1 # Drifter and dye tracks reveal dispersal processes that can affect

2 # phytoplankton distributions in shallow estuarine environments

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28 **ABSTRACT**

29 In shallow estuarine environments, the time scales of hydrodynamic processes that control

30 particle distribution may outpace the time scales of phytoplankton patch formation through

31 reproduction. Consequently, physical processes can dominate the distribution of the

32 phytoplankton, but these processes and their dynamics are not well understood. Here we used

33 flow measurements with a bottom mounted Acoustic Doppler Current Profiler (ADCP),

34 shipboard hydrographic transects, drifter releases, and Rhodamine dye to characterize the small-

35 scale flow environment and its effect on dispersion processes in a shallow estuarine environment,

36 Apalachicola Bay, Florida. Spatial spectra of salinity and chlorophyll followed a power law

37 behavior of -3 at length scales of 250 m – 5 km. The ADCP data revealed the presence of a

38 vertically sheared flow that was strongly modulated by tides and bottom topography. Tidal flows

39 had a characteristic magnitude of 20 – 40 cm s^{-1}, with durations of flow reversals between the

40 near-surface and bottom flows. Drifter triplets indicated shear and strain rates on the order of 10^{-3}

41 – 10^{-4} s^{-1}, and single particle dispersion rates (diffusivity) of 0.1 m^2 s^{-1}. The area evolution of the

42 dye patch observed by a drone corresponded to eddy diffusivity comparable to those estimated

43 from drifters, or about 0.1 m^2 s^{-1}. The dye patch experiments demonstrate how physical

44 processes at scales of 1-100 m can affect the shape and development of phytoplankton patches in

45 the bay. Vertical shear, produced by wind directions deviating from flow direction, can broaden

46 and divide a plankton patch by transporting different depths of a patch in different directions.

47 When winds and currents are aligned, shear leads to elongation and narrowing of the patch. The

48 results indicate that the small-scale flow environment in estuaries can be pivotal in controlling

49 the distribution and dispersal of planktonic organisms and thereby becomes a decisive factor for

50 the development and breakdown of phytoplankton communities.

51

1. Introduction

When favorable conditions promote growth of a phytoplankton organism, rapid

reproduction ensues leading to the local accumulation of cells, a phytoplankton patch, and to first

order phytoplankton cells can be considered passive particles. Hydrodynamics, grazing and

plankton behavior then may control the spatial and temporal development of the phytoplankton

patch. Phytoplankton organisms therefore are not distributed uniformly, and individuals occur

more frequently together than predicted from a random distribution. Satellite images reveal

large-scale patterns of phytoplankton distribution, and the light emission spectrum of chlorophyll

can be used for estimates of phytoplankton abundance and biomass. Although small amounts of

free chlorophyll released by decaying plant material may occur in the water column, chlorophyll

concentration is a useful and widely accepted proxy for phytoplankton and its patchy distribution

(Behrenfeld et al., 2005;Behrenfeld and Falkowski, 1997;Jimenez et al., 1987;Longhurst, 1995).

Owing to its spatial and temporal dynamics, describing phytoplankton patchiness is complex,

and typical methods include Fourier power spectra analysis, multi-point correlation functions,

wavelet analysis, and multifractal analysis. Because the spatial heterogeneity influences

productivity, diversity, and foodweb stability, understanding the controls of phytoplankton

patchiness is prerequisite for assessing the functioning of the marine ecosystem (Martin, 2003).

Phytoplankton communities are dispersed by water currents and associated turbulence

(Mackas et al., 1985; Okubo, 1978; Prairie et al., 2011). Characterizing the fluid flow and its

variability therefore is essential for developing models of phytoplankton distribution and

ecology. Spatial heterogeneity of environmental factors, such as current velocities, salinity,

nutrients, light, and grazers, modulate the patchy distribution of phytoplankton (Harris, 1986;

74 Haury et al., 1978), which has implications for the productivity and trophodynamics of an

75 ecosystem (Martin, 2005; Roman et al., 2005; Wetz et al., 2011). Most studies characterizing

76 phytoplankton patchiness and the processes influencing it have been conducted in the coastal and

77 open oceans (Mahadevan, 2016; Martin, 2003); and phytoplankton patchiness in estuarine

78 environments thus has remained poorly understood.

79   A characterizing feature of most estuaries is that they are relatively shallow, typically

80 with average water depth of less than 10 m (Bricker et al., 2008), which sets their hydrodynamics

81 and controls of phytoplankton distribution apart from those of deeper marine systems. Factors

82 controlling patchiness in estuaries that differ from those in the open ocean include river

83 discharge, steep density and nutrient gradients, local winds, and the morphology of the estuary

84 (Dustan and Pinckney, 1989; Lucas et al., 1999; Mortazavi et al., 2000; Roman et al., 2005). The

85 dominant temporal scales of estuarine processes/mixing also tend to be shorter than those in the

86 open ocean because of the smaller spatial scales (including water depth) and strong tidal

87 influences (Fischer, 1976; Geyer and Signell, 1992). In Apalachicola Bay/Florida, Geyer et al.

88 (2018) observed small-scale patches of phytoplankton with widths of 0.1 – 4.5 km and steep

89 chlorophyll *a* (Chl *a*) gradients. These small patches, which often were associated with density

90 fronts, accounted for about 10% of the phytoplankton biomass along the sampled transects and

91 thus were significant. However, the mechanisms shaping these patches were not clear.

92   The present lack of understanding of the processes controlling this patchiness limits the

93 ecological conclusions that can be drawn from these observations and ultimately quantification

94 of estuarine phytoplankton. This lack is significant as estuaries are among the most productive

95 environments in the oceans (Boynton et al., 1982; Cloern et al., 2014), are of exceptional

96 ecological and economical importance (Bundy, 1992; Day et al., 2012; Mansur et al., 2016), and

4

97   now are disproportionally threatened by climate change, sea level rise, nutrient input and other

98   human activities (Camargo and Alonso, 2006; Rabalais et al., 2009; Rabouille et al., 2001).

99   To address this lack in understanding, we initiated a study designed to characterize and

100   assess physical processes that contribute to the controls of small-scale estuarine phytoplankton

101   distribution patterns as those observed in Apalachicola Bay estuary in August 2011 (Geyer et al.

102   2018). Our process study uses flow tracer and drifter deployments as well as current

103   measurements to assess the influence of estuarine flow characteristics on transport and dispersion

104   of mock-phytoplankton patches represented by inert dye tracer patches. Main goals were to

105   quantify lateral advection and dispersion processes on short timescales (minutes to hours) in the

106   dynamic surface layer (< 1 m), and to compare the purely physically controlled distribution

107   characteristics of tracer dye patches with distribution characteristics of phytoplankton patches we

108   observed in the same estuary in 2011. This comparison suggests that small scale physical flow

109   and mixing processes can dominate phytoplankton distributions in estuarine settings, and

110   underscores the combination of synoptic high-resolution physical, biological, and chemical

111   measurements required for untangling the controls of phytoplankton bloom evolution in these

112   productive key environments.

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114

115   **2. Methods**

116

117   The in-situ work in Apalachicola Bay/Florida utilized Lagrangian drifters, aerial drone

118   photography, boat-mounted flow-through sensors, and fixed current meters to investigate the

119   transport of the mock-phytoplankton patches (i.e., dye) and the adjacent waters. Lagrangian

120   drifter data (Davis, 1991) allow an improved determination of diffusion coefficients (Pal et al.,

121   1998) that affect the distribution of phytoplankton, nutrients, or pollutants. Likewise, passive dye

122   tracers are a powerful tool for studying transport and circulation patterns in the open and coastal

123   ocean (Garrett, 1983; Sundermeyer and Ledwell, 2001; Sundermeyer et al., 2005; Watson and

124   Ledwell, 2000; Yu et al., 2016), estuaries (Bailey, 1966; Chant et al., 2007), and near-shore

125   environment (Brouwer et al., 2016; Clark et al., 2014). Fluorescent dyes stain the water and

126   allow the movement of the water (and thereby substances within, such as plankton, nutrients, or

127   pollutants) to be traced visually or with fluorometers. Since the dye is a practically inert tracer,

128   dye patches offer insight into the non-biological drivers of patch development. The recent

129   developments of remotely controlled drones with onboard cameras facilitates inexpensive aerial

130   observation of the dye tracer movement (Brouwer et al., 2016; Tauro, 2016).

