

30 **Abstract**

31 In this paper we develop and test a rigorous modeling framework, based on Duhamel's
32 Theorem, for the unsteady one-dimensional transport and mixing of a solute across a flat
33 sediment-water interface (SWI) and through the benthic biolayer of a turbulent stream.
34 The modeling framework is novel in that it allows for depth-varying diffusivity profiles,
35 accounts for the change in porosity across the SWI and captures the two-way coupling
36 between evolving solute concentrations in both the overlying water column and
37 interstitial fluids of the sediment bed. We apply this new modeling framework to an
38 extensive set of previously published laboratory measurements of turbulent mixing across
39 a flat sediment bed, with the goal of evaluating four diffusivity profiles (constant,
40 exponentially declining, and two hybrid models that account for molecular diffusion and
41 enhanced turbulent mixing in the surficial portion of the bed). The exponentially
42 declining profile is superior (based on RMSE, coefficient of determination, AICc, and
43 model parsimony) and its reference diffusivity scales with a dimensionless measure of
44 stream turbulence and streambed permeability called the Permeability Reynolds Number,
45 Re_K . The diffusivity's dependence on Re_K changes abruptly at $Re_K = 1$, reflecting
46 different modes of mixing below (dispersion) and above (turbulent diffusion) this
47 threshold value. The depth-scale over which the diffusivity exponentially decays is about
48 equal to the thickness of the benthic biolayer (2 to 5 cm), implying that turbulent mixing,
49 and specifically turbulent pumping, may play an outsized role in the biogeochemical
50 processing of nutrients and other contaminants in stream and coastal sediments.

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54 **Plain Language Summary**

55 How far and fast pollutants travel downstream is often conditioned on what happens in a
56 thin veneer of biologically active bottom sediments called the benthic biolayer. However,
57 before a pollutant can be removed in the benthic biolayer it must first be transported
58 across the sediment-water interface and through the interstitial fluids of these surficial
59 sediments. In this paper we investigate, through mathematical modeling and an
60 evaluation of previously published experimental data, the role that water column
61 turbulence plays in transporting solutes into and through the benthic biolayer of streams
62 and coastal sediments.

63

64 **1 Introduction**

65 Many physical and biological processes in aquatic ecosystems depend on, or are strongly
66 affected by, turbulent fluid motions at the sediment-water interface (SWI) (Franca and
67 Brocchini, 2015; Grant and Marusic, 2011). Stream turbulence drives the vertical
68 transport of dissolved constituents through the water column (Tomasek et al., 2018;
69 O'Connor and Hondzo, 2008; Hondzo, 1998) imposing an upper limit on the rate that
70 reactive constituents, nitrate for example, can be assimilated and removed by the
71 streambed (Grant et al., 2018a). Stream turbulence also facilitates the transport and
72 mixing of dissolved and fine particulate materials and energy across the SWI and in the
73 benthic biolayer, the upper 5 cm of the streambed where much of the microbial biomass,
74 as well as nutrient and pollutant processing, is concentrated (Tomasek et al., 2018; Knapp
75 et al., 2017; Caruso et al., 2017; Trauth et al., 2014; Harvey et al., 2013; Zarnetske et al.,
76 2011; Battin et al., 2008; Dahm et al., 2002).

77 At the scale of the benthic biolayer, stream turbulence facilitates mixing in at least
78 two ways (**Figure 1a**): (1) “turbulent pumping” occurs when spatially coherent eddies
79 spawn pressure waves that travel along the SWI and drive temporally oscillating laminar
80 flow across the interface (Kim et al., 2020; Zhong et al., 2016; Boano et al., 2011;
81 Higashino et al., 2009); and (2) “turbulence penetration” occurs when turbulent eddies
82 drive intermittent transport of mass and momentum across the interface (Kim et al., 2020;
83 Roche et al., 2018; Voermans et al., 2017; Reidenbach et al., 2010; Packman et al., 2004).
84 If ripples and dunes are present on the streambed surface, mixing across the benthic
85 biolayer is also facilitated by the advective transport of solutes across the SWI in spatially
86 isolated upwelling and downwelling zones, “bedform pumping” (Azizian et al., 2018;

87 Grant et al., 2014; Fleckenstein et al., 2010; Cardenas et al., 2008; Elliot and Brooks,
 88 1997a,b; Thibodeaux and Boyle, 1987)) and the entrapment and release of interstitial
 89 fluids associated with bedform migration, “bedform turnover” (Wolke et al., 2020; Zheng
 90 et al., 2019; Elliot and Brooks, 1997a,b)). Solute mixing in the streambed is also
 91 controlled by molecular diffusion (which smooths out the steep concentration gradients
 92 generated by the above transport mechanisms, Hester et al., 2017), bio-diffusion (in
 93 which pore fluids and sediment are “pumped” through benthic macrofauna and plants,
 94 Thibodeaux et al., 2011), and the streambed’s permeability and porosity fields (which can
 95 vary temporally and spatially, Laube et al., 2018; Newcomer et al., 2016; Stewardson et
 96 al., 2016; Salehin et al., 2004; Herzog et al., 2018).

97 At present, only under highly idealized experimental or modeling conditions is it
 98 possible to resolve the spatially and temporally complex eddy advection pathways
 99 generated by the transport mechanisms described above. Due to its simplicity,
 100 mathematical tractability and prior agreement with scaling studies, a common alternative
 101 is to approximate mass transport across the SWI, and through the benthic biolayer, as a
 102 flux-gradient diffusive process (Voermans et al., 2018):

$$103 \quad J(y,t) = -D_{\text{eff}}(y) \frac{\partial(\theta C_s)}{\partial y} \quad (1a)$$

104 Here, $J(y,t)$ [$\text{M L}^{-2} \text{T}^{-1}$] and $C_s(y,t)$ [M L^{-3}] are the horizontally and temporally averaged
 105 vertical flux and interstitial solute concentration at depth y [L] and time t [T] in the
 106 sediment. The effective diffusivity, D_{eff} [$\text{L}^2 \text{T}^{-1}$], for solute mixing in the surficial
 107 sediments of a stream or coastal sediment sums over contributions from the tortuosity-
 108 modified molecular diffusion, D'_m [$\text{L}^2 \text{T}^{-1}$], dispersion, D_d [$\text{L}^2 \text{T}^{-1}$], and turbulent

109 diffusion, D_t [$L^2 T^{-1}$] (Roch et al., 2019; Roche et al., 2018; Grant et al., 2018b; Voermans
110 et al., 2018; Voermans et al., 2017; Chandler et al., 2016; Zhong et al., 2016; Boano et al.,
111 2014; Grant et al., 2012; Boano et al., 2011; Grant and Marusic, 2011; Reidenbach et al.,
112 2010; O'Connor and Harvey, 2008; O'Connor and Hondzo, 2008; Packman et al., 2004;
113 Nagaoka and Ohgaki, 1990; Richardson and Parr, 1988):

$$114 \quad D_{\text{eff}} = D'_m + D_d + D_t \quad (1b)$$

115 Dispersion arises from spatial correlations between the time-averaged vertical velocity
116 component and the local mean solute concentration, while turbulent diffusion arises from
117 temporal correlations between the turbulent vertical velocity component and the
118 instantaneous turbulent concentration field (Voermans et al., 2018).

119 The use of equations (1a) and (1b) to describe mixing in the benthic biolayer
120 raises three questions. First, given that the mean flow and shear (turbulence) fields
121 responsible for mixing across the benthic biolayer are damped out with depth by viscous
122 dissipation (He et al., 2019; Roche et al., 2018; Voermans et al., 2017; Pokrajac and
123 Manes, 2009; Bruegem et al., 2006) how should the effective diffusivity be structured
124 vertically, and does the answer depend on the nature of the transport mechanism under
125 consideration (e.g., dispersion or turbulent diffusion)? Second, once an appropriate
126 vertical structure is selected for the effective diffusivity, how well does the flux-gradient
127 diffusive model (equation (1a)) represent solute transport through the streambed? Third,
128 how do we extrapolate effective diffusivities measured in the laboratory to streams and
129 coastal sediments? In this paper we focus on addressing these three questions in the
130 context of turbulent mixing across a flat sediment bed. Complementary efforts are

131 underway to address mixing in the benthic biolayer by bedform pumping and bedform
132 turnover.

133 The paper is organized as follows. In **Section 2** we demonstrate, through an
134 application of Duhamel’s Theorem (Perez et al., 2013), that solute concentration in the
135 interstitial fluids of the sediment bed can be represented as the convolution of solute
136 concentration in the water column and a Green’s function for mass transport in the
137 interstitial fluids of the streambed (Leij et al., 2000). This starting point leads to a set of
138 explicit solutions, valid for any Green’s function, for the spatiotemporal evolution of
139 solute concentration in the water and sediment columns of a closed system. Notably,
140 these solutions capture the two-way coupling of evolving solute concentrations above and
141 below the SWI, whereby mass transfer out of the streambed alters mass concentration in
142 the overlying water column which, in turn, alters mass transfer into the streambed, and so
143 on (**Figure 1b**). We then derive six Green’s functions for two choices of the lower
144 boundary condition (finite or semi-infinite sediment domain) and four functional forms of
145 the diffusivity depth profile. In **Section 3** we demonstrate how this theory can be used to
146 simulate unsteady mass transfer in a stirred tank experiment, and in **Section 4** apply it to
147 previously published measurements of turbulent mass transfer across a flat
148 unconsolidated sediment bed in a well-stirred tank (Chandler et al., 2016; Chandler,
149 2012). We address the three questions raised above in **Section 5** and present our
150 conclusions in **Section 6**.

151 [Figure 1 goes about here]

152 **2 Analytical Modeling Framework**

153 **2.1 Governing Equations for Turbulent Mixing in the Benthic Biolayer**

154 Double averaging over the turbulent timescale (e.g., estimated from the ratio of water
 155 depth and stream velocity (Pope, 2012)) and the horizontal plane of the SWI, and
 156 assuming that streambed porosity does not change appreciably through the benthic
 157 biolayer (i.e., the variable θ appearing in equation (1a) is a fixed constant in the upper 5
 158 cm or so of the streambed (Knapp et al., 2017)), we obtain the following mass
 159 conservation equation for one-dimensional diffusion of a solute through the interstitial
 160 fluids of the sediment bed:

$$161 \quad \frac{\partial C_s}{\partial t} = \frac{\partial}{\partial y} \left(D_{\text{eff}}(y) \frac{\partial C_s}{\partial y} \right) \quad (2a)$$

162 Equation (2a) equates the accumulation of mass in a differential horizontal slice of the
 163 sediment beneath a turbulent stream (left hand side) to the vertical diffusive transport
 164 (right hand side) of a conservative (non-reactive and non-adsorbing) solute (Incropera et
 165 al., 2007). The coordinate y increases with depth into the streambed and its origin (at
 166 $y=0$) is positioned at the horizontal plane of the SWI (**Figure 1a**).

167 In this study we explore four functional forms of the effective diffusivity depth
 168 profile, $D_{\text{eff}}(y) = D_{\text{eff},0} f(y)$, where the variable $D_{\text{eff},0}$ is the surficial effective diffusivity (at
 169 the SWI, $y=0$) and $f(y)$ (unitless) is a piecewise continuous function that equals unity at
 170 the SWI (i.e., $f(0)=1$). After substituting this functional form for the effective diffusivity,
 171 equation (2a) can be rewritten in dimensionless form where the new dependent variable,
 172 c_s [-], incorporates the solute's initial concentration in the interstitial pore fluids of the
 173 sediment bed and the overlying water column (C_{s0} and C_{w0} [M L⁻³], respectively):

174
$$\frac{\partial c_s}{\partial \bar{t}} = \frac{\partial}{\partial \bar{y}} \left(f(\bar{y}) \frac{\partial c_s}{\partial \bar{y}} \right) \quad (2b)$$

175
$$c_s(\bar{y}, \bar{t}) = \frac{C_s(\bar{y}, \bar{t}) - C_{s0}}{C_{w0} - C_{s0}}, \quad c_s(\bar{y}, \bar{t}) \in [0, 1] \quad (2c)$$

176
$$\bar{t} = t/t_T \geq 0, \quad t_T = 1/(D_{\text{eff},0} a^2), \quad \bar{y} = ay \geq 0 \quad (2d)$$

177 The constant a [L^{-1}] is an inverse depth-scale (the definition of which depends on the
 178 choice of diffusivity profile, see later) and t_T is a time constant for solute mixing in the
 179 benthic biolayer. Given the definition of the dimensionless concentration group (equation
 180 (2c)), the initial condition for equation (2b) becomes:

181
$$c_s(\bar{y}, \bar{t} = 0) = 0 \quad (3a)$$

182 At the upper boundary (at the SWI, $y=0$) we require that the interstitial tracer
 183 concentration equals the tracer concentration in the overlying water column,
 184 $c_s(y=0, t) = C_w(t)$ [$M L^{-3}$], which depends only on time; i.e., solute concentration in the
 185 water column is well-mixed. Expressed in dimensionless form the upper boundary
 186 condition becomes:

187
$$c_s(\bar{y}=0, \bar{t}) = \frac{C_w(\bar{t}) - C_{s0}}{C_{w0} - C_{s0}} H(\bar{t}) = c_w(\bar{t}) H(\bar{t}) \quad (3b)$$

188
$$c_w(\bar{t}) = \frac{C_w(\bar{t}) - C_{s0}}{C_{w0} - C_{s0}}, \quad c_w(\bar{t}) \in [0, 1], \quad H(\bar{t}) = \begin{cases} 0, & \bar{t} < 0 \\ 1, & \bar{t} > 0 \end{cases} \quad (3c)$$

189 The Heaviside function $H(\bar{t})$ [-] appearing on the righthand side of equation (3b) ensures
 190 that the upper boundary condition is zero for $\bar{t} < 0$ (this detail becomes important for the
 191 application of Duhamel's Theorem below). By expressing the upper boundary condition
 192 in this way, we are assuming that mass transfer across the SWI is rate-limited by the

193 mixing of solute within the streambed and not by convective mixing across the
194 concentration boundary layer above the streambed (Grant et al., 2018). Put another way,
195 we are assuming that the Biot Number—which expresses the ratio of timescales for
196 diffusive mixing in the streambed and convective mass transfer across the turbulent
197 boundary layer above the streambed—is much greater than unity (Incropera et al., 2007).

