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Velocity skew controls the flushing of a tracer in a system of shallow bays with multiple inlets

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Key Points:

- Wind currents and differences in tidal phase and amplitude along the inner shelf are an effective mechanism for flushing tracer out of a coastal bay with multiple inlets
- Velocity skew at the inlets directly controls the tracer flushing time
- A simple, single-parameter exponential model is used to determine the decrease in tracer mass in the coastal bays

22 **Abstract**

23 The exchange of dissolved constituents between a shallow bay and the ocean is governed by
24 regular tidal fluxes as well as by wind generated storm surges and currents. These mechanisms
25 regulate the rate at which pollutants and nutrients are removed from the lagoons. Here we
26 determine the main hydrodynamic drivers controlling the removal of a conservative tracer from a
27 bay with multiple inlets. The transport of the tracer is simulated using the numerical model
28 Delft3D in a system of shallow lagoons along the coast of the Delmarva Peninsula, Virginia. The
29 tracer flushing time is evaluated using the Eulerian approach and the decay of the tracer
30 concentration in time is approximated with an exponential curve. We assess the influence of tidal
31 amplitude, local winds, and time of release of the tracer with respect to the tidal cycle on
32 flushing time. Results show that wind-driven fluxes are a prevailing factor controlling the tracer
33 transport and, therefore, the tracer concentration within the lagoons. Variations in tidal phase and
34 amplitude along the inner shelf also promote the flushing of the tracer out of the bays, while the
35 time of tracer release with respect to the tidal phase has been found to play a relatively negligible
36 role. The tracer flushing time is proportional to a velocity skew index, accounting for the
37 asymmetry of the ebb-flood velocities at the inlets, while the tidal prism has minimal effect on
38 flushing time. Our simulations revealed that the average flushing time of these bays is around 24-
39 27 days, decreasing to 21 days if favorable wind conditions exist. Finally, a simplified approach
40 is presented to compute the decay of tracer concentration in time as a function of hourly variable
41 wind characteristics as well as seasonal changes in meteorological conditions, without the need
42 of large scale simulations.

43 **1 Introduction**

44 Estuaries and bays are affected by the release of pollutants and nutrients that deteriorate water
45 quality and put ecosystems at risk (e.g. Cavalcante et al. [2012]). Therefore, there is a need to
46 determine the fate of these pollutants and quantify how long it takes for the pollutant to be
47 exported to the open ocean. The decay rate and the flushing time of contaminants injected into a
48 bay is regulated by coastal hydrodynamics [Braunschweig et al., 2003].

49 Fugate et al. [2006] estimated the spatial distribution of residence time in Hog Island Bay,
50 located at the center of the Virginia Coast Reserve (USA). Residence time in the lagoon was
51 found to be sensitive to variations in wind, tidal range, and tidal stage. Moreover, their results

revealed that in lagoons regulated by shallow friction the residence time has low values near the inlets, while it becomes higher near the mainland. Safak et al. [2015] found that the residence time near the inlets is influenced by the tidal phase at the moment of particles release, whereas it is affected by wind in the inner parts of the bay. Finally, the exchange of water volume in the same bays was derived from remote sensing images and in situ tracers by Allen et al. [2011]. They found that about half of the water volume is flushed out of the bay during each tide cycle, while the other half is left inside the system and it is exported at a slower rate.

However, a comprehensive and systematic analysis of all processes determining the flushing of a tracer is still lacking. The main purpose of this study is to understand the hydrodynamic conditions under which a tracer is flushed out at the fastest rate. Particular attention is devoted to water circulation induced by wind and variations in tidal phase and amplitude along the shelf, which are shown to be the most effective processes flushing this system.

We compute the flushing time of the bays in order to compare different physical drivers. Flushing time, water age, exposure time, and residence time are timescales that describe the exchange and transport of water and dissolved materials in a coastal sea [Rayson et al., 2016; Monsen et al., 2002; Takeoka, 1984]. The residence time quantifies the retention time of water within a defined control volume, and it is a frequently used metric to evaluate the transport of substances [Braunschweig et al., 2003] and to determine the ability of tides to remove pollutant from a semi-enclosed water body [Patgaonkar et al., 2012]. This parameter depends on tidal range, bathymetry, stratification, wind, and freshwater runoff and is considered to be a local measure that varies in time [Choi and Lee, 2004]. Another commonly used transport time scale is the flushing time which measures the average time that water and its constituents spend within a reservoir before being flushed out [Du et al., 2018]. Flushing time is a bulk parameter that quantifies the overall flushing of the system, and it only describes the renewal capability of a waterbody without taking into account the physical processes that regulate the exchange of material and their spatial distribution [Du et al., 2018; Rayson et al., 2016; Monsen et al., 2002].

The second goal of the paper is to determine whether tidal prism and variations in tidal phase and amplitude at the inlets control the flushing time of the bays. A higher tidal prism during spring tides or wind setup increases the exchange of water between bays and the ocean, likely exporting more tracer. Tidal phase differences along the coast lead to asymmetric fluxes, promoting tracer flushing. This is because a tidal phase lag between inlets produces surface

water gradients in the bays that drive residual circulation. This residual circulation decreases flushing time [Herrling and Winter, 2015].

The third goal of this paper is to reduce the description of the system dynamics to a single-parameter model, in order to facilitate the comparison of flushing time values under different external agents. A single-parameter model allows the fast determination of flushing time as a function of tidal and wind conditions without the need of expensive 2D simulations. Lagrangian (particle tracking) and Eulerian (tracer patch) methods are typically used to analyze transport time scales in estuaries. The Lagrangian approach explicitly simulates the trajectories of individual tracer particles and registers the time when they leave the domain providing spatial and temporal variations in residence time; the residence time is defined for each water parcel, and this approach is thus frequently employed to relate the time-varying position of particles to the point of their initial release [Rayson et al., 2016; Andutta et al., 2013; Braunschweig et al., 2003; Monsen et al., 2002]. On the other hand, the Eulerian approach might be more suitable for the residence time evaluation in the entire domain [Aikman and Lanerolle, 2004]. Eulerian formulations can be more easily translated into a single-coefficient parametrization to describe the decay of mass within a lagoon [Cucco et al., 2009]. Moreover, the Eulerian approach properly models the advection and diffusion processes [Deleersnijder et al., 2001]. Flushing time can also be estimated with the freshwater fraction method. This approach involves the calculation of the total volume of freshwater found in an estuary and it estimates the flushing time as the time taken to replace the existing freshwater in the estuary at the same rate of the freshwater input [Guo and Lordi, 2000; Williams, 1986; Dyer, 1973]. The application of the Lagrangian and Eulerian methods to study the hydrodynamics of lagoons and the relationship to particle residence time has been presented in several studies. Safak et al. [2015] used the Lagrangian particle tracking method to evaluate both the residence time of neutrally buoyant particles and their exchange between the bays of the Virginia Coast Reserve system. The residence time obtained with this approach was low, varying from a few hours to a few weeks depending on the release location; this result was probably due to a relatively short simulation period (two months), and to the fact that residence time estimates the time particles need to reach the boundary of the domain without reentering once they exit the inlets. Defne and Ganju [2014] employed a Lagrangian model for the analysis of particle trajectories and employed different methods for the evaluation of the residence time in an estuary. They found that, when the

simulation period is too short, the residence time given by ensemble averaging the residence time of each particle in the domain does not account for the particles that remain in the domain. Burwell et al. [2000] compared the Eulerian and Lagrangian methods in predicting the residence time of neutrally buoyant particles in a bay. The Eulerian method gave a good description of the residence time variations in the bay, but it was affected by diffusion which produced lower residence times compared to those produced by the Lagrangian method. In addition, the Lagrangian approach was better suited to describe the spatial distribution of residence time, but it presented limitations due to its sensitivity to the number of particles simulated in the model and the eddy diffusivity value, which is set a priori and it is assumed to remain constant in time. Delhez and Deleersnijder [2006] developed an Eulerian procedure that is equivalent to the Lagrangian method and allows estimation of residence time as a function of space and time. This adjoint method presents the advantage of delineating the fine spatial distribution of the particles in the domain in space and time, while reducing the difficulties associated with the Lagrangian representation of a spatially variable diffusivity.

Herein, we follow results from Cucco et al. [2009] who state the Eulerian approach is more suitable to investigate the long-term flushing of substances from a tidal embayment. We evaluate the fluxes of a conservative tracer within a system of shallow coastal bays along the Atlantic coast of the USA, in the Virginia Coast Reserve (VCR). The transport of the tracer and its flushing time are examined under different external forcing conditions, such as of tides and wind induced currents. The tidal flow and the circulation of water and tracer are simulated using the hydrodynamic model Delft3D-FLOW [Lesser et al., 2004; Roelvink and Van Banning, 1994]. The Eulerian method is applied to evaluate the time evolution of the spatially-averaged flushing time.

2 Study area

The Virginia Coast Reserve (VCR) is a system of shallow coastal bays that extends for 100 km along the Atlantic Coast of the southern portion of the Delmarva Peninsula, USA (Figure 1).

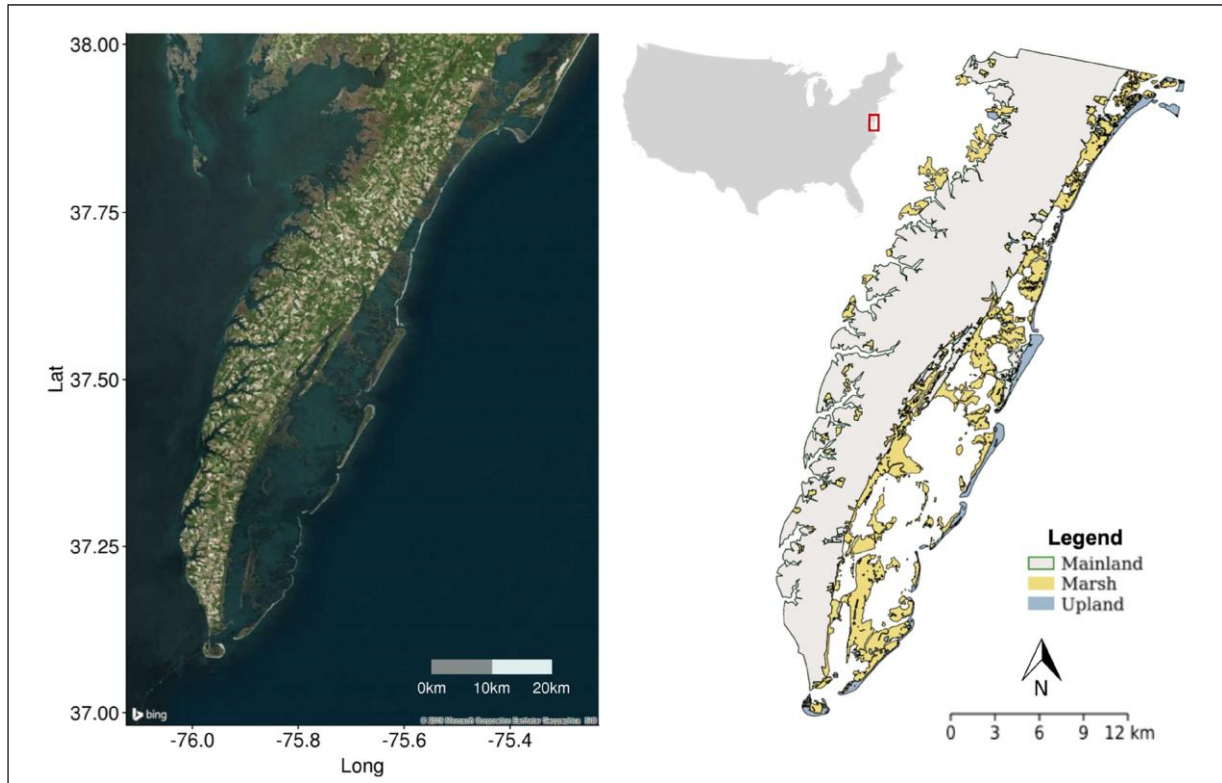


Figure 1. On the left, Virginia Coast Reserve along the Delmarva Peninsula. On the right, main land cover types found in the study area (<https://www.vcrlter.virginia.edu/home2/>).

