerical  
 607 models where the salinity and velocity are highly resolved in space and time. In this anal-  
 608 ysis *TEF* was calculated using various methods to sub-sample realistic numerical mod-  
 609 els in order to understand the sensitivity of *TEF* and to develop recommendations for  
 610 minimal sampling thresholds that accurately reproduce the exchange flow. Three dif-  
 611 ferent estuaries were examined, including San Diego Bay, the Columbia River, and the  
 612 Salish Sea exchange through the Strait of Juan de Fuca. These examples span a range  
 613 of estuary types (bay, salt-wedge, and fjord, respectively), scales, depths, and channel  
 614 bathymetries. Evenly distributed sample locations across the channel, representative of  
 615 moorings, was the most efficient way to capture the *TEF*. Three to four moorings were  
 616 typically the minimum lateral sample distribution required to capture  $\geq 90\%$  of the ex-  
 617 change transport rate  $Q_{in}$  and  $Q_{out}$ . In most cases, the exchange volume transport was  
 618 the limiting parameter and fewer moorings were required to capture  $\geq 90\%$  of the exchange  
 619 flow salinities  $S_{in}$  and  $S_{out}$ . The minimum vertical resolution to capture  $\geq 90\%$  of the  
 620 *TEF* was similar,  $N \geq 4$ , and was also limited by  $Q_{in}$  and  $Q_{out}$ . Although the exchange  
 621 calculated by these methods is also dependent on resolving the salinity, the *TEF* was  
 622 less sensitive to resolving salinity at the same vertical resolution as velocity and less sen-  
 623 sitive to the salinity interpolation method in systems where there was less vertical strat-  
 624 ification (e.g. the San Diego Bay, as compared to the Columbia River or the Strait of  
 625 Juan de Fuca). The *TEF* could be reproduced by resolving the currents throughout the  
 626 water column and only sampling salinity near the surface and bottom. In comparison  
 627 to geometrically distributing moorings across the channel, strategic sampling based on  
 628 capturing the temporal exchange flow variance did not improve the ability to capture  
 629 the exchange flow magnitude, likely a result of the fact that locations of strongest ex-  
 630 change flow are often not the locations with the highest variance. This method also re-  
 631 quires *a priori* knowledge of the flow field, and is ambiguous depending on which aspect  
 632 of the *TEF* (i.e.  $Q_{in,out}$  or  $S_{in,out}$ ) is used to calculate the variance and is therefore not  
 633 a recommended approach for estuary sampling methodology. Overall the results presented  
 634 here are promising suggesting that *TEF* can be captured well with a reasonable num-  
 635 ber of cross-sectionally distributed moorings and sampling depths and can be used to  
 636 estimate the sign and magnitude of errors. Future work to examine the exchange flow  
 637 through a wider range and greater number of estuaries and channels could be useful to

638 further refine the number of moorings and the approach to determine the best salinity  
639 interpolation.

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659 and analysis code used in this analysis are being organized to be hosted at the UCSD  
660 Library Digital Collections. This is PMEL Contribution 5370.

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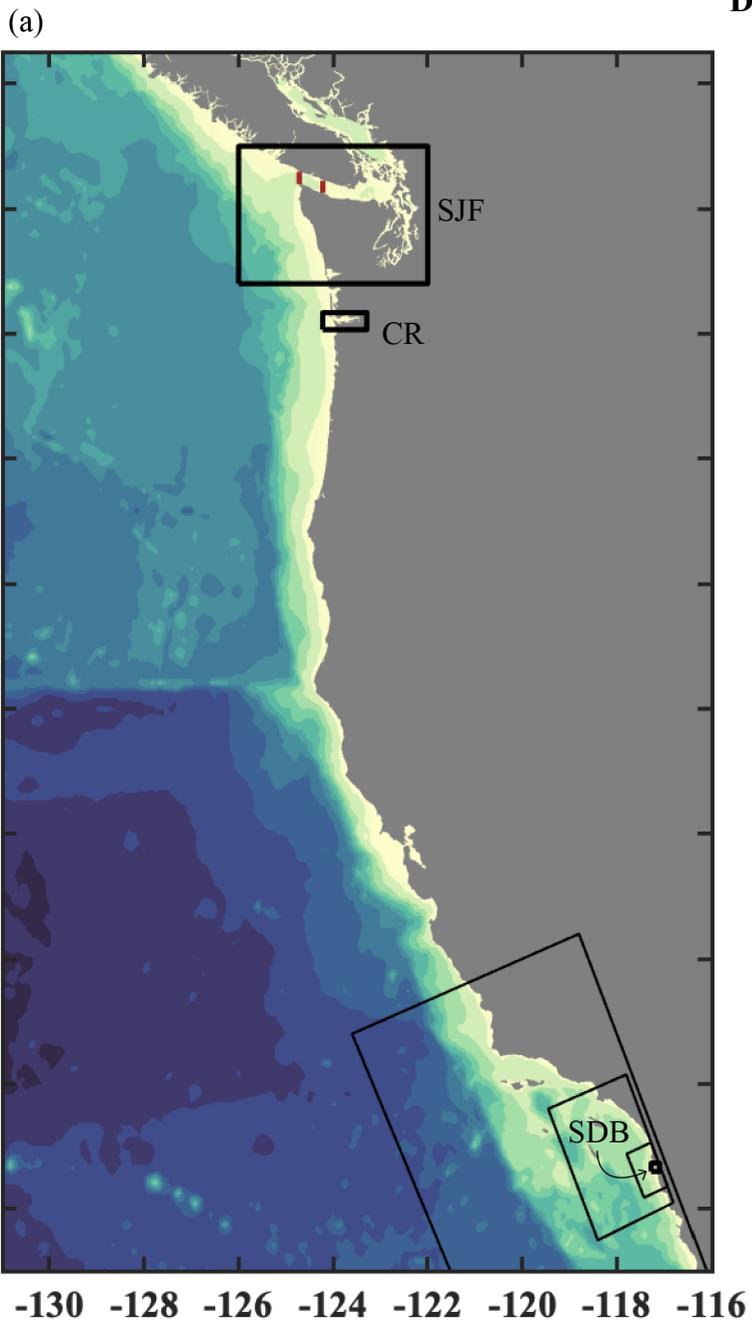
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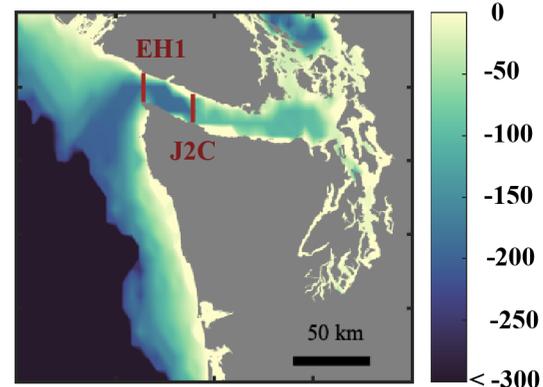
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Figure 1.

