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Exchange across the sediment-water interface quantified from porewater radon  
profiles

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21 HIGHLIGHTS:

- 22 • The radon profile method for estimating porewater exchange is reviewed
- 23 • A simple recirculation model is presented to aid in profile interpretation
- 24 • Uncertainties in the approach are discussed

25

## ABSTRACT

Water recirculation through permeable sediments induced by wave action, tidal pumping and currents enhances the exchange of solutes and fine particles between sediments and overlying waters, and can be an important hydro-biogeochemical process. In shallow water, most of the recirculation is likely to be driven by the interaction of wave-driven oscillatory flows with bottom topography which can induce pressure fluctuations at the sediment – water interface on very short timescales. Tracer-based methods provide the most reliable means for characterizing this short-timescale exchange. However, the commonly applied approaches only provide a direct measure of the tracer flux. Estimating water fluxes requires characterizing the tracer concentration in discharging porewater; this implies collecting porewater samples at shallow depths (usually a few mm, depending on the hydrodynamic dispersivity), which is very difficult with commonly used techniques. In this study, we simulate observed vertical profiles of radon concentration beneath shallow coastal lagoons using a simple water recirculation model that allows us to estimate water exchange fluxes as a function of depth below the sediment-water interface. Estimated water fluxes at the sediment water interface at our site were 0.18 – 0.25 m/day, with fluxes decreasing exponentially with depth. Uncertainty in dispersivity is the greatest source of error in exchange flux, and results in an uncertainty of approximately a factor-of-five.

**KEYWORDS:** porewater exchange, recirculated seawater, benthic flux, lake, tracers, radon

## 1. INTRODUCTION

Water recirculation through permeable sediments enhances the exchange of solutes and fine particles between sediments and overlying waters. In particular, it allows for a continuous supply of oxidants and fine particulate and dissolved matter (e.g. dissolved nutrients, bacteria, viruses, phytoplankton) into sediment porewaters, while enhancing the release of degradation products and organisms into overlying waters (Huettel and Rusch, 2000; Huettel et al., 1996). As a consequence, porewater exchange is considered a major contributor to the biogeochemical cycling of surface sediments and overlying waters, particularly in the coastal zone (Anschutz et al., 2009; Huettel et al., 2014, 1998; Jahnke et al., 2005). Porewater exchange increases in importance in highly-permeable sandy sediments, which cover >40% of coastal and shelf areas worldwide (Riedl et al., 1972), where this advective transport of solutes can exceed fluxes driven by molecular diffusion by several orders of magnitude (Huettel and Webster, 2001). This advective exchange between porewaters and overlying waters is caused by pressure gradients at the sediment-water interface, which might be forced by several mechanisms spanning a range of spatial and temporal scales, including wave and tidal pumping, interaction of bottom currents and seafloor topography, density instabilities or pumping activities of benthic fauna (bio-irrigation, Huettel et al., 2014; Santos et al., 2012). For instance, the passage of waves can produce oscillatory flows that interact with bottom topography (e.g. ripples), producing local increases of pressure that drive fluid exchange across the sediment-water interface. In addition, the passage of wave crests and troughs creates pressure

gradients over the seafloor that also enhance porewater exchange (Riedl et al., 1972; Rutgers van der Loeff, 1981; Webster, 2003).

Despite the importance of porewater exchange in coastal biogeochemical cycles, it is still not easy to quantify the advective flux of water and solutes in permeable sediments (Boudreau et al., 2001; Rocha, 2008). Common methods to estimate the rate of porewater exchange across the sediment-water interface in permeable sediments include (1) deploying automated seepage meters to monitor the porewater flow into overlying waters (e.g. Jahnke et al., 2000; Cable et al., 2006), (2) constructing mass balances in overlying waters for a tracer supplied by porewater inputs (e.g. Stieglitz et al., 2013; Rodellas et al., 2017), (3) injecting artificial tracers or dye into or above the sediments to trace fluid advection across the sediment-water interface (e.g. Reimers et al., 2004; Precht and Huettel, 2004), and (4) modeling the depth profiles of temperature, electrical conductivity or dissolved species in sediments to evaluate the sediment-water exchange rates (e.g. Cable and Martin, 2008; Savidge et al., 2016). Among these dissolved compounds, short-lived, naturally-occurring radionuclides (e.g.  $^{224}\text{Ra}$  and  $^{222}\text{Rn}$ ) have been widely used, mainly because their half-lives are well suited to the common timescales of these porewater exchange processes (e.g. Colbert and Hammond, 2008; Cable and Martin, 2008; Cai et al., 2014). Radium-224 (half-life = 3.6 d) is produced by radioactive decay of  $^{228}\text{Th}$ , but partitions to aquifer solids which reduces the sensitivity of dissolved  $^{224}\text{Ra}$  activities to porewater exchange fluxes. As a noble gas, radon ( $^{222}\text{Rn}$ ; half-life = 3.8 d) is an excellent tracer of porewater exchange because it is not affected by chemical and biological processes occurring within sediments. In this study, we review the application of radon porewater

profiles in sediments to estimate the porewater exchange across the sediment-water interface driven by pressure fluctuations on short time-scales, such as those produced by wave action. We simulate observed radon profiles using a numerical model of water recirculation that allows us to estimate porewater exchange rates as a function of depth below the sediment-water interface, and to explore sensitivity of estimated fluxes to model parameters. An alternative simpler approach is also presented, where water fluxes are estimated based on changes in the observed radon concentration gradient with depth.

## 2. THEORY

Radon-222 is a natural environmental tracer that has been used for quantifying groundwater inflows to streams and estuaries (e.g. (Cook et al., 2006; Genereux et al., 1993)) and the ocean (e.g. (Cable et al., 1996)) for almost three decades. Radon is produced in the sediments by the radioactive decay of  $^{226}\text{Ra}$ , which is part of the  $^{238}\text{U}$  decay chain and it is found in the sediment solids and in porewater. After porewater containing radon discharges to surface water bodies, radon activities in the surface water decrease due to gaseous exchange with the atmosphere and radioactive decay. Radon concentrations in surface water are therefore always much lower than concentrations within porewaters. When surface water infiltrates, its dissolved radon concentration will increase, according to

$$\frac{\partial c}{\partial t} = \gamma - \lambda c \quad (1)$$

where  $c$  is the radon concentration,  $\gamma$  is the production rate of dissolved radon,  $\lambda$  is the radon radioactive decay coefficient ( $0.1818 \text{ d}^{-1}$ ), and  $t$  is time. After a residence time of a few weeks the concentration will reach secular equilibrium, in which the rate of production is exactly balanced by the rate of radioactive decay. The concentration at secular equilibrium is equal to  $\gamma/\lambda$ . If the radon production rate within the sediments is constant, then the radon concentration in porewater beneath the seafloor will increase with depth, up to a depth where secular equilibrium is reached. The depletion of radon at shallow depths may therefore be used to derive the exchange between porewater and overlying waters.

## 2.1. Radon deficit model

The most commonly applied approach to estimate the flux of radon across the sediment-water interface is based on the deficit of porewater radon relative to radon concentrations at secular equilibrium, i.e. the radon concentration that would occur without solute exchange (e.g. Martin et al., 2007; Cable and Martin, 2008). This deficit must be equal to the total net flux of radon into overlying waters. More generally, and assuming 1D flow, the net radon flux at depth  $z'$  can be written as:

$$J_{Rn} = \int_{z'}^{\infty} (\gamma - \lambda c) \theta dz \quad (2)$$

where  $\theta$  is sediment porosity and  $c$  is the radon concentration at each depth  $z$ . Integrating from  $z'=0$  to  $\infty$  gives the net radon flux across the sediment-water interface.



