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1 **High frequency, continuous measurements reveal strong diel and seasonal cycling of**
2 **pCO$_2$ and CO$_2$ flux in a mesohaline reach of the Chesapeake Bay**
3

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11 **Key Points:**

12 • Automated pCO$_2$ measurements capture daily cycles and anomalous events in estuaries
13   where pCO$_2$ changes rapidly and across a wide range.

14 • Rhode River is net autotrophic (Dec-May), net heterotrophic (Jun-Nov), NEP is near
15   balanced annually, but can reverse status during a single day.

16 • Year-round continuous measurements reveal that pCO$_2$ and CO$_2$ flux are mediated by
17   temperature effects on biological activity and are inverse to the physical solubility of
18   CO$_2$.
19

21 **ABSTRACT**

22 We estimated hourly air-water gas transfer velocities (k_{600}) for carbon dioxide in the Rhode

23 River, a mesohaline subestuary of the Chesapeake Bay. Gas transfer velocities were calculated

24 from estuary-specific parameterizations developed explicitly for shallow, microtidal estuaries in

25 the Mid-Atlantic region of the United States, using standardized wind speed measurements.

26 Combining the gas transfer velocity with continuous measurements of pCO$_2$ in the water and in

27 the overlying atmosphere, we determined the direction and magnitude of CO$_2$ flux at hourly

28 intervals across a 3 yr record (01 July 2018 to 01 July 2021). Continuous year-round

29 measurements enabled us to document strong seasonal cycling whereby the Rhode River is

30 primarily autotrophic during cold-water months (Dec–May), and largely net heterotrophic in

31 warm-water months (Jun–Nov). Although there is inter-annual variability in CO$_2$ flux in the

32 Rhode River, the annual mean condition is near carbon neutral. Measurement at high temporal

33 resolution across multiple years revealed that CO$_2$ flux and apparent trophic status can reverse

34 during a single 24 hr period. pCO$_2$ and CO$_2$ flux are mediated by temperature effects on

35 biological activity and are inverse to temperature-dependent physical solubility of CO$_2$ in water.

36 Biological/biogeochemical carbon fixation and mineralization are rapid and extensive, so

37 sufficient sampling frequency is crucial to capture unbiased extremes and central tendencies of

38 these estuarine ecosystems.

40 **1. Introduction**

41 Understanding the air–sea exchange of gases and establishing methodologies for accurate

42 measurements has been a decades-long focus of atmospheric scientists, oceanographers, and

43 biogeochemists seeking to understand interactions between oceans and the atmosphere and how

44 these interactions contribute to the global carbon cycle (Broecker et al., 1979; Wanninkhof,

45 1992, 2013). Coastal oceans and estuaries are ecosystems of interest for understanding the

46 complex nature and contribution of the land–sea interface to lateral mass transport of carbon

47 (Abril & Borges, 2005; Cai & Wang, 1998; Frankignoulle et al., 1998; Song et al., 2023) but also

48 with respect to the role these ecosystems play as both atmospheric CO$_2$ sources and sinks (Abril

49 & Borges, 2005; Chen et al., 2020; Dai et al., 2022; Jiang et al., 2008). The exchange of carbon

50 dioxide, methane, and other greenhouse gases (GHGs)between Earth's atmosphere and inland

51 waters, estuaries, coastal oceans are well-documented but not fully quantified (Abril & Borges,

52 2005; Cai, 2011; Laruelle et al., 2017; Raymond & Cole, 2001; Raymond et al., 2013; Van Dam

53 et al., 2019). CO_2 evasion from estuaries alone has been estimated at 15–17% of the total CO_2

54 input from oceans to the atmosphere (Chen et al., 2020; Laruelle et al., 2017), indicating the

55 regional and global significance of estuaries (Bauer et al., 2013; Frankignoulle et al., 1998; Jiang

56 et al., 2008). Yet, there is still great uncertainty surrounding the true net contributions of coastal

57 oceans, estuaries, and inland water bodies to the atmospheric loading of GHGs (Borges, 2005;

58 Chen et al., 2020; Herrmann et al., 2020; Joesoef et al., 2015; Laruelle et al., 2017; Raymond et

59 al., 2013; Van Dam et al., 2019).