198 One of two lower boundary conditions can be selected, depending on whether the
199 sediment bed is assumed to be finite (equation (3d)) or semi-infinite (equation (3e)) in
200 extent.

$$201 \left. \frac{\partial c_s}{\partial \bar{y}} \right|_{\bar{y}=\bar{d}_b} = 0 \quad (3d)$$

$$202 c_s(\bar{y} \rightarrow \infty, \bar{t}) = 0 \quad (3e)$$

203 Equation (3d) enforces a no-flux boundary condition at scaled depth $\bar{d}_b = a d_b$ [-] where d_b
204 [L] is the depth of the sediment bed (**Figure 1**). Equation (3e) implies that, very deep into
205 the bed ($\bar{y} \rightarrow \infty$), the interstitial concentration is maintained at its initial state.

206 **2.2 Duhamel's Theorem and Green's Functions**

207 We seek an explicit solution to the above system of equations that is valid for any time-
208 varying solute concentration in the overlying water column, any piecewise continuous
209 diffusivity profile, and either a finite or semi-infinite streambed. To this end we invoke
210 Duhamel's Theorem, an analytical approach for solving the diffusion equation in cases
211 where one boundary is a piece-wise continuous function of time (Perez Guerrero et al.,
212 2013). Proofs of this theorem typically assume that the diffusion coefficient is constant.
213 However, as demonstrated in the Supplemental Information (**Text S1**), the theorem also
214 applies in cases where the diffusion coefficient varies solely as a function of depth.

215 For the version of Duhamel's Theorem adopted here, three conditions must be
 216 met (Myers, 1971): (1) the system must have a zero initial state; (2) the differential
 217 equation and boundary conditions must be homogeneous with the exception of a single
 218 time-dependent boundary condition or source/sink term in the differential equation; and
 219 (3) the single nonhomogeneous term should be initially equal to zero. By design, our
 220 system meets all three requirements. Accordingly, Duhamel's Theorem allows us to
 221 express the interstitial solute concentration in the sediment bed as a convolution of the
 222 time-derivative of the water column concentration $c_w(\bar{t})$ and a so-called *auxiliary*
 223 *function* $c_s^A(\bar{y}, \bar{t})$ where v is a dummy integration variable (Meyers, 1971):

$$224 \quad c_s(\bar{y}, \bar{t}) = \int_0^{\bar{t}} c_s^A(\bar{y}, \bar{t} - v) \frac{d}{dv} [c_w(v)H(v)] dv \quad (4a)$$

225 The auxiliary function is, by definition, a solution to the same system of equations
 226 described above for the interstitial solute concentration (equations (2b) - (3e)) but with
 227 the nonhomogeneous term replaced by unity (compare equations (3b) and (4d)):

$$228 \quad \frac{\partial c_s^A}{\partial \bar{t}} = \frac{\partial}{\partial \bar{y}} \left(f(\bar{y}) \frac{\partial c_s^A}{\partial \bar{y}} \right) \quad (4b)$$

$$229 \quad c_s^A(\bar{y}, \bar{t} = 0) = 0, \bar{y} \geq 0 \quad (4c)$$

$$230 \quad c_s^A(\bar{y} = 0, \bar{t}) = H(\bar{t}) \quad (4d)$$

231 The auxiliary function's lower boundary condition depends on whether the sediment bed
 232 is finite or semi-infinite in extent:

$$233 \quad \left. \frac{\partial c_s^A}{\partial \bar{y}} \right|_{\bar{y}=\bar{a}_b, \bar{t}} = 0 \quad (4e)$$

$$234 \quad c_s^A(\bar{y} \rightarrow \infty, \bar{t}) = 0 \quad (4f)$$

235 Duhamel's Theorem (equation (4a)) can also be expressed as a convolution of the
 236 dimensionless water column concentration, $c_w(\bar{t})$, and a so-called Green's function,
 237 $G(\bar{y}, \bar{t})$ [T^{-1}], scaled here by the mixing timescale introduced earlier, $\bar{G}(\bar{y}, \bar{t}) = t_r G(\bar{y}, \bar{t})$ (see
 238 **Text S2** for details):

$$239 \quad c_s(\bar{y}, \bar{t}) = \int_0^{\bar{t}} \bar{G}(\bar{y}, \nu) c_w(\bar{t} - \nu) d\nu \quad (5a)$$

$$240 \quad \bar{G}(\bar{y}, \bar{t}) = \frac{\partial c_s^A}{\partial \bar{t}} \quad (5b)$$

241 According to equation (5a), solute concentration in the interstitial fluid of the sediment
 242 bed at any depth and time, $c_s(\bar{y}, \bar{t})$, depends on the entire prior history of solute
 243 concentration in the water column, $c_w(\bar{t})$, filtered through the Green's function, $\bar{G}(\bar{y}, \bar{t})$.
 244 The Green's function, in turn, is a fundamental solution to the diffusion equation
 245 (equation (2b)) that characterizes the response of solute concentration in the interstitial
 246 fluids of the streambed to an impulsive injection of mass at the SWI at $\bar{t} = 0$, $c_w(\bar{t}) = \delta(\bar{t})$,
 247 where $\delta(\bar{t})$ [-] is the Dirac Delta function. Because the Green's function is calculated
 248 from the auxiliary function (equation (5b)) its functional form will depend on the vertical
 249 structure of the diffusivity profile, $f(\bar{y})$, and the lower boundary condition (either
 250 equation (4e) or (4f)). Five Green's functions, corresponding to different combinations of
 251 the diffusivity profile and lower-boundary condition, are derived in **Section 2.4**.

252 **2.3 Two-Way Coupling Across the SWI in a Closed System**

253 In a typical application of Duhamel's Theorem, the time-dependence of the
 254 nonhomogeneous boundary condition is stipulated in advance. In our case, however, the

255 nonhomogeneous boundary condition (i.e., the water column concentration, equation
 256 (3b)) depends on mass flux across the SWI through a mass balance over the water
 257 column. For a closed system with a well-mixed water column, such as the stirred tank in
 258 **Figure 1b**, the water column mass balance takes on the following form:

$$259 \quad A_b h_w \frac{dC_w}{dt} = A_b \theta D_{\text{eff},0} \left. \frac{\partial C_s}{\partial y} \right|_{y=0,t} \quad (6a)$$

260 In this equation, the change of solute mass in the water column (left hand side) equals the
 261 rate of mass transfer across the SWI by turbulent mixing (right hand side). New variables
 262 appearing here include the interfacial area, A_b [L^2], of the streambed and the height of the
 263 water column, h_w [L]. Streambed porosity, θ , is included on the right-hand side of the
 264 equation to account for the streambed's porosity; specifically, the abrupt change in area
 265 over which solute mass transport occurs above and below the SWI (Grant et al., 2012).
 266 Expressing equation (6a) using the dimensionless variables introduced earlier, we obtain
 267 equation (6b) where the dimensionless variable, \bar{h}_w (equation (6c)), is a scaled form of
 268 the water column depth.

$$269 \quad \frac{dc_w}{d\bar{t}} = \frac{1}{\bar{h}_w} \left. \frac{\partial c_s}{\partial \bar{y}} \right|_{\bar{y}=0,\bar{t}} \quad (6b)$$

$$270 \quad \bar{h}_w = \frac{a h_w}{\theta} \quad (6c)$$

271 Two-way coupling across the SWI manifests mathematically as a dependence of
 272 the water column concentration (left hand side, equation (6b)) on the streambed's
 273 interstitial solute concentration (right hand side, equation (6b)) and, simultaneously, the
 274 dependence of the streambed's interstitial solute concentration (left hand side, equation
 275 (5a)) on the solute concentration in the water column (through its convolution with the

276 Green's function, right hand side equation (5a)). This two-way coupling can be solved
 277 exactly by manipulating the water and sediment mass balance equations in the Laplace
 278 Domain. As demonstrated in Supplemental Information (Text S3), the result is a set of
 279 fully coupled solutions for solute concentration in the water and sediment columns of a
 280 closed system with two-way coupling across the SWI "turned on":

$$281 \quad c_w(\bar{t}) = (C_{w0} - C_{s0}) \mathcal{L}^{-1} \left[\frac{1/\bar{s}}{1 - \frac{1}{\bar{s}h_w} \left(\frac{\partial \tilde{G}}{\partial \bar{y}} \right)_{\bar{y}=0, \bar{s}}} \right] + C_{s0} \quad (7a)$$

$$282 \quad c_s(\bar{y}, \bar{t}) = (C_{w0} - C_{s0}) \mathcal{L}^{-1} \left[\frac{\tilde{G}(\bar{y}, \bar{s})/\bar{s}}{1 - \frac{1}{\bar{s}h_w} \left(\frac{\partial \tilde{G}}{\partial \bar{y}} \right)_{\bar{y}=0, \bar{s}}} \right] + C_{s0} \quad (7b)$$

283 In these solutions, the symbol $\mathcal{L}^{-1}[\cdot]$ represents the inverse Laplace Transform, $\bar{s} = st_T$ [-]
 284 is a dimensionless form of the Laplace Transform variable s [T⁻¹], and \tilde{G} is the Laplace
 285 transform of the dimensionless Green's function which, in turn, depends on the
 286 diffusivity depth profile $f(\bar{y})$ and bottom boundary condition (through the solution to the
 287 auxiliary function (equation (5b)). A corresponding set of solutions can also be derived
 288 for when two-way coupling across the SWI is "turned off"; i.e., the diffusion equation's
 289 upper boundary condition is held at its initial state, $c_s(\bar{y}=0, \bar{t}) = C_{w0}$:

$$290 \quad c_w(\bar{t}) = (C_{w0} - C_{s0}) \mathcal{L}^{-1} \left[\frac{1}{\bar{s}} \left(\frac{1}{\bar{s}h_w} \frac{\partial \tilde{G}}{\partial \bar{y}} \Big|_{\bar{y}=0, \bar{s}} + 1 \right) \right] + C_{s0} \quad (7c)$$

$$291 \quad c_s(\bar{y}, \bar{t}) = (C_{w0} - C_{s0}) \mathcal{L}^{-1} \left[\frac{\tilde{G}(\bar{y}, \bar{s})}{\bar{s}} \right] + C_{s0} \quad (7d)$$

292 [Figure 2 goes about here]

293 Depending on the choice of Green's function, the inverse Laplace transforms appearing
 294 in equations (7a) - (7d) were either determined analytically, or numerically evaluated
 295 using Gaussian Quadrature implemented in the Mathematica package authored by U.
 296 Graf (Graf, 2004).