131

132   *2.1. Study Site*

133

134   The in-situ study was conducted over 4 days, from March 30 – April 2, 2015, in the

135   western region of Apalachicola Bay (AB), a bar-built, microtidal estuary located in the

136   Northeastern Gulf of Mexico (Figure 1). The estuary has an east-west length of 63 km, and a

137   north-south width of 12 km at its widest points, with an average depth of 2 – 3 m (Edmiston,

138   2008). The bay receives freshwater from the Apalachicola River, the largest river in Florida by

139   flow volume, and opens to the ocean through four inlets: East Pass, Sikes Cut, West Pass, and

140   Indian Pass. Aside from intermittent strong winds, the energetic currents in the bay are driven to

141   first order by differences in tidal amplitudes between these different passes, with most of the

142   inflow being through East Pass. Local reversals can be seen in the western region when the flood

6

143 tide pushes water in through the western passes (Huang et al., 2002b). Sustained along-estuary

144 winds can modulate and even reverse the tidal surface currents (Huang and Foo, 2002; Huang et

145 al., 2002a). Long-term average annual wind velocity is 3.5 ± 0.4 m s^{-1}, with the main direction

146 November-June from the east, turning more north during July to October. Mean tidal range in

147 AB is 65 ± 16 cm. In 2015, the average water temperature in the center of the bay (Cat Point)

148 was 23.4 °C (range: 9.7- 33.7°C), salinity 22.3 ± 6.5 (3.4 – 33.9) and river discharge (573 ± 579

149 m^3 s^{-1} (178 – 5040 m^3 s^{-1}). This discharge creates a residual flow controlling flushing times of the

150 bay, which ranges from 6 – 12 days (Dulaiova and Burnett, 2008; Morey and Dukhovskoy,

151 2012). For more detailed descriptions of the system with respect to its temperature, salinity and

152 flow dynamics the reader is referred to (Huang et al., 2002a; Huang et al., 2002b; Huang and

153 Foo, 2002; Morey and Dukhovskoy, 2012)

154

155 *2.1.1 Environmental Characteristics at the study site: Winds, Tidal Amplitudes and River*

156 *Discharge*

157

158 Local wind velocity (5 s data averaged over 15 min intervals) and water level data were

159 collected by the Apalachicola National Estuarine Research Reserve at the Dry Bar monitoring

160 station (Figure 1a). The wind gage was located 10 meters above the water surface. During the

161 study, winds primarily blew from the west on the first 3 days, gradually increasing in strength

162 before weakening and shifting to blow from the south and southeast on Day 4 (Figure 1b).

163 Tides were followed through monitoring water depth, measured by a YSI 6600 EDS

164 multiparameter sonde affixed to the piling 0.3 m from the sediment surface in approximately 2 m

165 of water. The tides were mixed on the first day, gradually trending towards a more semi-diurnal

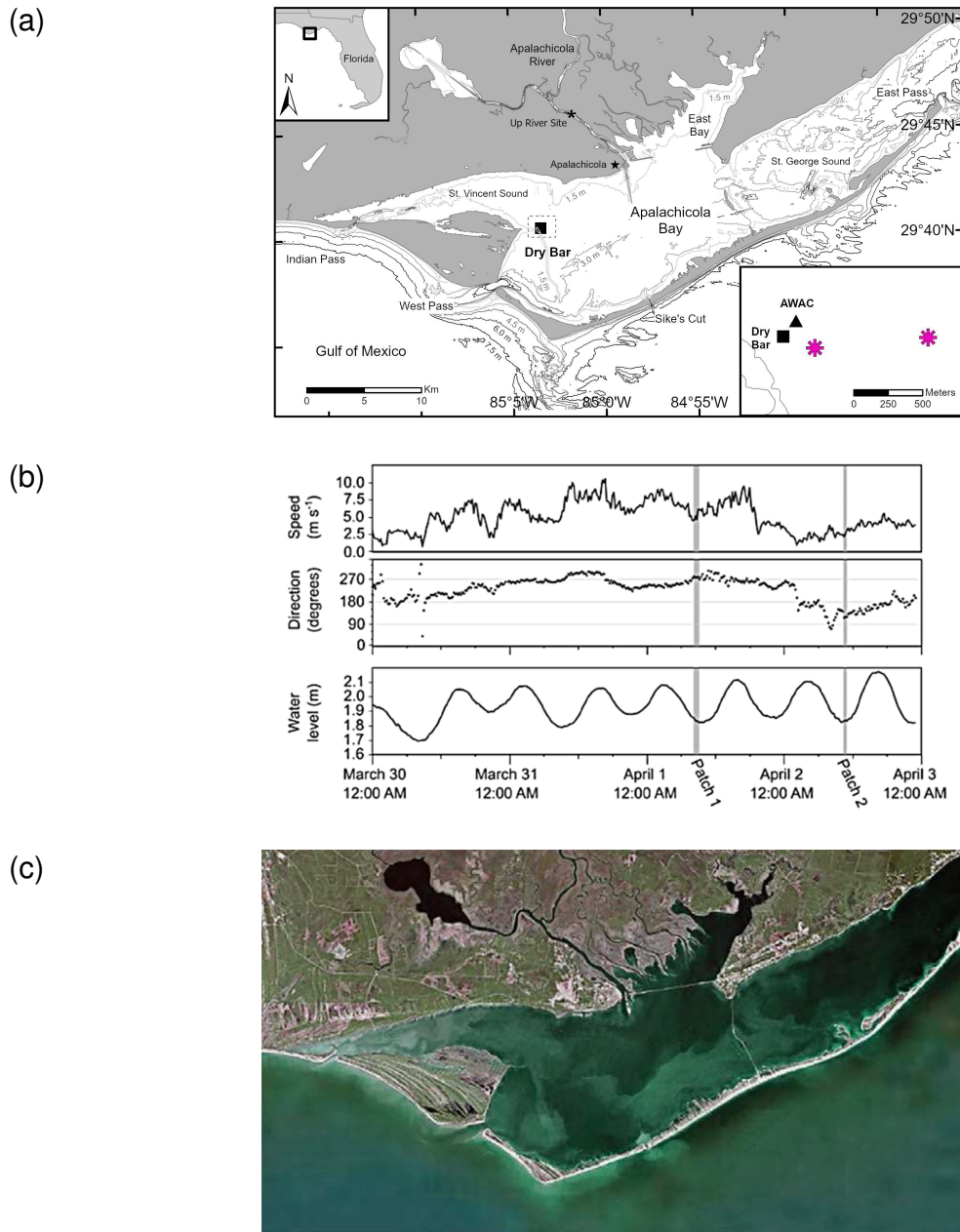166 characteristic towards the end of the experiment (Figure 1b).

Figure 1 (a) Study site in Apalachicola Bay, Florida, and its location in the Northeastern Gulf of Mexico (inset top left). Inset (bottom right) details the dashed square labeled Dry Bar in the main figure and depicts the relative locations of the Dry Bar piling (solid square), AWAC (solid triangle) and dye release sites (pink asterisks). (b) Wind speed and direction (180° indicates wind from the south), and water level at Dry Bar. Grey bars indicate times of dye experiments. (c) Satellite image showing the uneven distribution of phytoplankton and suspended matter in the bay (image: NASA).

167

168   Daily river discharge was retrieved from the USGS river gage (02359170) at Sumatra,

169   Florida (http://waterdata.usgs.gov/nwis/dv), 33 km upstream from the Apalachicola River mouth.

170   During the course of the experiment, river discharge was 635 ± 12 m^3 s^{-1}, which, according to

171   Dulaiova and Burnett (2008), results in a flushing rate of approximately 10 days. River

172   discharge and inflow of Gulf water into the bay during flood generate a complex flow

173   environment (Figure 1c) as well as strong salinity gradients and stratification in AB.

174   Stratification during the study period was primarily controlled by salinity and varied significantly

175   based on location and tidal cycle. A 3-year study conducted in AB by Mortenson (2013) revealed

176   temperature variations with depth $< 1^\circ$ C confirming vertical density variation dominated by

177   salinity differences.

178

179   *2.2.   Analyses of previous data to assess small-scale spatial structure of physical and*

180   *biological tracers in AB*

181

182   To demonstrate how physical processes may contribute to the controls of small-scale

183   estuarine phytoplankton distribution patterns in estuarine settings, we compare the observed dye

184   distribution patterns caused by physical processes to distribution patterns observed in natural

185   phytoplankton distributions. The latter were extracted from data we collected in AB in 2011

186   (Geyer et al., 2018). A direct causal link between physical processes and phytoplankton

187   distribution may only be established through simultaneous measurements of these processes and

188   associated distributions. We therefore emphasize that while our comparison between dye patch

189   evolution measured in 2015 and phytoplankton patches observed in 2011 can provide indications

9

190 of the influence of physical factors affecting estuarine phytoplankton distribution, the 2015

191 numerical results cannot be applied directly to the 2011 AB observations as the time periods of

192 physical and biological measurements differ.

193   In the 2011 measurements, the horizontal distributions of salinity and Chl *a* were

194 recorded at scales of 10's to 1000's meters (0.5 m below surface, 5 s intervals) using a flow-

195 through profiling instrument (DataFlow, Madden and Day, 1992). The georeferenced data had a

196 spatial resolution of approximately 50 m. We analyzed the power spectra of salinity and Chl *a* of

197 five of these transects that each had a length of approximately 23 km and were sampled over a

198 period of approximately 30-45 minutes, each (Figure 2a, b). The tracers measured along the

199 transects first were linearly interpolated to a uniform grid with 50 m spacing, then a slowly

200 varying mean, which was estimated using a moving average with a 3 km window, was removed

201 from the data to minimize the influence of a slowly varying background signal (Franks 2005) or

202 tidal aliasing. Slowly varying background signals were caused by tidal movements and river

203 outflow, which can be seen as anomalously fresh signals in some of the salinity transects (Figure

204 2b). The power spectra were estimated using the high-pass data, after the slowly varying mean

205 had been removed, and a multi-taper spectral estimation technique. These spectra are

206 representative of the spectral characteristics of features smaller than about 3 km in size, which is

207 the scale where we anticipate the largest signatures of the tracer stirring. We used the mspec

208 function in the jSpectral module of the jLab package (http://www.jmlilly.net/doc/jLab.html) for

209 the spectral calculations.