60

61 To better understand the effects of estuaries on atmospheric GHG exchange and accumulation, it

62 is imperative that we understand their capacity and function as carbon sources and sinks and

63 ultimately how estuaries factor into the planet's overall global carbon budget (Herrmann et al.,

64 2020; Laruelle et al., 2017; Van Dam et al., 2019). Many attempts to characterize CO_2 flux in

65 estuaries and nearshore oceans (Chen et al., 2013; Herrmann et al., 2020; Rosentreter et al. 2021)

66 have relied on direct measurements using floating domes, tracer gases or, more recently, eddy

67 covariance methods (Laruelle et al, 2017; Van Dam et al., 2019). Because flux measurements are

68 time intensive, they tend to be temporally and spatially limited (Herrmann et al., 2020; Klaus &

69 Vachon, 2020). Using direct flux measurements to derive accurate gas transfer velocity

70 constants (K_o, the velocity of gas crossing the air-water boundary) enables models to be

71 parameterized to estimate K_o and compute gas flux. Thus, correlative models that incorporate

72 simultaneous environmental measurements such as wind and/or water velocity, factors that affect

73 turbulence at the air-water interface and promote gas exchange, have aided in the widespread

74 accumulation of gas flux estimates (Raymond & Cole, 2001; Van Dam et al., 2019; Wanninkhof,

75 2014). Gas transfer velocity constant models vary according to the habitat/system being observed

76 and chemical, physical, and biological factors present in each (e.g., lakes, rivers/streams,

77 estuaries, and oceans; Herrmann et al., 2020; Ho et al., 2016; Raymond & Cole, 2001; Van Dam

78 et al., 2019; Wanninkhof, 1992). To reduce uncertainty of computed gas fluxes, it is critical that

79 the appropriate K_o models are matched to a targeted ecosystem.

80

81 Coastal oceans and estuaries are exceptionally complex, frequently characterized by their relative

82 shallowness and how their freshwater inputs (riverine, surface, and groundwater) mix with salt

83   water (Chen et al., 2020). High nutrient and pollutant loading, due to urbanization and

84   eutrophication by humans, also have important effects on estuaries and coastal oceans (Freeman

85   et al., 2019). High spatial and temporal variability are hallmarks of estuaries.

86

87   Here we present a 3 year data set that combines high frequency (1 min interval) measurements of

88   dissolved and atmospheric CO_2 with co-located and continuous measurements of salinity, water

89   temperature, tidal cycling, and wind velocity, recorded at the Smithsonian Environmental

90   Research Center (SERC) dock, in the Rhode River, Maryland. To estimate hourly, daily,

91   seasonal, and annual CO_2 flux rates, we applied a CO_2 gas velocity constant model developed by

92   Van Dam et al. (2019) for the New River, North Carolina. This model is expressly designed for

93   application to shallow, well-mixed, microtidal estuaries located in the Mid-Atlantic coast of the

94   United States.

95

96   **2. Methods**

97   2.1 Study Location

98   The Rhode River is a tributary and subestuary of the Chesapeake Bay, a drowned river valley,

99   coastal plain estuary (Fig. 1). The Rhode River has been studied extensively by SERC staff and

100   colleagues for over 4 decades: nutrient chemistry (Jordan & Correll, 1991; Jordan et al., 1991),

101   phytoplankton ecology (Gallegos et al., 2010), color dissolved organic matter distribution

102   (Tzortziou et al., 2008; Tzortziou et al., 2011), and more recently, modeling of dissolved organic

103   carbon (DOC) input from freshwater and tidal marsh sources (Clark et al., 2020). Located on the

104   Bay's northwestern shore (38º52'N, 76º32'W), the Rhode River is bounded at its head by Muddy

105   Creek and at its mouth by the mainstem of the Chesapeake Bay. The Rhode River is a shallow

106   (mean depth = 2 m, max depth = 4.1 m), mesohaline (0 to 18 ppt), well-mixed, eutrophic

107   tributary with a length of approximately 5 km; its surface area is approximately 5 km^2 with a