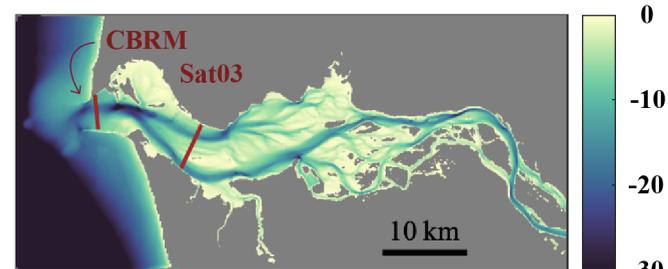
Depth (m)



(b) SJF



(c) CR



(d) SDB

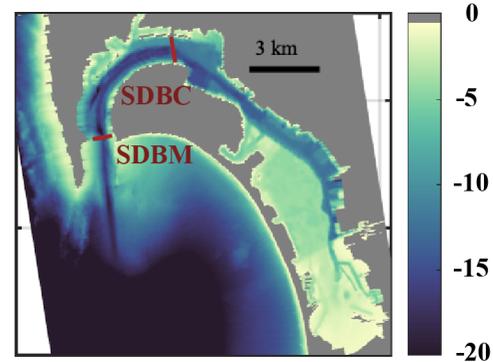


Figure 2.

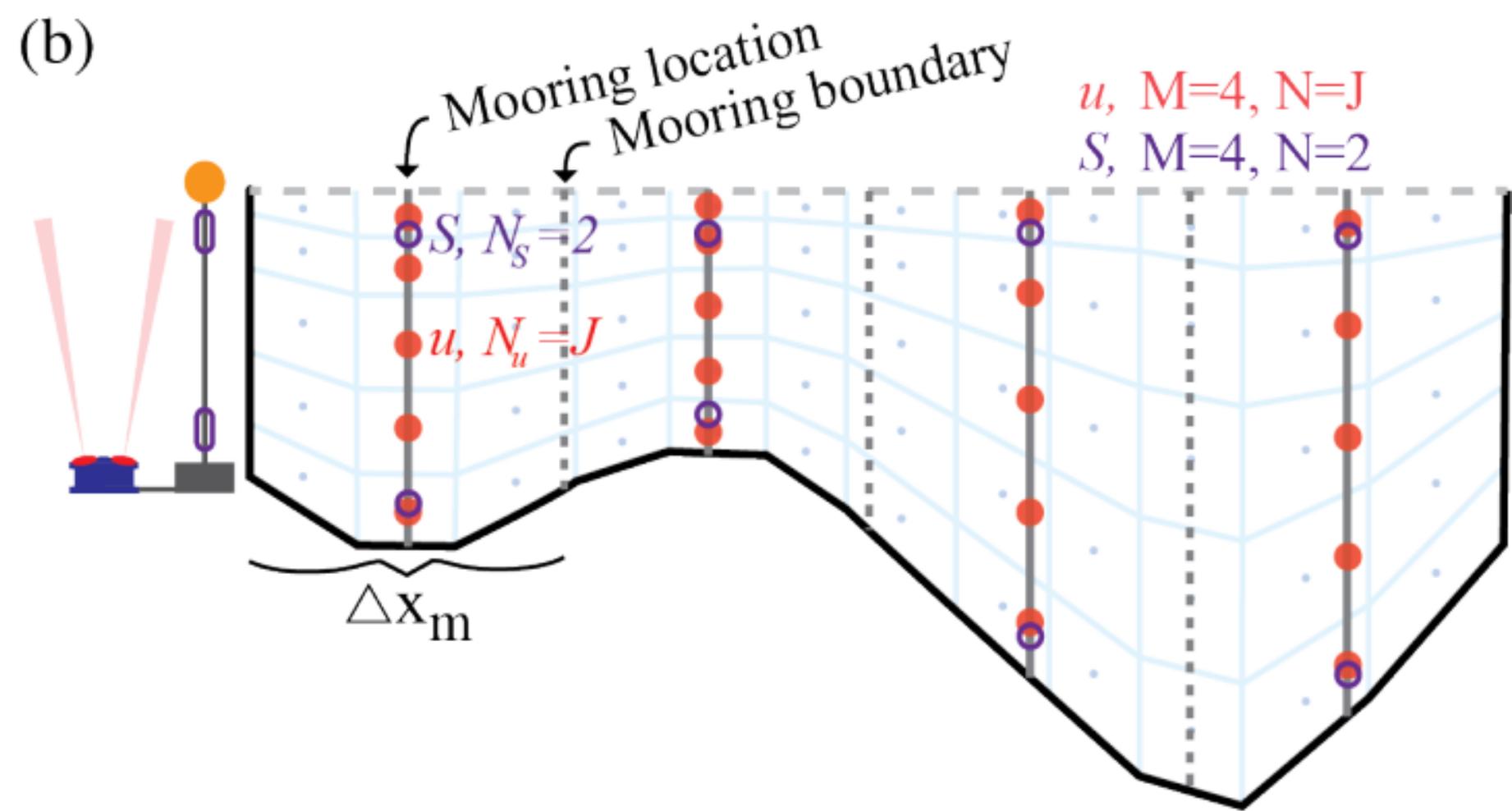
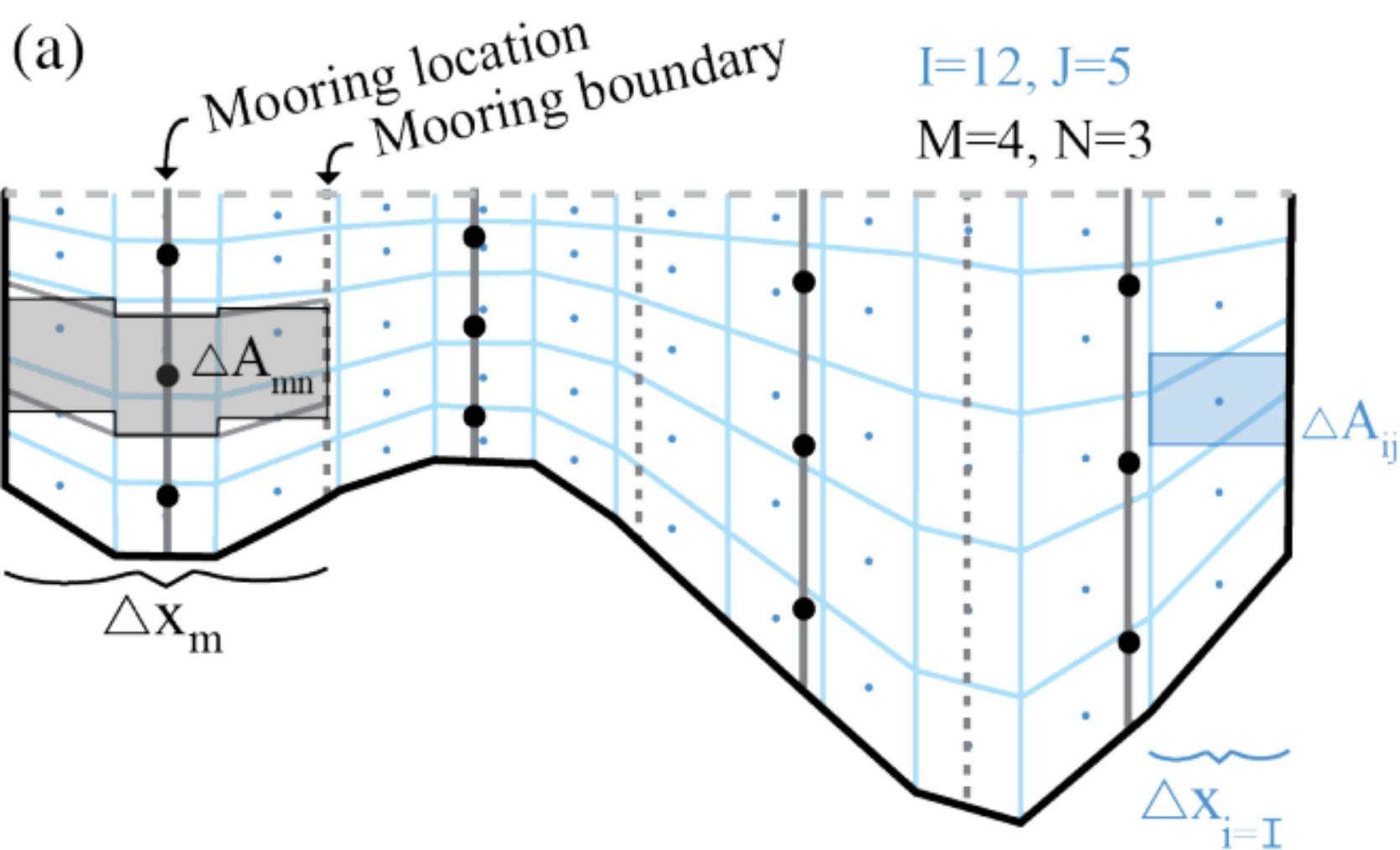