210

211 ***2.3. Observational platforms, data processing, and analysis techniques***

212

213 *2.3.1. Flow measurements by AWAC, Eulerian frequency spectra*

214

215   An Acoustic Current Profiler (Nortek AWAC) was moored for 50 hours (Days 2 – 4;

216 March 31 – April 2, 2015) about 150 m northeast of the Dry Bar Station (29°40.48′ N, 85°03.50′

217 W) in 3 m of water (Fig. 1). The AWAC recorded current velocities once every 5 minutes with a

218 bin height of 0.5 m, providing 4 velocity measurements in a vertical profile between depths of

219 0.5 – 2.5 m. The deepest bin was about 0.5 m above the sediment. The number of samples per

220 burst was 1024 sampled at 4 Hz.

221   The power spectrum, in frequency domain, was calculated using the same methods as

222 those used for calculating wavenumber spectra of tracers. In contrast to the spatial spectrum, this

223 analysis characterizes the variability of the flows over different time periods. For the velocity, a

224 constant time mean and linear trend were subtracted before calculating the spectrum. In addition,

225 we calculated a running average with half window length of 30 min (using the

226 "nanmoving_average" function in Matlab) to present the slowly varying part of the flow.

227

228 *2.3.2. Drifters, Flow Kinematics, and Single Particle Diffusivities*

229

230   **Three Davis-style drifters** (Lumpkin et al., 2017) were deployed and recovered during

231 the flood tide (~08:00 to 13:00 EST) each day at the locations shown in Figure 4. The drifters

232 were 0.5 m wide and reached 0.5 m deep into the water, allowing them to be advected with the

233 top layer of the water column with minimal wind drag. All three drifters were outfitted with a

234 NOAA TrackPack (https://comet.nefsc.noaa.gov/ioos/drift/driftdesign.html#General_Mission)

235 that was set to transmit its GPS coordinates every half hour via satellite communications.

236   Two Garmin DC40 units (3 s refresh rate) were also used to augment the temporal

237   resolution of GPS tracks of two drifters and gather higher temporal resolution data. However, we

238   were unable to recover complete trajectories from drifters due to an error in device setup, thus

239   the higher resolution data from these units were only available for part of the drifter deployment.

240   **Kinematic properties of the horizontal flow** – shear, strain, divergence and vorticity –

241   were estimated using the technique described in Molinari and Kirwan Jr (1975). This method

242   assumes that locally the flow can be described as a mean flow and contributions from linear

243   gradients in velocities (first two terms in a Taylor series). The mean flow is estimated using the

244   mean drift of a cluster of drifters, and the gradients are estimated using a least-squares fit to the

245   differences in velocities of the different drifters. These velocity gradient estimates inform the

246   different kinematic measures of the flow. This method requires simultaneous measurements from

247   at least 3 drifters, and we used the drifter positions from the TrackPacks for these estimates.

248   Further details of this calculation are presented in the Supplementary Material A

249   The velocity gradients that stretch fluid parcels apart are also indicative of the processes

250   active in dispersing the fluid. A measure of the rate of stretching, the longitudinal second order

251   velocity structure functions ($S2_l$), can be calculated from pairs of drifters and quantifies the

252   magnitudes of the stretching velocity as a function of the distance between the fluid parcels

253   (Babiano et al., 1990).

254
$$S2_l(r) = < \delta u_l (r)^2 >$$

255

256   Where <.> indicates averaging over all pairs of drifters that are within a separation distance, r,

257   from each other. δu_l is the longitudinal velocity difference, where longitudinal is in the direction

258   of the axis joining the two drifters.

259   **Single particle diffusivities** were estimated using the method outlined in Suara et al.

260   (2016) following the techniques from Davis (1991). The velocity from a single drifter is

261   decomposed into the large-scale flow component (which can be a spatial or temporal mean) and

262   the residual eddy flow (which is the component whose effects are meant to be represented by an

263   eddy diffusivity). The residual eddy velocity time series are used to calculate the characteristic

264   eddy energy - velocity variance (u'^2), and a Lagrangian time scale (T_L) from the lagged

265   autocorrelation. The eddy diffusivity is defined as $K = (u'^2) T_L$. Further details of the diffusivity

266   calculation and sensitivity to definition of mean flow and drifter position processing are

267   presented in the Supplementary Material B.

268

269

270   *2.3.3. Dye experiments*

271

272   A neutrally buoyant solution of Rhodamine WT (RWT) was released and monitored with

273   aerial photography to visualize a simulated phytoplankton patch. RWT is a bright pink dye

274   commonly employed to trace and visualize physical transport in water on short spatiotemporal

275   scales (Bogucki et al., 2005; Clark et al., 2014; Mirfenderesk et al., 2007). This fluorescent dye

276   was chosen because of its relatively low toxicity, favorable chromatic properties, low particle

277   affinity and slow photolysis rates (Smart and Laidlaw, 1977; Suijlen and Buyse, 1994). To keep

278   the dye in the surface layer that had a density of $1.007 - 1.019$ g cm^{-3}, dry RWT dye powder

279   (39% RWT by weight) was mixed with fresh water at a ratio of 3 g L^{-1} to yield a solution with a

280   density of 1.003 g cm^{-3}. Two dye release experiments were conducted on Days 3 and 4. Each

281   experiment was initiated by partially submerging a bucket of dye solution to allow the dye to

282   enter the surface water with minimal mixing during the release. A total of 68 L dye solution was

283   deployed with 4 buckets within ~2 minutes. Two drifters were released concurrently with the dye

284   solution; one at the beginning of the release and one after all the dye was deployed.

285   A DJI Phantom 2 Vision+ drone captured aerial photographs of the spreading dye patch

286   beginning at 1 – 6 min after completion of the dye releases, and then at 30-second time intervals.

287   The 14 megapixel camera was angled downward to zero degrees and recorded images in RAW

288   format. Drone flight software provided altitude data, and the onboard GPS allowed the drone to

289   maintain its position with a 0.8 m vertical and 2.5 m horizontal hover accuracy (DJI, 2017). All

290   images were stamped with a center point GPS location in WGS84 datum (error ± 3 m). Aerial

291   imagery was pre-processed in Adobe Lightroom and ImageJ, then georeferenced in ArcGIS

292   (details of processing are in Supplementary Material C).

293   The scaling of the drone camera photographs was calibrated using images of markers

294   spaced at known distances. The drifters were visible in the dye patch pictures, providing an in-

295   situ spatial calibration reference.

296   Dye traveling distances were measured using the displacement of the visually distinct

297   leading edge of the dye patch in consecutive pictures. Lateral advection velocity of the dye patch

298   was calculated by dividing dye travel distance by the length of the time interval between

299   pictures. The mean advection calculated with this method agreed with the mean advection

300   calculated using the approximate centroid of the dye patch.

301   During the dye experiments, salinity at 0.5 m depth was measured with the Dataflow

302   flow-through instrument. The boat speed during the measurements was ~3.5 km h^{-1} and the

303   transect lengths across the dye patch were 200 to 400 m long producing data with a spatial

304 resolution of approximately 5 m. The analysis of the georeferenced data was completed in

305 ArcGIS. In addition, vertical salinity profiles were conducted with a YSI 6600 Sonde.

306

307 **3. Results**

308

309   In this section, we first present results from the analysis of the data reported by Geyer et

310 al. (2018) with focus on salinity and Chl *a* small-scale spatial distributions in AB. We then

311 explain current characteristics revealed by the AWAC measurements before addressing the

312 results of the Lagrangian experiments. These drifter and dye experiments reveal details of the

313 small-scale processes that influence the development of patchiness.