Aside from the sediment porosity and the radon concentrations at the different depths, this approach requires knowledge of the production rate in sediments ( $\gamma$ ), which can either be calculated from slurry-equilibration experiments (Colbert and Hammond, 2008) or derived from deep porewater radon concentrations, which are assumed to be unaffected by porewater exchange (Cable and Martin, 2008).

## **2.2. Advection cycling model**

Exchange of solutes between rivers and lakes and underlying porewaters has frequently been simulated using mass balance models that represent the porewater zone as a perfectly mixed reservoir of constant depth (Bencala, 1983; Cook et al., 2006; Gooseff et al., 2003; Stieglitz et al., 2013). Multiple reservoirs sometimes have been used, although such reservoirs usually operate in parallel so that each reservoir is directly connected to the surface water (Choi et al., 2000). In most cases, the focus of these studies has been on reproducing exchange fluxes rather than porewater concentrations, although Lamontagne and Cook (2007) used a single reservoir model to relate water fluxes to radon concentrations within porewater of a river hyporheic zone that was assumed to be perfectly mixed.

As the objective of the present study is to interpret vertical radon profiles rather than mean porewater concentrations, we consider a series of vertically-stacked reservoirs rather than a single reservoir. This type of model, sometimes referred to as a compartmental mixing model, has been widely applied in soil water and groundwater transport simulations (Adar et al., 1988; Harrington et al., 1999; Kirk and Campana, 1990), but not previously in the context of porewater exchange.

Water and solutes are assumed to be perfectly mixed within each cell, and only advective fluxes between the cells are considered. Thus hydrodynamic dispersion is simulated implicitly through mixing within the cells rather than explicitly included in the governing equations. Compartmental mixing cell models have been shown to produce similar results to advective-dispersive models provided that advection is the dominant transport process and that the size of the mixing cell is appropriately chosen (Xu et al., 2007).

The model presented here represents a two-dimensional recirculation cell, in which flow reverses periodically. This type of flow system might be produced by the passage of waves across the water surface, with surface water moving into the sediments beneath the wave peaks (high pressure), and exiting beneath the wave troughs (low pressure). We assume that the vertical scale of the recirculation cell is much greater than the horizontal scale, so that horizontal travel times are negligible (this is discussed further below). The two-dimensional model thus collapses into two one-dimensional profiles that exchange water and solutes. The fully saturated porewater zone is discretized into a number of layers (cells), each of which is assumed to be perfectly mixed. Each cell is assumed to continually exchange water with the cells immediately above and below it (the uppermost cell representing the surface water), and also to exchange water with the cell at the equivalent depth in the adjacent profile (Figure 1). The downward flow in one profile is thus balanced by the upward flow in the second profile, and the change in vertical water flux with depth determines the exchange flux between the two profiles. For each profile, we solve

$$\frac{\partial(\theta c)}{\partial t} = -\frac{\partial}{\partial z}(q_v c) + \gamma\theta - \lambda\theta c - S \quad (3)$$

where  $q_v$  is vertical water flux (which is a function of depth), and  $S$  is a mass flux term that allows for flow between the upwelling and downwelling profiles.  $q_v$  is defined so that downward fluxes are positive and upward fluxes are negative. Dispersion is not explicitly simulated, but it is implicitly simulated based on the size of the mixing cells. The relationship between cell size ( $\Delta z$ ) and implicit dispersivity ( $\alpha$ ) is given by (Kirchner, 1998; Shanahan and Harleman, 1984):

$$\alpha = \frac{\Delta z}{2} \quad (4)$$

The water flux decreases with depth in the downwelling profile, so that for the first phase of the recirculation cycle, water moves downward on the left hand side and upwards on the right hand side, as shown in Figure 1. Horizontal flows occur from left to right, and are given by

$$q_h(i) = q_v(i - 1) - q_v(i) \quad (5)$$

where  $q_h(i)$  is the horizontal water flux at the depth represented by cell  $i$ ,  $q_v(i-1)$  is the downward water flux into cell  $i$  from the overlying cell and  $q_v(i)$  is the vertical water flux from cell  $i$  into the underlying cell. Note that both  $q_v$  and  $q_h$  are volumetric fluxes per square area of the sediment surface, and so  $q_h$  is not a Darcy velocity in the traditional sense. Note also that the term  $S$  in Equation 3 can be obtained by multiplying  $q_h(i)$  by the concentration in cell  $i$  of the downwelling profile.  $S$  will therefore be positive for downwelling profiles ( $-S$  is negative, representing a mass loss) and negative for upwelling profiles ( $-S$  is positive, representing a mass gain).

The solute mass balance equations are:

$$\Delta c^a(i) = \Delta t \left[ \frac{q_v^a(i-1)}{\theta} \left( \frac{c^a(i-1) - c^a(i)}{\Delta z} \right) + \gamma - \lambda c^a(i) \right] \quad (6)$$

211

$$\Delta c^b(i) = \Delta t \left[ \frac{-q_v^b(i)}{\theta} \left( \frac{c^b(i+1) - c^b(i)}{\Delta z} \right) + \frac{(q_v^b(i) - q_v^b(i-1))}{\theta} \left( \frac{c^a(i) - c^b(i)}{\Delta z} \right) + \gamma - \lambda c^b(i) \right] \quad (7)$$

213 where  $c$  is concentration,  $\Delta c$  is change in concentration,  $q_v$  and  $q_h$  are the vertical  
 214 and horizontal water fluxes,  $\Delta z$  is the vertical cell dimension,  $\Delta t$  is the temporal  
 215 discretization and the superscripts  $a$  and  $b$  denote the fluxes and concentrations in  
 216 the downward-flow and upward-flow cells respectively (symmetry of the  
 217 recirculation cell requires that  $q_v^a = -q_v^b$ ).

218 After a period of time ( $t_r/2$ ), the flow reverses, and so the directions of all arrows  
 219 shown in Figure 1 reverse. The calculations are then repeated, with superscripts  $a$   
 220 and  $b$  switched in Equations 6 and 7. This cycle is then repeated. Because the  
 221 direction of flux changes during the recirculation cycle, the mean upward and  
 222 downward water flux at each depth across a complete recirculation cycle ( $\overline{q_v}$ ) is  
 223 calculated by dividing  $q_v$  by two (i.e.  $q_v = 2\overline{q_v}$ ).

224 The upper boundary condition is constant concentration ( $c=c_0$ ). The lower  
 225 boundary is the concentration of radon in equilibrium with the rate of production  
 226 ( $c=\gamma/\lambda$ ), although in practice the lower boundary is set to be sufficiently deep so  
 227 that it does not affect simulation results. The key parameters in the model are the  
 228 concentration in the overlying surface water ( $c_0$ ), the sediment characteristics ( $\theta$ ,  
 229  $\gamma$ ), and the characteristics of the recirculation, which include the time for a  
 230 completed cycle ( $t_r$ ), and the velocity profile  $q_v(i)$ .