Figure 3.

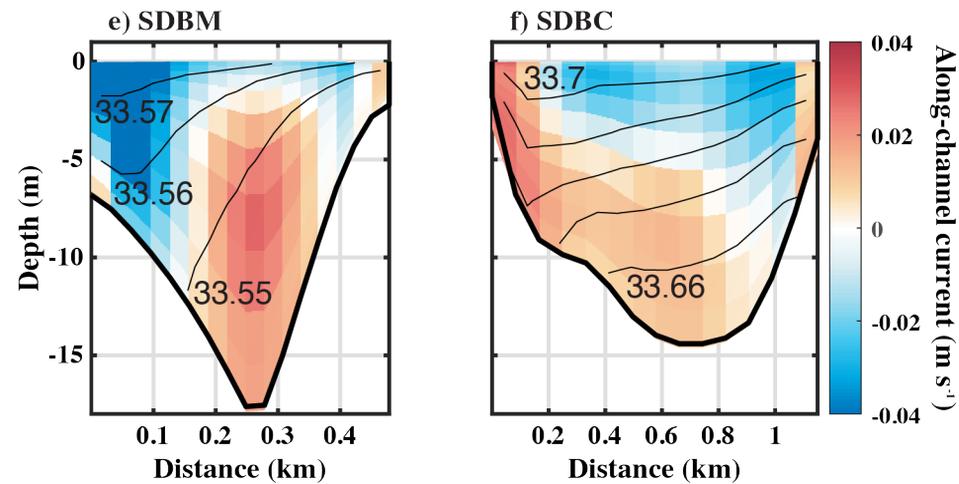
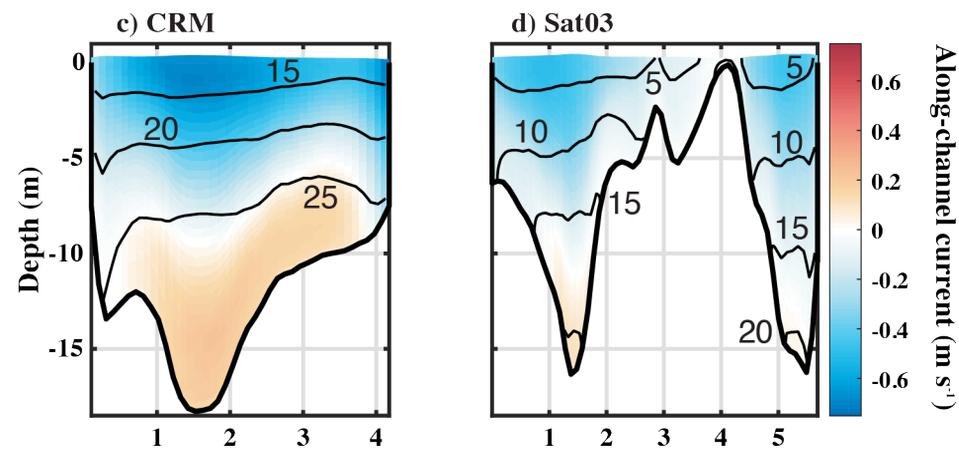
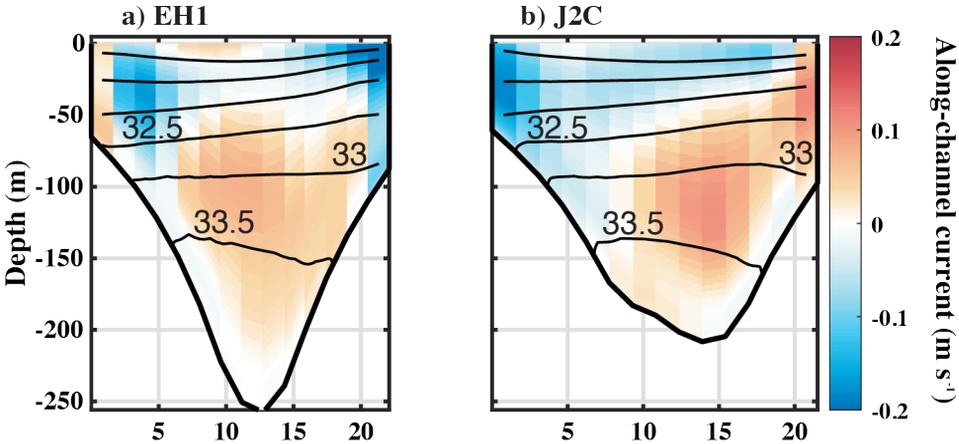
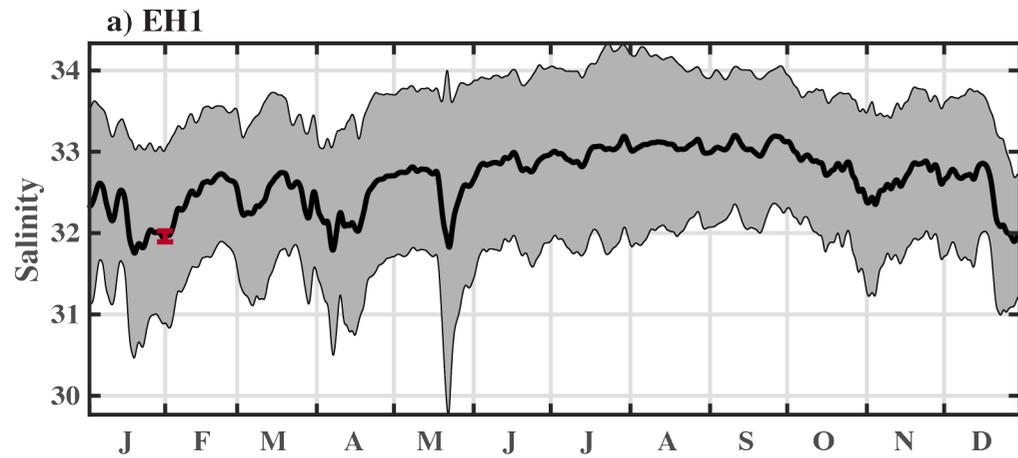