108   shoreline perimeter of 39 km (Breitburg et al., 2008; Clark et al., 2018). A 0.21 km^2 tidal marsh

109   (Kirkpatrick Marsh) fringes the estuary at the mouth of Muddy Creek (Fig. 1). Tides are semi-

110   diurnal with a mean amplitude of approximately 30 cm, but water height can be strongly affected

111   by wind and weather events. Muddy Creek is the main freshwater source of the Rhode River and

112   has a maximum flow rate of 10.42 $m^3 \cdot s^{-1}$ and mean flow rate 0.18 $m^3 \cdot s^{-1}$ (mean flow = 15,552

113    $m^3 \cdot d^{-1}$; Clark et al., 2020; Clark et al., 2018; Jordan et al., 1986). The mean daily volume of

114   freshwater inflow from Muddy Creek is approximately 0.5% of the mean daily tidal exchange

115   volume, based on the Rhode River's area and mean tidal amplitude. In the absence of

116   measurements of the pH or pCO$_2$ of the freshwater entering the Rhode River from Muddy Creek

117   or other lesser freshwater inputs to the estuary, we are unable to report these pCO$_2$ or pH values.

118   However, given the exceedingly small overall volume of freshwater input to the Rhode River

119   from its surrounding watershed, it is not considered a river-dominated estuary so is not expected

120   to be substantially influenced by the chemical characteristics of this input. This is not to say there

121   is no freshwater influence, only that such influences are likely quite local when mixing with far

122   larger volumes of water from the Chesapeake Bay and therefore beyond the resolution of this

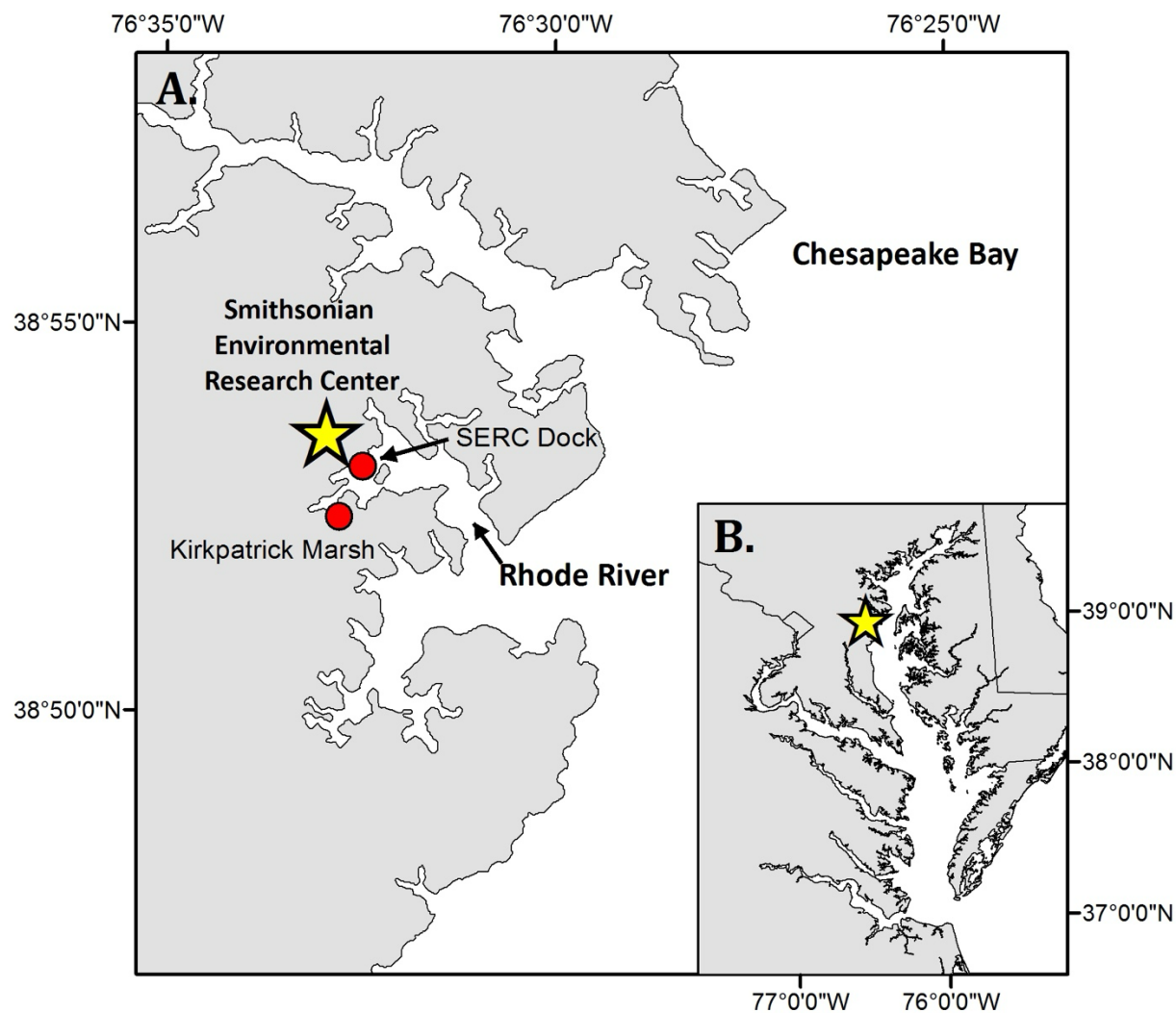
123   study.