231 It should be noted that as the period of the recirculation ( $t_r$ ) becomes small, radon  
 232 concentrations derived from the oscillating flow model are equivalent to those  
 233 observed in a simple 1D steady state model with flow occurring simultaneously in  
 234 both directions. This is given by

$$235 \quad \frac{|q_v(i-1)|}{\theta} \left( \frac{c(i-1)-c(i)}{\Delta z} \right) - \frac{|q_v(i)|}{\theta} \left( \frac{c(i)-c(i+1)}{\Delta z} \right) + \gamma - \lambda c = 0 \quad (8)$$

236

### 237 2.3. Dispersion model

238 For small recirculation times ( $t_r$ ), the solute profiles in a reversing flow field can  
 239 also be expressed in terms of an enhanced dispersion coefficient, rather than by  
 240 directly simulating advection (Qian et al., 2009). The flux of radon at any depth can  
 241 be expressed

$$242 \quad J_{Rn} = D_e \theta \frac{\partial c}{\partial z} \quad (9)$$

243 where  $J_{Rn}$  is the radon flux and  $D_e$  is the enhanced dispersion coefficient ( $\text{m}^2 \text{d}^{-1}$ ).  
 244 Qian et al. (2009) developed a two-dimensional hydraulic model to examine the  
 245 effect of wave action on porewater solute profiles, and showed that the value of the  
 246 enhanced dispersion coefficient decreased exponentially with depth, and could be  
 247 approximated by

$$248 \quad D_e(z) = \frac{5\alpha Ka}{L\theta} \exp \left[ \frac{-6.15z}{L} \right] \quad (10)$$

249 where  $K$  is the sediment hydraulic conductivity,  $a$  is the half-wave amplitude and  $L$   
 250 is the wavelength (m).

251

### 252 3. METHODS

#### 253 3.1. Radon sampling: study site, sampling and analyses

254 Porewater profiles for radon analysis were collected at La Palme lagoon, located  
255 on the western French Mediterranean coastline. La Palme is a small (5 km<sup>2</sup> surface  
256 area), shallow lagoon, with mean and maximum water depths of 0.6 and 1.5 m  
257 respectively. It is connected with the Mediterranean Sea through a small opening  
258 in the coastal sand spit and it receives continuous fresh groundwater inputs  
259 mainly from a regional karst aquifer (Stieglitz et al., 2013). The internal mixing of  
260 the lagoon and its exchange with coastal waters is driven primarily by the strong  
261 north-westerly winds characteristic of the region (regularly exceeding 10 m/s).  
262 Tidal forcing plays a minor role on the hydrodynamic functioning of this lagoon  
263 (tidal variations in the Mediterranean Sea are usually less than 0.4 m and exchange  
264 between the lagoon and the sea is highly restricted). Most of the lagoon is covered  
265 by coarse-grained highly-permeable sediments. A recent study conducted by  
266 Stieglitz et al. (2013) revealed that wind-driven horizontal pressure gradients at  
267 the sediment-water interface produce the recirculation of large amounts of lagoon  
268 water through surface sediments. Indeed, they estimated that the equivalent of the  
269 volume of the entire lagoon recirculates through the sediments every 25 days. La  
270 Palme is thus an ideal site to evaluate the exchange of porewater across the  
271 sediment-water interface by using radon porewater profiles.

272 Porewater samples for radon analysis were collected at 2 different locations (Pz1  
273 at 42.9741°S, 3.0163°E and Pz2 at 42.9391°S, 3.0248°E) using a drive point

piezometer. At each location, samples were collected at depths of 0.05 (only Pz1), 0.10, 0.15, 0.20, 0.30, 0.50, 0.80 and 1.30 (only Pz2) m below the sediment – water interface. 10-mL porewater samples were collected using a gas-tight syringe coupled to the piezometer tubing (minimizing water-air contact) and transferred to 20-mL vials prefilled with a 10-mL high-efficiency mineral oil scintillation cocktail (Cable and Martin, 2008). Concentrations of radon were analyzed by liquid scintillation counting on a Quantulus 1220 with alpha-beta discrimination counting (background of 0.2-0.4 cpm; efficiency of 1.6-2.2, depending on the quenching factor of the sample). Surface water samples (2 L) were collected using a small submersible pump to minimize the gas loss and analyzed using the radon-in-air monitor RAD7 coupled to an extraction system. Samples were decay-corrected to the time of collection.

### 3.2. Model

The advective compartmental mixing cell model was programmed into Fortran 95. We used a cell size of  $\Delta z = 0.01$  m, which is equivalent to an implicit dispersivity of  $\alpha = 0.005$  m. The latter is consistent with a flow length of approximately 0.5 m (Gelhar et al., 1992). (This is the approximate depth of radon depletion apparent in the measured profiles, and hence also the apparent depth of recirculation.) Temporal discretization was  $10^{-7}$  days (0.0086 seconds), but identical results were obtained using smaller values. The model was run for at least 20 days, so that radon concentrations reach dynamic equilibrium and are unaffected by initial

conditions. Sediment porosity and radon production rate were both assumed to be constant (i.e., do not vary with depth).

#### 4. RESULTS

The two porewater profiles Pz1 and Pz2 showed radon concentrations increasing rapidly with depth up to around 30 - 50 cm depth. Radon concentrations below these depths were relatively constant, with maximum concentrations of approximately 5000 Bq m<sup>-3</sup> and 2500 Bq m<sup>-3</sup> for Pz1 and Pz2 respectively, which likely reflect concentrations reaching secular equilibrium. The lower value observed at 0.8 m for PZ1 is likely due to analytical or sampling uncertainty or may reflect a spatial variation in the radon production rate. The difference between the measured radon concentrations at shallow depth and these equilibrium concentrations sustained by the production rate indicates that there is a significant exchange of radon between porewaters and overlying waters.

##### 4.1. Deficit model

The radon production rates for the two sites are approximately 900 and 450 Bq m<sup>3</sup> d<sup>-1</sup> for Pz1 and Pz2, respectively, as derived from the maximum concentrations measured at each site (assuming constant production rates over depth). By applying equation 2 and using a porosity ( $\theta$ ) of 0.4, we estimated a total net flux of radon into overlying waters of 58 and 38 Bq m<sup>-2</sup> d<sup>-1</sup> for Pz1 and Pz2, respectively.



The radon flux estimated here refers to the total net loss of radon from sediments into surface waters and thus includes also the flux of radon supplied by molecular diffusion. The net advective-dispersive flux of radon from sediments can be calculated as the difference between the estimated total flux and the diffusive flux, which can be approximated using Fick's First Law and radon diffusion coefficients corrected for both temperature and tortuosity ( $\sim 1 \times 10^{-4} \text{ m}^2 \text{ d}^{-1}$ ; (Boudreau, 1997; Li and Gregory, 1974)). By using the radon gradients over depth measured in the shallowest porewaters (above 20 cm), the diffusive flux is on the order of  $10^{-1} \text{ Bq m}^{-2} \text{ d}^{-1}$ , which is 2 orders of magnitude lower than the total radon flux. Therefore, radon diffusive fluxes in the studied profiles can be neglected and the total  $^{222}\text{Rn}$  flux can be attributed to advective-dispersive fluxes.

#### 4.2. Advection cycling model

Observed radon profiles were simulated using constant surface water concentrations of 30 (Pz1) and 90 (Pz2)  $\text{Bq m}^{-3}$ , and  $q_v(z)$  and  $t_r$  were varied in a trial-and-error fashion, until good fits with observed profiles were obtained. It was found that best-fits to radon profiles were produced with water fluxes that decreased exponentially with depth, and so this was adopted in all simulations.

The simulated radon profile for Pz1 is shown in Figure 3, and closely matches the observed profile to a depth of 0.5 m. This simulation uses a recirculation time of  $t_r/2 = 10^{-5}$  days (0.864 seconds), although identical profiles are produced for recirculation times on the order of  $<10^{-2}$  days. For longer times, different upwelling and downwelling profiles are obtained (Figure 4). In this case, the

340 profile that has just completed its downwelling phase (orange line in Figure 4) has  
341 much lower concentrations than the upwelling profile (blue line). This essentially  
342 represents differences that would be observed depending on the sampling time in  
343 relation to the phase of the cycling.