297 **2.4 Laplace Domain Solutions for the Green's Function**

298 In **Table 1** we present five Laplace domain solutions for the Green's function given four
 299 different choices of the diffusivity depth profile and two different choices of the bottom
 300 boundary condition (finite or semi-infinite sediment bed) (derivations in Supplemental
 301 Information, **Text S4**). Together with equations (7a) through (7d), these five Green's
 302 functions provide a total of 20 different solution combinations for solute concentration in
 303 the water and sediment column of a closed system with two-way coupling across the SWI
 304 turned on (equations (7a) and (7b)) or off (equations (7c) and (7d)). The four diffusivity
 305 depth profiles we evaluated include (**Figure 2**): (1) constant (*C Profile*, equation (8a));
 306 (2) exponentially declining (*E Profile*, equation (8b)); (3) exponentially declining to a
 307 tortuosity-modified molecular diffusion coefficient (*E2M Profile*, equation (8c)); and (4)
 308 constant to exponentially declining diffusivity (*C2E Profile*, equation (8d)).

$$309 \quad f_c(\bar{y}) = 1, \quad \bar{y} = a_c y \quad (8a)$$

$$310 \quad f_E(\bar{y}) = e^{-\bar{y}}, \quad \bar{y} = a_E y \quad (8b)$$

$$311 \quad f_{E2M}(\bar{y}) = \begin{cases} e^{-\bar{y}}, & 0 \leq \bar{y} \leq \bar{\ell}_m \\ \bar{D} = D'_m / D_{\text{eff},0}^{E2M}, & \bar{y} > \bar{\ell}_m \end{cases}, \quad \bar{\ell}_m = -\ln \bar{D}, \quad 0 < \bar{D} < 1, \quad \bar{y} = a_{E2M} y \quad (8c)$$

$$312 \quad f_{C2E}(\bar{y}) = \begin{cases} 1, & 0 \leq \bar{y} \leq \bar{\ell}_t \\ e^{-(\bar{y}-\bar{\ell}_t)}, & \bar{y} > \bar{\ell}_t \end{cases}, \quad \bar{\ell}_t = a_{C2E} \ell_t, \quad \bar{y} = a_{C2E} y \quad (8d)$$

313 Most laboratory (Marion and Zaramella, 2015; O'Connor and Harvey, 2008; Grant et al.,
 314 2012) and field (Wörman, 2000) studies of diffusive mixing across the SWI adopt the C
 315 Profile. However, several studies (He et al., 2019; Roche et. al, 2019; Chandler et al.,
 316 2016; Nagaoka and Ohgaki, 1990) have shown that turbulent mixing in the sediment bed
 317 declines exponentially with depth, and a recent numerical modeling study concluded that
 318 the E Profile is consistent with experimental breakthrough curves measured in the
 319 laboratory and field (Bottacin-Busolin, 2019). The E2M profile is a natural extension of
 320 the E Profile that accounts for the fact that the tortuosity-modified molecular diffusion
 321 coefficient D'_m [$L^2 T^{-1}$] imposes a lower-bound on the effective diffusivity. E2M's
 322 mathematical representation (equation (8c)) includes two new variables: the
 323 dimensionless depth at which the diffusivity profile transitions from exponentially
 324 declining to the tortuosity-modified molecular diffusion coefficient, $\bar{\ell}_m = a_{E2M} \ell_m$ [-] where
 325 ℓ_m [L] is the transition depth and $\bar{D} = D'_m / D_{eff,0}^{E2M}$ [-] is a dimensionless form of the
 326 tortuosity-modified molecular diffusion coefficient. These two parameters are
 327 mathematically related as follows: $\bar{\ell}_m = -\ln \bar{D}$. Finally, the C2E Profile captures enhanced
 328 mixing at the top of the streambed by extending the surficial effective diffusivity $D_{eff,0}^{C2E}$ to
 329 a depth $y = \ell_t$ [L] below the SWI. For depths greater than the constant mixing thickness,
 330 $y > \ell_t$, the diffusivity profile declines exponentially. The dimensionless form of the
 331 enhanced mixing thickness ℓ_t is defined in the usual way, $\bar{\ell}_t = a_{C2E} \ell_t$.

332 [Table 1 goes about here]

333 **3 Example of the Theory's Application to Mixing Across the SWI in a Stirred Tank**

334 As a tangible example of the theory presented in **Section 2**, consider an experiment
 335 (along the lines examined in detail later, see **Section 4**) in which a mass M of a
 336 conservative solute is added to the interstitial fluids of a sediment bed in an otherwise
 337 solute-free stirred tank. Adopting the notation indicated in **Figure 1b**, the initial
 338 interstitial solute concentration in the sediment bed will be: $C_{s0} = M / (d_b A_b \theta)$. At time $t = 0$
 339 the impeller is turned on causing solute concentration in the overlying water column to
 340 increase as solute turbulently mixes out of the bed. If the experiment runs long enough,
 341 the solute concentration in the water column and interstitial fluids of the sediment bed
 342 will approach a final (well-mixed) equilibrium concentration, C_{eq} [$M L^{-3}$]:

$$343 \quad \frac{C_{eq}}{C_{s0}} = \frac{\bar{d}_b}{1 + \bar{d}_b}, \quad \bar{d}_b = d_b \theta / h_w \quad (9a)$$

344 How does solute concentration in the overlying water column and interstitial fluids of the
 345 sediment bed evolve from their initial state (where all solute mass is located in the
 346 sediment bed) to the final (well-mixed) state given by equation (9a)? Within the context
 347 of our modeling framework, the answer depends on the depth-dependence of the
 348 diffusivity profile, whether two-way coupling across the SWI is turned “on” or “off”, and
 349 whether the sediment bed is assumed to be semi-infinite or finite in extent.

350 As an illustration we adopt the C Profile and derive from the theoretical
 351 framework in **Section 2** the following three explicit solutions for the evolution of solute
 352 concentration in the water column (see **Text S5** for details): (1) a “null model” which
 353 assumes two-way coupling across the SWI is turned off and the sediment bed is infinitely
 354 deep (equation (9b)); (2) an “infinite bed” model that assumes two-way coupling is
 355 turned on and the sediment bed is infinitely deep (equation (9c)); and (3) a “finite bed”

356 model that accounts for both two-way coupling across the SWI and the finite depth of the
 357 sediment bed (equation (9d)).

$$358 \quad \frac{C_{w,null}^c(\bar{t}_c)}{C_{s0}^c} = 2\sqrt{\frac{\bar{t}_c}{\pi}}, \quad \bar{t}_c = \theta^2 D_{eff,0}^c t / h_w^2 \quad (9b)$$

$$359 \quad \frac{C_{w,\infty}^c(\bar{t}_c)}{C_{s0}^c} = 1 - e^{-\bar{t}_c} \operatorname{erfc}\left(\sqrt{\bar{t}_c}\right) \quad (9c)$$

$$360 \quad \frac{C_{w,finite}^c(\bar{t}_c)}{C_{s0}^c} = 1 - \mathcal{L}^{-1} \left[\frac{1/\bar{s}}{1 + \tanh(\bar{d}_b \sqrt{\bar{s}}) / \sqrt{\bar{s}}} \right] \quad (9d)$$

361 The corresponding set of solutions for solute concentration in the interstitial fluids of the
 362 sediment bed are as follows:

$$363 \quad \frac{C_{s,null}^c(\bar{y}_c, \bar{t}_c)}{C_{s0}^c} = \operatorname{erfc}\left(\frac{\bar{y}_c}{\sqrt{4\bar{t}_c}}\right), \quad \bar{y}_c = \theta y / h_w \quad (10a)$$

$$364 \quad \frac{C_{s,\infty}^c(\bar{y}_c, \bar{t}_c)}{C_{s0}^c} = 1 - e^{-\bar{t}_c + \bar{y}_c} \operatorname{erfc}\left(\frac{2\bar{t}_c + \bar{y}_c}{\sqrt{4\bar{t}_c}}\right) \quad (10b)$$

$$365 \quad \frac{C_{s,finite}^c(\bar{y}_c, \bar{t}_c)}{C_{s0}^c} = 1 - \mathcal{L}^{-1} \left[\frac{\cosh((\bar{d}_b - \bar{y}_c)\sqrt{\bar{s}})}{\bar{s} \cosh(\bar{d}_b \sqrt{\bar{s}}) + \sqrt{\bar{s}} \sinh(\bar{d}_b \sqrt{\bar{s}})} \right] \quad (10c)$$

366 The superscript ‘‘C’’ indicates that these solutions are specific to the C Profile.

367 The null model predicts that solute concentration in the water column increases
 368 without bound and in proportion to the square root of time (i.e., solute concentration in
 369 the water column increases linearly when plotted against $\sqrt{\bar{t}_c}$, thick black dashed line,
 370 **Figure 3a**). The infinite bed model rises with the null model initially (i.e., until around
 371 $\sqrt{\bar{t}_c} \approx 0.25$) but then slows as two-way coupling reduces the rate of mass transfer across
 372 the SWI (thin black solid curve in **Figure 3a**). The finite bed model exhibits three phases

373 (colored dashed lines in the figure): tracking the null solution early on, transitioning to
374 the infinite bed solution at intermediate times, and stabilizing at a final equilibrium
375 concentration at long times (thin horizontal colored lines).

376 [Figure 3 goes about here]

377 Similar patterns are evident for model-predicted solute concentration in the
378 interstitial fluids of the streambed (**Figure 3b**). For these simulations we focused on the
379 temporal evolution of interstitial solute concentration in the shallow portion of the bed,
380 $\bar{y}_c = 0.05$; this particular depth was chosen so that we would not violate the inequality
381 requirement, $\bar{y}_c \leq \bar{d}_b$, for the smallest dimensionless bed depth of $\bar{d}_b = 0.1$ (light blue
382 horizontal lines in **Figure 3**). The null model (thick black dashed curve) predicts a steep
383 drop in solute concentration initially (as solute in the upper portion of the streambed
384 mixes into a solute-free water column) followed by a gradual decline with time (as solute
385 from deeper in the bed mixes upward). The infinite bed solution (thin black curve)
386 declines with the null model initially, but then rebounds as two-way coupling slows mass
387 transfer across the SWI. Indeed, this rebound closely approximates the rise in water
388 column concentration predicted by the infinite bed model (compare thin black curves in
389 **Figures 3a** and **3b**) implying that, when two-way feedback is turned on, solute
390 concentrations in the interstitial fluids near the top of the sediment bed are similar to the
391 solute concentration in the overlying water column. Interstitial solute concentrations
392 predicted by the finite bed model (colored dashed curves in **Figure 3b**) exhibit the same
393 three phases noted above for the water column; namely, they track the null model at early
394 times, follow the infinite bed model at intermediate times, and stabilize at an equilibrium
395 concentration at long times. For the model-data comparisons described in the next section

396 we employ a set of infinite bed models derived for each of the four diffusivity profiles (C,
397 E, E2M, and C2E) and restrict the experimental time window to well before the onset of
398 equilibrium conditions.

399 **4 Applying the Theory to Previously Published Measurements**

400 **4.1 Chandler et al.'s Experiments**

401 Next we turn to an extensive set of previously published measurements of turbulent
402 mixing of a conservative tracer (Rhodamine WT) across the SWI in a stirred tank with a
403 flat sediment bed, along the lines of the experimental set-up illustrated in **Figure 1b**
404 (Chandler et al., 2016; Chandler, 2012). Chandler et al.'s study is notable for several
405 reasons. First, it is one of the few where tracer concentrations were simultaneously
406 measured in the water and sediment columns allowing us to directly compare mixing
407 parameters estimated from data collected exclusively above or below the SWI. Second,
408 the twenty (out of twenty-six total) Chandler et al. experiments included in this study
409 cover a range of bed shear velocities ($u_* = 0.01$ to 0.04 m s⁻¹), mean grain diameters ($d_g =$
410 0.15 to 5.00 mm), and sediment permeabilities ($K = 0.18$ to 223 m²). Finally, by
411 conceptually dividing the sediment bed into a series of layers and fitting the diffusion
412 equation (with a constant diffusivity profile) to each layer separately, Chandler et al.
413 concluded that the diffusivity declines exponentially with depth. Nagaoka and Ohgaki
414 (1990), who pioneered this approach, reached a similar conclusion. Thus, there is already
415 a strong indication that the diffusivity in their system declines with depth. One caveat is
416 that Chandler et al.'s range of shear velocities is on the low side for some streams; e.g.,
417 the 72 headwater streams included in the U.S. Lotic Intersite Nitrogen eXperiment (LINX
418 II) had shear velocities ranging from 0.02 to 0.48 m s⁻¹ and only seven streams had shear

419 velocities in the range interrogated by Chandler et al. (Hall et al., 2009). However, when
420 the shear velocity and permeability for each of Chandler et al.’s experiments are
421 combined to calculate a Permeability Reynolds Number, this master variable spans both
422 dispersive and turbulent diffusive mixing regimes (see **Section 5.1**).

423 Details of Chandler et al.’s experiments and the approach we used for parameter
424 estimation and model performance evaluation are briefly described here (see **Text S6** for
425 details). In brief, the sediment column, which had a depth of $d_b = 0.2$ m and a porosity of
426 $\theta = 0.38$ to 0.39 , consisted of randomly packed single-sized spherical soda glass spheres.
427 In all experiments, the initial state was a Rhodamine WT saturated sediment bed
428 (concentration of $C_{s0} = 100$ ppb) and a Rhodamine-free water column ($C_{w0} = 0$ ppb)
429 although the actual concentrations varied somewhat by experiment (**Table S1**).

430 Experiments were initiated when the impeller motor was turned on, whereupon tracer
431 concentrations were fluorometrically monitored in the water column and at 5 depths in
432 the sediment column ($x = 0.015, 0.049, 0.083, 0.117, \text{ and } 0.151$ m below the SWI) at a
433 frequency of 0.1 Hz over a period of hours to days. Diffusivity profile parameters were
434 inferred by minimizing the root mean squared error (RMSE) calculated from
435 observations and profile-specific infinite bed model predictions using the non-linear least
436 squares algorithms implemented within the “NonlinearModelFit” command in the
437 Mathematica computing package (v. 11.20, Wolfram Research, Inc.). Several model
438 performance metrics were also generated, including RMSE, coefficient of determination
439 (R^2 value), and the corrected Akaike Information Criterion (AICc). The latter was used to
440 rank the performance of the four diffusivity profile models, accounting for the trade-off
441 between model fit and model complexity. The top-ranked (most parsimonious) model has

442 the smallest AICc value (Aho et al., 2014). Because model predictions were based on
443 “infinite bed models” specific to each profile (see **Section 3**), we restricted the
444 experimental window to periods when the bottom interstitial Rhodamine WT
445 concentration, as measured by the deepest probe, was within 90% of its initial value (see
446 t_{final} values in **Table S1**). By this approach, diffusivity profile parameters were estimated
447 for all four diffusivity profiles (C, E, E2M, and C2E) and for 20 of Chandler et al.’s
448 experiments (inferred parameter values, estimated errors, and model performance metrics
449 are summarized in **Tables S2-S7**); six experiments were excluded due to missing data or
450 other issues.