Figure 4.

## Section mean salinity



## Stratification Index

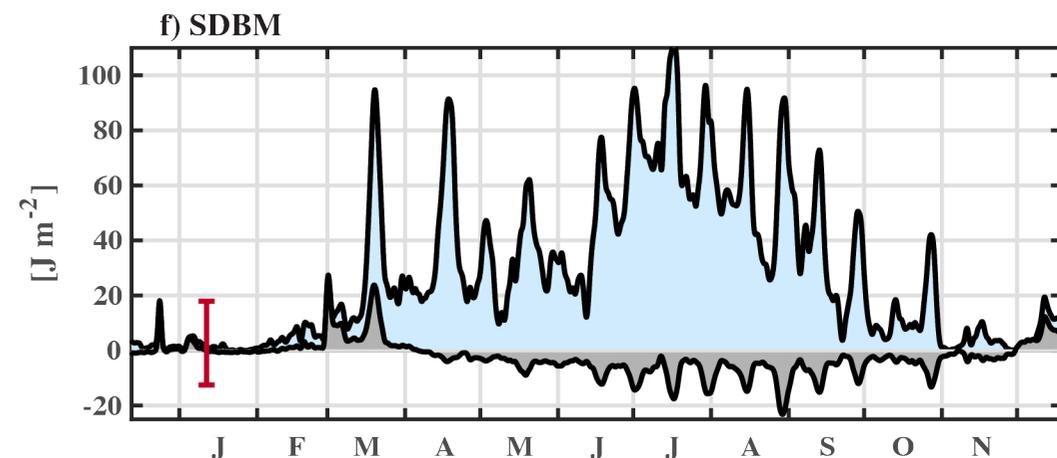
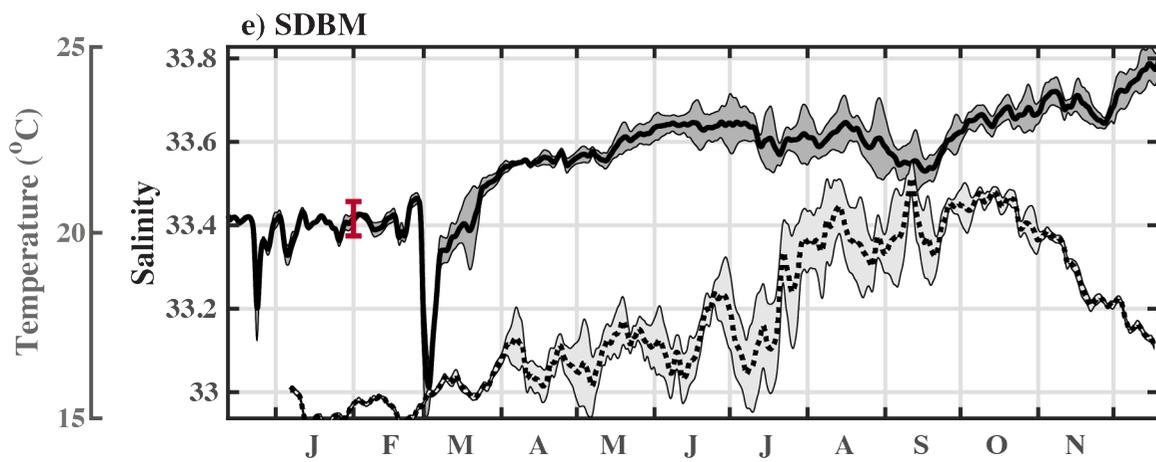
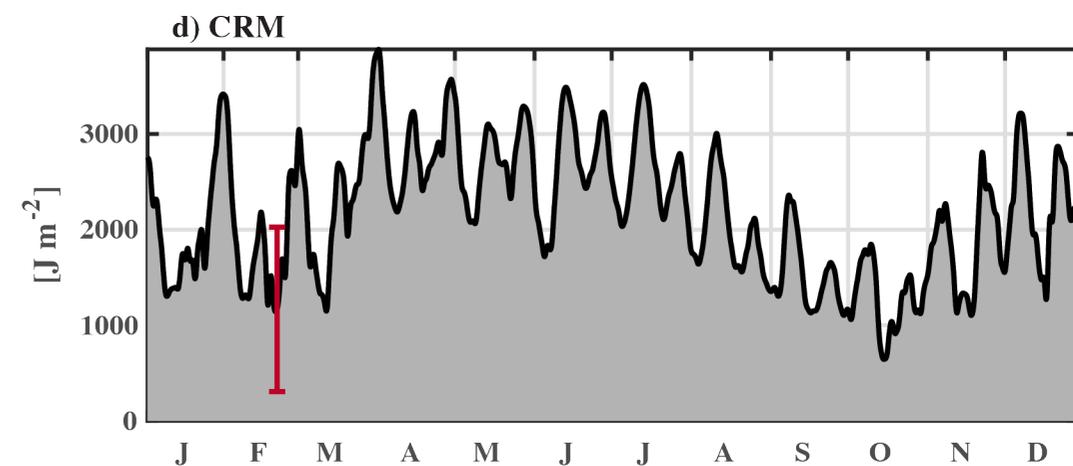
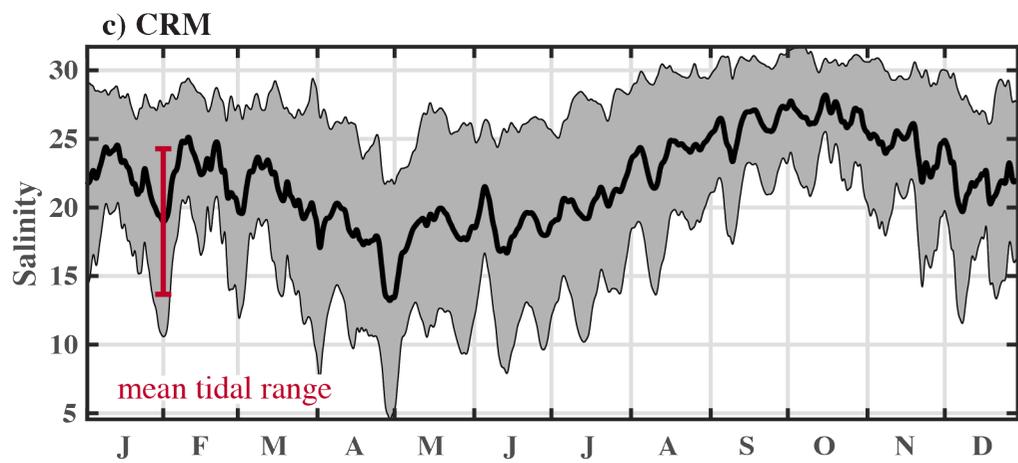
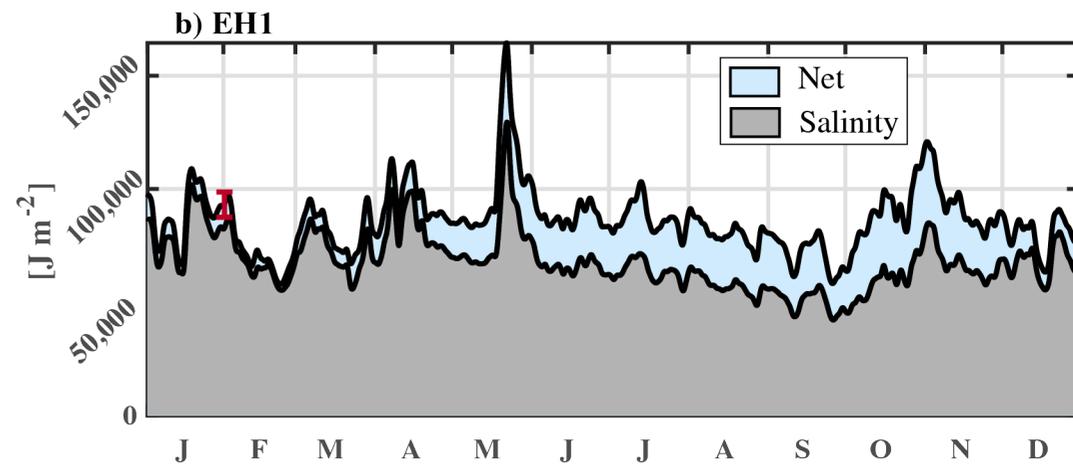