344 The mean upwelling or downwelling water flux across the sediment – water  
345 interface in Pz1 is  $\overline{q_v}(0) = 0.25 \text{ m d}^{-1}$  ( $3 \times 10^{-6} \text{ m s}^{-1}$ ). The radon flux from the  
346 surface into the underlying cell is  $\overline{q_v}(0)c(0)$ , which is equal to  $7.5 \text{ Bq m}^{-2} \text{ d}^{-1}$ . The  
347 radon concentration in this second cell (immediately below the surface water  
348 layer) is  $240 \text{ Bq m}^{-3}$ , and so there is also an upward radon flux into the surface  
349 water of  $\overline{q_v}(0)c(1) = 61 \text{ Bq m}^{-2} \text{ d}^{-1}$ . The net radon flux into the surface water is  
350 therefore  $\overline{q_v}(0)[c(1) - c(0)] = 54 \text{ Bq m}^{-2} \text{ d}^{-1}$ . This is essentially the same as the  
351 radon flux for Pz1 calculated from Equation 2 ( $58 \text{ Bq m}^{-2} \text{ d}^{-1}$ ).

352 Figure 5 depicts the sensitivity of the radon profiles to variations in surface water–  
353 sediment exchange flux ( $\overline{q_v}(0)$ ), attenuation of water flux with depth ( $k$ ), and  
354 implicit dispersivity. The latter is investigated by varying the cell dimensions in the  
355 model. The simulations show, that if dispersivity is fixed, then if sampled with  
356 sufficient resolution, the radon profiles allow unique determination of  $\overline{q_v}(0)$  and  $k$ ,  
357 as these parameters affect the profiles in different ways. In particular, varying  $k$   
358 only affects the deeper parts of the profile. However, dispersivity and velocity  
359 affect radon profiles similarly, indicating model non-uniqueness. Assuming a value  
360 for dispersivity is therefore essential for estimating the water flux. In our case,  
361 fitting the radon profiles with dispersivities of  $0.0025$  and  $0.01 \text{ m}^2 \text{ d}^{-1}$  (cell sizes of  
362  $0.005$  and  $0.02 \text{ m}$ , respectively) would have resulted in mean surface water fluxes  
363 of  $0.5$  and  $0.125 \text{ m d}^{-1}$  respectively.

Additionally, the uncertainties associated with radon analytical measurements allow for accommodating different radon profiles (for a fixed dispersivity), and thus yield different simulated  $k$  and  $\overline{q_v}$  values (Figure 6). Thus, within the error bars of the radon measurements, surface water fluxes of  $\overline{q_v}(0) = 0.2 - 0.35 \text{ m d}^{-1}$  and flux attenuation rates of  $k = 0.5 - 15 \text{ m}^{-1}$  are possible. However, the lowest values of  $k$  are only possible if the two shallow radon values ( $< 0.15 \text{ m}$  depth) are towards the upper limit of the analytical uncertainty range, and the four deeper values ( $0.15 - 0.5 \text{ m}$ ) are towards the lower limit. The reverse applies for the higher values of  $k$ . This is extremely unlikely to be the case (probability of  $2^{-6}$ ; or less than 2%). Thus, the true uncertainty in  $k$  is much less than this, and probably closer to  $2.0 < k < 10 \text{ m}^{-1}$ . Values of  $k$  are most sensitive to radon concentrations at  $0.2 - 0.4 \text{ m}$  depth (Figure 5), and increasing the precision of these measurements would significantly improve the accuracy of the flux attenuation rate estimate.

Radon concentrations in profile Pz2 are reproduced using a production rate of  $\gamma = 450 \text{ Bq m}^{-3} \text{ d}^{-1}$ , and surface flux  $\overline{q_v}(0) = 0.18$ , but with other parameters identical ( $\theta = 0.4$ ;  $t_r/2 = 10^{-5} \text{ d}$  ( $0.86 \text{ s}$ ); with  $k = 5 \text{ m}^{-1}$ ).

#### 4.3. Dispersion Model

It is also possible to represent the observed radon data in terms of enhanced dispersion coefficients rather than explicitly considering advective fluxes. By equating Equations 2 and 9, we obtain

$$D_e \theta \left. \frac{\partial c}{\partial z} \right|_{z=z'} = \int_{z'}^{\infty} (\gamma - \lambda c) \theta dz \quad (11)$$

386 The dispersivity profile can therefore be obtained by approximating  $\frac{\partial c}{\partial z}$  using the  
387 observed changes in radon concentration between sampling depths and solving  
388 Equation 11.

389 The profile of enhanced dispersion coefficient calculated for profile Pz1 in this way  
390 is depicted in Figure 7a. The dispersion coefficient is related to the flux by  
391  $D_e = |v|\alpha$ , where  $v$  is the water velocity. As dispersion will occur under both  
392 upward and downward flow, it is therefore related to the recirculation flux  
393 according to

$$394 \quad D_e = \frac{q_v \alpha}{\theta} = \frac{2\bar{q}_v \alpha}{\theta} \quad (12)$$

395 where  $q_v$  is the upward or downward flux that occurs during respective upwelling  
396 and downwelling phases, and  $\bar{q}_v$  is the mean upward or downward flux averaged  
397 across the two phases. Thus we can calculate the average flux  $\bar{q}_v$  by rearranging  
398 equation 12. This is shown in Figure 7b, where it is also compared with the flux  
399 profile obtained from the advection cycling model. The surface flux calculated  
400 using Equations 11 and 12 is  $0.31 \text{ m d}^{-1}$ , which is similar to that calculated from the  
401 mixing cell model ( $0.25 \text{ m d}^{-1}$ ).

402

## 5. DISCUSSION

### 5.1. Model sensitivity

The results from both the advection cycling and the dispersion model are a function of the selected longitudinal dispersivity (expressed as cell size for the advection cycling model). Changing the selected dispersivity will result in proportional changes on the estimated water fluxes. Since dispersivity is a scale dependent parameter, longitudinal dispersivity ( $\alpha$ ) has been often related to the length of the flow path at the field scale (Gelhar et al., 1992; Neuman, 1990). Assuming a flow path length of 0.5 m, we would obtain an approximate dispersivity of 0.005 m, which is the value used in this study. The estimated dispersivity would largely depend on the length scale selected, which may be difficult to define when dealing with porewater fluxes. As an alternative approximation, Qian et al. (2008) suggested that pore-scale dispersivities in a sediment bed can be approximated by the average particle size. Since the sediment of the study site are mainly composed by coarse-grained particles, the mean particle size is likely ranging from half to a few millimeters, and the expected range of  $\alpha$  would therefore be on the order of 0.001 m. This gives a factor-of-five uncertainty in  $\alpha$ , and hence also in water flux.

Aside from longitudinal dispersivity, which is the single most important control on model results, the advection cycling model is also sensitive to other input parameters. Although exchange flux ( $\overline{q_v}$ ) is proportional to porosity ( $\theta$ ), this parameter is relatively easy to estimate in unconsolidated sediments, and it is not expected to vary in a wide range. The radon production rate ( $\gamma$ ) is constrained by the radon concentrations measured in deep porewaters, and so it is important for

samples to be collected from sufficient depth. In our example,  $\gamma$  was assumed to be constant with depth, although variations in  $\theta$  and  $\gamma$  with depth can be introduced to the model if they are expected to occur. In this case,  $\gamma$  would need to be derived from equilibration experiments, and previous studies have shown that these measurements can have significant uncertainties (Key et al., 1979; Berelson et al., 1982). Model-derived water fluxes in the uppermost layers of the sediment are relatively insensitive to the selected production rate ( $\gamma$ ), but  $\gamma$  exerts an important control on the water fluxes simulated for deeper layers, where radon concentrations approach secular equilibrium. Accurate estimation of water fluxes at depth, and the rate (and functional form) of flux attenuation, is also highly sensitive to the analytical precision of radon measurements. We assumed that  $\overline{q_v}$  decreased exponentially with depth, as this is commonly assumed (e.g. Qian et al., 2009; Fram et al., 2014; Wilson et al., 2016) and because the exponential decrease produced the best fit to the data. Finally, the model is not very sensitive to variations in the time for each phase for the cycle when  $t_r$  is on the order of  $<10^{-2}$  days. For a given value of dispersivity, the uncertainty in all these input parameters and analytical precision result in an uncertainty in the water flux across the sediment – water interface ( $\overline{q_v}(0)$ ) of approximately a factor-of-two. Thus, as previously discussed, uncertainty in dispersivity dominates the uncertainty in surface water flux.