451 **4.2 Experimental Evaluation of the C Profile**

452 Significant bias is evident when tracer concentrations predicted by the C Profile’s infinite
453 bed model are compared to Chandler et al.’s experimental data (**Figure 4a**). In this figure
454 we have plotted Chandler et al.’s Rhodamine WT measurements against the square root
455 of dimensionless time, $\sqrt{\bar{t}_c} = \sqrt{t/t_c}$, where the time-constant, $t_c = h_w^2 / (\theta^2 D_{\text{eff},0}^c)$, varies from
456 experiment-to-experiment depending on the inferred value of the C Profile’s effective
457 diffusivity (summarized in **Table S2**); the other two parameters, h_w and θ , varied little
458 across Chandler et al.’s twenty experiments. When plotted in this way Chandler et al.’s
459 data can be compared directly to a single model-predicted curve for the time evolution of
460 Rhodamine WT concentration in the water column (solid black curve, equation (9c), top
461 panel in **Figure 4a**) and at two depths (15 or 151 mm below the SWI) in the sediment
462 column (solid black curves, equation (10b), bottom panel in **Figure 4a**). For clarity,
463 Rhodamine WT measurements at the three intermediate depths (4.9, 8.3, and 117 cm

464 below the SWI) were not included this figure, although these data were included in the
465 model optimization studies described in **Section 5.1**.

466 [Figure 4 goes about here]

467 By virtue of the way Chandler et al.'s experiments were conducted, the
468 Rhodamine WT concentration in the water column is proportional to the cumulative
469 Rhodamine WT mass transferred from the sediment bed to the water column over time.
470 The results plotted in the top panel of **Figure 4a** therefore imply that the C Profile model
471 under- and over-estimates mass transfer out of the sediment bed at short and long times,
472 respectively. A comparison of model-predicted and measured Rhodamine WT
473 concentrations in the sediment bed reveals the problem (bottom panel, **Figure 4a**): the C
474 Profile model under-estimates mixing in the surficial portion of the bed at early times
475 (i.e., predicted concentrations exceed measured concentrations at 15 mm below the SWI)
476 and over-estimates mixing deeper in the bed at later times (i.e., predicted concentrations
477 are less than measured concentrations at 151 mm below the SWI).

478 **4.3 Experimental Evaluation of the E Profile**

479 The model bias described above is reduced substantially when the effective diffusivity
480 decays exponentially with depth (**Figure 4b**). The E Profile's infinite bed model was
481 constructed by substituting its Green's function (equation (T3)) into the water and
482 sediment mass balance solutions (equations (7a) and (7b)) for a closed system with two-
483 way coupling across the SWI "turned on" (the superscript "E" indicates that these
484 solutions are specific to the E Profile):

$$485 \quad C_{w,\infty}^E(\bar{t}_E) = (C_{w0} - C_{s0}) \mathcal{L}^{-1} \left[\frac{\bar{h}_w K_1(2\sqrt{s})}{\bar{s} \bar{h}_w K_1(2\sqrt{s}) + \sqrt{s} K_0(2\sqrt{s})} \right] + C_{s0} \quad (11a)$$

486
$$C_{s,\infty}^E(\bar{y}, \bar{t}_E) = (C_{w0} - C_{s0}) \mathcal{L}^{-1} \left[\frac{\bar{h}_w \sqrt{e^{\bar{y}}} K_1(2\sqrt{e^{\bar{y}} \bar{s}})}{\bar{s} \bar{h}_w K_1(2\sqrt{\bar{s}}) + \sqrt{\bar{s}} K_0(2\sqrt{\bar{s}})} \right] + C_{s0} \quad (11b)$$

487
$$\bar{t}_E = t/t_E, \quad t_E = 1/(a_E^2 D_{\text{eff},0}^E), \quad \bar{h}_w = a_E h_w / \theta, \quad \bar{y} = a_E y \quad (11c)$$

488 Because the E Profile model has two unknown parameters ($D_{\text{eff},0}^E$ and a_E , as opposed to the
 489 C Profile's one unknown parameter, $D_{\text{eff},0}^C$) there is no longer a single "master" curve
 490 against which all of Chandler et al.'s water column data, for example, can be compared to
 491 (as there was for the C Profile in the top panel of **Figure 4a**). Instead, for the E Profile
 492 (and the C2E Profile described below) model-data comparisons must be conducted on an
 493 experiment-by-experiment basis. In **Figure 4b** we therefore focus our model-data
 494 comparison on a typical experiment (red symbols in the figure, experiment ID
 495 #20110613). To generate model predictions for this experiment we: (1) fit equation (11a)
 496 to the experiment's water column data yielding estimates for the inverse depth-scale and
 497 effective diffusivity ($a_E = 50 \pm 1.2 \text{ m}^{-1}$ and $D_{\text{eff},0}^E = (5.6 \pm 0.5) \times 10^{-6} \text{ m}^2 \text{ s}^{-1}$); and (2) plotted the
 498 model against the same abscissa used in **Figure 4a** after applying the following
 499 dimensionless time transformation, $\bar{t}_E = (\bar{h}_w^2 D_{\text{eff},0}^E / D_{\text{eff},0}^C) \bar{t}_C$, and substituting experiment-
 500 specific values for the dimensionless water depth ($\bar{h}_w = a_E h_w / \theta = 33 \pm 14$) and the C Profile's
 501 effective diffusivity ($D_{\text{eff},0}^C = (3.4 \pm 0.2) \times 10^{-7} \text{ m}^2 \text{ s}^{-1}$). Equation (11a) closely reproduces the
 502 experiment's water column concentrations of Rhodamine WT (compare solid curve and
 503 red points in **Figure 4b**, top panel), although some model bias is evident for $\sqrt{\bar{t}_C} < 0.05$.
 504 The E Profile's infinite bed model (equation (11b)) also captures the fast and slow mixing
 505 of Rhodamine WT out of the shallow and deep portions of the sediment bed (compare

506 black curves and red points, lower panel, **Figure 4b**). The last result is notable given that
507 the parameter values used to generate these predictions were inferred from the water
508 column data; i.e., by fitting the E Profile's infinite bed model to measurements of
509 Rhodamine WT in the water column we can predict the evolution of Rhodamine WT in
510 the interstitial fluids of the sediment bed with reasonable fidelity.

511 **4.4 Experimental Evaluation of the E2M Profile**

512 Setting the tortuosity-modified molecular diffusion coefficient, D'_m , as a lower bound on
513 the effective diffusivity does not improve the E Profile's model performance. We arrived
514 at this conclusion by estimating a tortuosity-modified diffusion coefficient for Chandler
515 et al.'s experiments ($D'_m = 1.31 \times 10^{-10} \text{ m}^2\text{s}^{-1}$) from the ratio of the molecular diffusion
516 coefficient for Rhodamine in water at 21°C ($D_m = 2.9 \times 10^{-10} \text{ m}^2\text{s}^{-1}$) (Chandler 2012) and
517 the tortuosity of the sediment bed ($\tau = 2.22$), estimated by substituting the measured bed
518 porosity ($\theta = 0.39$) into an empirical formula proposed by Iversen and Jorgensen (1992)
519 for sandy beds: $\tau = 1 + 2(1 - \theta)$. Values of the inverse depth-scale and effective diffusivity,
520 a_{E2M} and $D_{\text{eff},0}^{\text{E2M}}$, were then inferred by fitting the E2M Profile's infinite bed model
521 (constructed by substituting equations (T5a,b,c) in **Table 1** into equations (7a) and (7b))
522 to Chandler et al.'s water column Rhodamine WT measurements. Repeating this process
523 for all twenty of Chandler et al.'s experiments reveals that the E2M's inverse depth-
524 scales and effective diffusivities are nearly identical to those inferred from the E Profile
525 model (**Figure S1**, also compare **Tables S4** and **S6**), implying that these two profiles are
526 functionally equivalent. This result can be rationalized by noting that the timescale for
527 mixing through the sediment column by molecular diffusion ($t_m = \ell_m^2 / D'_m \approx 11$ months,

528 where the mixing depth is $\ell_m = -\ln(D'_m/D_{\text{eff},0}^{\text{E2M}})/a_{\text{E2M}}$, see equation (8c) and discussion
529 thereof) is much longer than the time period Chandler et al. conducted their experiments
530 (days to weeks). While the E2M Profile did not add value (relative to the E Profile) to our
531 analysis of Chandler et al.'s experiments, it might prove useful in settings with very low
532 levels of turbulence and relatively impermeable sediments (i.e., very small values of the
533 Permeability Reynolds Number, see **Section 5**).

534 **4.5 Experimental Evaluation of the C2E Profile**

535 The C2E Profile's infinite bed model provides a near perfect representation of Chandler
536 et al.'s water column data (**Figure 4c**). The C2E's infinite bed model was constructed by
537 substituting its Green's function (equations (T6a) - (T6c) in **Table 1**) into equations (7a)
538 and (7b). We then followed the same two-step procedure outlined in **Section 3.3**, by (1)
539 fitting the model to Rhodamine concentrations measured in the water column during Exp
540 ID# 20110613 ($D_{\text{eff},0}^{\text{C2E}} = (1.5 \pm 0.07) \times 10^{-6} \text{ m}^2 \text{ s}^{-1}$, $\ell_t = 0.04 \pm 0.002 \text{ m}$, $a_{\text{C2E}} = 66 \pm 3.1 \text{ m}^{-1}$); and
541 (2) plotting the model against the same abscissa used in **Figures 4a,b** after applying the
542 following dimensionless time transformation, $\bar{t}_{\text{C2E}} = (\bar{h}_w^2 D_{\text{eff},0}^{\text{C2E}} / D_{\text{eff},0}^{\text{C}}) \bar{t}_{\text{C}}$, and substituting
543 experiment-specific values for the dimensionless water depth ($\bar{h}_w = a_{\text{C2E}} h_w / \theta = 44 \pm 19$) and
544 the C Profile's effective diffusivity ($D_{\text{eff},0}^{\text{C}} = (3.4 \pm 0.2) \times 10^{-7} \text{ m}^2 \text{ s}^{-1}$). As noted above, the
545 C2E Profile's infinite bed model provides a near perfect match to Rhodamine WT
546 concentrations measured in the water column (compare solid curve and red points in
547 **Figure 4c**). However, when these same parameter values are used to estimate Rhodamine
548 WT concentration in the interstitial fluids of the bed, there is some loss of model
549 performance compared to the E Profile; i.e., the C2E model systematically

550 underestimates mixing in the streambed, especially for Rhodamine measurements at 15
551 mm below the SWI (compare lower panels in **Figures 4b** and **4c**).

552 **5. Discussion**

553 Here we expand our focus to include all twenty of Chandler et al.'s experiments with the
554 goal of answering the three questions raised in **Section 1**.

555 **5.1 How is the Effective Diffusivity Structured Vertically?**

556 From refractive index matched particle image velocimetry (RIM-PIV) studies of
557 turbulent motions across the SWI of a permeable streambed, Voermans et al. (2017)
558 concluded that the flux of mass and momentum is dominated by different transport
559 mechanisms (molecular diffusion, dispersion, turbulent diffusion) depending on the
560 magnitude of the Permeability Reynolds Number, $Re_K = u_* \sqrt{K} / \nu$, a dimensionless
561 arrangement of the shear velocity u_* [$L T^{-1}$], sediment bed permeability K [L^2], and
562 kinematic viscosity of water ν [$L^2 T^{-1}$]. Specifically, molecular diffusion dominates
563 mixing across the SWI for small values of this dimensionless number ($Re_K < 0.01$),
564 dispersion for intermediate values ($0.01 < Re_K < 2$), and turbulent diffusion for large values
565 ($Re_K > 2$).

566 Chandler et al.'s dataset allows us to quantitatively compare the performance of
567 the three diffusivity profiles (C, E, and C2E) over a Permeability Reynolds Number range
568 (0.2 to 4.34) that spans dispersive and turbulent diffusive mixing regimes (note that the
569 E2M profile is not included in this list because, as noted in **Section 4.4**, this profile is
570 functionally equivalent to the E Profile in the context of Chandler et al.'s dataset). Based
571 on our earlier analysis of Chandler et al.'s Exp ID 20110613 we concluded that the E
572 Profile represents a substantial improvement over the C Profile (**Section 4.3**). This

573 conclusion extends to the rest of Chandler et al.’s experiments as well (compare blue and
574 red circles in **Figure 5**, inferred parameter values summarized in **Tables S2 – S5**). In all
575 cases, the E Profile’s infinite bed model captures a larger fraction of data variance ($R^2 >$
576 99.5% , **Figure 5a**) and has substantially smaller RMSE values (**Figures 5b**). The E
577 Profile’s AICc is also >10 units lower than the C Profile’s AICc (**Figure 5c**) implying
578 that, despite its increased complexity (i.e., the C and E Profiles have one and two
579 unknown parameters, respectively) the E Profile is the more parsimonious model (Weijts
580 and Ruddell, 2020; Aho et al., 2014).