The peninsula is bounded by Chesapeake Bay to the West and by the Atlantic Ocean to the East. VCR includes 14 barrier islands protecting bays of different size. The small watersheds found in the upland part of the study area supply a negligible amount of fluvial freshwater and sediment to the lagoons [Nardin et al., 2018; Wiberg et al, 2015; Fugate et al., 2006], and the water fluxes among different bays as well as the connection between bays and the ocean varies significantly within the system [Wiberg et al., 2015]. The shallow bays present an average depth of 1.0 m below mean sea level and are connected to the Atlantic Ocean by a network of channels which have a mean depth of 5 m and become more than 10 m deep at the inlets [Nardin et al., 2018, Mariotti et al., 2010]. Tides are semidiurnal and the mean tidal range is about 1.2 m [Safak et al., 2015]. The principal wind directions are from SSE-SSW and N-NE and the highest wind speeds are observed during the winter season [Fagherazzi and Wiberg, 2009].

Freshwater inputs to the VCR bays are minimal, and measurements of freshwater discharge in three representative streams (out of 54) indicate an average discharge of $0.03 \text{ m}^3/\text{s}$ in each stream

[Anderson et al., 2009], so that the estimated total freshwater flux to the bays is less than 0.3% the tidal prism. As a result, salinity variations in the bays are negligible. Salinity measurements along two transects that run from the inlet to the mainland in the southern part of the Delmarva Peninsula indicate average salinity of 31.0 ± 1.7 and 31.0 ± 1.1 ppt at the inlets, and 29.7 ± 2.7 and 29.0 ± 3.8 ppt at the landward shore of the bay [McGlathery et al. 2018]. Tidal exchange dominates transport processes in these bays and it has been the subject of previous studies [e.g. Wiberg et al., 2015; Fagherazzi and Wiberg, 2009; Fugate et al., 2006], but many mechanisms, such as those responsible for the flushing of pollutants and temporal variations in flushing time, remain unclear.

3 Methods

3.1 Model description

The hydrodynamic model Delft3D-FLOW [e.g. Lesser et al., 2004; Roelvink and Van Banning, 1994] was used to reproduce the hydrodynamics and transport of tracers in the VCR bays under tidal and wind forcing. The domain of the study area was delineated by a numerical grid of dimension 459x200, and cell size of 250x250m. Three open boundaries were defined along the East (Atlantic Ocean), South, and North sides of the domain (Figure 2). Given that VCR has no significant fluvial sources of freshwater and the salinity is overall uniform, the domain of the study area was composed of one vertical layer (2D), as in several other previous studies conducted over the same region [Nardin et al., 2018; Wiberg et al., 2015; Mariotti et al., 2010].

The water level was defined on all open boundaries by superimposing the various tidal harmonics with their corresponding phases and amplitudes. In a first set of simulations we used as boundary conditions the tidal harmonics obtained from the NOAA Wachapreague Station (8631044, Wachapreague, Virginia, USA). The amplitude and phase were adjusted to account for the propagation and amplification/dissipation of the tide in the lagoons and to properly reproduce the measured tides at the location of the Wachapreague station.

In a second set of simulations the phase and amplitude were assumed to vary along the open boundaries in order to account for spatial differences of the tidal signal in the Middle Atlantic Bight. These differences were small, but they can affect residual tidal currents. Specifically, the North and South boundaries (Figure 2) were divided into five segments, while the Atlantic

boundary was split into 10 segments. The amplitude and phase of the end points of each segment were derived from the ADCIRC Tidal Database for the Atlantic Ocean [Mukai et al., 2002], and then linearly interpolated by Delft3D so that the model could reproduce the regional tidal forcing.

The presence of vegetation was accounted by means of a spatially varying Chezy coefficient for bed roughness. The coefficient was set equal to $65 \text{ m}^{1/2}/\text{s}$ in the marshes and $45 \text{ m}^{1/2}/\text{s}$ in the rest of the grid [Safak et al. 2015; Wiberg et al. 2015]. The tracer introduced in the model was non-active and had no settling velocity. Sediment transport processes, as well as possible morphological changes were neglected. The initial tracer concentration was equal to one at every grid cell of the domain inside the lagoons and to zero for ocean and upland cells, with a clamped condition (zero tracer concentration) at the boundaries.

When wind is present, a spatially constant wind speed was applied to the entire domain. The wind exerts drag that was implemented as a surface shear stress term in the momentum equations. The magnitude of the shear stress was computed as $\tau = \rho C_d U_{10}^2$, where U_{10} is the wind speed 10 meter above the free surface, ρ the density of air, and C_d the wind drag coefficient, which was assumed not to vary with wind speed and to be equal to 0.00723.

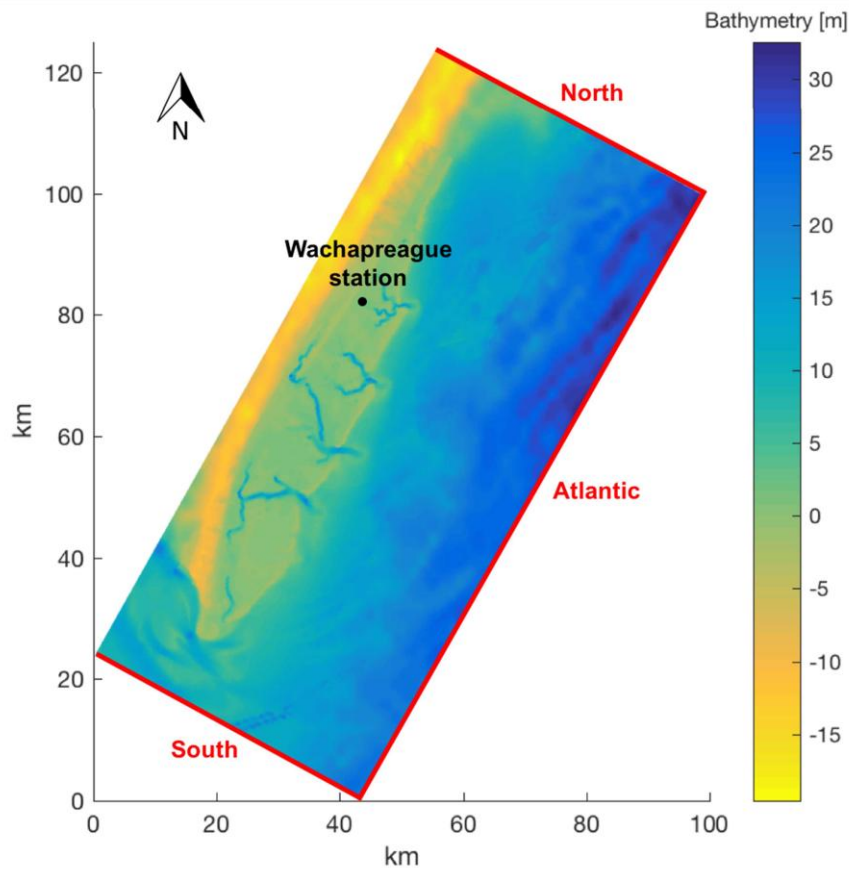


Figure 2. Bathymetry, boundaries of the domain of the study area, and location of NOAA Wachapreague station (8631044, Wachapreague, Virginia, USA).

3.2 Model validation

Following Cunge [2003], model calibration relies on careful adjustment of the tidal boundary conditions. This is consistent with the results of Abbott and Cunge [1975] who showed that adjusting the boundary conditions, rather than adjusting friction, was a more reliable approach to the calibration of estuarine models. The same approach was adopted by Wiberg et al. [2015] and Mariotti et al. [2010] for the VCR bays. The calibration of the boundary conditions was done by comparing simulated water levels with water levels measured at Wachapreague station, for the period from April 1st (time 00:00:00) to April 5th, 2015 (time 00:00:00). The model performance was evaluated with two statistics: the model efficiency, ME, and the root mean squared error, RMSE [Mariotti et al., 2010]. Both phase and amplitude at the boundaries were varied until reaching the maximum model efficiency. The values calculated were 0.97 for ME and 0.26 m for RMSE. The amplitude and phase at Wachapreague station before (observed values) and after the

calibration (calibrated values) are presented in Table S1.

To test the reliability of the model we computed the difference in amplitude and phase between the main harmonic constituents measured by NOAA at the Wachapreague station and the harmonic constituents simulated by Delft3D with the ADCIRC boundary conditions (Figure S1). The extraction of the harmonic constituents from the modeled water level was carried out using the T_tide Harmonic Analysis Toolbox [Pawlowicz et al., 2002]. The model validation gave excellent results, with amplitude differences of less than 2 cm and phase differences of less than 12 degrees (*VARYING TIDE* scenario, Figure S1). For comparison, we also plotted the difference in amplitude and phase using the modified Wachapreague tidal signal along all boundaries (*SPRING&NEAP* tide scenario, Figure S1). This model performed very well, although it led to slightly higher phase errors due to the neglect of phase variations along the Atlantic boundary.

3.3 Simulation scenarios

The performed simulations are meant to explore whether the flushing time and the decay of the tracer can be altered by one the following: i) differences in the time of release of the tracer with respect to the tidal cycle, ii) tidal amplitude, iii) variations in tidal phase and amplitude along the inner shelf, iv) shelf currents and varying tide along the boundaries, and v) wind conditions.