A limitation of the model is that it neglects horizontal travel times between downwelling and upwelling profiles. This may be reasonable for small-scale recirculation driven by wave action, as at any time upwelling and downwelling zones are likely to be separated by only a few centimetres. The assumption is less

likely to be valid for recirculation systems driven by processes operating over longer timescales, such as seiches or tides. For these, porewater concentrations (such as those shown in Figure 4) are likely to be underestimated, particularly for upwelling profiles.

## 5.2. Quantification of water exchange across the sediment-water interface

A number of previous studies have estimated net radon fluxes across the sediment–water interface by integrating the radon deficit in sediment porewater (Equation 2) (e.g. Cable and Martin, 2008; Martin et al., 2007). However, it is not straightforward to calculate the water flow from this data. The common approach to convert this radon flux into specific porewater discharge is by dividing the mass flux by the radon concentration in the shallowest porewater sample (Cable and Martin, 2008; Martin et al., 2007). Given that it is extremely difficult to collect porewater samples for radon analysis in the top centimeters of the sediment, the nearest-surface porewater samples are commonly collected at 0.05-0.1 m depth. In the case of the profiles collected in La Palme lagoon, the radon concentration in the uppermost porewater sample in PZ1, which was collected at a depth of 0.05 m, is 930 Bq m<sup>-3</sup>. Had we followed this approach to estimate the water exchange between porewater and overlying waters, we would have obtained a flow of 0.06 m d<sup>-1</sup>, which would underestimate the water flow derived from the advection cycling model ( $\overline{q_v}(0) = 0.25 \text{ m d}^{-1}$ ) by a factor of approximately 4 (or by a factor of 8 if the shallowest sample was collected at 0.1 m depth). The main difference

between these estimates is that the midpoint of the uppermost cell for the advection cycling model is at a depth of 0.005 m (with a simulated radon concentration of 240 Bq m<sup>-3</sup>), which is a depth virtually impossible to sample for radon analysis. Although the concentration difference between the uppermost cell and the surface water must be considered to estimate the water exchange rate, the surface water concentrations are often small when compared with porewater concentrations. Thus, whilst the commonly applied radon deficit approach allows estimating net radon fluxes across the sediment-water interface, water fluxes are more accurately quantified by modeling the radon distribution with depth. An alternative approach is to calculate the dispersion coefficient by dividing the net radon flux at each depth by the concentration gradient. Even if samples are collected from below 5 cm, the dispersion coefficient calculated in this way does not appear to greatly underestimate the dispersion coefficient at the sediment – water interface. This is because  $\frac{\partial c}{\partial z}$  varies more slowly with depth than does  $c$ . The water flux can then be estimated by dividing the surface dispersion coefficient by the dispersivity.

One of the advantages of the radon approach described here is that it allows the estimation of water fluxes as a function of depth. However, as discussed above, an accurate estimation of exchange fluxes at deeper depths would require more precise radon measurements (e.g. counting deep radon porewater samples for longer times), increased sampling resolution and/or radon equilibration experiments (Colbert and Hammond, 2007) to provide independent estimates of the radon production rate. While most of the porewater exchange studies have focused on the fluxes across the sediment-water interface, determining the



498 exchange fluxes at deeper depths is important for understanding the  
499 biogeochemical cycles in sediments. The penetration depth of reactants (e.g.  
500 oxygen), for example, will depend on the advective porewater velocities, as well as  
501 on the consumption/production rates in the sediment layers (Precht et al., 2004).

502 The approaches described in this paper are most appropriate in those systems  
503 where the driving force generating horizontal pressure gradients at the sediment-  
504 water interface oscillates in relatively short temporal scales (seconds to hours).  
505 Larger recirculation times (hours to days) would result in profiles that would  
506 significantly change depending on the sampling time in relation to the phase of the  
507 advection cycle (upwelling or downwelling) (Figure 4). The proposed approaches  
508 are thus best suited to quantifying porewater exchange fluxes produced by the  
509 undulating pressure at the seafloor generated by gravity waves interacting with  
510 relatively flat sediment surfaces. These models implicitly include the effects of  
511 interaction between wave-driven oscillatory flows and seabed morphology, which  
512 may significantly enhance water recirculation through sediments, particularly in  
513 areas with a water depth shallower than half the wavelength of the wave (Precht  
514 and Huettel, 2003). However, if bedforms (e.g. ripples) are stable on timescales of  
515 hours or longer, this might give rise to stable zones of up- and downwelling, and so  
516 profiles would vary depending on the area of collection. Note that bottom  
517 topography can change significantly over short time scales (e.g. ripple migration),  
518 particularly during strong periodic events (e.g. storms) or in areas affected by  
519 strong bottom currents (Precht et al., 2004; Savidge et al., 2008). Therefore, zones  
520 of upwelling and downwelling porewater in permeable sediments would also  
521 propagate with ripple migration (Precht et al., 2004). This intensive lateral shifting

of up- and downwelling zones within the sediment together with horizontal diffusion and dispersivity may contribute to homogenizing the vertical profiles. In a similar manner, areas of preferential resuspension or deposition of sediments could also release or trap significant volumes of porewaters (Santos et al., 2012), and thus would also result in significantly different porewater profiles depending on the sampling area. Collecting different radon porewater profiles in the same area should provide additional information on the temporal and spatial scales of the driving forces, by identifying the stability of upwelling and downwelling zones.

The advection cycling and the dispersion models represent thus a reliable method to characterize water exchange across the sediment-water interface driven by pressure gradients reversing at short temporal scales. Radon has advantages over other porewater tracer approaches, as it is more sensitive to low exchange fluxes than temperature (Briggs et al., 2014), and is simpler than dye injection approaches. Other methods commonly applied to quantify porewater exchange are not well suited to the estimation of fluxes with such short residence times. In situ seepage meters may alter fluxes above and below the sediment interface due to the presence of the instrument. This might not be significant for fluxes driven by longer-scale pressure changes, but is likely to be important for processes operating on very short time-scales. Tracer mass balances in overlying waters require estimation of the concentration of exchanging water to convert the tracer mass balance to a water mass balance. The appropriate end-member concentration will depend on the hydrodynamic dispersivity for this method, as it does for porewater tracer methods. However, mass balances in overlying waters will have additional uncertainties due to the need to also define other components of the mass balance.



and solute fluxes across the sediment-water interface in different wind (wave) conditions.

Considering that the water depths of La Palme lagoon usually range from 0.3 to 1.5 m, the figures estimated in this study would imply that the entire lagoon volume would recirculate through sediments every few days. These recirculation fluxes may therefore have important implications for the functioning of this coastal lagoon, since they may enhance the exchange of oxygen, solutes and particle-associated compounds between sediment and the overlying water column (Anschutz et al., 2009; Huettel and Rusch, 2000; Huettel et al., 1996; Jahnke et al., 2005). Accurately evaluating these recirculation fluxes is therefore required to understand the role that this process may play on the biogeochemical cycles of lagoon water and sediments.