124

125   Although the Rhode River is a model ecosystem that has been studied intensively for several

126   decades across many dimensions (Clark et al. 2018; Correll et al., 1992; Gallegos et al., 1992;

127   Jordan et al. 1991; Rose et al. 2019), no work to date has expressly characterized the nature and

128   dynamics of CO$_2$ flux between the river and the atmosphere.

129
130 **Fig. 1.** Location of study site on the Rhode River, Edgewater MD (A), as situated in the
131 Chesapeake Bay (B). All pCO_2 and related water quality values reported were measured from the
132 SERC dock, that extends approximately 75 m from shore on Rhode River. Red circles indicate
133 location of dock and a tidal creek that drains the Kirkpatrick saltmarsh (marsh area = 0.21 km^2, 1
134 km up estuary from the dock).

135

136 2.2 *In Situ* Measurements, Calculated Parameters and Quantities

137 Continuous, automated environmental measurements were made in and above the Rhode River

138 during a 3 year period between 01 July 2018 and 01 July 2021. The purpose of these

139 measurements was to document fluctuations in aqueous pCO_2 on a fine time scale, from which

140 CO_2 flux between the water and atmosphere could be calculated.

141 2.2.1 Aqueous CO_2 (pCO_{2water})

142 To measure the CO_2 gradient ($\Delta C = pCO_{2water} - pCO_{2air}$) across the Rhode River surface waters

143 and its overlying atmosphere, measurements of pCO_2 were made with a non-dispersive infrared

144 (NDIR) detector. In the case of dissolved gas measurements, water was equilibrated continuously

145 with a spherical falling film equilibrator (Miller et al. 2019). Water from 1 m below the water's

146 surface was pumped and dispersed continuously over a 25.4 cm diameter sphere. The falling film

147 created on the sphere generates a gas exchange surface which forces CO_2 in the equilibrator

148 headspace into equilibrium with the water's CO_2 content (i.e. mole fraction = xCO_2 (μmol/mol).

149 Water exits the equilibrator via an airtight drain that prevents headspace contamination from

150 surrounding atmospheric air. Headspace gas circulates continuously in a closed loop through the

151 equilibrator, water trap and gas dehumidifier, past the NDIR, and back into the equilibrator.

152 Experimental observations concluded that spherical falling film equilibrators achieve 99%

153 equilibration of CO_2 within 10–15 mins, depending on whether step changes are from low to

154 high or high to low; details of the operation and performance of the falling film equilibrator are

155 described in Miller et al. (2019). Measurements were made at 1 min intervals at a pressure equal

156 to the ambient barometric pressure.