Figure 5.

# Transport by Dividing Salinities

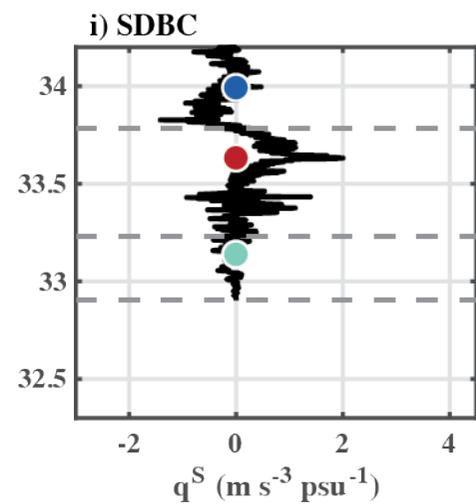
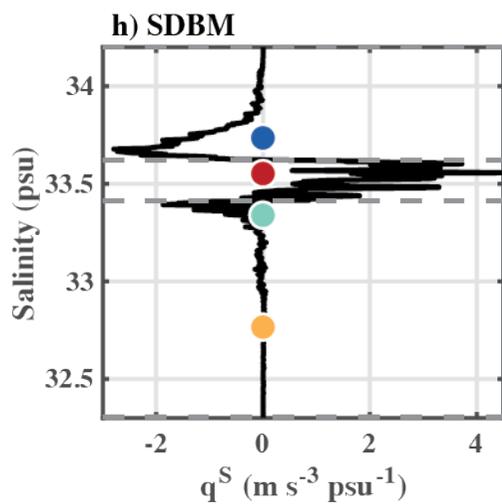
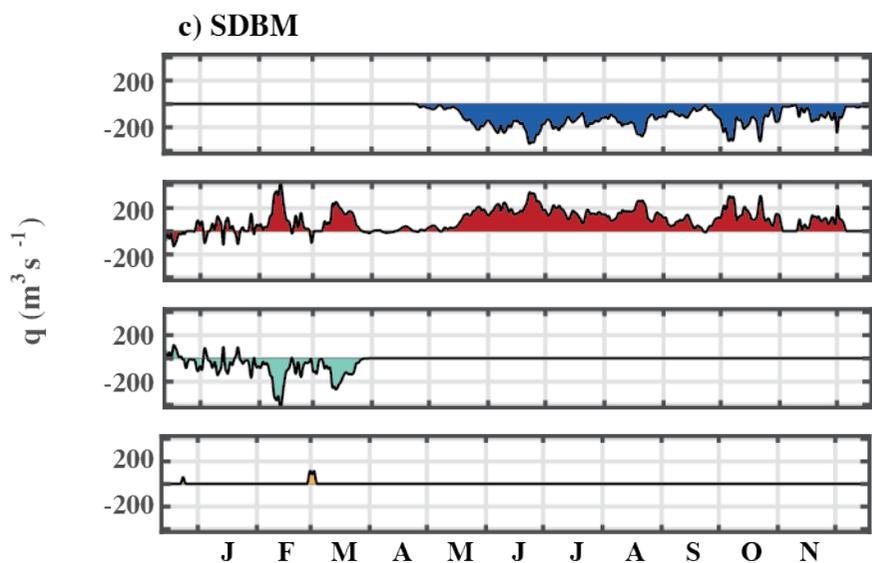
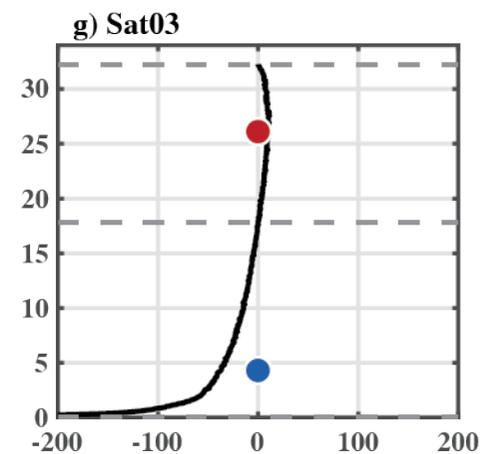
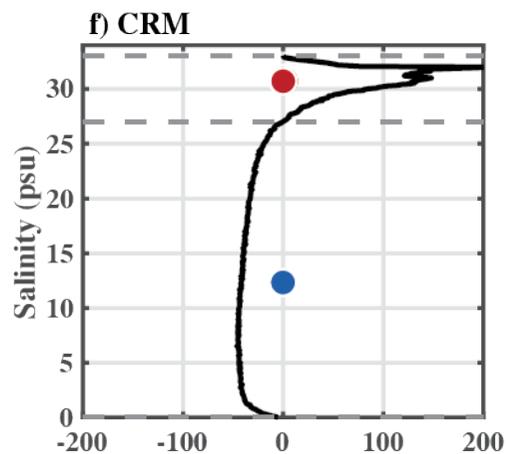