To study whether differences in the time of release affect the flushing of the tracer, a base scenario, called standard simulation (*STD*), was run based on the average values of tidal amplitude and frequency of the first six harmonics (M2, S2, N2, K1, M4, O1) at the open boundaries. The same simulation was run twice, once starting from ebb conditions (*STD EBB*) and once starting from flood conditions (*STD FLOOD*). For both cases, the initial water level was around mean sea level. In a second set of simulations the tracer was released in correspondence of high (*HIGH WL*) and low tide (*LOW WL*) at the open boundaries.

To determine whether tidal amplitude affects flushing time, we ran a simulation with the smallest amplitude associated to the N2 constituent, called *N2 AMPLITUDE*, and another simulation with tidal amplitude equal to the sum of the amplitudes of the harmonic constituents, combined to obtain the maximum tidal amplitude (*MAX AMPLITUDE*). Tidal amplitudes that are much different from the measured ones are unrealistic. However, we chose to run these hypothetical scenarios in order to expand the parameter space of the simulations, so that we can better understand the processes controlling the flushing of a tracer. Other two simulations were

run by accounting for the gradual alternation between the spring and neap tide, given by the combination of the first six harmonics (M2, S2, N2, K1, M4, O1). In one of these, the tracer was released during spring tide (*SPRING&NEAP*), and in the other the tracer was released during neap tide (*NEAP&SPRING*).

To account for the spatial variation of the amplitude and frequency of each harmonic constituent along the Virginia shelf we simulated the tracer mass decay under the effect of a realistic tidal forcing (*VARYING TIDE*). To determine the effect of currents on flushing time, we run a simulation (*CURRENT*) where we added a shelf current of 3 cm/s moving from northeast to southwest along the shore. The current velocity was imposed perpendicular to the north and south boundaries and was derived from field measurements of circulation over the Middle Atlantic Bight continental shelf [Lentz 2008].

To determine the effect of wind on flushing time, we run twenty-one simulations with a wind with constant speed and direction. Eight wind directions (every 45°, with north equal to 0°), and four values of wind speed, 5, 6, 10 and 12 m/s, were considered. The convention used to name each of the simulations that involve the wind effect is the following: the first part of the name is determined by the direction (e.g. the wind blowing from south is called *S*, and the wind coming from south-east is called *SE*), while the second part of the name is the wind speed in m/s (e.g. the wind blowing from north-west with a velocity of 5 m/s is labeled as *NW5*). *No Wind* indicates the scenario without wind. The characteristics of all the scenarios simulated in this study are described in Table 1. Wind constantly blowing from one direction for several weeks is unrealistic. Similarly to Scully et al. [2013], we also ran two simulations including wind data observed at the NOAA Wachapreague station during the winter and summer seasons of 2015 (*WINTER WIND* from December 21st, 2014 to March 20th, 2015, and *SUMMER WIND* from June 21st to September 23rd, 2015). The measured wind data (Figure 3) in the summer period included a block of missing values from September 1st to September 9th and these data were ignored in the evaluation of the decay parameter. No wind was prescribed during such period. All of the above simulations started at mean sea level during ebb, with the exceptions of the low and high water level ones.

278 **Table 1.** Parameters of the scenarios simulated with Delf3D.

279

Simulation ID	Initial time	Final time	Amplitude and Phase	Wind direction [°]	Wind speed [m/s]
STD EBB	01/04/2015	01/08/2015	Average amplitude*,	-	-
	03:18 (MSL, ebb)	03:18m	null phase		
STD FLOOD	01/04/2015	01/08/2015	Average amplitude*,	-	-
	21:19 (MSL, flood)	21:19m	null phase		
HIGH WL	01/04/2015	01/08/2015	Average amplitude*,	-	-
	00:12 (High tide, ebb)	00:12 m	null phase		
LOW WL	01/04/2015	01/08/2015	Average amplitude*,	-	-
	06:19 (Low tide, ebb)	06:19m	null phase		
MAX AMPLITUDE	01/04/2015	01/08/2015	Sum of amplitudes**,	-	-
	04:24 (MSL, ebb)	04:24m	null phase		
N2 AMPLITUDE	01/04/2015	01/08/2015	N2 amplitude,	-	-
	03:13 (MSL, ebb)	03:13m	null phase		
SPRING&NEAP	18/04/2015	18/08/2015	M2, S2, N2, K1, M4,	-	-
	17:23 (MSL, ebb)	17:23m	O1		
NEAP&SPRING	12/04/2015	12/08/2015	M2, S2, N2, K1, M4,	-	-
	05:48 (MSL, ebb)	05:48m	O1		
VARYING TIDE	01/04/2015	01/08/2015	M2, S2, N2, K1, M4,	-	-
	02:47 (MSL, ebb)	02:47m	O1 from ADCIRC		
CURRENT	01/04/2015	01/08/2015	M2, S2, N2, K1, M4,	-	-
	02:47 (MSL, ebb)	02:47m	O1 from ADCIRC		
SE6	01/04/2015	01/08/2015	Average amplitude*,	135	6
	03:08 (MSL, ebb)	03:08m	null phase		
SE12	01/04/2015	01/08/2015	Average amplitude*,	135	12
	02:29 (MSL, ebb)	02:29m	null phase		
SW6	01/04/2015	01/08/2015	Average amplitude*,	225	6
	03:19 (MSL, ebb)	03:19m	null phase		
SW12	01/04/2015	01/08/2015	Average amplitude*,	225	12
	03:28 (MSL, ebb)	03:28m	null phase		
NO WIND	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	-	-
	00:00 (MSL, ebb)	23:00m	O1		

[

N5	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	0	5
	00:00 (MSL,ebb)	23:00m	O1		
N10	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	0	10
	00:00 (MSL,ebb)	23:00m	O1		
NE5	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	45	5
	00:00 (MSL,ebb)	23:00m	O1		
NE10	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	45	10
	00:00 (MSL,ebb)	23:00m	O1		
E5	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	90	5
	00:00 (MSL,ebb)	23:00m	O1		
E10	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	90	10
	00:00 (MSL,ebb)	23:00m	O1		
SE5	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	135	5
	00:00 (MSL,ebb)	23:00m	O1		
SE10	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	135	10
	00:00 (MSL,ebb)	23:00m	O1		
S5	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	180	5
	00:00 (MSL,ebb)	23:00m	O1		
S10	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	180	10
	00:00 (MSL,ebb)	23:00m	O1		
SW5	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	225	5
	00:00 (MSL,ebb)	23:00m	O1		
SW10	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	225	10
	00:00 (MSL,ebb)	23:00m	O1		
W5	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	270	5
	00:00 (MSL,ebb)	23:00m	O1		
W10	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	270	10
	00:00 (MSL,ebb)	23:00m	O1		
NW5	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	315	5
	00:00 (MSL,ebb)	23:00m	O1		
NW10	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	315	10
	00:00 (MSL,ebb)	23:00m	O1		
WINTER WIND	21/12/2014	20/03/2015	M2, S2, N2, K1, M4,	Observed data	Observed data
	00:00 (MSL,ebb)	23:00m	O1		
SUMMER WIND	21/06/2015	23/09/2015	M2, S2, N2, K1, M4,	Observed	Observed

* Average value of the amplitudes of the first six harmonic constituents, equal to 0.139425 m

** Sum of the amplitudes of the first six harmonic constituents, equal to 0.83655 m

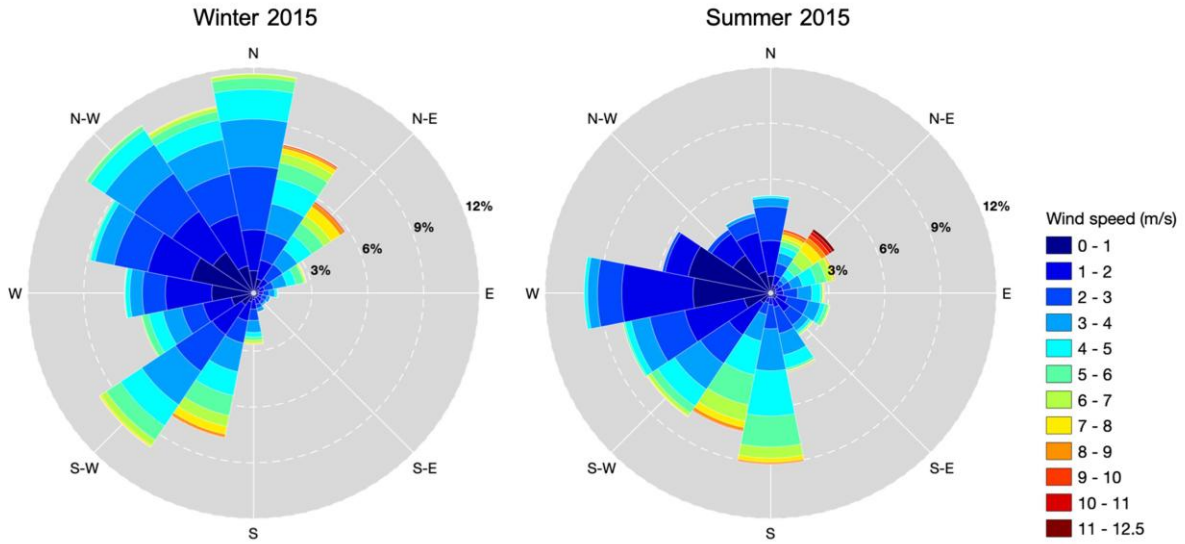


Figure 3. Distribution of wind speed and direction observed at NOAA Wachapreague station, Virginia, USA (tidesandcurrents.noaa.gov). On the left, the winter season from December 21st, 2014 to March 20th, 2015. On the right, the summer season from June 21st to September 23rd, 2015. Colors indicate the wind speed, as specified in the legend. The radial axis shows the frequency expressed as percentage of the total measurements of each combination of wind speed and direction.

3.4 Decay of tracer concentration and flushing time

The flushing time is a system-level measure that quantifies the effectiveness of tidal flushing in removing a dissolved substance from a water body through its open boundaries and provides a single value for the whole system [Choi and Lee, 2004]. A water body connected to the sea, such as a coastal embayment, is characterized by a flushing time that is well approximated by a double-exponential decay curve [Defne and Ganju 2014; Periañez et al. 2013; Choi and Lee, 2004]. In this study the time variation of the tracer mass fraction that remains in the lagoon in the absence of wind was approximated with a double-exponential function as it presents higher correlation coefficients compared to a single exponential curve. The interpolation was

implemented on the curve obtained by averaging the mass fraction within the system of bays every tidal cycle, as shown in Figure 4.