157

158 Measured raw CO_2 mole fractions (μmol/mol) were converted to partial pressures (μatm) using

159 equation 1. Minute-over-minute values were rounded down to the nearest hour and averaged to

160 provide hourly means. The mole fractions were then evaluated with corresponding water

161 temperature and salinity measurements following the methodology of Zeebe and Wolf-Gladrow

162 (2001) where saturation vapor pressure of water is calculated according to Weiss and Price

163 (1980) to determine pCO_{2water}.

164

165 $$pCO_{2water} = xCO_2 \cdot (p - pH_2O) \qquad (1)$$

166

167 where

168 pCO_2 = partial pressure of CO_2 of water (μatm)

169 xCO_2 = mole fraction of CO_2 in water (μmol/mol)

170   p = total pressure = 1 atm

171   pH_2O = saturation vapor pressure of water (μatm)

172

173   2.2.2 Atmospheric CO_2

174   Every six hours, the sample gas stream was automatically diverted with programmed solenoid

175   control valves from the equilibrator to an atmospheric port located approximately 5 m above the

176   pier deck. During atmospheric sampling, 15 1-min interval measurements were made. To

177   account for inaccuracies during the transition period from equilibrator to atmospheric sampling,

178   the final eight measurements were averaged and the first seven were discarded. Similarly, the

179   first 30 measurements following switchover from atmospheric port to equilibrator were

180   discarded, to ensure measurements were fully equilibrated with water. For these atmospheric

181   measurements, the contribution of the vapor pressure of water to the total atmospheric pressure

182   of the open-air environment was considered negligible (i.e. $pH_2O = 0$ and $p = 1$), such that

183     $pCO_{2atm} = xCO_{2atm}$. As such, any potential differences are expected to fall well within the

184   measurement accuracy of the instrument (see below).

185

186   One advantage to using a shared NDIR sensor for aquatic and atmospheric samples is that any

187   minor effects of instrument drift will be reflected in both data streams, as opposed to two sensors

188   that drift independently of one another. Likewise, significant and sustained deviation from

189   typical local atmospheric variability will be captured during atmospheric sampling and can signal

190   the need for recalibration and assist with QA/QC of corresponding data from both streams. A

191   disadvantage of using a common sensor for both dissolved and atmospheric CO_2 measurements

192   is that it results in a mismatch in sampling frequency of the two. With this limitation in mind, we

193   chose a higher sampling frequency for aquatic measurements to better describe the inherently

194   higher variability in dissolved CO_2 in water vs. that in the atmosphere (Fig. 2).