314

315 *3.1. Characteristics of spatial Salinity and Chl a distribution based on data reported by Geyer*

316 *et al. (2018)*

317

318   The measurements of physical, chemical, and biological variables collected along repeat

319 transects in the bay by Geyer et al. (2018) revealed distinctive spatial and temporal variability of

320 these variables. Here we present a wave-number (1/ wave-length), or inverse length-scale based

321 characterization of the spatial variability of salinity and chlorophyll along the longest repeat

322 transect, which extended from East Bay to Dry Bar (0 km to 23 km, respectively, in Figure 2a-b).
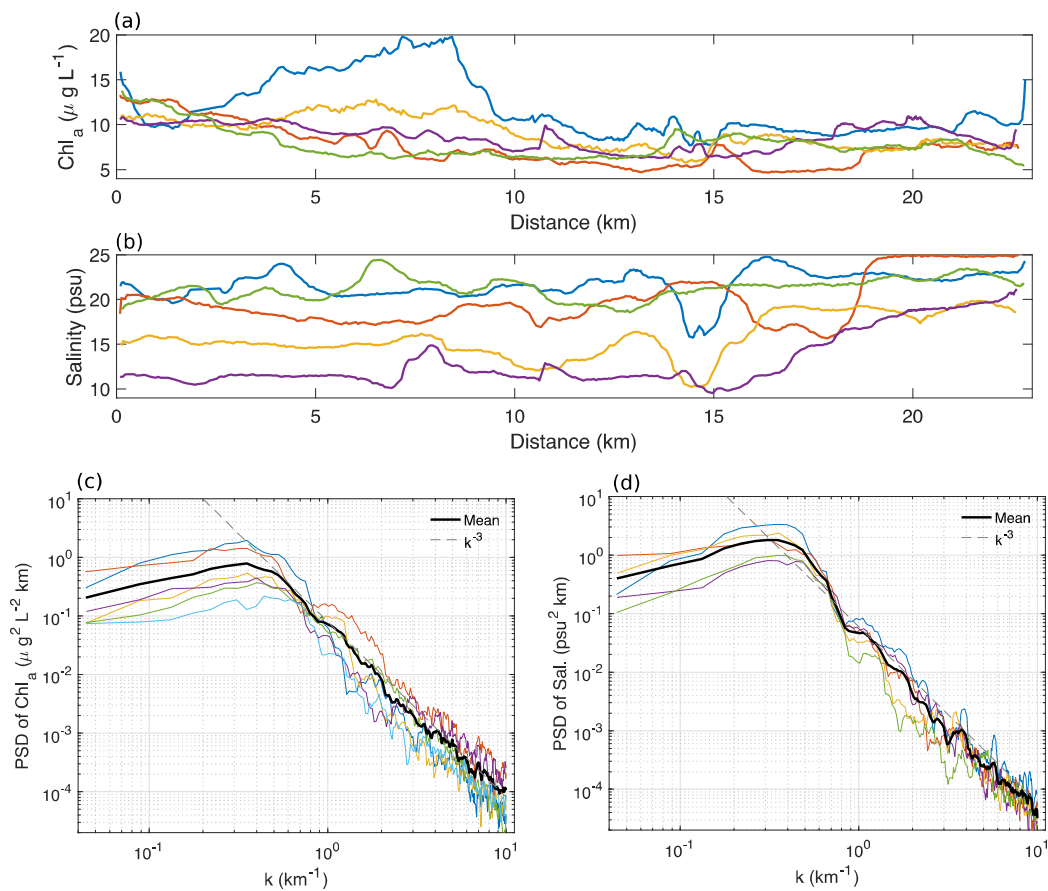
Figure 2 (a, b) Data recorded by (Geyer et al. 2018) of chlorophyll and salinity along the transects from East Bay (0 km) to Dry Bar (23 km) that were analyzed for this study. (c, d) Spectra of Chl *a* and salinity; thin colored lines are the power spectra from individual transects and thick black line is an average.

Both salinity and Chl *a* have a large scale structure defined by the fresher and more chlorophyll rich water in East Bay (Geyer et al., 2018). The small scale variability of this large scale structure can be characterized through a wavenumber power spectrum (Figure 2c-d), which here follows a power law behavior, with an exponent of -3, at scales of 0.1 – 3 km. The slight elevation in the salinity spectrum above the power law behavior, at scales around 2 – 3 km, is caused by the presence of the freshwater river plume in the transects.

335　　　　Stirring by the flow in the estuary cascades tracer variance from large scales to small

336　　scales, breaking large filaments into smaller filaments. At the smallest scales, these tracer

337　　variances are removed by molecular diffusion. Simple models of turbulent flows, which assume

338　　homogeneity, isotropy, and statistical stationarity, predict the tracer variance to have power law

339　　behavior with slopes in the log-log plot (exponent of the power law), between -2 to -1. Steeper

340　　tracer spectrum slopes are indicative of the flow having more kinetic energy at smaller scales,

341　　which is reflected in a flatter kinetic energy spectrum encompassing higher wavenumbers before

342　　dropping off. The spectral slope of -3 is relatively steep, suggesting that flows at small scales in

343　　the bay are energetic. In an inhomogeneous Bay environment, mean flows with complex spatial

344　　structures may result from the presence of bottom topography at small scales and complex lateral

345　　coastal boundaries, with localized tracer sources and sinks.

346　　　　Simple models that rely on homogeneity, isotropy, and smoothly varying background

347　　tracer structures, therefore, might not be applicable in AB. Nevertheless, spectral behaviors of

348　　salinity (passive tracer) and Chl *a* (reactive tracer) collected along the repeat transects in the bay

349　　by Geyer et al. (2018) were very similar, suggesting that at these length scales the dominant

350　　controls structuring tracer distributions were physical.

351

352　***3.2. Currents***

353

354　　　　The AWAC data provided information about the general properties of the currents at the

355　　study site (Figure 3a-b). The time series underline the strong imprint of the tidal modulation,

356　　with a typical amplitude around 30 cm s^{-1}. There was substantial vertical shear in the flow, with

357　　numerous instances when the near surface and near bottom flows moved in opposite direction.

358   During the tracer experiment, the flow at 1.5 m above the bottom was preferentially

359   towards the north-north-west, with signs of a strong south-south-east return flow at heights of 2.0

360   and 2.5 m. This slight preference for north-south flow may be a result of bottom topography,

361   which has contours running north-south near this location, orienting the flow in the same

362   direction.

363   The velocity variance generally decreases with depth/increases towards the surface

364   (Figure 3c). The variance of flow increased slightly in the bottom bin, due to the generation of

365   turbulence in the bottom boundary layer. Variance in the north-south motions was stronger than

366   east-west motions, likely a sign of topographic orientation of the flow.

367   In the frequency power spectrum of the currents (Figure 3d), tidal motions dominated

368   with time scales longer than approximately $6 - 7$ h (frequency: $5\times10^{-5}\,\mathrm{s^{-1}}$). There is a steep drop

369   in power at periods around $4 - 6$ h, followed by a semblance of a -5/3 power law at periods in the

370   range of $\sim1 - 4$ h ($8\times10^{-5}\,\mathrm{s^{-1}}$ to $3\times10^{-4}\,\mathrm{s^{-1}}$). This steep drop off supports that the slower motions,

371   tidal and low-mode wind driven, are significantly dominant (at least 1 order of magnitude in

372   power) over processes that might be generating motions at faster time scales, such as waves,

373   hydraulic bores, 3D turbulence, etc. This is at least true at 2.0 m (shown here) and 2.5 m (not

374   shown); the drop off was less steep at 1.5 m (not shown).
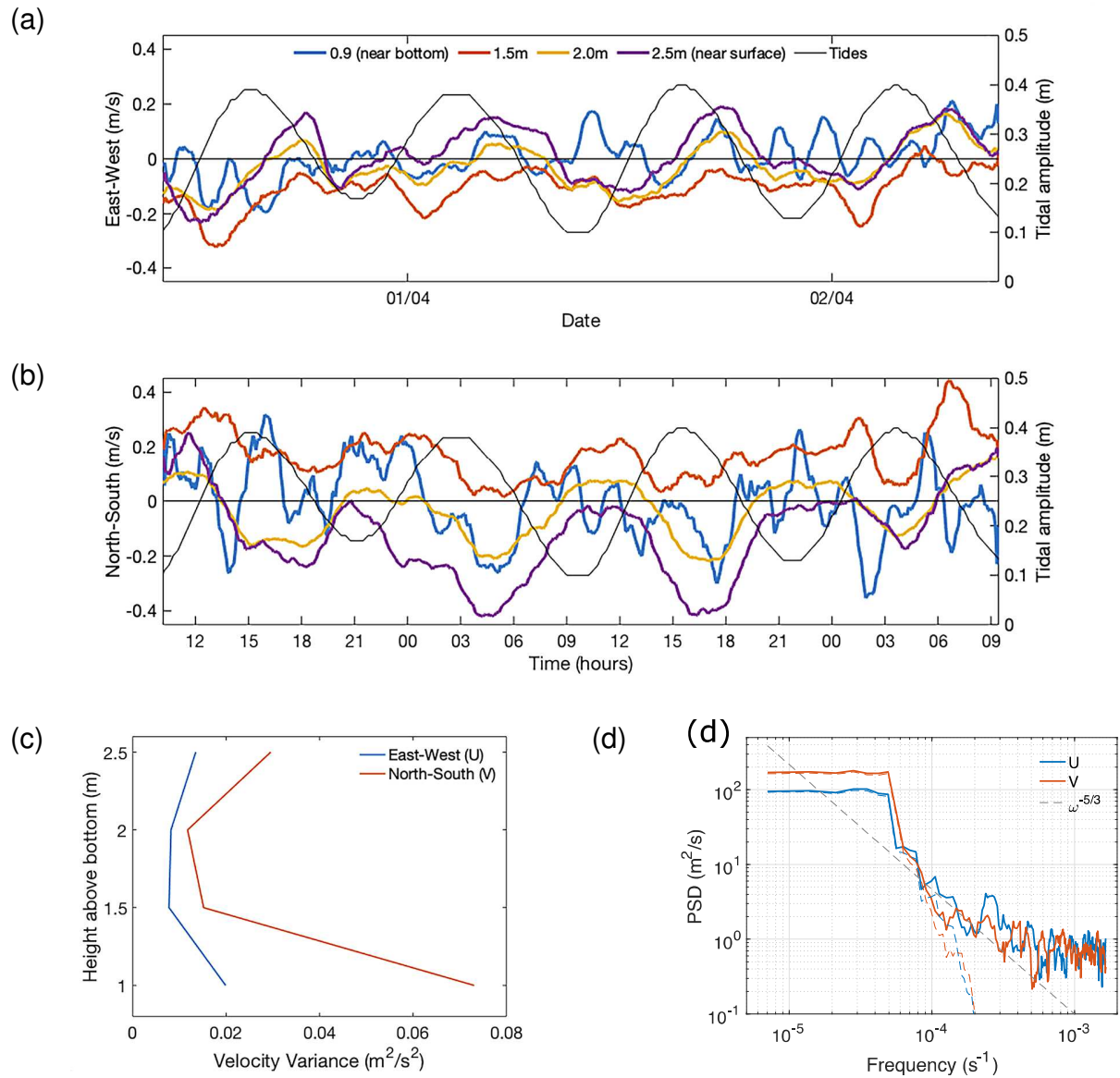
375