#### 5.4. Extension to Other Radionuclide Tracers

Within the last few years, a number of studies have used the  $^{224}\text{Ra}/^{228}\text{Th}$  ratio in coastal sediments to calculate the  $^{224}\text{Ra}$  flux and the corresponding water flux across the sediment-water interface (Cai et al., 2012, 2014, 2015). Unlike  $^{222}\text{Rn}$ ,  $^{224}\text{Ra}$  and  $^{228}\text{Th}$  will partition onto the solid phase, and so the  $^{224}\text{Ra}$  deficit must be calculated from the total exchangeable  $^{224}\text{Ra}$  and  $^{228}\text{Th}$  activities, rather than from  $^{224}\text{Ra}$  and  $^{228}\text{Th}$  dissolved in pore water. However, calculation of water fluxes from  $^{224}\text{Ra}$  and  $^{228}\text{Th}$  profiles also requires information on  $^{224}\text{Ra}$  activities in the dissolved phases, as only this component will be transported with the water. In addition, the partitioning of  $^{224}\text{Ra}$  between water and solid phases is largely

dependent on the chemical composition of porewaters (e.g. salinity, pH, temperature, redox potential) and the characteristics of the sediments (e.g. grain size, fraction of exchangeable sites, content of iron and manganese), requiring an appropriate characterization of the  $^{224}\text{Ra}$  solid-solution partitioning coefficients (Beck and Cochran, 2013; Gonner et al., 2008). Finally, it should be noted that partitioning to the solid phase means that depletion of  $^{224}\text{Ra}$  in pore water will be much shallower than for  $^{222}\text{Rn}$  (as depleted  $^{224}\text{Ra}$  in pore water will be replaced by  $^{224}\text{Ra}$  released from the sorbed phase), and so this requires much finer resolution sampling.

Recent studies using the  $^{224}\text{Ra}/^{228}\text{Th}$  method have been mainly in finer grained sediments than those that have used the  $^{222}\text{Rn}$  method, and in these environments diffusion often forms a significant component of the tracer flux. The advection cycling model presented in this paper is less amenable to situations in which molecular diffusion is a significant component of the tracer flux. However, notwithstanding the above limitations, in advection-dominated systems, numerical models similar to that presented in this paper could potentially be used to estimate the variation in water flux with depth based on measured  $^{224}\text{Ra}$  profiles.

## 6. CONCLUSIONS

We simulated radon porewater profiles using an advective cycling numerical model to improve estimates of the flux of radon and water between sediments and overlying waters. This model is based on a series of radon mass balances in vertically-stacked reservoirs and where radon profiles are governed by advective

614 porewater fluxes that reverse periodically. The model allows estimation of water  
615 fluxes at different depths, which may provide some insights on the overall  
616 penetration depth of recirculation processes and the biogeochemical cycling in  
617 sediments. A simpler approach, based on the estimation of dispersion coefficients  
618 from the radon concentration gradient with depth, can also provide reasonable  
619 estimates of the advective water flux.

620 The proposed approaches are well suited to evaluate porewater fluxes driven by  
621 pressure gradients reversing at short temporal scales (up to hours), such as those  
622 produced by waves and semidiurnal tides, and in areas with no permanent  
623 bedforms that create preferential flow cells. Other methods commonly applied to  
624 quantify benthic fluxes (e.g. tracer mass balance in overlying waters, seepage  
625 meters) are not well suited to the estimation of fluxes with such short time-scales.

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820 Universitat Autònoma de Barcelona for radon sample analyses.

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## 9. FIGURE LEGENDS

Figure 1. Schematic representation of advective mixing cell model. Arrows denote flow directions during the first phase the recirculation cycle, in which flows are downwards on the LHS and upward on the RHS. (The flow direction is reversed during the second phase of the cycle.) Vertical water fluxes into and out of cell  $i$  are  $q_v(i-1)$  and  $q_v(i)$ , where  $i = 1, \dots, n$ , where  $n+1$  is the number of cells in the vertical dimension. The horizontal flux between downwelling and upwelling profiles is denoted  $q_h(i)$ . Concentrations in cell  $i$  are  $c_a(i)$  and  $c_b(i)$  on LHS and RHS respectively.  $c(0)$  is the concentration in surface water.

Figure 2. Observed radon profiles at two different locations within La Palme Lagoon. The error bars represent the analytical uncertainties ( $1\sigma$ ) for radon (liquid scintillation counting).

Figure 3. (a) Mean vertical water velocity (upward or downward), as a function of depth, and resulting radon concentration profile for  $t_r/2 = 10^{-5}$  d (0.86 s) for Pz1. The mean vertical water velocity is described by an exponential decrease in depth, according to  $\overline{q_v}(z) = \overline{q_v}(0)e^{-kz}$ . The best-fit to the data is produced with  $k = 5 \text{ m}^{-1}$  and  $\overline{q_v}(0) = 0.25 \text{ m d}^{-1}$ . Other parameters are  $\gamma = 900 \text{ Bq m}^{-3} \text{ d}^{-1}$  and  $\theta = 0.4$ .

Figure 4. Sensitivity of radon profiles to recirculation times ( $t_r$ ). Upwelling and downwelling profiles following complete cycles of (a)  $t_r = 0.2$  days (reversing every 0.1 d), (b)  $t_r = 2$  days and (c)  $t_r = 40$  days are represented. Note that the results obtained from a recirculation time of 40 days would be equivalent to those

obtained for larger cycles. Note also that the inflection point in the downwelling profile is similar to the estimated water flux ( $q_v(0)$ ) multiplied by the period of downwelling ( $t_r/2$ ). Other model parameters are the same as in Figure 3.

Figure 5. Sensitivity of radon profiles in Pz1 to (a) surface water – sediment exchange flux ( $\overline{q_v}(0)$ ), (b) attenuation of water flux with depth ( $k$ ), and (c) implicit dispersivity ( $\alpha$ ).

Figure 6: Simulated  $\overline{q_v}$  at different depths from 4 different radon profiles that could be accommodated within the analytical uncertainties of radon measurements of Pz1. Orange and grey dots represent lower and upper bounds ( $1\sigma$ ), respectively, of the radon measurements. Model parameters are  $q=0.35$ ,  $k=0.5$ ,  $\gamma=950$  (Black line),  $q=0.35$ ,  $k=2$ ,  $\gamma=1000$  (Blue),  $q=0.2$ ,  $k=0.5$ ,  $\gamma=820$  (Red),  $q=0.15$ ,  $k=15$ ,  $\gamma=820$  (Green). All simulations use  $\theta = 0.4$ .

Figure 7. a) Calculated enhanced dispersion coefficient as a function of depth for Pz1 and b) comparison of the water fluxes ( $\overline{q_v}$ ) at different depths derived from the dispersion approach (circles) and the advective cycling model (solid line).