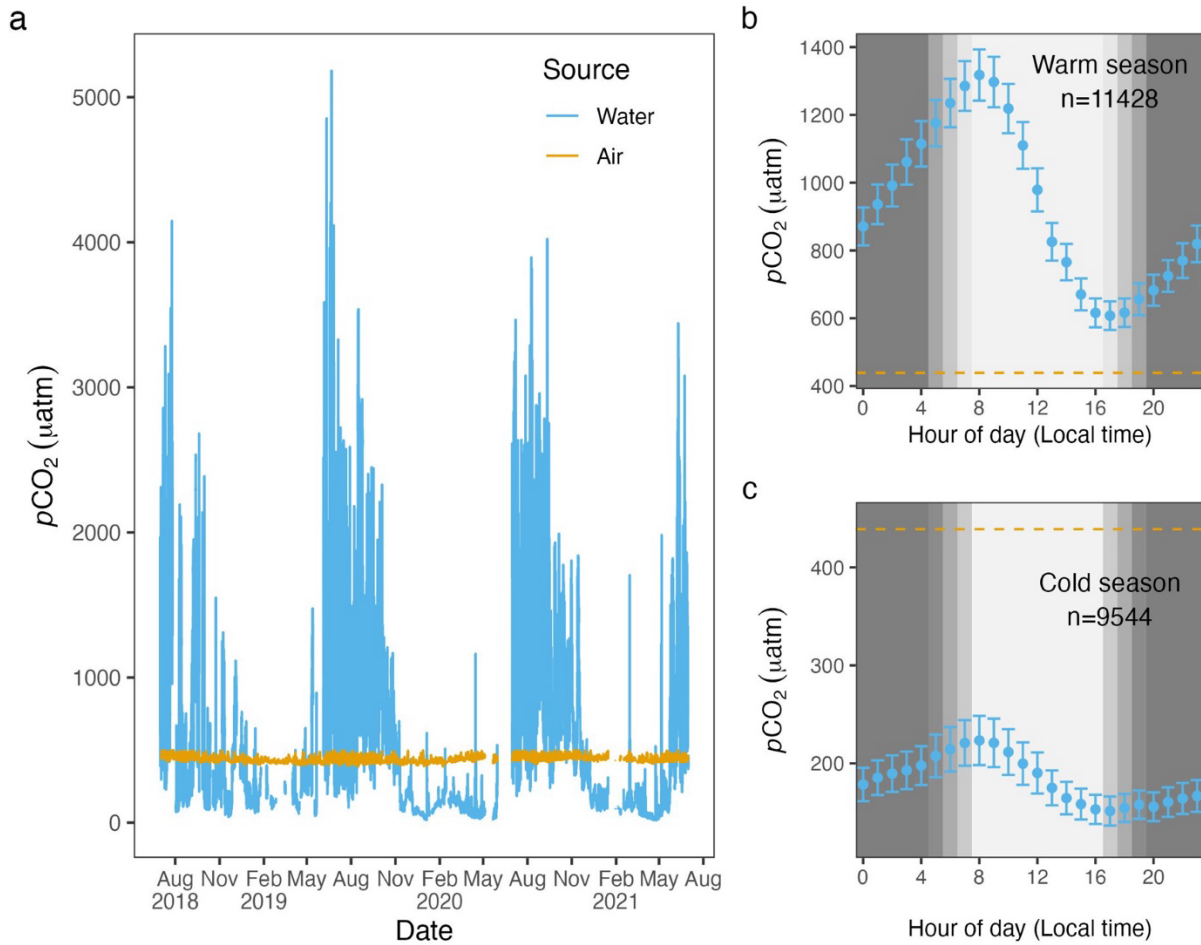



Fig. 2. Hourly pCO_{2water} (blue) and pCO_{2air} (goldenrod) values from 01 July 2018 to 01 July 2021 (panel a). The air-water CO_2 gradient, $\Delta C = pCO_{2water} - pCO_{2air}$ describes the directionality of gas diffusion. Negative ΔC values (pCO_{2water} values falling below goldenrod demarcation) represent gas movement from air to water and vice versa. Panels b and c depict mean pCO_{2water} values (95% CI) for each hour of the day for warm and cold seasons, with the dashed lines equal to the mean 3 yr value of pCO_{2air}.

Given the 3 yr time series and strong diel cycling of pCO_{2water} (and dissolved oxygen (DO), see Figs. S1 and S2) in the Rhode River, we chose to aggregate aqueous minute-over-minute measurements to mean hourly averages. Owing to the relative lack of short-term variability in local atmospheric CO_2 concentrations (Fig. 2), we used linear interpolation to impute atmospheric CO_2 concentrations during hours in between actual readings (6 hr gaps between atmospheric measurements), which we assumed to be more realistic and reliable than Last Observation Carried Forward (LOCF) methods, where the last observation is repeated for all gaps until the next measurement is encountered, a method that has fallen out of favor, especially

211   for environmental time series data (Lachin, 2016). To determine if any inadvertent bias was

212   introduced by linear interpolation procedure, summary statistics of actual atmospheric readings

213   and actual readings + imputed CO_2 values were compared statistically. This approach enabled us

214   to take advantage of >25,000 time points throughout the 3 yr period of observation, providing

215   hourly resolution. Mean pCO_{2air} = 437 ± 20.0 μatm (Table 1), variability that falls well within

216   manufacturer's specifications (see section 2.2.4).