581 [Figure 5 goes about here]

582 The C2E Profile’s infinite bed model also performs quite well, but is a problem
583 with the inferred values of its inverse depth-scale. Compared to the E Profile, the C2E
584 Profile’s infinite bed model has consistently lower RMSE and AICC values (compare
585 crosses and red circles in **Figures 5b** and **5c**) and a slightly improved coefficient of
586 determination ($R^2 > 99.8\%$, **Figure 5a**). However, these improvements come at the cost of
587 a new parameter (the C2E’s inverse depth scale, a_{c2E}) whose inferred values are poorly
588 constrained (coefficient of variation in excess of 40 for some experiments, see **Table S7**),
589 highly variable (varying over 1000-fold from experiment-to-experiment, see distribution
590 of values represented by the violin plot in **Figure 6a**) and, in some cases, not physically
591 meaningful (e.g., the implied depth over which the effective diffusivity decays for $y > \ell_t$
592 is $1/a_{c2E} \approx 20$ microns for the largest value of a_{c2E} indicated in **Figure 6a**). Inferred
593 values of the C2E Profile’s other two parameters (effective diffusivity, $D_{eff,0}^{c2E}$, and depth of
594 constant mixing, ℓ_t) are strongly correlated ($R^2 = 0.93$ and 0.76) with the effective

595 diffusivity, $D_{\text{eff},0}^E$, and decay length-scale, $1/a_E$, inferred from the E Profile (**Figure 6b**,
596 **6c**), implying that both models capture the same basic information about the vertical
597 structure of the effective diffusivity; namely that the diffusivity is larger at the SWI and
598 lower at depth. In summary, of the four profiles evaluated in this study and across the full
599 range of Permeability Reynolds Numbers evaluated here (including dispersive and
600 turbulent diffusive transport regimes, see horizontal axis of **Figure 5**) the E Profile is the
601 most parsimonious descriptor of the effective diffusivity's vertical structure.

602 [Figure 6 goes about here]

603 **5.2 Is the Flux-Gradient Diffusive Model an Accurate Representation of Turbulent** 604 **Solute Transport Through the Streambed?**

605 We have already shown that the E Profile's infinite bed model captures a large
606 percentage of the variance in Chandler et al.'s water column Rhodamine measurements
607 ($R^2 > 99.5\%$, **Figure 5a**) but this assessment is based on the same datasets used for model
608 calibration. A more rigorous test of the E Profile in particular, and the flux-gradient
609 diffusive model more generally, can be stated as follows: *are the same E Profile*
610 *parameter values obtained when the model is optimized with Rhodamine measured in the*
611 *water column versus when the model is optimized with Rhodamine measured in the*
612 *interstitial fluids of sediment bed? Put another way, can the evolution of solute*
613 *concentrations in the interstitial fluids of the sediment bed be inferred from the evolution*
614 *of solute concentration in the water column, and vice versa?*

615 The answer is a qualified "yes". Across all 20 of Chandler et al.'s experiments,
616 effective diffusivities $D_{\text{eff},0}$ estimated from the water and sediment column data are
617 strongly correlated over a 1000-fold change in the magnitude of this parameter (**Figure**

618 **7a**, Pearson’s correlation coefficient, $R=0.867$). Values of the inverse depth-scale
619 inferred from the water and sediment column data are much less variable and not
620 significantly correlated (**Figure 7b**), but their respective log-means ($a=10^{1.61\pm0.18}$ m⁻¹ and
621 $a=10^{1.70\pm0.08}$, respectively) are equal within error. The corresponding mean values
622 ($a_e=44.0\pm18.5$ m⁻¹ and $a=51.4\pm9.71$ m⁻¹, respectively, obtained by applying equations
623 (S25a) and (S25b) in Supplemental Information) are also equal, within error, to the
624 inverse depth-scale estimated by Chandler et al. in their original publication ($a=55$ m⁻¹)
625 (Chandler et al., 2016). This inverse depth-scale corresponds to a “1/e-folding depth”
626 (i.e., the depth at which the E Profile’s effective diffusivity declines to $1/e\approx0.37$ of its
627 surficial value) of approximately 2 cm, which comports with field and laboratory
628 estimates for the thickness of the benthic biolayer (2 to 5 cm below the SWI) (Tomasek et
629 al., 2018; Knapp et al., 2017; Krause et al., 2017; Caruso et al., 2017; Trauth et al., 2014;
630 Harvey et al., 2013; Kessler et al., 2013; Zarnetske et al., 2011; Battin et al., 2008; Dahm
631 et al., 2002). Thus, turbulent mixing may play an outsized role in the biogeochemical
632 processing of nutrients and other contaminants in streambed and coastal sediments.

633 [Figure 7 goes about here]

634 While effective diffusivities inferred from data collected above and below the
635 SWI are strongly correlated, bias is evident for experiments with Permeability Reynolds
636 Numbers greater than the approximate threshold for a fully turbulent SWI, $Re_K > 10^{0.3} = 2$
637 (Voermans et al., 2017) (**Figure 7a**). One possible explanation focuses on how Chandler
638 et al. measured Rhodamine concentrations in the interstitial fluids of the sediment bed.
639 Chandler et al.’s in-bed measurements of Rhodamine WT are taken at a “point” (sensing
640 volume ca., 0.23 cm³) with fiber optic fluorometers, whereas our analytical framework

641 assumes that interstitial solute concentrations are horizontally averaged (**Section 2**).
642 Indeed, Chandler (2010) describes how measured time series of concentration in the
643 sediment were sensitive to fluorometer location and these differences appeared consistent
644 over time; i.e. tracer appeared to mix out of the streambed faster on one side of the tank
645 than on the other (ibid. pg. 173). These authors also document distinct and persistent
646 patterns of mean flow velocity within the tank (ibid. pg. 118) which would lead to
647 heterogeneous turbulence intensities and corresponding heterogeneous efflux across the
648 SWI at high Permeability Reynolds Numbers. To the extent that Chandler et al.’s point
649 measurements are not equal to horizontally averaged concentrations, the effective
650 diffusivities inferred from these data will suffer accuracy problems. Indeed, Chandler
651 (2010) noted an order of magnitude discrepancy in the timescale over which interstitial
652 Rhodamine WT concentration declined on opposite sides of the stirred tank—a
653 discrepancy that could induce order-of-magnitude inaccuracy in estimated sediment-side
654 diffusivities (the scale of disagreement seen in **Figure 7a**).

655 An alternative explanation is that the flux-gradient diffusive model (equation (1a))
656 is an imperfect descriptor of turbulent mass transfer in the interstitial fluids of the
657 sediment bed at high Permeability Reynolds Number—a conclusion supported by the
658 systematic loss of model performance (i.e., higher RMSE and lower AICc and R^2 values)
659 observed at large values of this dimensionless number (i.e., $Re_k > 1$, **Figure 5**). Indeed,
660 the flux-gradient description of momentum transfer across other types of “porous”
661 boundaries, such as vegetation canopies, can break down; i.e., significant momentum flux
662 can occur in the absence of a velocity gradient, the analog of a concentration gradient for
663 mass transfer (Ghisalberti and Nepf, 2002). However, such observations typically occur

664 in the dispersive transport regime ($0.01 < \text{Re}_k < 2$) where our estimates of effective
665 diffusivity above and below the SWI are in general concordance (**Figure 7a**). It should
666 also be noted that, even for the highest Permeability Reynolds Numbers trialed here and
667 so long as the diffusivity is allowed to decay exponentially with depth, the flux-gradient
668 diffusive model still explains a large fraction of variance in water column measurements
669 of Rhodamine WT ($R^2 > 0.995$) (**Figure 5a**).

670 In summary, based on Chandler et al.'s laboratory measurements we can conclude
671 the flux-gradient diffusive model is a reasonable representation of turbulent mass transfer
672 in the interstitial fluids of the sediment bed, *provided that the vertical structure of the*
673 *effective diffusivity is correctly specified (e.g., with the E Profile)*. However, a systematic
674 loss of model performance is observed with increasing Permeability Reynolds Number in
675 the turbulent diffusive regime ($\text{Re}_k > 2$) possibly reflecting non-idealities associated with
676 Chandler et al.'s experimental measurements, a more fundamental breakdown in the flux-
677 gradient diffusive model, or some combination thereof. Further studies along these lines,
678 particularly in field settings, are warranted.

679 **5.3 Can Laboratory Measurements of Turbulent Mixing across the SWI be** 680 **Extrapolated to Stream and Coastal Sediments?**

681 Translation of these results to the field requires “scaling relationships” from which the E
682 Profile's two parameters—the effective diffusivity and inverse depth-scale—can be
683 estimated. Over the years a number of researchers have reported that effective
684 diffusivities (inferred by fitting the C Profile's null model to flume measurements of
685 turbulent mixing across flat sediment beds) exhibit a quadratic dependence on the
686 Permeability Reynolds Number, $D_{\text{eff},0}^c \propto \text{Re}_k^2$ (Richardson and Parr, 1988; O'Connor and

687 Harvey, 2008; Grant et al., 2012; Voermans et al., 2018; Grant et al., 2018). As noted
688 earlier, the Permeability Reynolds Number is calculated from the shear velocity, u_* ,
689 sediment bed permeability, K , and the kinematic viscosity of water, ν . Permeability can
690 be estimated from the grain diameter and porosity of unconsolidated sediments (e.g.,
691 using the Kozeny-Carmen equation (Kamaan et al., 2007)) while the kinematic viscosity
692 of water is determined primarily by temperature (Rumble, 2019). Several methods are
693 available for measuring the shear velocity (c.f., Johnson and Cowen, 2017) including a
694 force-balance approach that provides spatially averaged estimates from the depth, h_w , and
695 slope, S [-], of a stream: $u_* = \sqrt{gh_w S}$ where $g = 9.81 \text{ m s}^{-2}$ is gravitational acceleration.
696 Thus, if the E Profile's two parameters can be expressed in terms of the Permeability
697 Reynolds Number, such relationships might pave the way for a direct translation of
698 laboratory measurements to field applications.

699 [Figure 8 goes about here]

700 When log-transformed effective diffusivities are plotted against the log-
701 transformed Permeability Reynolds Number, a significant change in slope and intercept
702 (as represented by non-overlapping 95% confidence intervals in equation (12a)) is
703 evident around $Re_K = 1$ (**Figure 8a**).

$$704 \log_{10}(D_{\text{eff},0}^E, \text{ m}^2\text{s}^{-1}) = \begin{cases} (-5.31 \pm 0.09) + (2.53 \pm 0.17) \times \log_{10} Re_K, & R^2 = 0.905, Re_K < 1 \\ (-5.57 \pm 0.06) + (0.99 \pm 0.15) \times \log_{10} Re_K, & R^2 = 0.440, Re_K > 1 \end{cases} \quad (12a)$$

705 The power-law exponent for the surficial effective diffusivity declines from 2.53 ± 0.17
706 in the dispersive regime ($Re_K < 1$) to 0.99 ± 0.15 in the turbulent diffusive regime
707 ($Re_K > 1$). Likewise, the scaling behavior of the inverse depth-scale transitions from a

708 constant value ($a = 10^{1.74 \pm 0.01} \text{ m}^{-1}$) to a weak inverse dependence on the Permeability

709 Reynolds Number ($a \propto \text{Re}_K^{-0.32 \pm 0.08}$) (**Figure 8b**):

$$710 \quad \log_{10}(a_E, \text{ m}^{-1}) = \begin{cases} 1.74 \pm 0.01, & \text{Re}_K < 1 \\ (1.69 \pm 0.02) - (0.32 \pm 0.08) \times \log_{10} \text{Re}_K, & R^2 = 0.18, \text{Re}_K > 1 \end{cases} \quad (12b)$$

711 In the dispersive mixing regime ($\text{Re}_K < 1$) the implied mixing depth, $1/a_E \approx 2 \text{ cm}$, is
712 between 4 and 130 times larger than the diameter of the glass spheres that make up the
713 sediment bed ($0.150 \leq d_g \leq 5 \text{ mm}$, depending on the experiment) and about 20 times larger
714 than an estimate of the depth to which the time-averaged turbulent velocity boundary
715 layer penetrates into the streambed called the Brinkman Layer thickness, $\delta_b \approx 20\sqrt{K} = 0.9$
716 mm (Voerman et al.'s 2017). It is also about ten-fold less than the depth of the sediment
717 bed ($d_b = 20 \text{ cm}$), implying that the inferred inverse depth scale is not a proxy for bed
718 depth. These comparisons raise the question: *what is the physical interpretation of the*
719 *inverse depth-scale?* Based on a model for mass exchange across the SWI by turbulent
720 pumping, Higashino et al. (2009) reported that, at depths of around 2 cm and for shear
721 velocities on the lower end of the range employed by Chandler et al. ($u_* = 0.01$ and
722 0.016 m s^{-1}), the root mean square vertical velocity of the interstitial pore fluids are
723 $>10\%$ of their value at the SWI (ibid, Figure 3). Thus, one plausible interpretation is that
724 the inverse depth-scale represents the surficial region of streambed over which tracer
725 mass is vigorously mixed by turbulent pumping.