867 Table 1. Measured values of porewater salinity and radon concentration.

868 Uncertainty for radon concentration represents one standard deviation.

Pz1			Pz2		
Depth (m)	Salinity (g/kg)	<sup>222</sup> Rn (Bq m <sup>-3</sup> )	Depth (m)	Salinity (g/kg)	<sup>222</sup> Rn (Bq m <sup>-3</sup> )
0	30.6	25 ± 10	0	37.2	89 ± 28
0.05	32.9	933 ± 152	0.10	38.1	1238 ± 185
0.10	30.8	1753 ± 218	0.15	38.1	1807 ± 216
0.15	31.3	2895 ± 280	0.20	38.3	1953 ± 231
0.20	30.8	3477 ± 327	0.30	38.6	2228 ± 247
0.30	30.7	4258 ± 393	0.50	38.9	1925 ± 372
0.50	37.7	4985 ± 496	0.80	51.6	2665 ± 276
0.80	84.5	4022 ± 478	0.13	62.2	2600 ± 275

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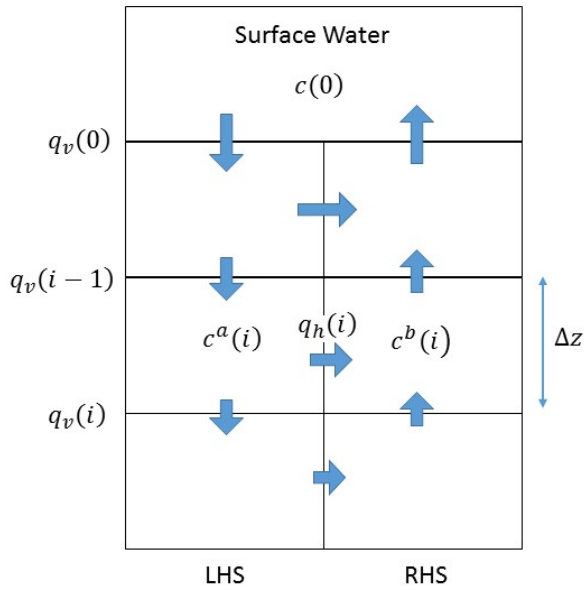


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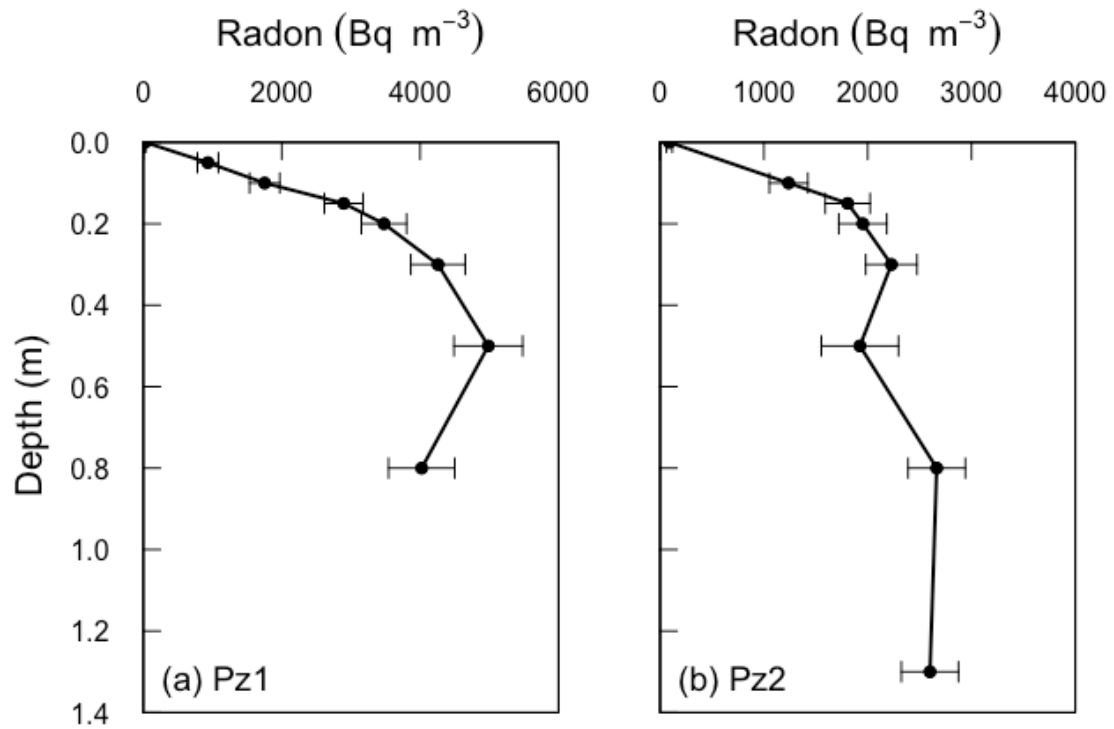


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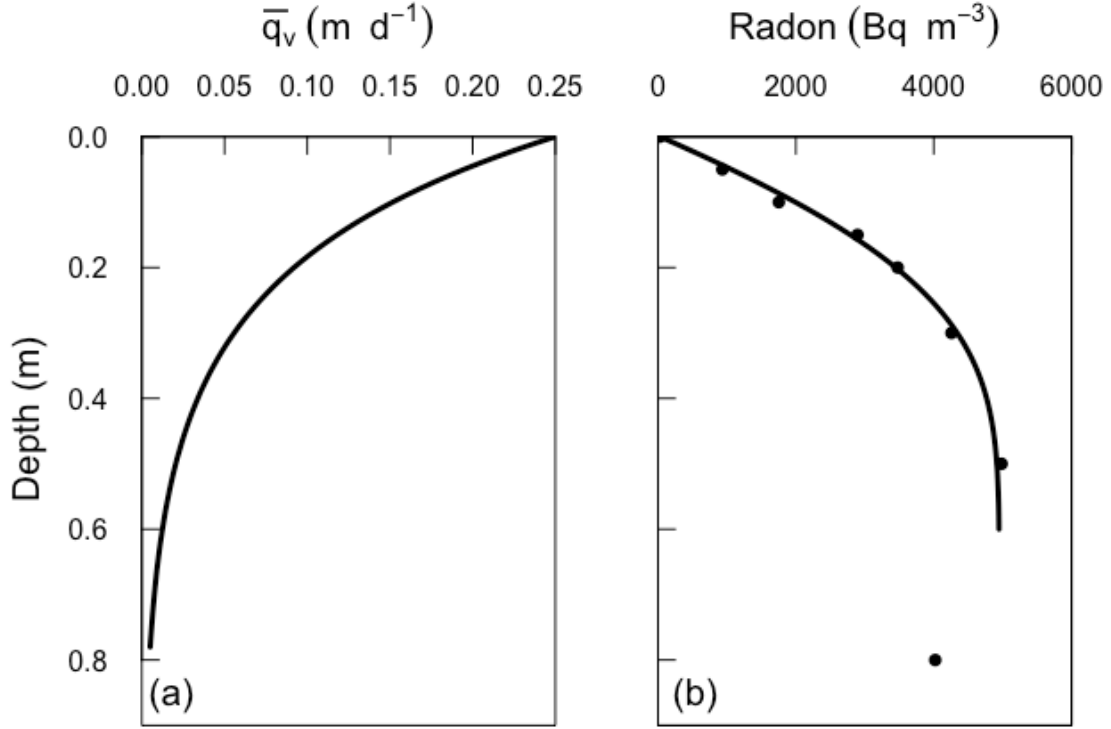