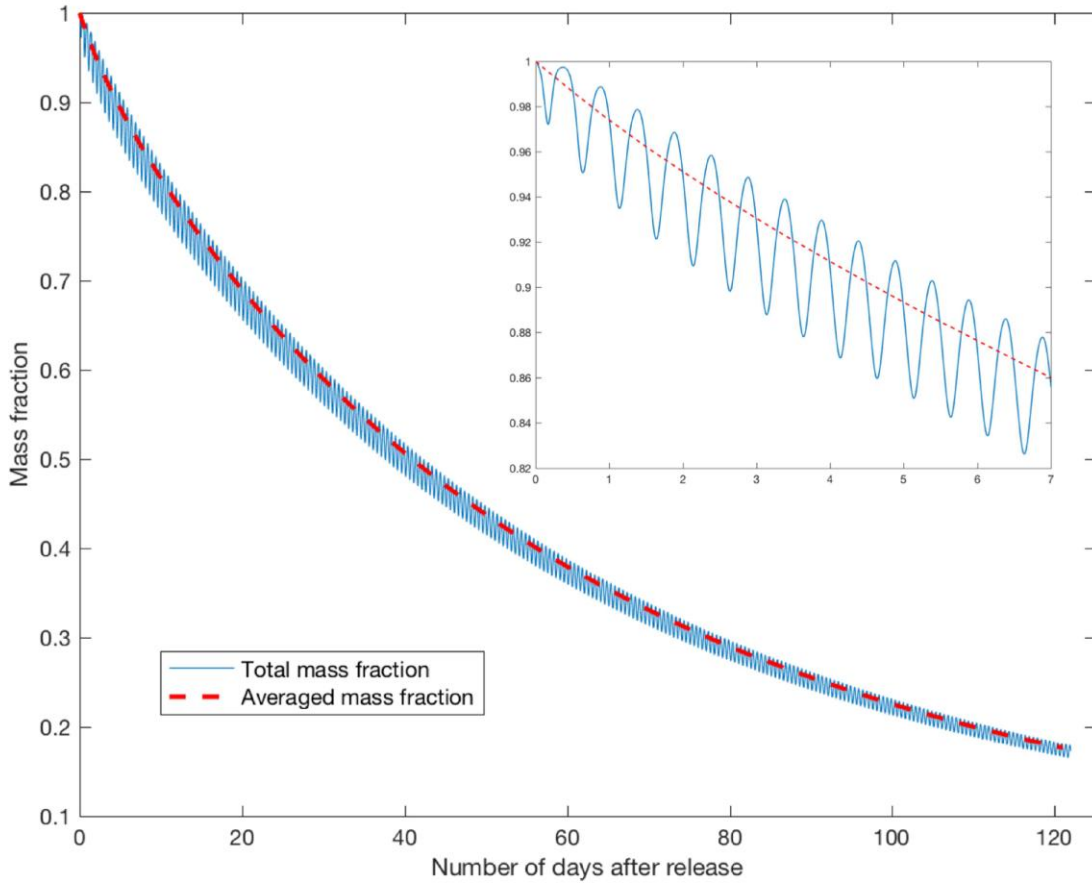


Figure 4. Mass fraction decay for the *STD EBB*. The blue line represents the decaying mass fraction calculated as integral over the entire bay. The red line is the corresponding average. In the smaller panel a magnification of the first simulation days is reported.

The decay function is made of two exponential terms, and is defined by the three parameters a , b , e :

$$\frac{m(t)}{m_0} = ae^{-bt} + (1 - a)e^{-et} \quad (1)$$

where $m(t)$ is the tracer mass at time t , m_o is the tracer initial mass. This function describes a decay process that is faster at the beginning and gets slower with time [Defne and Ganju, 2014]. The corresponding flushing time, T_f , was determined using the following equation [Choi and Lee, 2004]:

$$T_f = \frac{a}{b} + \frac{1-a}{e} \quad (2)$$

When a strong, constant wind is included in the model, the flushing process is faster, and the flushing time obtained from a single-exponential curve is similar to the value obtained from a double-exponential curve (Table S2). Therefore, in the simulations including wind, the decay of the tracer mass was interpolated with a simple exponential curve:

$$\frac{m(t)}{m_o} = ke^{-\lambda t} \quad (3)$$

where k (≈ 1) and λ are the parameters of the exponential function. In this case, the decay rate λ was employed to evaluate the flushing time, T_f , of the tracer [Choi and Lee, 2004]:

$$T_f = \frac{1}{\lambda} \quad (4)$$

An example of the different interpolations of the mass decay produced by the single and double exponential functions for the *STD EBB* scenario can be found in Figure S2.

3.5 Velocity skew at the inlets

In order to determine what process is responsible for the flushing of the tracer, the flushing time computed from Equations 2 and 4 was compared to the monthly averaged tidal prism of the bays and the velocity skew at the inlets. The tidal prism accounts for the total amount of water exchanged in a tidal cycle between the bays and the ocean. A larger tidal prism means larger fluxes and possibly a decrease in flushing time. The velocity skew at each inlet however accounts for the asymmetry of the tidal flow. Ebb velocities greater than flood velocities would prevent the return of the tracer into the bays, reducing the flushing time. This is particularly true

for a bay with several inlets, in which water can enter from one inlet during flood but then exits from a different inlet during ebb. Following Nidzieko and Ralston [2012], the velocity skew is computed as:

$$S = \frac{\int_0^T v^3 dt}{(\int_0^T v^2 dt)^{3/2}} \quad (5)$$

where v is the maximum velocity in the inlet and T is the tidal period.

We also defined a total skew index, which accounts for the combined skew of each inlet and provides a metric of asymmetry for the entire complex of bays:

$$S_{\text{tot}} = \sum_i |S_i| \quad (6)$$

where S_i is the velocity skew of one inlet and n is total number of inlets.

3.6 One-dimensional model for computing flushing time under different wind conditions

Simulations with constant wind speed and direction were used to construct a polynomial function that allows the determination of the exponential decay of the tracer mass using the least number of parameters. The goal of this method is to collapse complex, time-consuming 2D simulations in a simple point model based on a parameterized polynomial. This simple point model can then be effectively used to rapidly determine the flushing time of the system as a function of wind conditions.

In the presence of wind, the decay parameter $\dot{\Lambda}$ was evaluated as a function of wind direction and wind speed. Specifically, the wind speed, v (m/s), and direction, θ (rad), and the corresponding decay rate, λ , for the exponential curve (Equation 3) were related using the following polynomial regression function:

$$\dot{\Lambda}(\theta, v) = p_0 + p_1 \theta + p_2 v + p_3 \theta^2 + p_4 \theta v + p_5 v^2 + p_6 \theta^3 + p_7 \theta^2 v + p_8 \theta v^2 + p_9 v^3 + p_{10} \theta^4 + p_{11} \theta^3 v + p_{12} \theta^2 v^2 \quad (7)$$

where $p_0, p_1, p_2, p_3, p_4, p_5, p_6, p_7, p_8, p_9, p_{10}, p_{11}, p_{12}$ are the coefficients of the interpolating function. This polynomial model presents 4 degrees in the θ variable and 2 degrees

in the v variable, and is the regression function that gave the best fit. Results from these ensemble simulation runs with eight different constant directions and two constant wind speeds were used to determine the coefficients of Equation 5. Specifically, for each simulation a value of λ was determined for the given values of θ and v ; these points were then used to fit the polynomial surface represented by Equation 7.

Since the decay exponent was calculated as a function of wind speed and intensity (Equation 7), we could simulate the mass decay for any time period characterized by a variable wind speed. Specifically, we divided the time period in constant intervals, Δt , and for each interval we computed the decay exponent assuming an average wind speed. Then we compounded the results for the whole period. Note that because the decay was exponential, the decay of concentration in time was simply obtained by adding all the exponents:

$$\frac{m(n\Delta t)}{m_0} = k e^{-\sum_{i=1}^n \lambda_i \Delta t} \quad (8)$$

where n is the number of time intervals, λ_i is the decay exponent during the interval $i\Delta t$, $(i+1)\Delta t$.

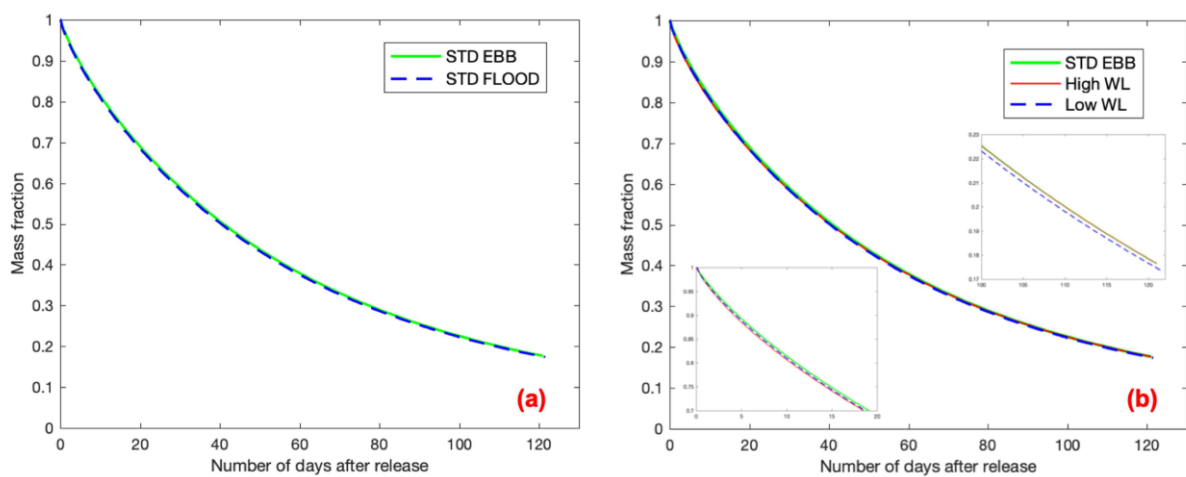
This equation allowed us to reproduce the time-varying mass decay produced by observed values of wind direction and speed. Even if the decay rates for different wind speeds and directions were obtained assuming a constant wind within every time interval, in our model each decay rate is used for only few hours, mimicking the natural wind variability of the Virginia Coast Reserve.

To test the model (Equation 8) we used wind data observed every hour at the Wachapreague station from 21st June to 23rd September 2015 (Figure 3). We derived two additional wind data sets from the hourly observed wind data, by averaging the data every 12 and 24 hours and applying Equation 8 with the same time interval of 12 and 24 hours. The obtained exponential curves were compared to the mass decay determined by a full scale Delft3D simulation during the same time period using the observed water levels and wind data registered at the Wachapreague station.

399 **4 Results**

400 4.1 Factors influencing the flushing time of tracer

401 To determine whether the timing of the tracer injection is important, two standard simulations
402 were carried out, one starting at mean sea level during ebb (*STD EBB*) and one starting at mean
403 sea level during flood (*STD FLOOD*). These two simulations presented a similar decay of the
404 tracer mass (Figure 5a).