726 **6 Conclusions**

727 In this paper we developed and tested a rigorous one-dimensional modeling framework,
728 based on Duhamel's Theorem, for predicting mass transfer across the SWI and in the

729 benthic biolayer of a turbulent stream. The framework allows for depth-varying
730 diffusivity profiles, accounts for the change in porosity across the SWI, and encodes two-
731 way coupling across the SWI, in which mass transfer into the water column from the
732 sediment bed alters the water column concentration which, in turn, alters the mass flux
733 from the water column to the sediment bed, and so on. We applied this theory to an
734 extensive set of previously published measurements of turbulent mixing across a flat
735 sediment bed in a closed stirred tank (Chandler et al., 2016) with the goal of evaluating
736 the performance of four diffusivity depth profiles (C, E, E2M, and C2E Profiles). Key
737 findings include: (1) the flux-gradient diffusive model is a reasonable representation of
738 turbulent mass transfer across the SWI and in the interstitial fluids of the sediment bed,
739 *provided that the vertical structure of the effective diffusivity is correctly specified*; (2)
740 Chandler et al.'s experiments are most consistent with an exponentially declining
741 diffusivity profile (i.e., the E Profile); (3) values of the E Profile's two parameters
742 (effective diffusivity at the SWI, $D_{\text{eff},0}^E$, and inverse depth-scale, a_E) vary with the
743 Permeability Reynolds Number, Re_K , a dimensionless number that incorporates bed shear
744 stress and sediment permeability; (4) the dependence on the Permeability Reynold
745 Number changes abruptly at $Re_K = 1$, reflecting different modes of mixing below
746 (turbulent dispersion, $D_{\text{eff},0}^E = 10^{-5.31 \pm 0.09} Re_K^{2.53 \pm 0.17}$ and $a_E = 10^{1.74 \pm 0.01}$) and above (turbulent
747 diffusion, $D_{\text{eff},0}^E = 10^{-5.57 \pm 0.06} Re_K^{0.99 \pm 0.15}$ and $a_E = 10^{1.69 \pm 0.02} Re_K^{-0.32 \pm 0.08}$) this threshold value; (5)
748 the effective diffusivity's $1/e$ -folding depth is approximately 2 cm, consistent with the
749 hypothesis that solute mixing in the interstitial fluids of the sediment bed is facilitated
750 primarily by turbulent pumping; and (6) this $1/e$ -folding depth is also concordant with

751 field and laboratory measurements of the benthic biolayer thickness, implying that
752 turbulent mixing, and in particular turbulent pumping, may play an outsized role in the
753 biogeochemical processing of nutrients and other contaminants in streambed and coastal
754 sediments. Studies are presently underway to extend these findings to mixing across the
755 SWI by bedform pumping and bedform turnover with the ultimate goal of informing
756 pollutant fate and transport in coastal sediments and streams at catchment-to-continental
757 scales (Schmadel et al., 2019; Grant et al., 2018; Gomez-Velez et al., 2015; Gomez-Velez
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779 **References.**

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991 surface in the smooth open channel flow. *Water Resources Research*, 52, 3563–3578.
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993 **Figure Captions.**

994

995 **Figure 1. (a)** An illustration of how water column turbulence can influence mass
996 transport in the benthic biolayer. In this diagram, the benthic biolayer consists of a flat
997 coarse-grained streambed subject to turbulent pumping (traveling pressure wave, dashed
998 blue line), turbulence penetration (red eddies), and a time-averaged turbulent velocity
999 boundary layer that crosses the sediment-water interface (envelope of black arrows). The
1000 vertical mass flux $J(y)$ arising from these turbulence-linked phenomena is assumed to
1001 follow the flux-gradient diffusive model (equation (1a)). **(b)** Turbulent mixing across the
1002 SWI can be measured in the laboratory using closed systems, such as a stirred tank. Two-
1003 way coupling across the SWI is indicated by the two circular arrows.

1004 **Figure 2.** Four functional forms of the effective diffusivity profile $f(y)$ trialed in this
1005 study (equations (8a) through (8d)). Variables represent the depth into the sediment bed
1006 (y), surficial effective diffusivity (at the SWI, $D_{\text{eff},0}$), a mixing length-scale ($1/a$), and
1007 the thickness of enhanced mixing at the surface of the sediment bed (ℓ_t).

1008 **Figure 3.** The influence of two-way coupling and finite bed depth on the evolution of
1009 solute concentration in the **(a)** water column and **(b)** interstitial fluids of the sediment
1010 bed, assuming solute is initially present only in the interstitial fluids of the sediment bed
1011 and the diffusivity profile is constant with depth (C Profile). See text for details.

1012 **Figure 4.** Chandler et al.'s water (top panels) and sediment (bottom panels) column data
1013 compared to infinite bed model predictions for the **(a)** C Profile, **(b)** E Profile, and **(c)** C2E
1014 Profile. Black curves are model-predicted solute concentration in the water column (top
1015 panel) and at two depths in the sediment bed (15 and 151 mm below the SWI, lower panel).
1016 Blue horizontal lines represent the well-mixed (equilibrium) concentration (equation (9a)).
1017 Data highlighted in red correspond to Chandler et al.'s Exp ID 20110613.

1018 **Figure 5.** Performance of the C, E and C2E infinite bed models across all 20 of Chandler
1019 et al.'s stirred tank experiments. Model performance metrics include: **(a)** coefficient of
1020 determination, R^2 ; **(b)** Root Mean Square Error, RMSE; and **(c)** Akaike's Information
1021 Criterion, AICc.

1022 **Figure 6.** An evaluation of C2E fitting parameters inferred from Rhodamine
1023 concentration in the water column for all twenty of Chandler et al.'s stirred tank
1024 experiments. (a) Values of C2E's inverse depth scale vary over three orders of
1025 magnitude, as illustrated here with a violin plot. The effective diffusivity (b) and depth of
1026 the constant mixing zone (c) inferred from the C2E profile model (vertical axes) are
1027 strongly correlated ($R^2 = 0.93$ and 0.76) with, respectively, the E Profile's effective
1028 diffusivity and inverse decay depth-scale (horizontal axes).

1029 **Figure 7.** A comparison of (a) effective diffusivities and (b) inverse depth-scales
1030 obtained by fitting the E Profile model to water column (vertical axis) or sediment
1031 column (horizontal axis) measurements of Rhodamine concentration. The points are
1032 colored to indicate the Permeability Reynolds Number.

1033 **Figure 8.** Permeability Reynolds Number scaling of the E Profile's two parameters
1034 estimated from Chandler et al.'s water column measurements. The surficial diffusivity (a)
1035 and inverse decay length-scale (b) follow different scaling relationships in the dispersive
1036 (blue lines) and turbulent diffusive (green lines) ranges. Dashed curves are 95%
1037 prediction intervals.

1038

1039

Table 1. Green's functions for various choices of the diffusivity depth profile and a finite or semi-infinite sediment bed.¹

C Profile, $f_c(\bar{y})=1$	
Semi-infinite Sediment Bed, $\bar{y}=a_c y > 0, \bar{s}=st_\tau$	
$\tilde{G}(\bar{y}, \bar{s})=e^{-\bar{y}\sqrt{\bar{s}}}, \bar{y} > 0$	(T-1)
Finite Sediment Bed, $\bar{y}=a_c y > 0, \bar{d}_b=d_b a_c > 0, \bar{y} < \bar{d}_b, \bar{s}=st_\tau$	
$\tilde{G}(\bar{y}, \bar{s}) = \frac{e^{-\bar{y}\sqrt{\bar{s}}}\left(e^{2\bar{d}_b\sqrt{\bar{s}}} + e^{2\bar{y}\sqrt{\bar{s}}}\right)}{1 + e^{2\bar{d}_b\sqrt{\bar{s}}}}$	(T-2)
E Profile, $f_E(\bar{y})=e^{-\bar{y}}$	
Semi-infinite Sediment Bed, $\bar{y}=a_E y > 0, \bar{s}=st_\tau$	
$\tilde{G}(\bar{y}, \bar{s}) = \sqrt{e^{\bar{y}}} \frac{K_1\left(2\sqrt{\bar{s}}e^{\bar{y}}\right)}{K_1\left(2\sqrt{\bar{s}}\right)}$	(T-3)
E2M Profile, $f_{E2M}(\bar{y}) = \begin{cases} e^{-\bar{y}}, & 0 \leq \bar{x} \leq \bar{\ell}_m \\ \bar{D}, & \bar{x} > \bar{\ell}_m \end{cases}$	
Semi-infinite Sediment Bed, $\bar{y}=a_{E2M} y > 0, \bar{\ell}_m = -\ln \bar{D}, 0 < \bar{D} < 1, \bar{D} = \frac{D'_m}{D_{eff,0}^{E2M}}, \bar{s}=st_\tau$	
$\tilde{G}(\bar{y}, \bar{s}) = \begin{cases} \tilde{G}_1(\bar{y}, \bar{s}), & \bar{y} \leq -\ln \bar{D} \\ \tilde{G}_2(\bar{y}, \bar{s}), & \bar{y} > -\ln \bar{D} \end{cases}$	(T-5a)
$\tilde{G}_1(\bar{y}, \bar{s}) = \sqrt{e^{\bar{y}}} \frac{I_1\left(2\sqrt{e^{\bar{y}}\bar{s}}\right)\left[K_0\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right) - K_1\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right)\right] + K_1\left(2\sqrt{e^{\bar{y}}\bar{s}}\right)\left[I_0\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right) + I_1\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right)\right]}{K_1\left(2\sqrt{\bar{s}}\right)\left[I_0\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right) + I_1\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right)\right] + I_1\left(2\sqrt{\bar{s}}\right)\left[K_0\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right) - K_1\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right)\right]}$	(T-5b)
$\tilde{G}_2(\bar{y}, \bar{s}) = \frac{e^{-(\bar{y} + \ln \bar{D})\sqrt{\bar{s}}}}{\sqrt{\bar{D}}} \frac{I_1\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right)K_0\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right) + I_0\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right)K_1\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right)}{K_1\left(2\sqrt{\bar{s}}\right)\left[I_0\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right) + I_1\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right)\right] + I_1\left(2\sqrt{\bar{s}}\right)\left[K_0\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right) - K_1\left(2\sqrt{\frac{\bar{s}}{\bar{D}}}\right)\right]}$	(T-5c)
C2E Profile, $f_{C2E}(\bar{y}) = \begin{cases} 1, & 0 \leq \bar{y} \leq \bar{\ell}_t \\ e^{-\bar{y}}, & \bar{y} > \bar{\ell}_t \end{cases}$	
Semi-infinite Sediment Bed, $\bar{y}=a_{C2E} y > 0, \bar{\ell}_t = a_{C2E} \ell_t > 0, \bar{s}=st_\tau$	
$\tilde{G}(\bar{y}, \bar{s}) = \begin{cases} \tilde{G}_1(\bar{y}, \bar{s}), & \bar{y} \leq \bar{\ell}_t \\ \tilde{G}_2(\bar{y}, \bar{s}), & \bar{y} > \bar{\ell}_t \end{cases}$	(T-6a)

$\tilde{G}_1(\bar{y}, \bar{s}) = \frac{K_1(2\sqrt{\bar{s}}) \cosh[\sqrt{\bar{s}}(\bar{y} - \bar{\tau}_t)] - K_0(2\sqrt{\bar{s}}) \sinh[\sqrt{\bar{s}}(\bar{y} - \bar{\tau}_t)]}{K_1(2\sqrt{\bar{s}}) \cosh(\bar{\ell}_t \sqrt{\bar{s}}) + K_0(2\sqrt{\bar{s}}) \sinh(\bar{\ell}_t \sqrt{\bar{s}})}$	(T-6b)
$\tilde{G}_2(\bar{y}, \bar{s}) = \frac{e^{(\bar{y} - \bar{\tau}_t)/2} K_1(2e^{(\bar{y} - \bar{\tau}_t)/2} \sqrt{\bar{s}})}{K_1(2\sqrt{\bar{s}}) \cosh(\bar{\ell}_t \sqrt{\bar{s}}) + K_0(2\sqrt{\bar{s}}) \sinh(\bar{\ell}_t \sqrt{\bar{s}})}$	(T-6c)

1042 ¹The functions K_0 , K_1 , and K_2 are Modified Bessel functions of the Second Kind, while
 1043 I_0 , I_1 , and I_2 are Modified Bessel functions of the First Kind.