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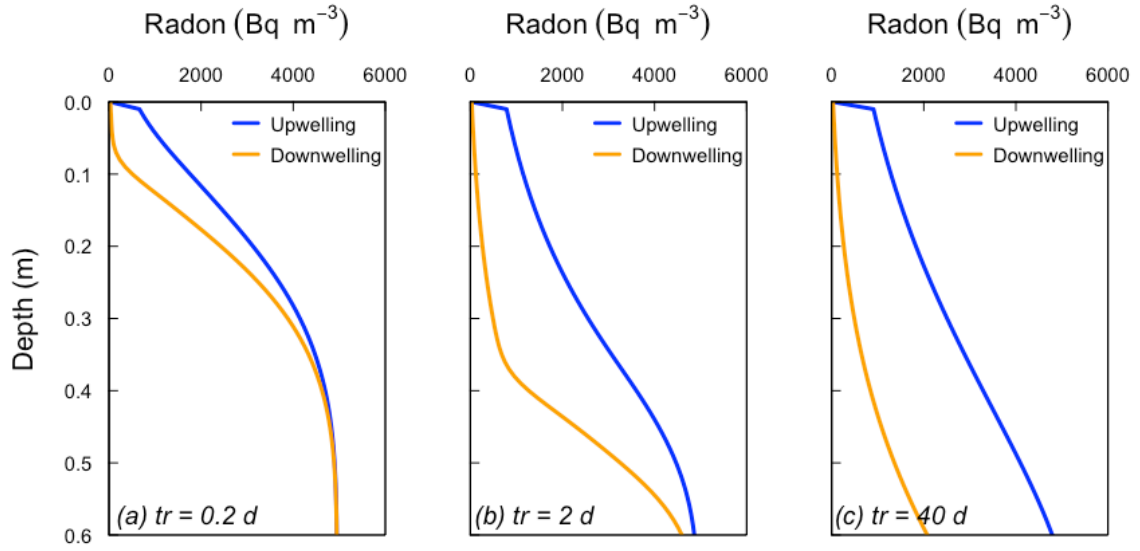
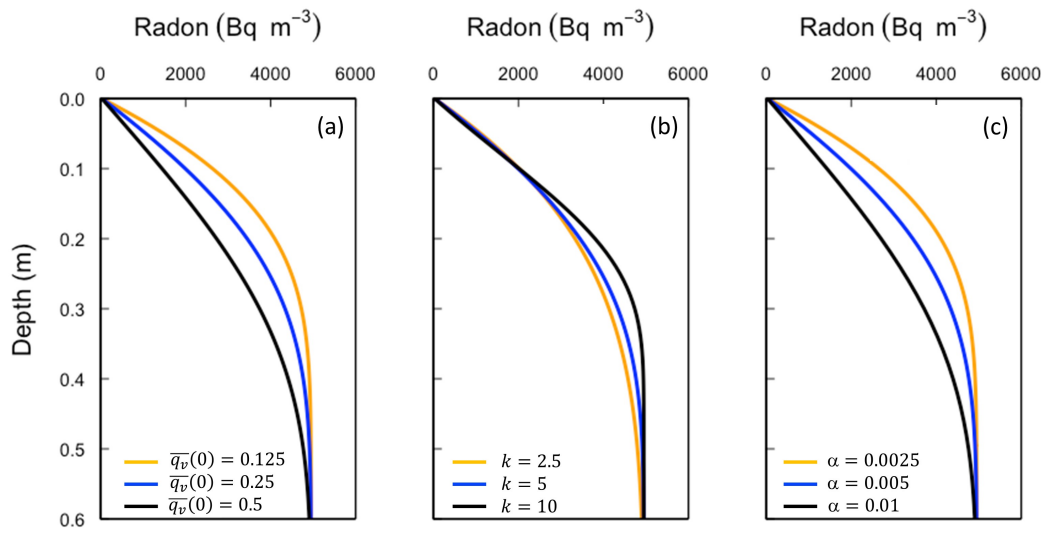


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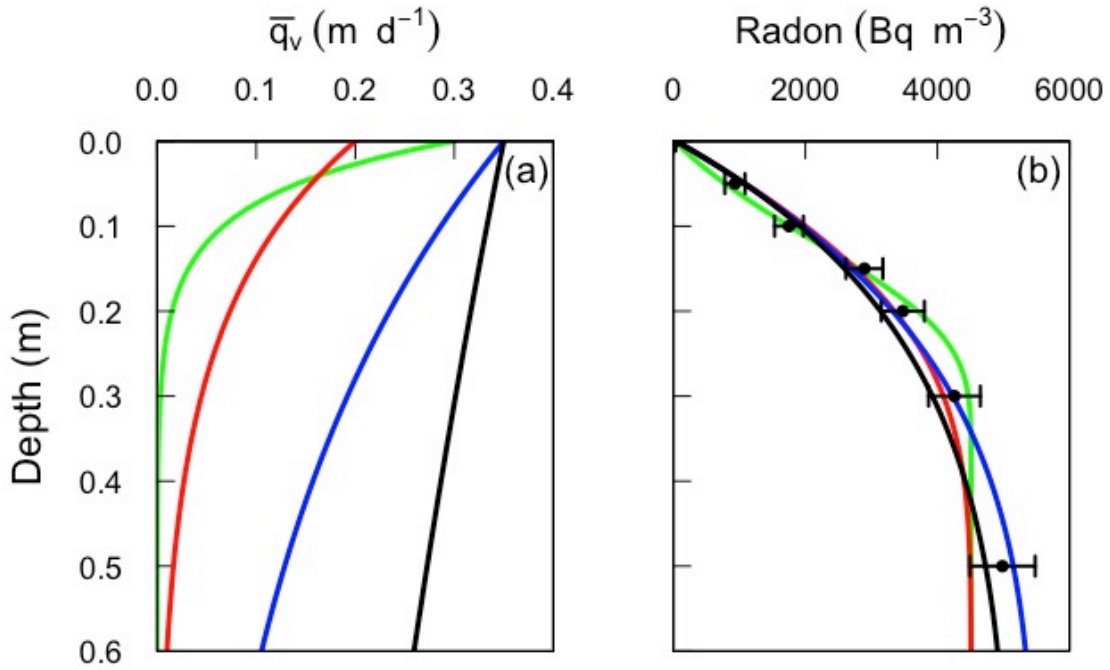


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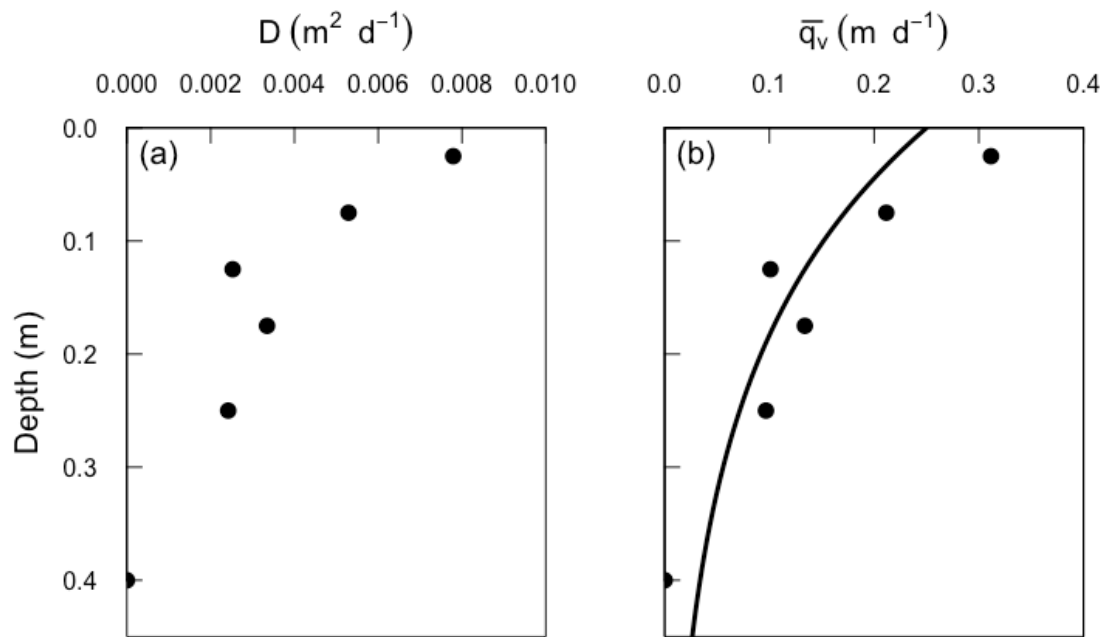


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