406
407
408 **Figure 5.** (a) Comparison between the mass fraction decay in the simulation for scenarios
409 starting at mean sea level during ebb (*STD EBB*, green line) and at mean sea level during flood
410 (*STD FLOOD*, blue dashed line). (b) Comparison between the mass fraction decay in the
411 simulation starting from mean sea level during ebb (*STD EBB*, green line) and tracer injected
412 during high (*High WL*, red line) and low water level (*Low WL*, blue dashed line). In the inset
413 panels we present a magnification of the first and last days of simulation.

414
415
416 **Table 2.** Values of the coefficients of the double exponential function and flushing time for
417 each simulation without wind. In bold are highlighted simulations that used realistic tide
418 conditions for the VCR bays.

Simulation ID	a	b	c	Flushing time
		(days ⁻¹)	(days ⁻¹)	(days)
STD EBB	0.18102	0.05950	0.01294	66.33
STD FLOOD	0.17953	0.06395	0.01306	65.63
HIGH WL	0.16253	0.07537	0.01319	65.68
LOW WL	0.28679	0.03676	0.01175	68.50
MAX AMPLITUDE	0.61253	0.14127	0.02301	21.17
N2 AMPLITUDE	0.12571	0.07032	0.01139	78.55
SPRING&NEAP	0.43649	0.17282	0.02500	25.06
NEAP&SPRING	0.52961	0.10923	0.02091	27.34
VARYING TIDE	0.64422	0.08857	0.02089	24.30
CURRENT	0.63233	0.08474	0.01960	26.22

Both simulations had a flushing time of about two months, 66.33 days for *STD EBB* and 65.63 days for *STD FLOOD* (Table 2). This result suggested that there is no significant difference whether a tracer is added to the system during ebb or during flood. In the next two simulations, the initial release of tracer was set to occur during high water level (*HIGH WL*) and low water level (*LOW WL*). Also in this case the flushing time did not show relevant differences (Figure 5b). In addition, the flushing time was similar to the one obtained for the standard simulation starting at mean sea level during ebb; a relatively small (two days) increase in flushing time occurred for the scenario starting at low water level.

Three simulations were then used to investigate the effect of tidal amplitude, and to see whether different tidal ranges influence the flushing time. In the *MAX AMPLITUDE* scenario, the decay of the tracer mass within the lagoon was faster than in the *N2 AMPLITUDE* scenario (Figure 6). The flushing time in the *N2 AMPLITUDE* simulation was about 79 days, approximately 60 days longer than the flushing time computed during the *MAX AMPLITUDE* scenario. Two more simulations were carried out to determine the difference in flushing time under Spring-Neap alternation: in one the tracer was released during a neap tide followed by a spring tide (*NEAP&SPRING*); in the second it was released during a spring tide followed by a neap tide (*SPRING&NEAP*). In this case the comparison presented a less pronounced

divergence. The difference in flushing time values was less than 2 days, further demonstrating that the timing of tracer injection had no significant impact in the flushing process (Figure 6).

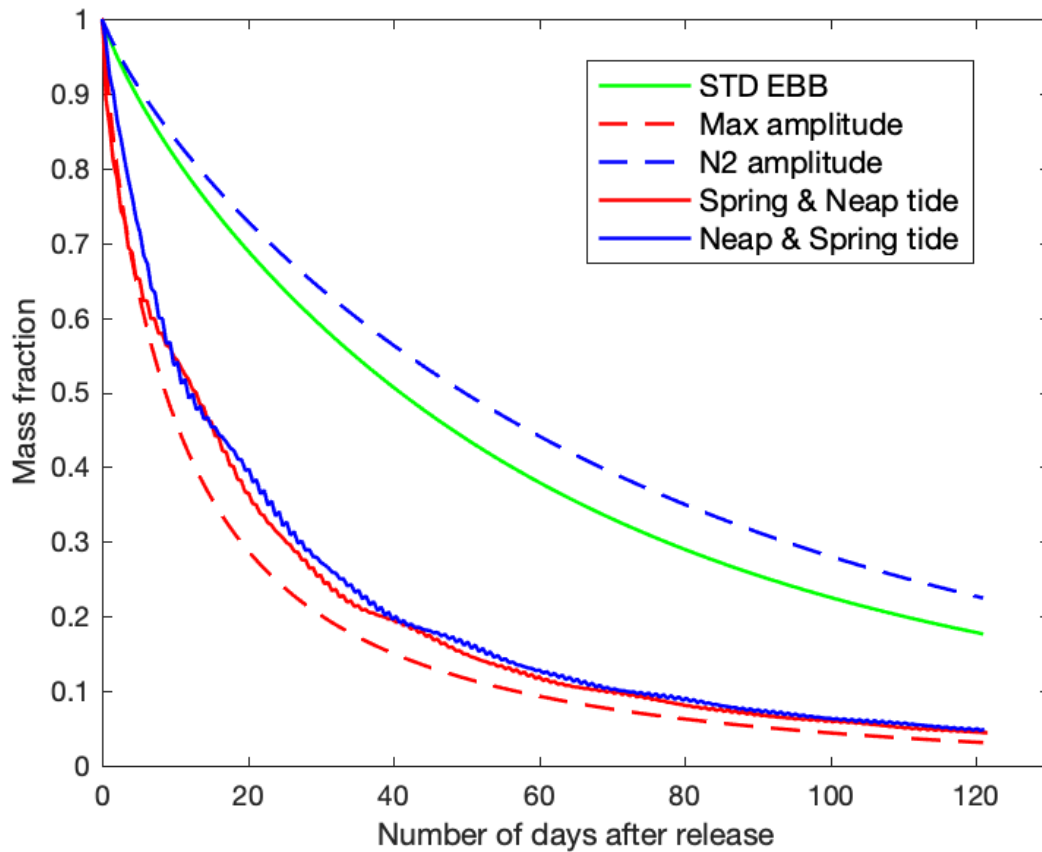


Figure 6. Comparison between the mass fraction decay in the standard simulation and scenarios including Spring and Neap tides. The mass fraction is plotted after one full tide.

To determine the effect of wind on tracer dispersion, hypothetical simulations with constant wind speed and direction were run. We considered the combination of two wind speeds and two wind directions, one perpendicular and one parallel to the inlets (Figure 7). We chose the two wind directions that produce the maximum (SW) and minimum effect (SE), in order to capture the full variability of this process. In these simulations the decay of the tracer mass fraction resulted to be faster than in the scenarios without wind, indicating that wind might be a strong driver of water exchange in the system. The decay of mass fraction due to the effect of wind can be interpolated with a simple exponential curve, since the use of a double exponential

interpolation yields the same values of flushing time and the correlation coefficient is high in both cases (Table S2). Table 3 shows the values of the flushing time evaluated for each simulation using a single exponential function to interpolate the decay. In particular, the simulations with the highest wind speed (12 m/s) were the most effective at flushing the system, and in the case of south-west wind the flushing time reached its lowest value (less than 4 days). Regardless of direction, as the speed doubles its value to 12 m/s, the flushing time decreases and becomes less than half the flushing time obtained when the velocity is 6 m/s.

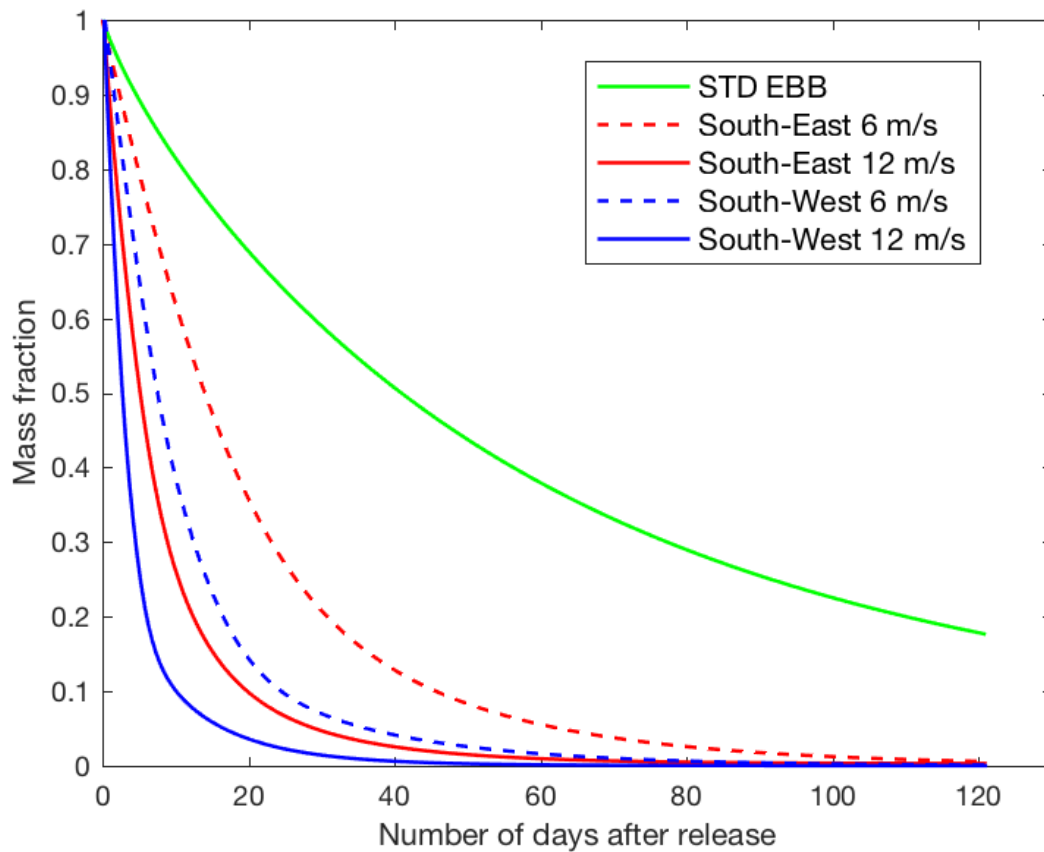


Figure 7. Comparison between the mass fraction decay in the standard simulation and scenarios including wind coming from south-east and south-west directions and with a velocity of 6 m/s and 12 m/s.

However, note that a constant wind lasting for several days is uncommon at the Virginia Coast Reserve, and more realistic conditions with variable wind should also be tested. Two simulations

were run with variable wind conditions measured at Wachapreague (Figure 8). When the variability of wind speed and direction was taken into account, the flushing time increased (Table 3, *WINTER WIND* and *SUMMER WIND* simulations). This is because most of the time the wind is weak (Figure 3). The flushing time computed with winter winds is similar to the flushing time without winds (25.06 days, see Table 3), whereas the flushing time decreases with summer winds to 21.07 days. This is likely due to strong winds from northeast that were present in Summer 2015 (Figure 3).