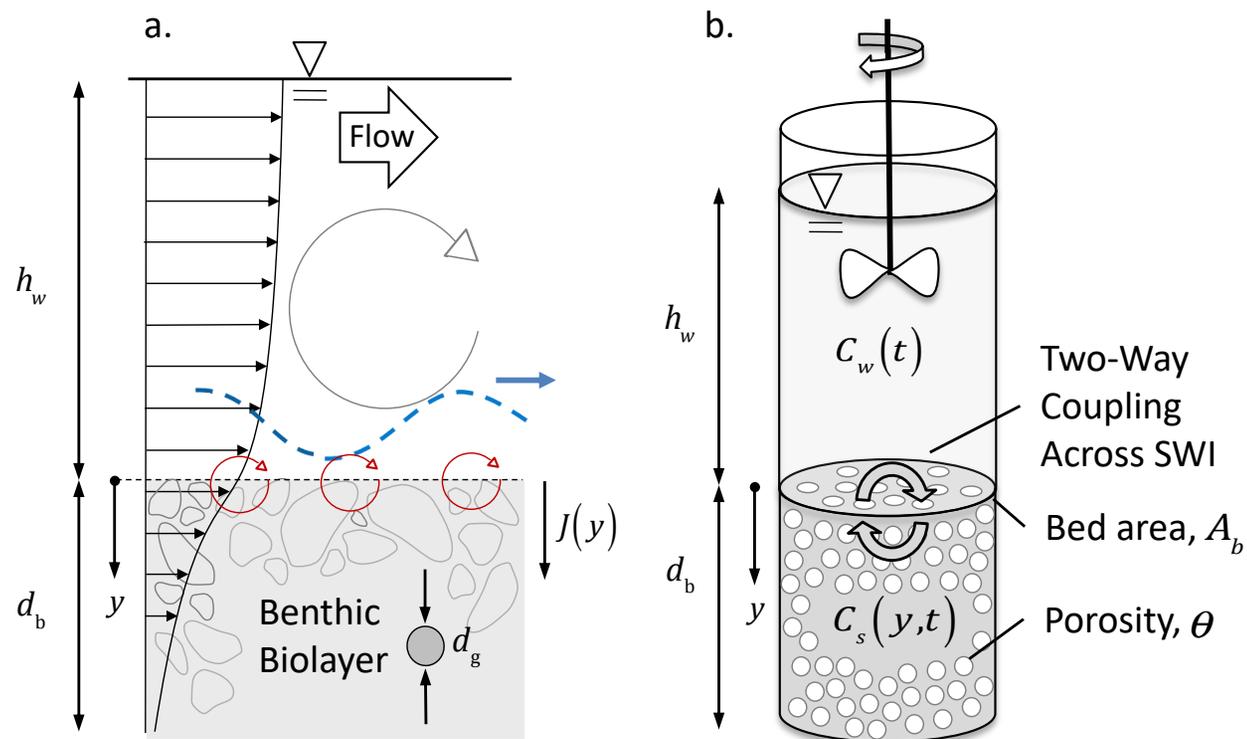


Figure 1. Grant et al. (2020)

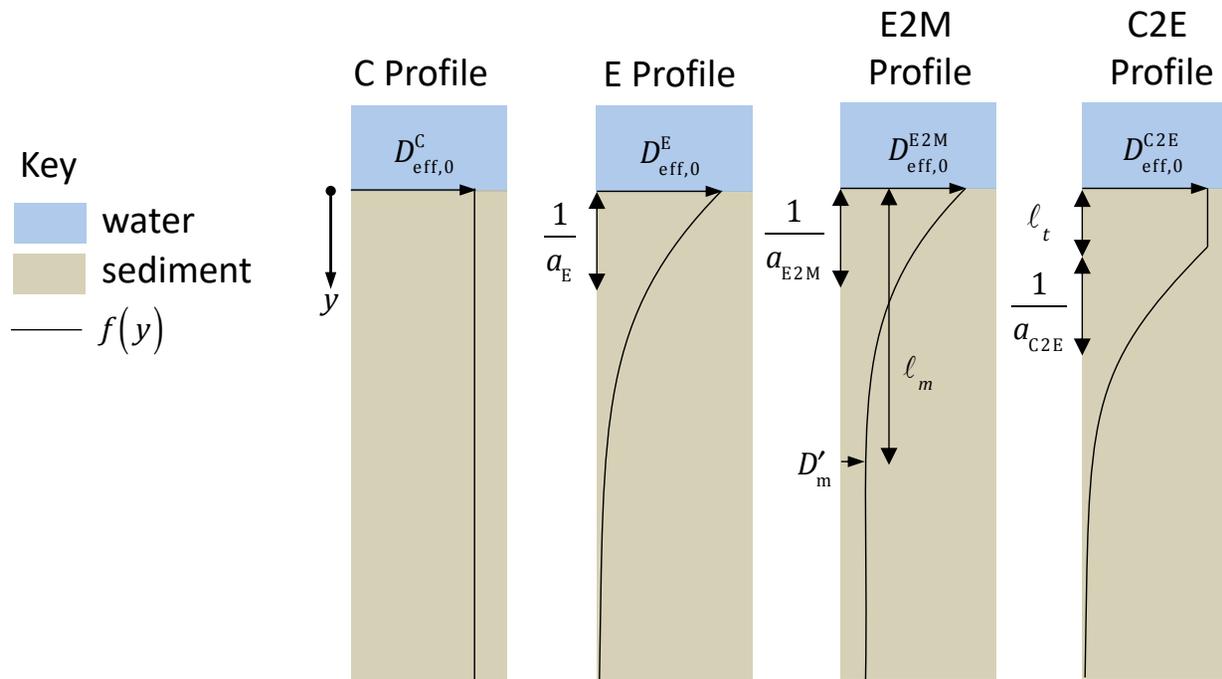


Figure 2. Grant et al. (2020)

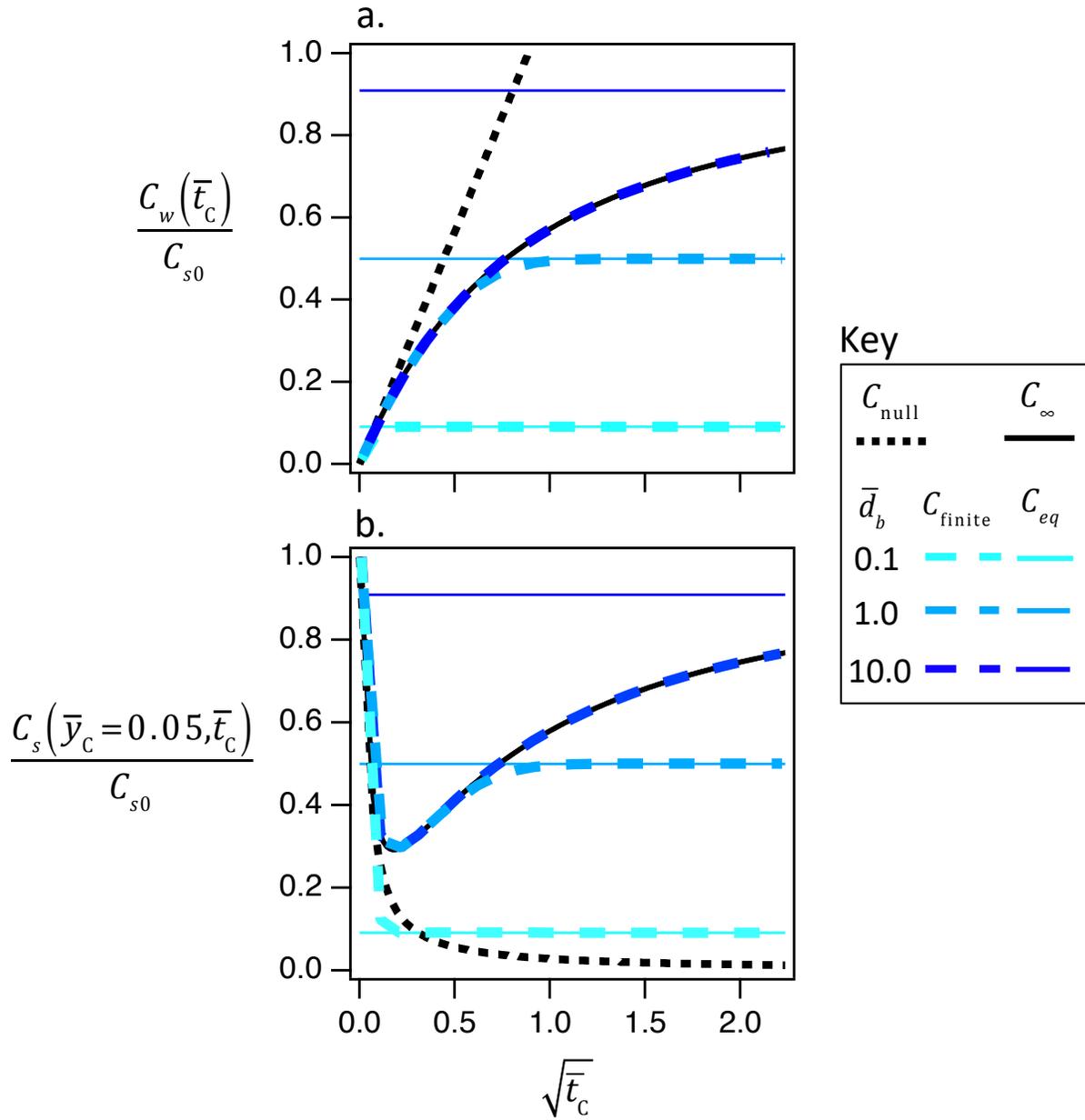


Figure 3. Grant et al. (2020)

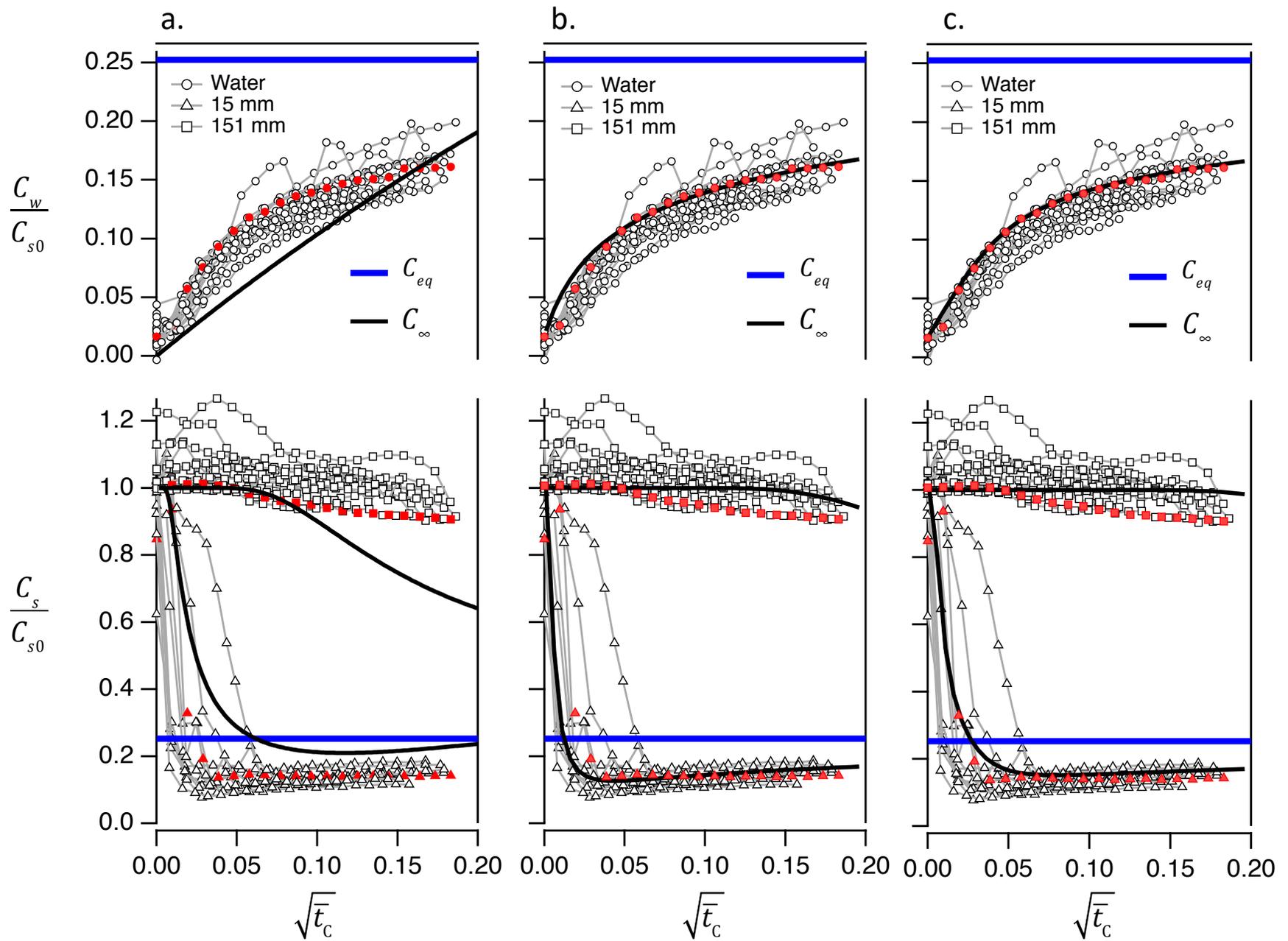


Figure 4. Grant et al. (2020)

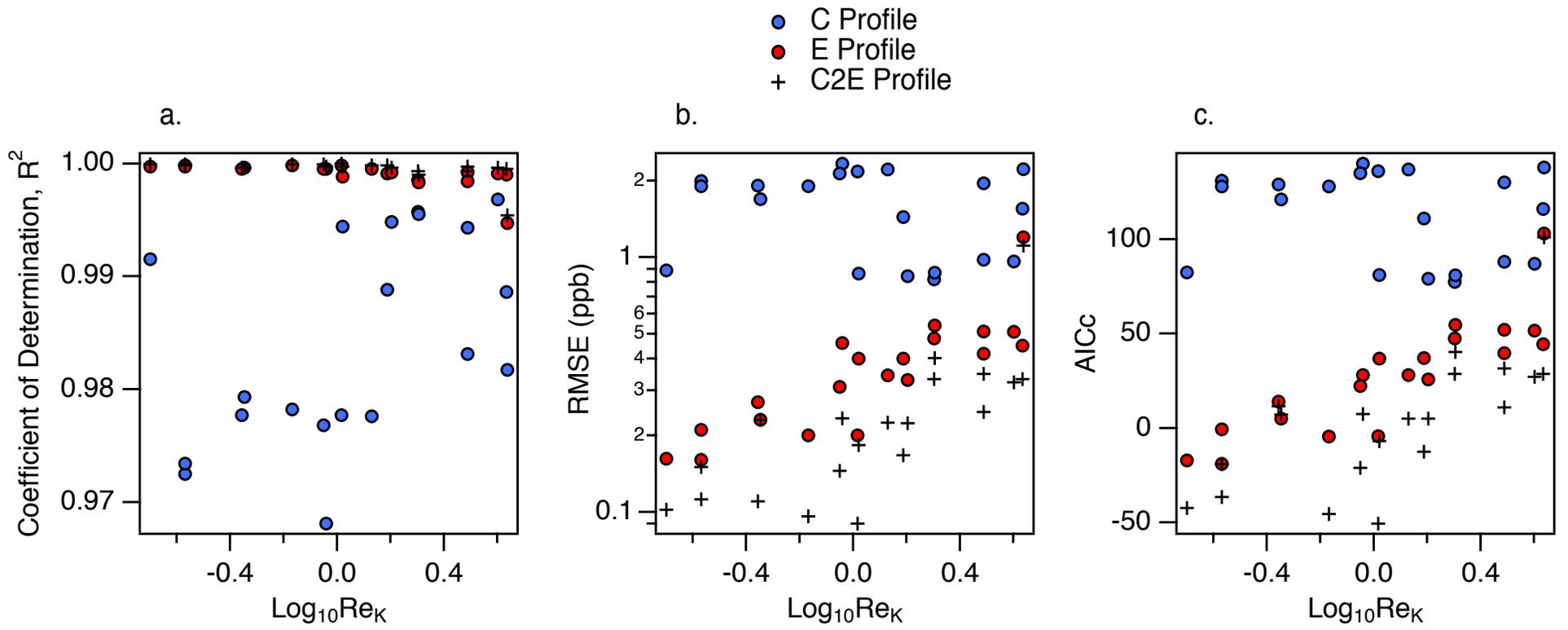


Figure 5. Grant et al. (2020)