217

218   2.2.3 CO_2 gradient (ΔC)

219   ΔC was determined by subtraction, pCO_{2water} – pCO_{2air}, where positive ΔC values correspond to

220   higher CO_2 concentrations in the water, tending toward movement from water to air (outgassing

221   or evasion, where Rhode River = CO_2 source), and negative values that signal CO_2 transport

222   from air to water (dissolution, where Rhode River = CO_2 sink). Figure 2 shows pCO_{2water} and

223   pCO_{2air} plotted on an hourly basis for the 3 yr period beginning 01 July 2018 and ending 01 July

224   2021. Across this period, ΔC was predominantly negative during cold months and predominantly

225   positive during warm months when pCO_{2water} tended to reach the highest values of the year, but

226   ΔC sometimes reversed sign due to occasional extreme day-time photosynthetic drawdown of

227   CO_2 (Fig. 2).

228

229   2.2.4 Accuracy of CO_2 measurements

230   Estimated accuracy of the spherical falling film equilibrator and NDIR sensor (SenseAir K30,

231   https://senseair.com/) combination were experimentally determined in the lab and found to

232   measure water equilibrated with known gas concentrations to be within the ±1% uncertainty

233   limits of the of certified standard gas mixtures used, and well within the published accuracy

234   specification of the SenseAir K30 (i.e., ± 30 ppmv ± 3% of instrument reading). Experimental

235   analysis by Martin et al. (2017) report even higher accuracy when relative humidity and

236   atmospheric pressure are controlled for. Details on performance of the spherical falling film

237   equilibrator, such as accuracy, precision, and time constants can be found in Miller et al. (2019).

238   Although SenseAir offers automated calibration via long term comparisons to atmospheric

239   readings, this feature was deactivated. The K30 NDIR was periodically validated using standard

240   zero CO_2 (nitrogen) and standard certified span gases at intervals of one to two months during

241   the study period. Although the K30 was never observed to drift beyond its factory specifications,

242 the sensor was occasionally re-calibrated in the lab, and measured values were accepted without

243 adjustment.

244

245 CO_2 measurements were downloaded to a database at approximately two-week intervals during

246 the observation period. Data were graphed and reviewed visually, in combination with twice

247 weekly observations of equilibrator function recorded in an accompanying notebook. Anomalous

248 data were flagged and excluded from data analysis (e.g., flooding or clogging events that

249 interrupted proper equilibration.)

250

251 2.3 Co-located water quality and atmospheric measurements

252 This water quality station at the SERC dock is a long-term node of the Maryland Department of

253 Natural Resources "Eyes on the Bay" Chesapeake Bay tidal water monitoring program, and has

254 been operated by the SERC since 1986. Water quality and atmospheric data are maintained by

255 the MarineGEO Upper Chesapeake Bay Observatory and can be accessed online (Benson et al.,

256 2023). A YSI EXO2 sonde was positioned 1 m below the water's surface and in proximity (~2.5

257 m distance) to the submerged water pump that fed the pCO$_2$ equilibrator. Sonde measurements

258 were made at 6 minute intervals and aggregated to 1 hr averages. The published accuracy

259 specifications for the YSI sonde are as follows: temperature: ±0.01 ºC (-5º to 35º C); salinity:

260 ±1% of reading or 0.1 ppt (0–70 ppt); dissolved oxygen: ±0.1 mg/L or 1% of reading (0 to 20

261 mg/L). Discrete measurements of temperature and salinity were made with a handheld YSI

262 Professional Plus 2030 with Quattro Cable instrument, with the following specifications:

263 temperature: ±0.02 ºC (-5º to 70º C); salinity: ±1% of reading or 0.1 ppt (0–70 ppt); dissolved

264 oxygen: ±0.2 mg/L or 2% of reading (0 to 20 mg/L). Equilibrator temperature was measured

265 with a probe (EDS model OW-TEMP-B3-12xA) accurate to ±0.5 ºC (-10º to 85 ºC). Discrete

266 measurements were routinely compared with the sonde to corroborate measurement agreement.

267 Wind speed measurements were made using a sonic anemometer (Vaisala WXT-520 weather

268 transmitter) mounted 7 m above the mean low tide height of the water and located directly above

269 the pCO$_2$ equilibrator.

270

271 2.4 Data Processing

272 Data included in this study span 01 Jul 2018 to 01 Jul 2021.

273   2.4.1 Gas-specific solubility

274   To determine the purely physical effects of temperature and salinity on CO_2 solubility, gas-

275   specific solubility values K_0 (mmol $\cdot$ m^{-3} $\cdot$ µatm^{-1}) were calculated across the 3 yr observation

276   period using water temperature and salinity measurements in combination with pCO_{2water} values,

277   according to Weiss and Price (1980) at 1 hour intervals.