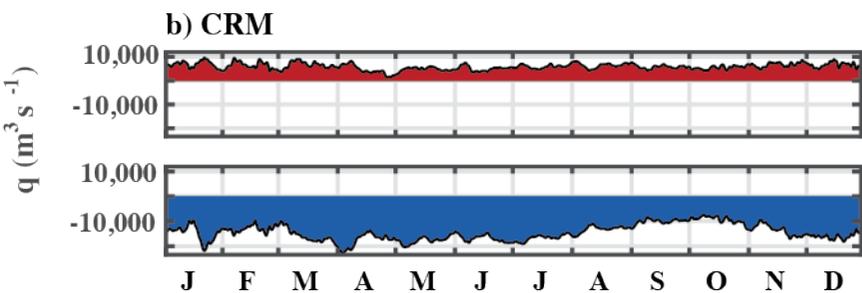
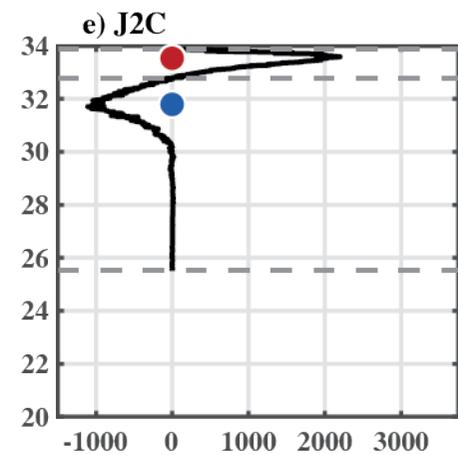
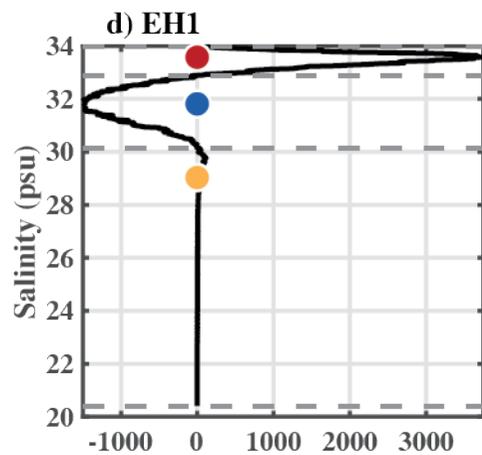
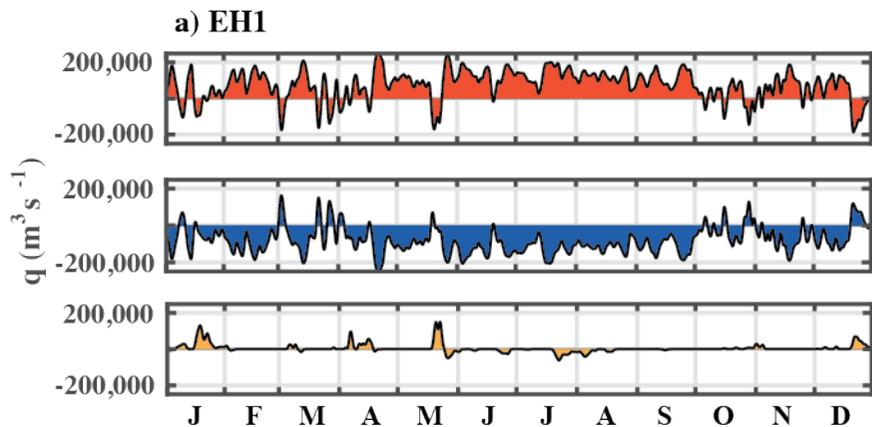
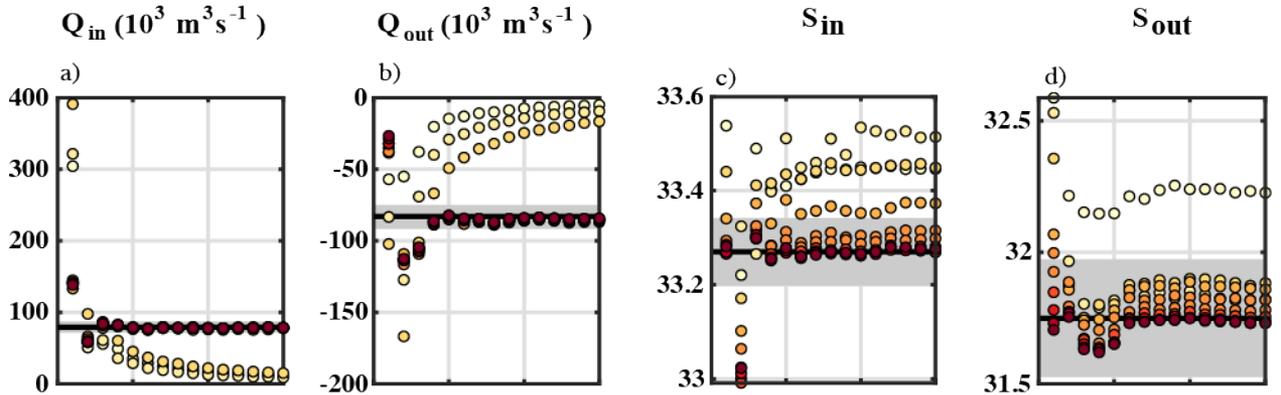
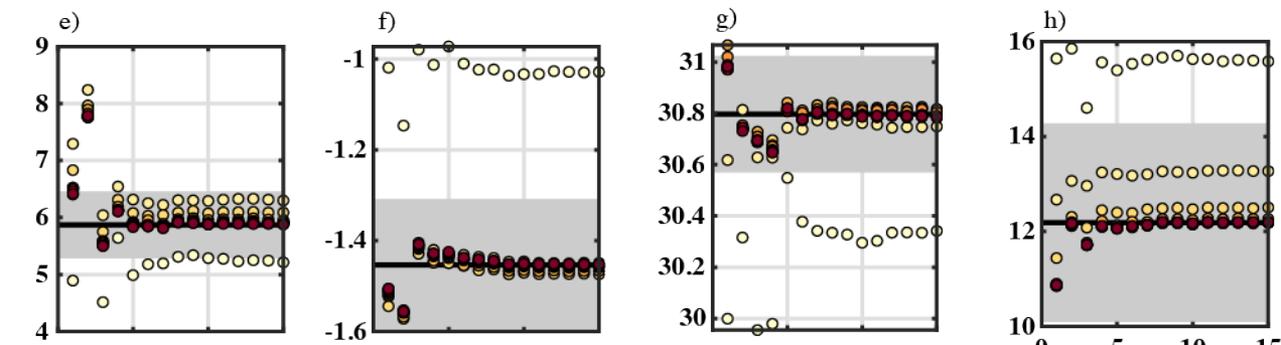