376

Figure 3 (a, b) East-West (U) and North-South (V) velocities at different depths near Dry Bar, calculated from ADCP data, with a 30 min running mean filter applied. The height above the bottom is noted in the legend. (c) Velocity variance as a function of height above bottom (distance from AWAC + 0.1 m). (d) Frequency spectrum of the U and V velocity at height of 2.0 m. Red and blue dashed lines show the power spectrum of the smoothed velocity time-series shown in panels a and b; the dashed gray line marks the -5/3 power law. The behavior was qualitatively similar at 1.5 m and 2.5 m.

377

378

379   The -5/3 power law behavior is the same as what would be expected for 3D turbulence,

380   but occurs at periods that are physically too long for 3D turbulence to be playing a role. This

381   power law behavior is probably indicative of internal waves riding the stratification in the bay,

382   which do produce qualitatively similar spectral slopes in the deep ocean (Garrett and Munk,

383   1972). At the highest resolved frequencies, or periods < 0.5 – 1 h, the spectrum flattens out,

384   indicating that the noise floor of the instrument as configured has been reached. A time of 0.5 h

385   (30 min) was hence used to smooth (moving average) the time-series and obtain the dominant

386   signal with minimal influence from noise for visualization. The spectra for the smoothed signal

387   in Figure 3a-b, is shown as dashed lines in Figure 3d, demonstrating that the smoothed (low-

388   pass) signal is primarily composed of the strong flows associated with tides and slow persistent

389   flows generated by the winds.

390

391   *3.3. Drifters*

392

393   *3.3.1. Flow Patterns*

394

395   The drifter tracks revealed large-scale circulation patterns during the time of the

396   experiment (Figure 4). On Days 1 – 3, the drifters moved to the east, then on Day 4, they drifted

397   northwestward, towards Indian Pass, with velocities ranged from 0.03 – 0.6 m s^{-1}. The smooth

398   nature of the drifter trajectories suggested that most of the kinetic energy was in the longer

399   spatial scales, rather than small-scale flow features. This spatial smoothness is complementary to

400   the Eulerian frequency spectra (Figure 3d), which showed a sharp drop off in energy at time

401   scales faster than 4 – 6 h. Similar to the frequency spectrum, this suggests that there are strong

402 flows steered by bathymetry, which have a large spatial scale structure that is forced by winds

403 and tides. Overlain are much weaker small-scale currents that are driven by wind driven waves,

404 stratification, and bathymetric interactions.

405

(a)

406

(b)   (c)

407
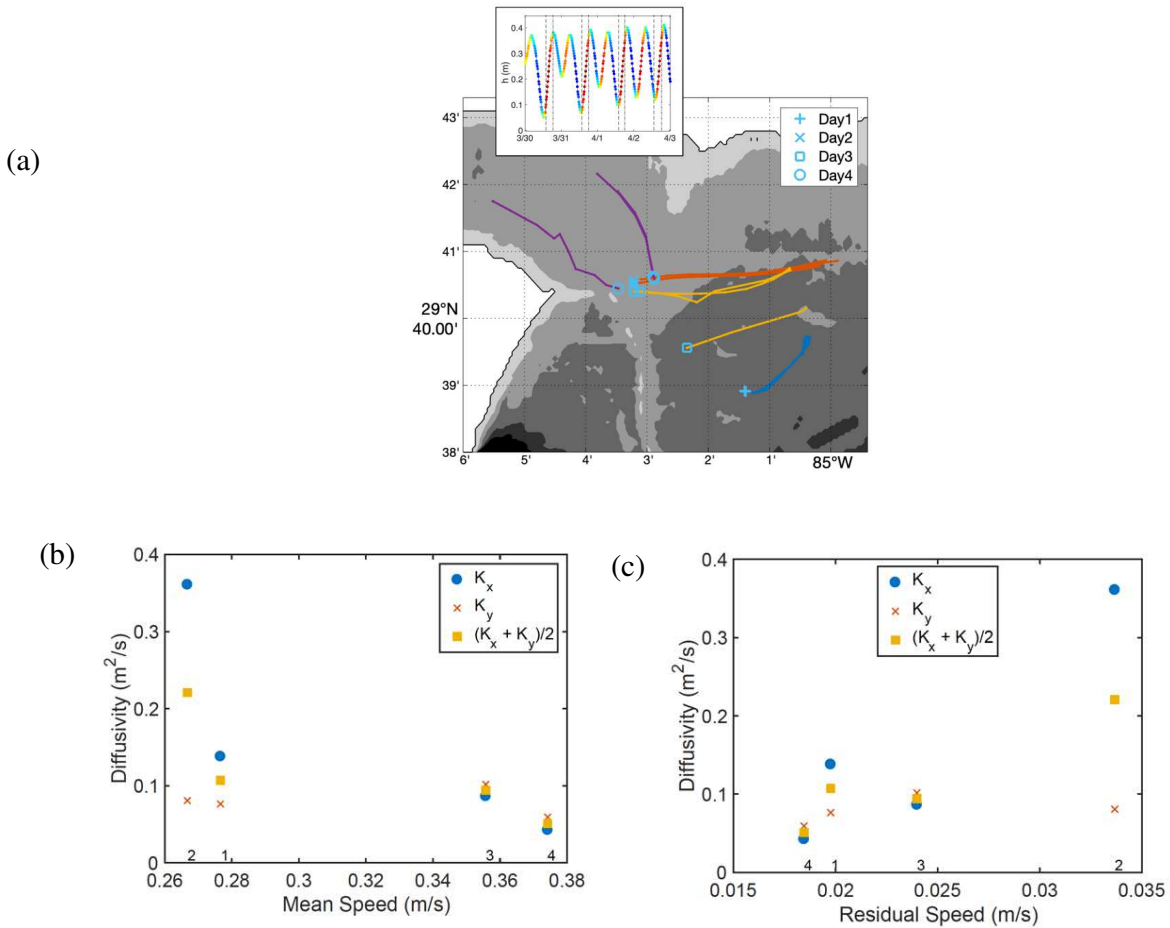
Figure 4. (a) Drifter trajectories plotted in color for different days, with release locations labeled by unique markers

for each day. Drifters 1 and 2 were from the sets of drifters on Days 2 and 3 that went to the east, and drifters 3 and

4 are the eastern pair of drifters released on Day 4 that went to the north (also see Figure B1). Bathymetric contours

(1 m levels) are shown in gray shading. Inset plot depicts the tidal signal, with the duration of the float releases

(always during the flood tide) marked by thin dashed vertical lines. (b, c) Drifter diffusivities from the GPS drifters,

as a function of mean speed (b) and residual speed (c). The drifter number is indicated at the bottom of the plots for

each data point. K_x and K_y are the diffusivities in the east-west and north-south directions.

3.3.2. Eddy Diffusivities

We used the four drifter trajectories with the higher resolution GPS sensors to estimate eddy diffusivity using the autocorrelation of the residual velocity time series (details of the calculation are presented in Supplementary Material B and high-resolution trajectories are shown in Figure B1). Figure 4 b,c shows the eddy diffusivity estimates as a function of the speed of the mean, or large-scale flow, and the residual or eddy speed. Here the mean was defined as a running average with half window width of 1 hour, further discussion of this choice and plots of mean and residual velocity are in Supplementary Material B. The characteristic magnitude of the eddy diffusivity was 0.1 m^2 s^{-1}. While the data set is limited, there is some suggestion that the eddy diffusivity increased with increasing eddy speeds and decreased with increasing mean flow speeds. Two of the four drifters (drifters 1 and 2), which traversed from west to east, show an enhanced zonal diffusivity in the direction corresponding to the mean flow, suggesting the influence of shear flow on dispersion. Drifters 3 and 4 show isotropic diffusivity, not enhanced in any particular direction. For these two drifters the two components of diffusivities remain isotropic, even if the diffusivities are decomposed in along and across mean flow directions (not shown).