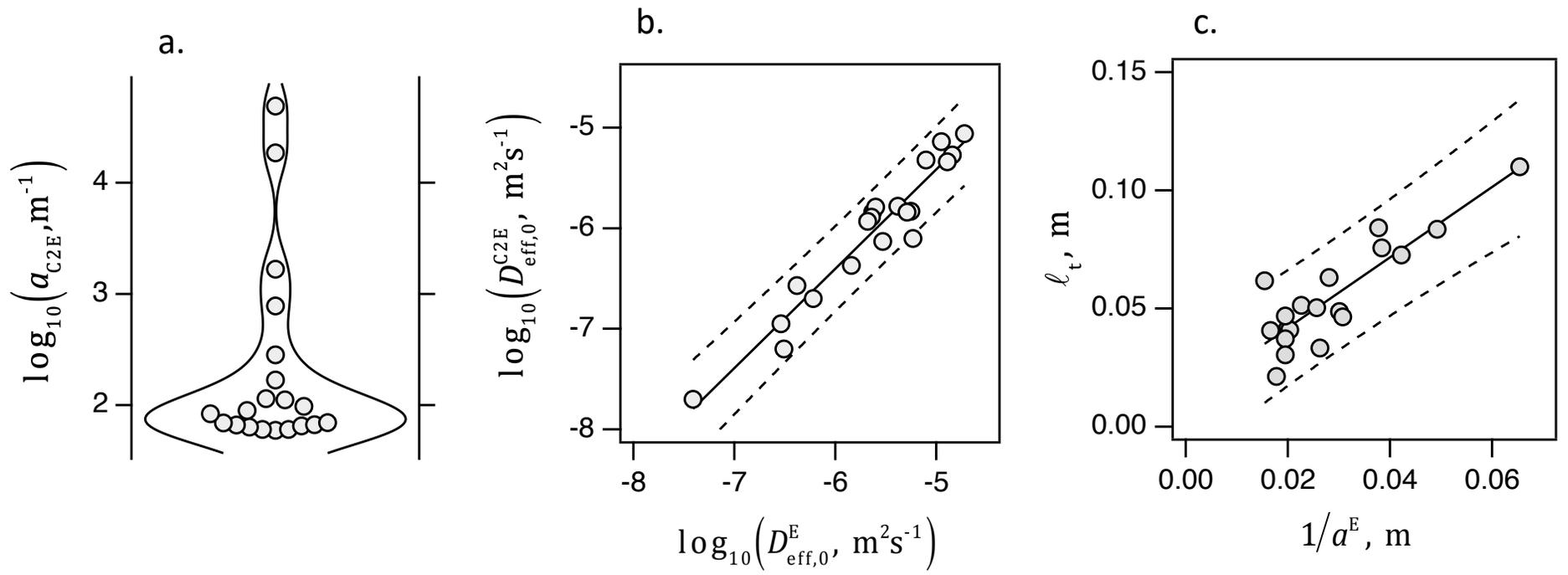


Figure 6. Grant et al. (2020)

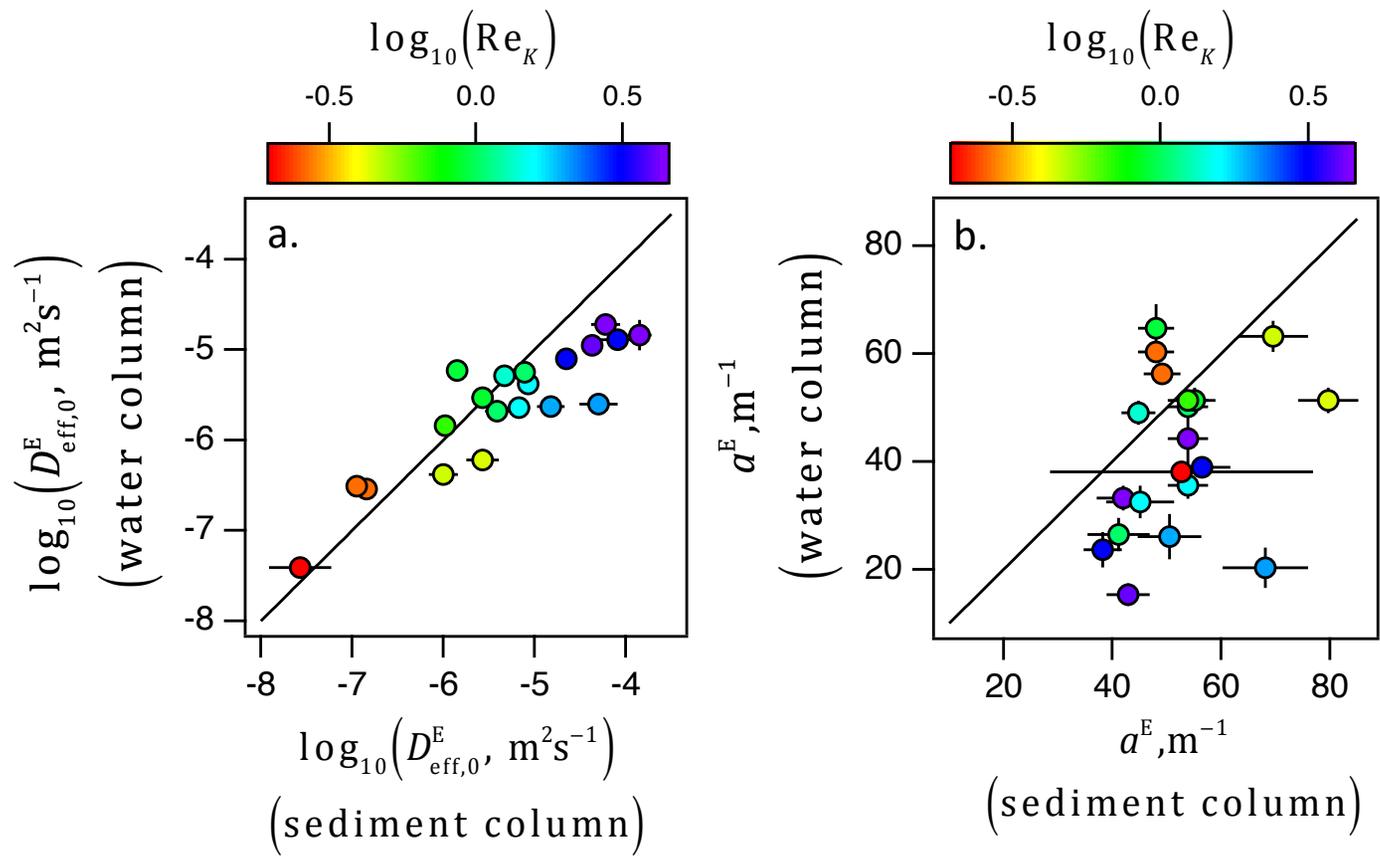


Figure 7. Grant et al. (2020)

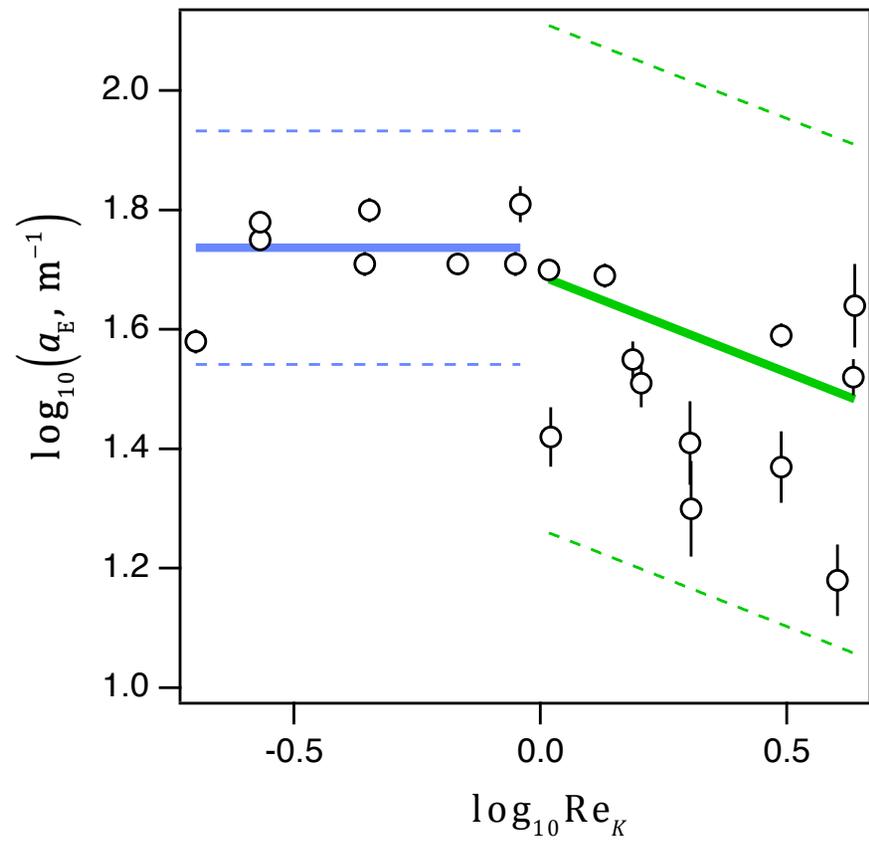
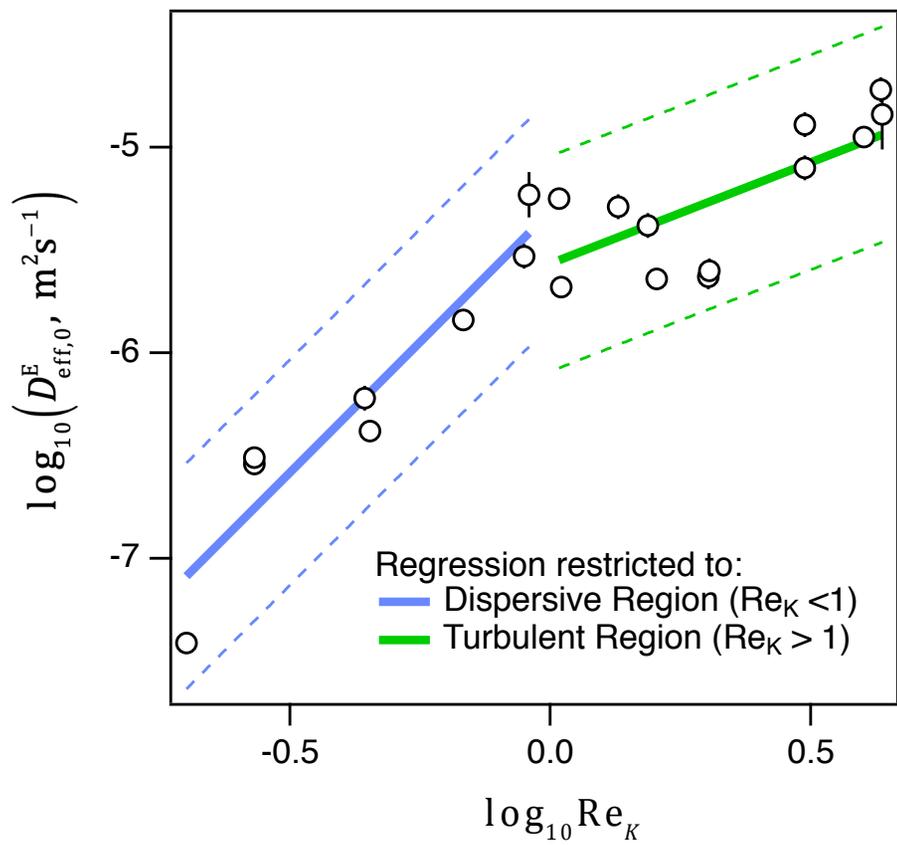


Figure 8. Grant et al. (2020)