Figure 6.

EH1



CRM



SDBM

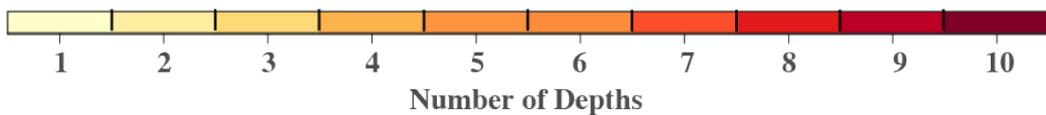
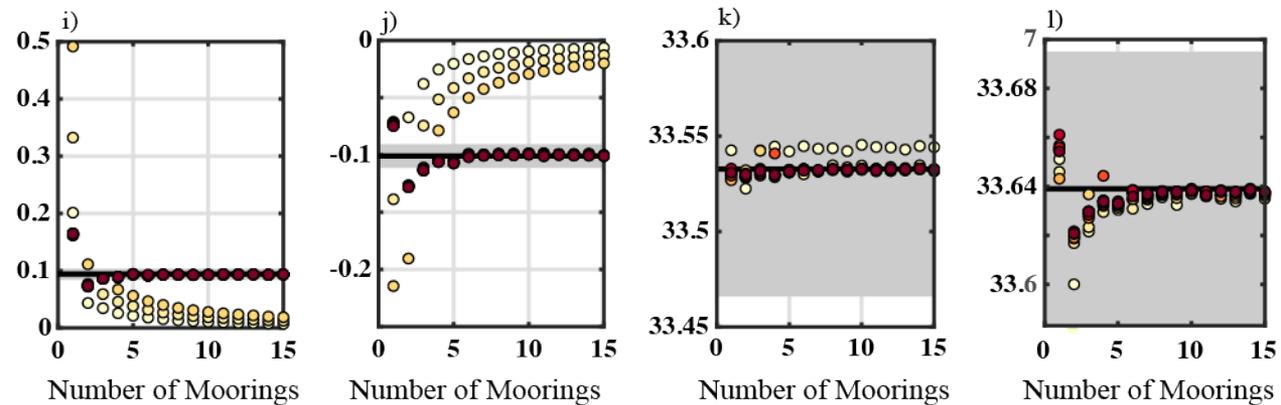
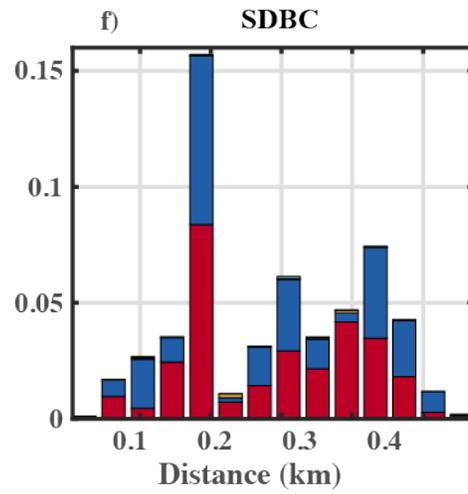
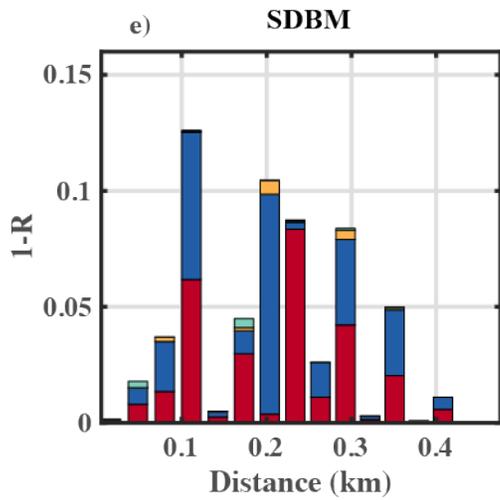
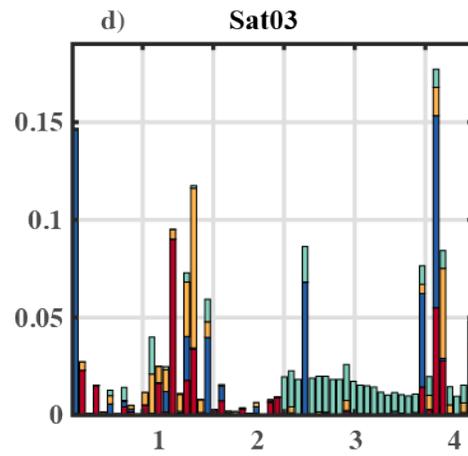
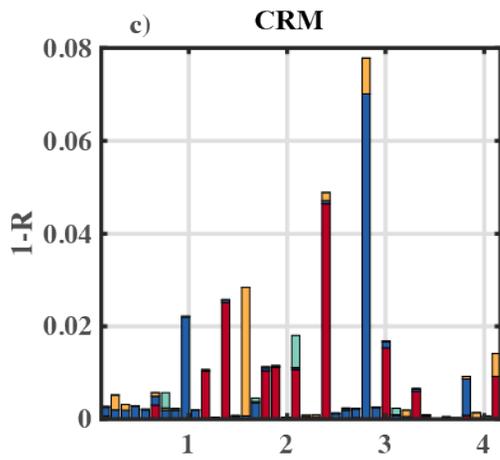
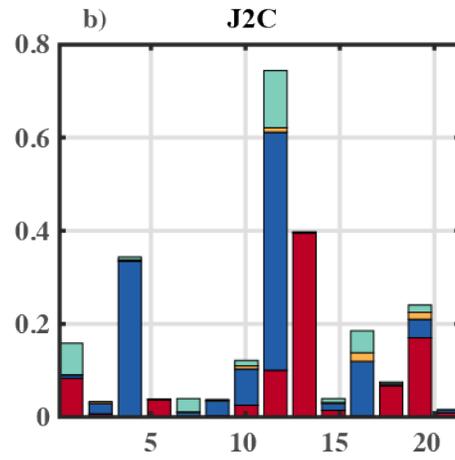
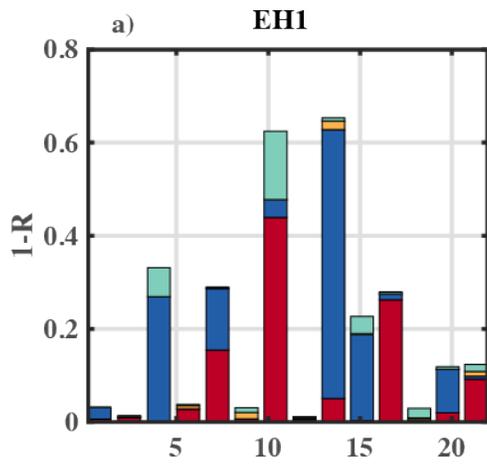