3.3.3. Kinematics

As the drifters were released in triplets, we were able to use a least squares approach to estimate the local velocity gradients, and estimate the shear, strain, vorticity and divergence (method described in Supplementary Material C). Histograms of these quantities are shown in

438   Figure 5a and are quite similar. A median value of 5×10^{-4} s^{-1} was observed, indicating that the

439   Rossby number (vorticity/f) in the bay is larger than 1 and the flow is not dominated by rotation

440   as is the case in the open ocean. However, the magnitudes of these gradients suggest that the

441   velocity gradients in the bay are not very strong relative to the size of the bay. For a sinusoidal

442   spatial distribution, a velocity scale of 0.2 m s^{-1} and a gradient of 5×10^{-4} s^{-1} (median of the

443   histogram) implies a wavelength of almost 2 km, which is about a quarter of the bay width. This

444   scale is close to that of the peak in the energy spectra.

445   The longitudinal second order velocity structure function, which quantifies the velocity

446   that stretches fluid particles apart, is shown in Figure 5b. As there were very few drifter

447   deployments, $S2_l$ is quite noisy but shows a modest increase as the separation distance increases.

448   This suggests that the drifters were being separated at an approximate velocity of 2 cm s^{-1} at

449   separations of 10 m, and 10 cm s^{-1} at separations of 500 m – 1 km.

450

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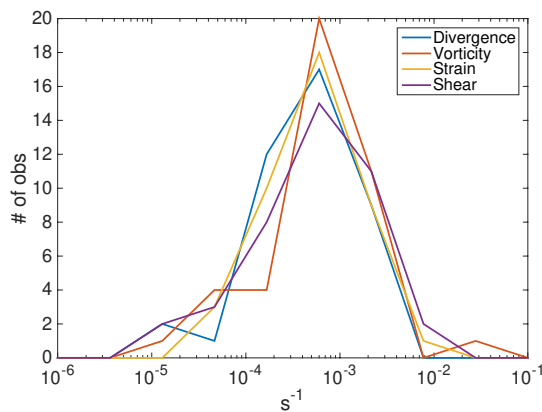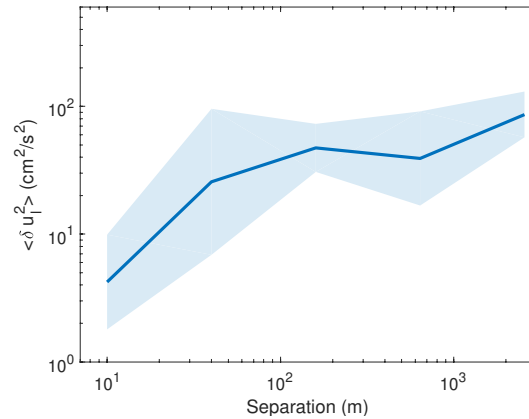
452   (a)   (b)

453   **Figure 5** (a) Histograms of divergence, vorticity, strain and shear from drifter tracks. (b) The longitudinal structure

454   function (longitudinal velocity difference squared) as a function of separation distance.

455

456   *3.4. Dye Transport*

457

458   *3.4.1. Patch 1*

459

460   The first dye release (Patch 1) was initiated on Day 3 (1 April 2015) in the immediate

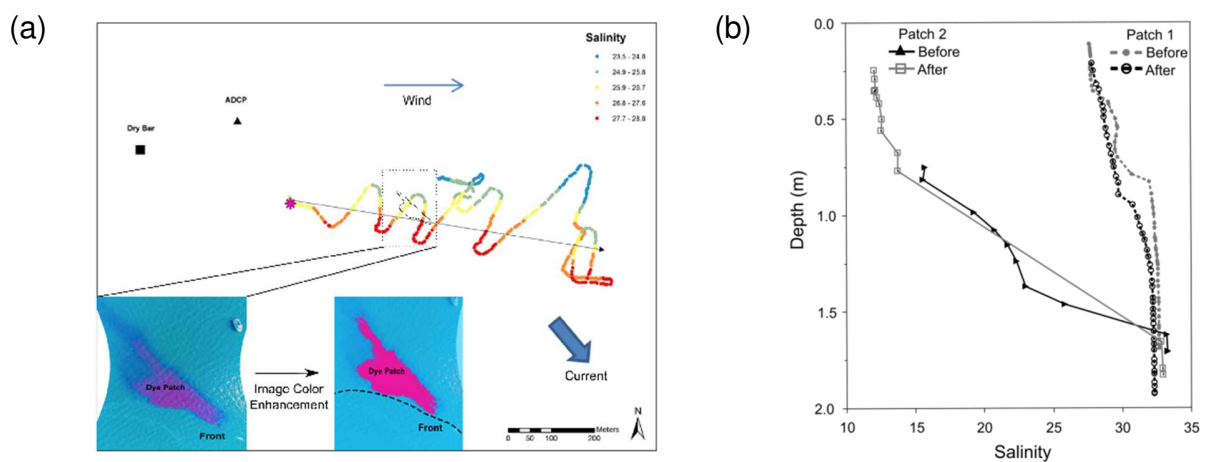461   vicinity of a visible flotsam line associated with an incoming tidal front (Figure 6a).

Figure 6 (a) Map of dye release experiment 1. Pink asterisk: Patch 1 release location. Long black arrow: advection trajectory of the dye patch. Multicolored line: Salinity of water 0.5 m below the surface, with color representing salinities as listed in the upper right corner of the map. Data were collected with an onboard flow-through instrument and indicated the presence of a salinity front. Left inset: Example of drone imagery. Right inset: Same image after contrast enhancement; dashed line denotes the position of flotsam line (visible in left inset) along the density gradient. The current and its direction were calculated from the AWAC data (b) Vertical stratification of the water column immediately before dye release and after drone monitoring of the dye patch concluded.

462

463   The surface salinity between opposite sides of the front differed by about 4 ppt. The

464   water column was vertically stratified with a salinity difference of 5 between surface (28) and

465   bottom (33 at 1.7 m) (Figure 6b). During the deployment, winds blew from the west at 5 – 6 m

24

466   s^{-1}, and surface currents were oriented towards the southeast at 0.14 ± 0.06 m s^{-1}. Flow at 1.5 m

467   above the bottom was preferentially towards the north-north-west, with signs of a strong south-

468   south-east return flow at 2.0 and 2.5 m from the bottom. This slight preference for north-south

469   flow may be a result of bottom topography, which has contours running north-south near this

470   location, orienting the deeper flow in the same direction.

471   The dye was released at 08:18 EST. Eleven minutes after the release (08:29), the area of

472   the dye patch measured approximately 380 m^2. The patch was slightly elongated on its NNW –

473   SSE axis, but it maintained a roughly ellipsoid shape throughout the experiment. The dye patch

474   moved towards the east, and streaks of dye radiated away from the flotsam line (Figure 6a,

475   insets). The dye patch continued to spread to the east and developed a sharp boundary on the

476   south-side of the patch along the flotsam line of the front.

477   Around 17 minutes after dye release (08:35), the dye patch started to separate from the

478   flotsam line, while remaining cohesive and parallel to it. By 08:49 the dye patch size had

479   increased to an area of approximately 540 m^2, corresponding to a growth rate or diffusivity of

480   0.134 m^2 s^{-1}. Over the course of the next 20+ minutes, the patch continued to disperse and

481   became less distinguishable. Drone photography ended 52 minutes after the dye release (09:10).