In the simulation with tidal phasing along the boundary (*VARYING TIDE*) the decay of the tracer concentration in the bays was fast (Figure 8). The flushing time (24.3 days) decreased with respect to *NEAP&SPRING* simulation with a constant phase (27.34 days, see Table 2). This is likely due to subtidal circulation triggered by differences in tidal phase among the inlets. In fact the residual Eulerian velocity enters the bays in the northern inlets, where the tidal signal arrives first, while the velocity exits the bays in the southern inlets, where the tidal signal arrives with a delay (Figure S3). Moreover, at each inlet the residual velocity is segregated in very distinct flood and ebb paths. Both processes increase the dispersion of tracer when it exits the bays, reducing the flushing time.

In the *CURRENT* simulation we considered a current along the shore from northeast to southwest, mimicking the mean shelf circulation over the Middle Atlantic Bight. The flushing time did not further decrease, and instead increased to 26.22 days (Figure 8, Table 2). The four scenarios *VARYING TIDE*, *CURRENT*, *WINTER WIND*, and *SUMMER WIND* used realistic boundary conditions and therefore the corresponding flushing time is more representative of the system.

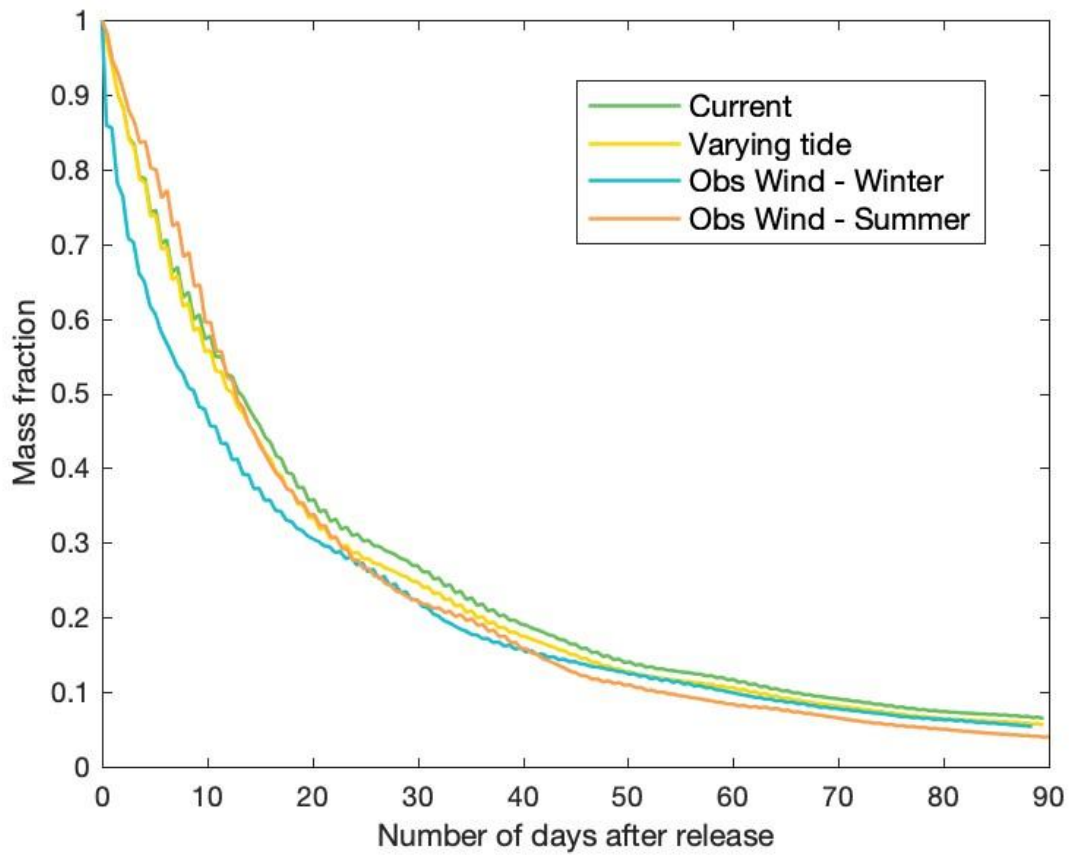


Figure 8. Comparison between the mass fraction decay in the scenario with a shelf current, the scenario with varying tides along the boundaries and the observed wind data in the winter and summer seasons of 2015.

Table 3. Values of the coefficients of the single exponential functions, flushing time, and adjusted R^2 and RSS of the interpolation for each simulation including wind. In bold are highlighted simulations that used realistic wind conditions for the VCR bays.

Simulation ID	K	d (days ⁻¹)	Flushing time (days)	Adjusted R^2	RSS
SE6	1.0160	0.05112	19.56	0.9990	0.0136
SE12	0.9967	0.12600	7.94	0.9950	0.0420

SW6	1.0360	0.09545	10.48	0.9972	0.0278
SW12	1.0230	0.25000	4.00	0.9921	0.0392
WINTER WIND	0.7568	0.03991	25.06	0.9632	0.2266
SUMMER WIND	0.9718	0.04745	21.07	0.9874	0.1295

Figure 9 shows the fraction of total concentration of the tracer during the first time steps of the simulation (after 24 and 48 hours on the first day of simulation) in the scenarios without wind (STD EBB) and with wind (south-west 12 m/s and south-east 12 m/s). Both comparisons demonstrate that wind pushes the tracer outside the lagoons and prevents the return of water and tracer inside the lagoons. Therefore, the concentration of tracer remaining in the lagoons decreases if wind is included in the simulation, and as time passes the concentration of tracer decreases faster than in the scenario without wind.

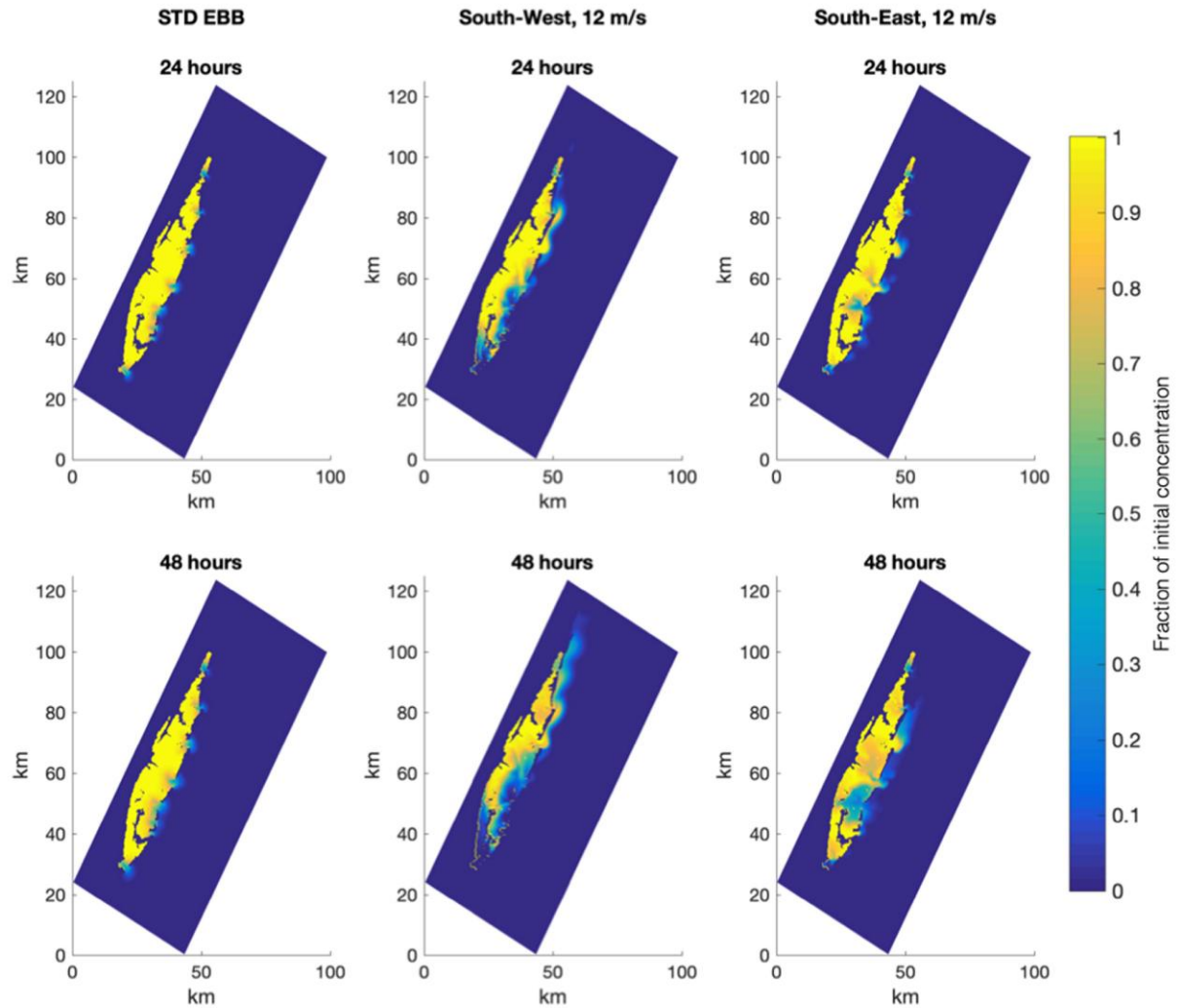


Figure 9. Fraction of total tracer concentration in the lagoon after the first 24 and 48 simulation hours in the scenario without wind (*STD EBB*), and in the scenarios with strong wind (*South-West, 12 m/s*, and *South-East, 12 m/s*).

4.2 Relationship between velocity skew at the inlets and flushing time

We then determined whether tidal prism or velocity asymmetry at the inlets affect the flushing time of tracer in the bays using the results of all simulations. Flushing time is significantly correlated to total skew index (Kendall's tau = -0.73, $p < 0.05$), but not to tidal prism (Kendall's tau = -0.26, $p = 0.06$, Figure 10). Moreover, a multiple linear regression model between the logarithm of the flushing time (response), and the logarithm of the total velocity skew and of the

tidal prism (predictors) has an adjusted R^2 of 0.77 and a p -value of $9.88\text{e-}09$. Specifically, the adjusted R^2 for a linear regression model between the logarithm of the flushing time and the logarithm of the total velocity skew is 0.70 with a p -value of $3.37\text{e-}08$ (see equation in Figure 10). On the contrary, the adjusted R^2 decreases to 0.13 when we analyze the correlation between the flushing time and the tidal prism. These results suggest that the total velocity skew explains 70% of the variance in flushing time in log-log space (Figure 10), while the tidal prism can capture only the 13% of the variance in flushing time; together they explain 77% of the variance in flushing time. The high total skew for the simulations with wind indicates that some of the water entering the system from one inlet exits from a different inlet.