Figure 7.



Figure 8.

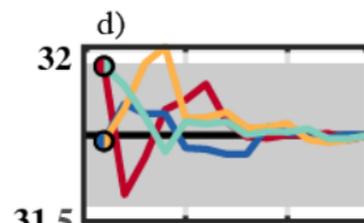
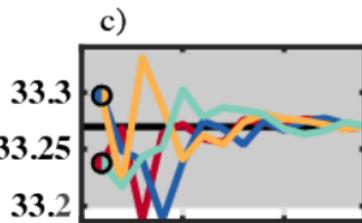
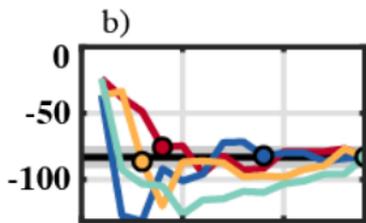
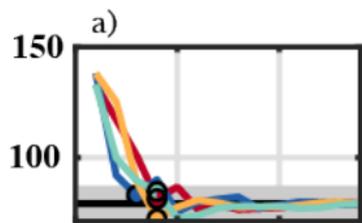


Parameter used to locate moorings

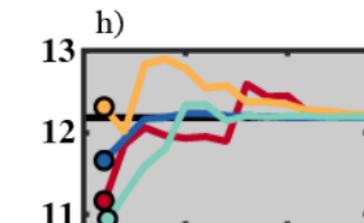
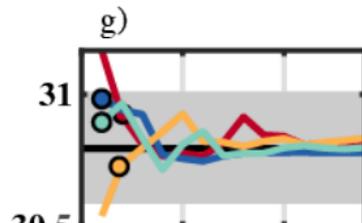
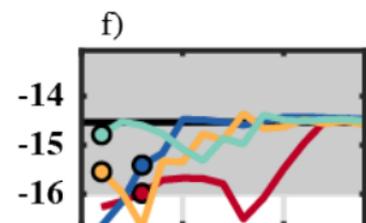
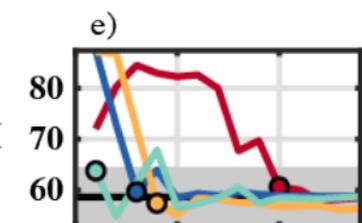
Figure 9.

$Q_{in} (10^3 \text{ m}^3 \text{ s}^{-1})$  $Q_{out} (10^3 \text{ m}^3 \text{ s}^{-1})$  $S_{in}$  $S_{out}$ 

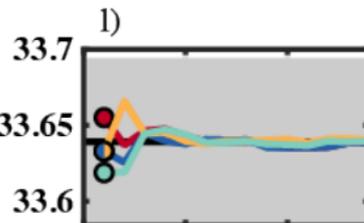
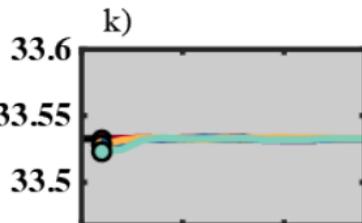
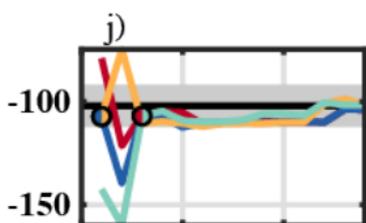
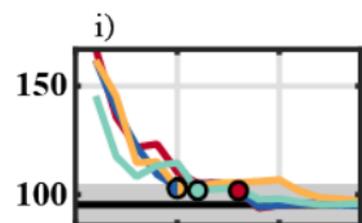
EH1



CBRM



SDBM



Number of Moorings

Number of Moorings

Number of Moorings

Number of Moorings



Parameter used to locate moorings

Figure 10.