482   Over the course of the observation, the dye patch moved ~ 770 m to the east in 52 minutes, at an

483   average speed of 0.3 ± 0.1 m s^{-1} (n = 7).
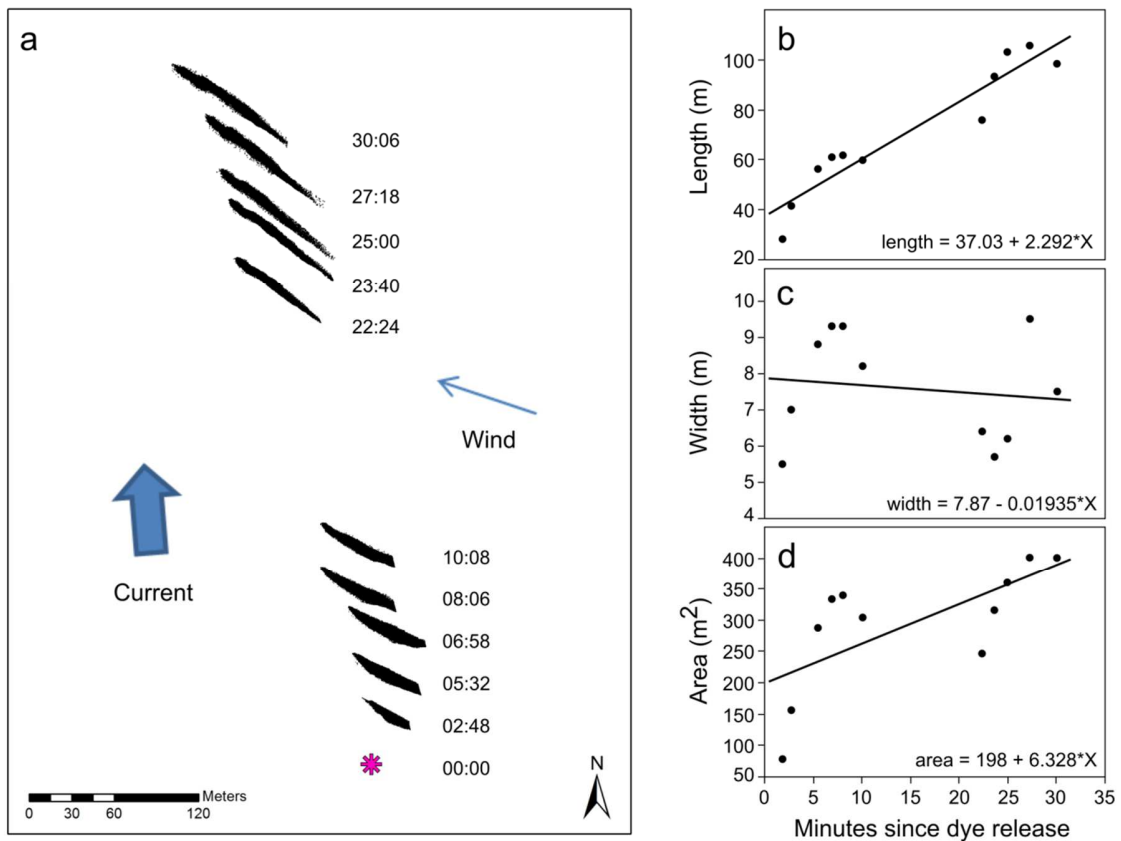
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489   *3.4.2. Patch 2*

490

491   The second dye release (Patch 2) started at 10:29 EST on Day 4 under relatively calm

492   conditions with wind from the ESE at 2 – 3 m s^{-1}. The water column was highly stratified –

493   surface salinity was 10 and bottom (2.0 m) salinity was 32 (Figure 6b).

494

495   **Figure 7** (a) Map of Patch 2 advection and dispersion. The dye was released at the location of the pink asterisk and

496   tracked to the NNW for 31 minutes. Black shapes represent the shape and location of the dye patch at the indicated

497   time (minutes: seconds) after the release. The gap in the middle of the time series was when the drone battery was

498   changed. The first five patch images had a small amount of the dye 'tail' cut off by the edge of the picture. Graphs b,

499   c, and d show changes in the patch dimensions over time. Changes in (b) length (p < 0.0001, R^2 = 0.9), (c) width (p

500   = 0.6, R^2 = 0.02), and (d) area (p = 0.02, R^2 = 0.47).

501

26

502   The AWAC was retrieved an hour before the Day 4 dye release; however, 1 - 2 hours

503   before the dye release surface currents were flowing towards the N – NNW at 0.17 ± 0.06 m s^{-1}

504   (range: $0.07 – 0.24$ m s^{-1}). Immediately after the dye release (10:33), the area of the dye patch

505   measured 78 m^2. The patch developed an ellipsoid shape with a major axis (30 m long) oriented

506   northwest to southeast (minor axis 6 m) (Figure 7a-c).

507   The patch began to stretch along that same axis leading to an elongated shape with a

508   dense kernel of dye at the northwest patch edge and a much larger but less dense filament

509   towards the southeast. Over the next 15 minutes, the dye patch continued to elongate as it was

510   advected to the NNW, while its visible width decreased (Figure 7b-c). The dye patch area

511   initially grew linearly at a rate (diffusivity) of 0.71 m^2 s^{-1} (43 m^2 min^{-1}) (Figure 7d). Drone

512   photography continued for 31 minutes and ended at 11:00 EST when the dye patch became

513   visibly poorly defined. This dye patch moved ~ 500 m to the NNW in 31 minutes, at an average

514   speed of 0.3 ± 0.1 m s^{-1} (n = 10).

515

516   **4. Discussion**

517

518   Our study demonstrates how hydrodynamics can control small-scale tracer distribution in

519   AB and provides insights on how these physical processes may influence phytoplankton patch

520   characteristics in shallow estuarine environments. Divergence, diffusivity, vorticity, shear and

521   strain rates produced from our measurements can be implemented in ecological models of AB

522   phytoplankton development and help us understand and quantify primary production in this

523   estuary. Likewise, these rates can be applied to estuaries with similar environmental settings

524   (e.g., the numerous northern Gulf estuaries) facilitating improved estimates of the contribution of

525 these estuaries to the productivity of the Gulf, which affects its fisheries, oxygen dynamics and

526 local economies.

527

528 *4.1. Effect of flow properties on patchiness*

529

530 By definition, plankton cannot move against water flow, and the influence of complex

531 estuarine current settings on phytoplankton distribution thus can be investigated with drifter and

532 dye deployments. The drifters simultaneously traced the currents at multiple locations within AB

533 and revealed how the flow can affect phytoplankton patchiness by stirring or stretching fluid

534 parcels apart. The drifter trajectories and velocity frequency spectra from the AWAC suggest

535 that flow in AB can be decomposed into a large-scale component driven by tides and winds,

536 superimposed by small scale flow variability resulting from density fronts, surface and internal

537 waves, 3D turbulence, bathymetry and coastlines. Ensuing stirring rates, determined from

538 horizontal shear and strain rates, were on the order of $10^{-4} - 10^{-3}\,s^{-1}$, or $2 - 3$ orders of magnitude

539 greater than typical open ocean strain rates ($10^{-6}\,s^{-1}$, e.g., Martin, 2003; Sundermeyer and

540 Ledwell, 2001). This substantially enhanced stirring underlines the difference between the

541 estuarine and ocean settings critical for phytoplankton patch dynamics. The stretching rates

542 ranged from $2 - 10$ cm s^{-1} at scales of $10 - 1000$ m, which can potentially separate two fluid

543 parcels over the entire width of AB in a single tidal cycle.

544 The diffusivity estimates from the drifters and dye ranged from 0.1 to 0.4 m^2 s^{-1} and 0.1

545 to 0.7 m^2 s^{-1}, respectively, corresponding to length scales of a few 100s of meters (dye patch

546 size) and time scales shorter than about an hour (following the definition of the mean path for the

547 drifters). The agreement of the values obtained with the two independent methods strengthens

548 and confines the calculated diffusivity. Bogucki et al. (2005) estimated a similar horizontal

549 diffusivity of 0.1 m^2 s^{-1} on scales of 10 m from aerial imagery of dye dispersion in an

550 embayment. Our estimates are also broadly consistent with the diffusivity compilation in Okubo

551 (1971), but smaller than canonical estimates from coastal (Rypina et al., 2016) and open ocean

552 regions (Balwada et al., 2016), for which larger values can be expected.

553 Despite the rapid stretching rates indicated by the strain field, the concentration

554 anomalies associated with small-scale fronts of environmental tracers (salinity and chlorophyll)

555 were relatively small compared to the gradients associated with large scale filaments and

556 gradients that develop as water mixes from the river, East Bay, and the Gulf of Mexico. This is

557 complementary to the results of Geyer et al. (2018), who found that the small chlorophyll

558 filaments correspond to about 10% of the biomass in the bay. This relatively small variability at

559 small scales may be explained by the rapid flushing times of the bay (6 – 12 d); i.e., the larger

560 filaments are flushed out of the bay before they can break into smaller filaments. High-resolution

561 modeling studies of the bay or a study that follows chlorophyll filaments over a longer time

562 period would be of value to test the validity of this hypothesis.

563

564 *4.2. Physical Dispersal Revealed by Dye Patch Evolution*

565

566 In the typically turbid estuarine waters where light may penetrate only a couple of meters,

567 the surface water layer is central for phytoplankton development, and our aerial imagery of dye

568 patch development provided insights into physical processes that can control plankton

569 distribution and dispersal (Bogucki et al., 2005; Clark et al., 2014). While the dye images

570 provide limited information on the vertical distribution of the dye (~ upper 10 cm), they allow

571 the exploration of lateral transport, stirring and diffusivities. Patch 1 demonstrated how two

572 physical processes affect patch evolution in the shallow estuarine environment in opposite ways:

573 shear – causing gradients in transport – produced filaments and diffuse patch boundaries, while

574 density fronts – limiting transport in across-front direction – caused sharp patch boundaries

575 (Figure 6a). The combination of these processes resulted in a patch with asymmetrical shape and

576 concentration gradients, and can help explaining the asymmetrical Chl *a* peak that was observed

577 in AB by Geyer et al. (2018) along the Apalachicola River plume front (Figure 8).