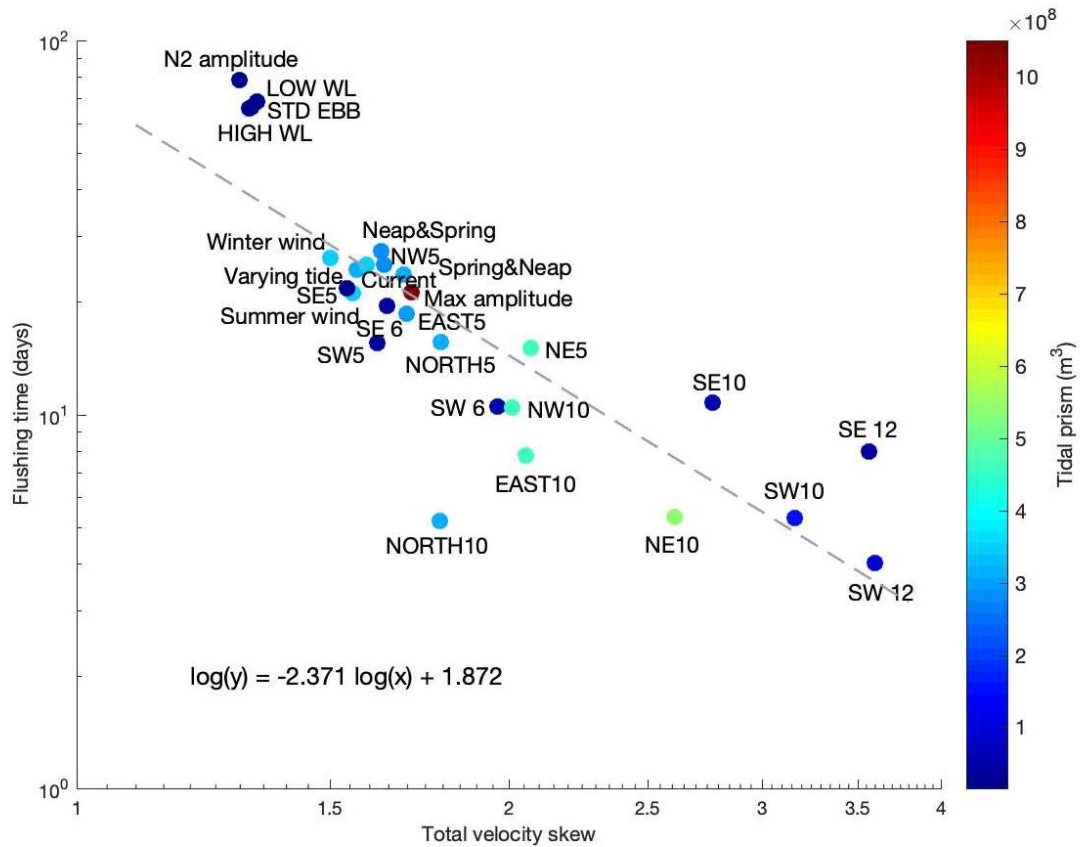


Figure 10. Flushing time of tracer as a function of tidal prism and total velocity skew. The grey line is the linear interpolation of the flushing time as a function of the total velocity skew. The equation of the linear model is indicated in the left bottom corner. Both x and y axes are in logarithmic scale.

4.3 One-dimensional model for computing flushing time under different wind conditions

Previous results refer to the most frequent conditions in terms of wind speed and direction. However, since wind appears to be an effective forcing for flushing, additional simulations were carried out for increments of wind speed of 5 m/s and every 45 degrees of wind direction (Table 4).

Table 4. Values of the coefficients of the single exponential function, flushing time, and R^2 and RSS of the interpolation for the simulations including wind (simulations used for the polynomial fit).

Simulation ID	K	λ (days ⁻¹)	Flushing time (days)	Adjusted R^2	RSS	Variance of residuals
No Wind	0.92330	0.03421	29.23	0.9795	0.1975	1.0000e-03
N5	1.02000	0.06403	15.62	0.9990	0.0102	5.8072e-05
N10	1.03200	0.19320	5.18	0.9980	0.0104	4.7768e-05
NE5	1.01700	0.06653	15.03	0.9991	0.0092	4.9934e-05
NE10	1.02400	0.18840	5.31	0.9990	0.0054	2.4527e-05
E5	1.00400	0.05378	18.59	0.9992	0.0085	4.5660e-05
E10	1.01600	0.12890	7.76	0.9993	0.0048	2.6462e-05
SE5	0.98530	0.04592	21.78	0.9985	0.0166	8.8675e-05
SE10	1.02700	0.09284	10.77	0.9994	0.0054	3.0003e-05
S5	1.02200	0.06421	15.58	0.9990	0.0112	6.1717e-05
S10	1.03900	0.18190	5.50	0.9983	0.0093	4.3930e-05
SW5	1.01400	0.06447	15.51	0.9990	0.0103	5.5191e-05
SW10	1.04400	0.18940	5.28	0.9984	0.0086	4.3941e-05
W5	1.00900	0.05372	18.61	0.9989	0.0123	6.5706e-05
W10	1.02000	0.11910	8.39	0.9995	0.0039	2.1251e-05
NW5	0.97810	0.04234	23.62	0.9974	0.0286	1.4887e-04
NW10	1.05000	0.09581	10.44	0.9985	0.0141	7.7408e-05

For each simulation, the mass fraction of the tracer inside the lagoons is fitted using a single exponential decay, as in the previous simulations including wind (Table 3). The decrease of mass fraction inside the lagoons becomes more rapid as the wind speed increases (Figure 11), with the fastest decay reached when the velocity is equal to 10 m/s and the wind is blowing in the direction parallel to the coast (northeast and southwest). Conversely, when the wind blows perpendicular to the barrier islands (northwest and southeast), it facilitates the reintroduction of the tracer that has left the lagoons through the inlets. The mass fraction decay triggered by winds from southwest and northeast presents an analogous exponential trend. Similarly, winds blowing from southeast and northwest yield a similar flushing time.

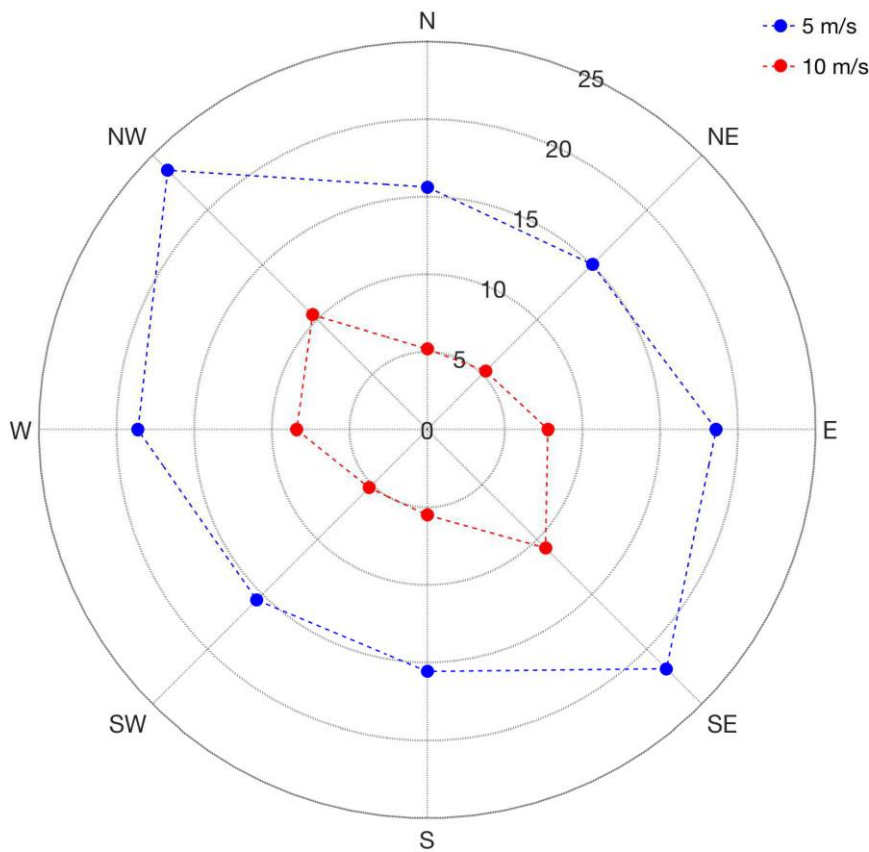


Figure 11. Flushing time distribution (days) as a function of the wind speed (in blue and red) and direction for the simulations with constant wind.

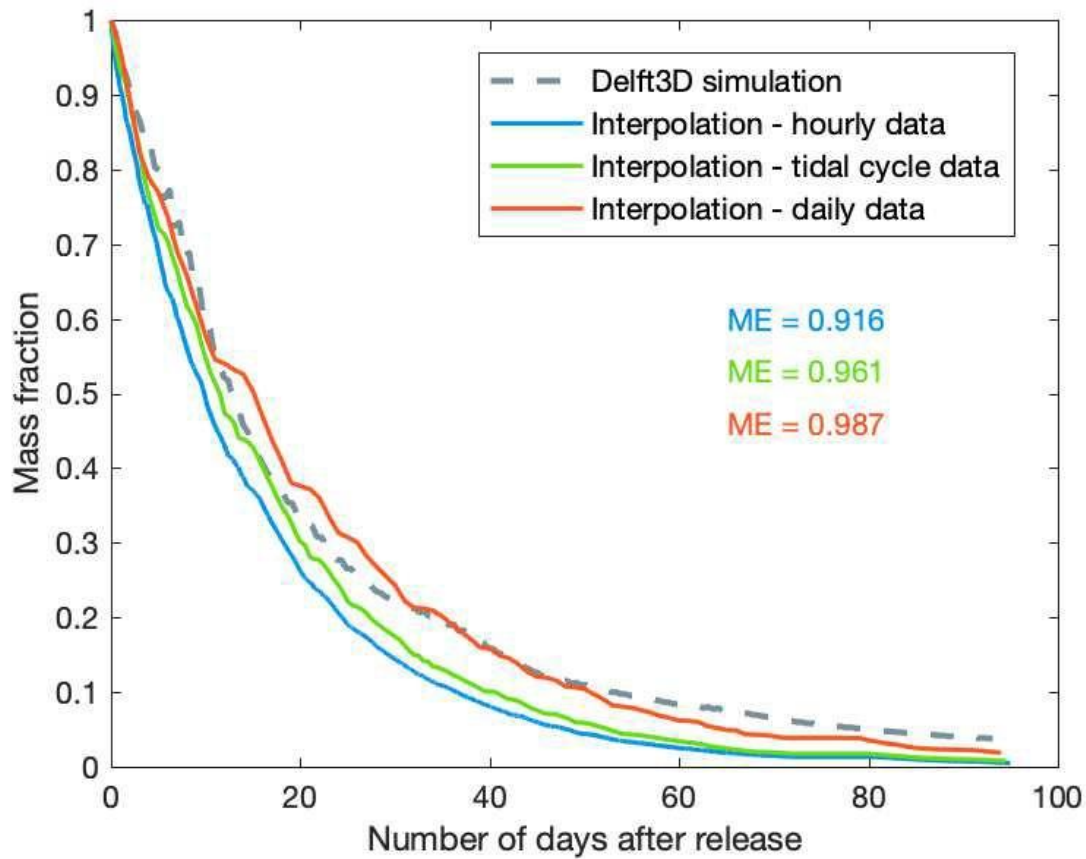


Figure 12. Comparison between the mass fraction decay derived from the Delft3D simulation and the mass fraction decay derived from the application of the polynomial function to the hourly, tidally and daily averaged wind data in the summer season of 2015.

The analytical method based on Equations 7 and 8 successfully reproduces the general mass decay behavior simulated by Delft3D (Figure 12). It is observed that the mass fraction retained in the system at the end of the simulation is better reproduced by the interpolation made with the daily averaged wind data, while tidally averaged, and hourly wind data are found to better interpolate the decay in the first days. The decay function varies depending on the interval over which the wind is averaged (one hour, one tide, one day) because the interpolating polynomial function is not linear.

572 **5 Discussion**

573 In the scenarios with Neap and Spring tides and without wind (Table 2, *SPRING&NEAP* and
574 *NEAP&SPRING*) the Eulerian approach adopted in this analysis produced a higher flushing time
575 than the values obtained with the Lagrangian method in Safak et al. [2015]: from 10 to almost 28
576 days in the case of the bays closer to the inlet. The reason for this is that the approach followed
577 by Safak et al. [2015] does not consider that a significant amount of particles that leave the
578 system during ebb can reenter the bays during flood, resulting in an underestimation of the actual
579 flushing time. On the other hand, coastal processes might facilitate the removal of the tracer once
580 it reaches the nearshore area outside the inlets, and thus, determine a faster decay of the tracer
581 mass. Here some of these processes, such as waves and wave-driven longshore currents, are not
582 included in the simulations and this might cause a change in flushing time. There is a difference
583 of approximately two days between the flushing time observed after the tracer is released during
584 high and low water levels, and this result seems to be in accordance with the median flushing
585 time evaluated in Safak et al. [2015] for almost all the bays having a direct connection to an inlet.
586 Also, the decrease in flushing time when wind is included in the model is in agreement with
587 Safak et al. [2015], and the magnitude of the decrease is a function of the velocity and direction
588 of the simulated wind. Specifically, the flushing time was lower during the summer of 2015
589 because a storm brought strong winds from northeast, a wind scenario that favors flushing
590 (Figure 11).

591 The particle tracking approach adopted by Safak et al. [2015] allows the evaluation of the
592 spatial distribution of the flushing time at each location, and shows that the parts of the bays
593 located near the inlets are more sensitive to tidal phase than wind forcing [Safak et al., 2015;
594 Fugate et al., 2006]. Here the Eulerian method provides a single value that defines the overall
595 flushing time of the tracer for all bays, regardless of the point of release. The entire system of
596 lagoons is more sensitive to wind forcing and differences in tidal phase along the coast than to
597 instant of release, indicating that the global flushing time is more affected by the flushing time
598 associated with the bays found in the interior region of the domain rather than in the areas
599 located near the inlets.

600 Defne and Ganju [2014] analyzed the influence of tides, coastal, riverine, and meteorological
601 processes on flushing time. They found that the percentage of particles removed from the domain
602 increases with the progressive addition of forcings to the scenario with only tides. However, in

that study the effects of each single factor were not investigated, and it is difficult to determine what mechanism caused the greatest reduction in flushing time. Defne and Ganju [2014] concluded that remote coastal forcings (i.e., tidally averaged water levels and currents added to the tidal oscillations at the open boundaries) and meteorological forcings (surface air pressure, wind speed and direction) were the most effective in the removal of particles.

In the scenarios of this study including the meteorological forcing, i.e. wind conditions, it has been observed that as wind speed increases the mass decay occurs at a faster rate, and therefore the flushing time of the tracer decreases. This is because wind-driven circulation produces asymmetric fluxes of water between the lagoon and the ocean. Note that our synthetic scenarios assume a wind speed that is constant in intensity and direction for an unrealistic long time, so the resulting mass decay should be seen as a theoretical value. On the other hand, in our proposed simplified model (Equation 7 and 8) the decay rate is changed every few hours to account for wind variability in a realistic manner.

Given the highest wind speed (12 m/s), the most effective wind direction in the removal of tracer particles is observed to be the southwest direction, corresponding to the direction parallel to the inlets of the lagoons. In this scenario the tracer leaving the lagoons is not able to re-enter into the system because the wind-generated currents are moving northwest. Moreover, the Ekman transport facilitates the exchange of water and particles when the wind blows parallel to the coast, increasing storm surges and extreme low tides [Fagherazzi et al., 2010]. Fagherazzi et al. [2010] indicate that Ekman transport can increase the water level in the VCR lagoons by 0.2-0.4 m, which roughly corresponds to an increase in tidal prism between 15% and 30%, and therefore an increase of the fluxes in and out the system. These fluxes also vary across the inlets, producing velocity asymmetry in flood and ebb and tracer dispersion (Figure 10).

On the other hand, in the scenario with a wind direction blowing landwards perpendicular to the coastline (southeast direction), the particles exit and re-enter the system several times through the inlets, producing a slower decay of the tracer concentration inside the lagoons. In addition, the tracer that remains in the lagoons is pushed toward the northern bays of VCR, where the inlets are smaller, and thus, the exchange of the tracer particles between the bays and the open ocean takes place at a lower rate.

A wind perpendicular to the barrier islands but blowing from land towards ocean (northwest direction) increases at first the flushing of the tracer by producing a setdown and a large outward

634 discharge. However, for continuity, a sizable water flux will also occur during the next flood
635 phase to replenish the lagoons with the water lost during the setdown. This phenomenon was also
636 observed by Fagherazzi and Priestas [2010] in the Louisiana coast. Wind blowing perpendicular
637 to the coast increased the flux of sediment toward the ocean at first, but then the sediment re-
638 entered the system during the following tidal cycle. The overall effect of a wind perpendicular to
639 the shore is an increase in water fluxes in and out of the system and a slight velocity skew in the
640 inlets, moderately increasing the flushing of the tracer.

641 A wind parallel to the shore produces instead a strong asymmetry in circulation, with flow
642 exiting the bays from one inlet and re-entering from a different one so that most of the tracer
643 exiting the system does not return (Figure 9). Field measurements and numerical simulations
644 carried out by Li [2013] in a bay with three inlets and wind blowing parallel to the shore indicate
645 that a net outward flow takes place at the downwind inlet, while a net inward flow occurs at the
646 inlet located upwind. As a result, fluxes during ebb are not rebalanced by symmetric fluxes
647 during flood within the same inlet, dramatically increasing the overall tracer removal from within
648 the lagoons.

649 When the variability in wind direction is accounted for, the flushing time becomes more
650 similar to the simulations without wind (Table 3). This is because most of the time the wind is
651 weak (Figure 3). We do notice a sharp decrease in flushing time with summer winds. While on
652 average the measured winds were higher in winter, in the summer of 2015 a storm occurred that
653 triggered winds above 10 m/s (Figure 3). The maximum wind blew from northeast, along the
654 shallow bays, which we showed being one of the directions more favorable for tracer flushing
655 (Table 4). Of all the simulations with realistic tidal and wind conditions, the one with summer
656 winds led to the lowest flushing time (21.07 days). We therefore conclude that winds do exert
657 control on the flushing of the system, but only if they blow along the bays and not
658 perpendicularly to them. Moreover, infrequent events with high wind speed are more important
659 than average wind conditions.

660 We conclude that processes triggering asymmetric fluxes between the bays and the ocean, such
661 as wind-driven subtidal circulation, and differences in tidal phase and amplitude along the shore
662 are the most successful at flushing tracer. Processes that augment the fluxes of water in and out
663 the bays, such as wind driven storm surges and setup caused by Ekman transport, also enhance
664 tracer flushing, but to a lower degree and partly through an increase in velocity skew.

Some of these results might be only valid for coastal bays with multiple inlets. In fact, if only one inlet is present, subtidal circulation driven by wind is likely absent, since all the water is entering and exiting from the same inlet. Wind can still trigger surges and Ekman transport, thus increasing the volume of water exchanged with the ocean and therefore flushing, but the lack of asymmetric flows will increase the overall flushing time of the tracer.

In this study we have not considered wind generated waves and long-shore currents. Based on our findings, it seems likely that wave-driven long-shore currents would favor the dispersion of tracer by creating asymmetric fluxes at the inlets, similarly to what wind-driven currents do.

More research is needed to determine the role of wave-breaking and long-shore currents on tracer dispersion.

6 Conclusions

Semi-enclosed water bodies, such as bays and lagoons, host ecosystems sensitive to the release of nutrients and pollutants due to human activities along the coast. Therefore, understanding the decay and flushing time of tracers has important environmental consequences. Given a system of bays, the mean flushing time associated to the entire area can be evaluated by using hydrodynamic models and an Eulerian based approach. In this paper, the transport of a tracer within the system of shallow bays in the VCR has been simulated using the hydrodynamic model Deflt3D. The decrease in time of the tracer mass follows an exponential decay function. Specifically, a double exponential function was found to better approximate the tracer decay when the effect of wind is neglected, while a single exponential function is a good approximation when the wind effect is included in the simulations. We further identified factors producing the shortest flushing time of the tracer within a system with multiple inlets. We show that asymmetry in ebb-flood velocity at the inlets is responsible for the rapid decay of the tracer mass in the lagoons. These asymmetries, computed with a velocity skew index, are generated by differences in tidal phase and amplitude along the inner shelf and by wind-driven circulation under non-storm conditions.

In general, the average flushing time of the VCR bays is around 25-27 days. When the difference in tidal phase along the Atlantic coast is accounted for, the flushing time decreases to 24 days, due to asymmetric tidal fluxes at the inlets. Winds can also decrease the flushing time, but only if they are strong and blow along the bays (from southwest or northeast). In the summer

of 2015, strong winds reduced the flushing time to 21 days. Currents triggered by shelf circulation do not seem to affect the flushing of the system.

A simple exponential decay function using wind-dependent parameters calculated with a polynomial regression (Equation 7 and 8) can be used to estimate the decay in concentration without the need of a hydrodynamic model. This function well predicted the tracer mass decrease during the summer months of 2015 in the VCR bays. This result demonstrates that the average flushing time of tracer within the entire system of bays in the Virginia Coast Reserve can be estimated from the evaluation of a single parameter, which controls the exponential decay of the tracer mass inside the lagoons.

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