578
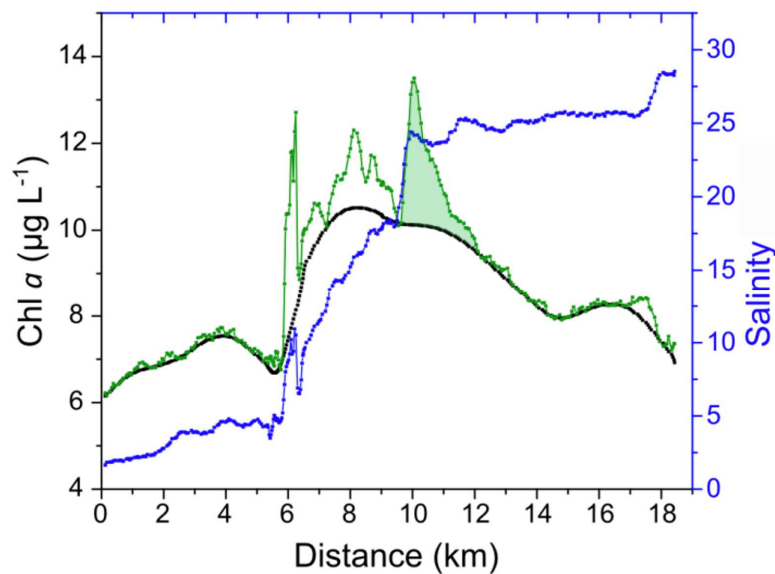
579

580 **Figure 8** Asymmetrical Chl *a* patch (green shaded area at distance 9.5 - 12 km) at a salinity front in AB. Chl *a*

581 (green) and salinity (blue) were measured along a transect from the Up River site (0 km) to Dry Bar (19 km)(see

582 Figure 1a for station locations). The black baseline delineates large-scale Chl *a* distribution. These Dataflow

583 measurements were conducted on August 29, 2011, within an associated research study reported by Geyer et al.

584 (2018).

585

586   Observations of such phytoplankton concentration gradients at fronts were explained with

587 passive accumulation or enhanced growth stimulated by shear-induced nutrient fluxes (Franks,

588   1992; Largier, 1993, Dustan and Pinckney, 1989). Our dye experiment demonstrated that

589   physical processes alone can produce these distribution patterns (Figure 6a).

590

591   **Patch splitting by vertical shear.** Patch 1 moved in an ESE direction despite a general

592   current flow towards the SE. This was caused by the westerly wind, moving the surface water

593   layer in ESE direction and generating shear between surface and subsurface layers that led to the

594   formation of dye tracer filaments at the northern trailing edge of the dye patch (Figure 6a insets).

595   This substantial vertical shear is also reflected in the AWAC velocity measurements (Figure 2a-

596   b). The feathering and filament formation seen on the photographs was partly caused by dyed

597   subsurface water moving in the SE direction, separating from the ESE moving surface layer.

598   Shear processes producing water layers moving in different directions thus can separate a well-

599   confined phytoplankton patch into two patches, one in the surface layer, and the other in the

600   subsurface layer. Such spreading due to vertical shear so far has not been addressed in AB

601   models as these models so far simulated vertically integrated flow (e.g., Huang et al., 2002a;

602   Huang et al., 2002b).

603   The evolution of Patch 2 underlines the role of wind-induced surface layer movement on

604   patch development. In contrast to Patch 1, Patch 2 was influenced by wind and current moving in

605   similar directions. The northward advection of Patch 2 by water currents was modulated by wind

606   from the ESE such that the patch was stretched into a long ribbon shape (Figure 7a). The leading

607   northwest edge of the dye patch remained a more intense red, tracing the surface layer, which

608   was pushed towards the WNW by the wind more quickly than the dye in the southeast 'tail' of

609   the patch that was entrained in slightly deeper water, again evidenced by the shift in color from

610   red to blue.

31

611   **Patch widths in straining flow.** The stretching of Patch 2 by coherent wind and current

612   movement led to a narrowing of the patch and the question arises whether this narrowing would

613   continue and what ultimately the limitations of this process are. A minimum phytoplankton patch

614   width has been theorized to exist for a patch filamented by a straining flow (Garrett, 1983;

615   Martin, 2000; Sundermeyer and Price, 1998), and this width can be determined if the diffusivity

616   and strain rate are known, such that

617     $$L \; = \; (K \, / \, \gamma)^{1/2} \; ,$$

618   where L is the minimum patch width, K is the horizontal eddy diffusivity, and γ is the

619   mean strain rate of the flow. Using the minimum observed Chl *a* peak width of O(100 m)

620   observed by Geyer et al. (2018) as the 'minimum' peak width and a strain rate on the order of 10^{-4}

621   or O(10^{-4}) derived here, we can solve for the effective diffusivity, which is estimated as O($1 \, m^2$

622   s^{-1}). This is slightly higher than K at a similar scale O($0.5 \, m^2 \, s^{-1}$) calculated from the drifter

623   dispersion and dye spreading, which is to be expected as the minimum widths observed by Geyer

624   et al. (2018) may be limited by the sampling spatial resolution.

625   The phytoplankton population growth rate, μ_{net}, does not affect the minimum patch

626   width, as long as $\mu_{net} / \gamma < 2.5$, otherwise growth rates are expected to affect the steepness of the

627   slope as it relates to diffusion and strain (Martin, 2000, McLeod et al., 2002). Growth rates in AB

628   have been estimated as $\mu_{net} = 0.08 – 1.92 \, d^{-1}$ (Putland and Iverson, 2007), suggesting that growth

629   does not have substantial effect on the slopes of the phytoplankton peaks in the straining flow we

630   observed (all combinations of $0.08 – 1.92 \, d^{-1} / 10^{-3} – 10^{-4} \, s^{-1} < 2.5$). This is also in agreement with

631   the spectral analysis of chlorophyll and salinity concentrations, showing that the spatial

632   structures of these tracers are potentially steered by similar dynamics. For μ_{net} / γ to be greater

633   than 2.5, μ_{net} would need to exceed $20 \, d^{-1}$ to overcome strain rates of $10^{-4} \, s^{-1}$. Whereas a

634   phytoplankton community with $\mu_{net} = 1.92 \text{ d}^{-1}$ divided by a smaller strain rate, such as that

635   observed in the open ocean (10^{-6} s^{-1}), would be expected to influence a patch's structure in a

636   straining field ($1.92 \text{ d}^{-1}/ 10^{-6} \text{ s}^{-1} > 2.5$). These values further emphasize the role of stirring on

637   estuarine phytoplankton at these scales.

638

639   *4.3. Effect of Timescales*

640

641   Most phytoplankton species' reproduction rates are on the scale of hours to days (Harris,

642   1986), indicating that dispersion at rates observed from our dye patch experiments are too fast

643   for phytoplankton growth to cause patch formation at this scale. Therefore, estuarine

644   phytoplankton patches O(10-100 m) are more likely to form when diffusivity is reduced (Koseff

645   et al., 1993), through passive accumulation at features such as fronts (Largier, 1993), in still

646   areas of the bay that are not rapidly flushed, or when larger patches are stretched or divided.

647   Zooplankton grazing rates ($0.07 - 1.94 \text{ d}^{-1}$) in AB can be similar to the phytoplankton growth

648   rates ($0.08 - 1.92 \text{ d}^{-1}$) (Putland and Iverson, 2007). To understand patch dynamics in estuaries,

649   the effect of physical processes on both phytoplankton and zooplankton populations thus needs

650   to be determined.

651

5. Conclusions

652

653

654   Our process study, designed to contribute to a better understanding of the physical

655   processes that influence spatial phytoplankton distribution, characterizes drivers of small-scale

656   distribution patterns as observed in our example data collected in AB in 2011 and also estuarine

33

657   phytoplankton distributions in general. Current measurements combined with dye patch and

658   drifter behavior offered insights into the non-biological processes impacting short-term

659   phytoplankton patch dynamics in estuarine settings. The results highlight differences between

660   controls of plankton patch development in shallow estuarine and deeper ocean settings. Dye

661   patches were transported and dispersed within shorter time scales than phytoplankton

662   reproduction rates; therefore, the formation and dispersion of estuarine phytoplankton patchiness

663   at this spatial scale (1 – 100 m) are strongly governed by physical processes. A similar

664   conclusion was also reached by comparing spatial variability in a passive tracer (salinity) and a

665   biologically-active tracer (chlorophyll). The formation and dispersal of smaller-scale patches

666   influences larger-scale spatial features (Levin, 1992; van Haren et al., 2004). Only through a

667   better understanding of the spatial and temporal distribution of phytoplankton can we design

668   measurement protocols that will allow producing realistic estimates of phytoplankton standing

669   stock and biomass dynamics. This study provides new insights into processes that determine

670   phytoplankton distribution in estuarine settings; however, the extent of the interactions between

671   features at different spatial scales, such as those observed in Geyer et al. (2018) requires further

672   investigation.

673

6. Acknowledgements

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687

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Flow Controls on Estuarine Phytoplankton Distribution