278

279   2.4.2 Gas transfer velocity estimation (k)

280   Given the similarities between the Rhode River and New River estuaries (e.g., shallow,

281   microtidal estuaries with slow water velocity and strong diel cycles in pCO_2 and DO), we chose

282   to parameterize gas transfer velocity k (cm $\cdot$ h^{-1}) standardized to the unitless Schmidt number

283   600 (k_{600}) according to the estuary-specific k parameterization model developed by Van Dam et

284   al. (2019). Van Dam et al. (2019) determined that k correlated with wind speed differently during

285   daytime versus nighttime hours (linear vs. parabolic relationships). Wind speed data were

286   collected during the 3 yr period from a sonic anemometer located on the SERC dock directly

287   above the equilibration system and approximately 7 m above the water's surface at mean low

288   tide height. For the analysis, windspeeds were standardized for a height of 10 m following a

289   power-law relationship, $U_{10} = U_7 * (10/7)^{0.15}$ (Saucier, 2003). Following Van Dam et al.,

290   wind speed data were binned to 1.5 m s^{-1} intervals for day and night readings and raw values

291   replaced by the mean wind speed for each bin. The median binned windspeed over the Rhode

292   River was 2.2 m s^{-1}, regardless of time of day or season. Recorded windspeeds never exceeded

293   10m s^{-1} and were dominated by much lower values (Fig. S1). Unlike the New River Estuary, the

294   Rhode River's windspeed profile does not differ much between day and night, nor across season.

295   For this reason, we chose to use the most conservative k_{600} formulation from Van Dam et al

296   (2019), that combines day and night winds to estimate k_{600}.

297

298   Wind speed was used to parameterize k_{600} as follows:

299

300                            $k_{600} = 1.5 * U_{10} + 4.2$ (2)

301

302   where U_{10} = mean of binned wind speed at 10 m above the water's surface (m $\cdot$ s^{-1}).

303

304 2.4.3 CO_2 flux

305 Using continuous, parallel 3 yr records (01 July 2018 to 01 July 2021) of dissolved and

306 atmospheric pCO_2, water temperature, salinity, and wind speed (at standard 10m height, U_{10}),

307 CO_2 flux was derived according to the equation:

308

309   CO_2 flux $= k_{600} \cdot K_0 \cdot \Delta C \cdot (600 / Sc)^{-0.5}$ (3)

310 where

311   CO_2 flux = the rate and direction of CO_2 mass moving between water and gas phases

312   (mmol $\cdot$ m^{-2} $\cdot$ hr^{-1})

313   k_{600} = gas transfer velocity (cm $\cdot$ hr^{-1}), normalized to a common Schmidt number

314   ($Sc = 600$)

315   K_0 = gas-specific solubility for CO_2 (mmol $\cdot$ m^{-3} $\cdot$ µatm^{-1})

316   ΔC = air-water concentration gradient (µatm)

317   Sc = Schmidt number

318

319 Note: CO_2 flux calculations require conversion from traditional k_{600} units (cm $\cdot$ hr^{-1}) to (m $\cdot$ hr^{-1})

320 from ΔC units (µatm) to (atm) prior to calculation.

321

322 2.4.4 Day/Night Designation

323 To differentiate daytime from nighttime hours, we used the position of the measurements

324 (latitude) in the Rhode River, combined with the local date and time. This approach enabled us to

325 uniformly designate various environmental measurements as happening during the day or night

326 (R package "LakeMetabolizer", Winslow et al., 2016).

327

328 2.4.5 Seasonality

329 We chose to break the year into two 6 month periods based seasonal water temperature shifts,

330 designating June–November as "warm-water months" when water temperatures averaged $23.2 \pm$

331 6.90 °C. (mean $\pm$ 1 sd) and December–May as "cold-water months", 10.9 ± 5.66 °C (Figs. S1and

332 S2).

333

334

335   2.4.6 Effect size

336   Owing to the large number of observations available for comparison in this study, the likelihood

337   of finding statistically significant results is quite high. Whether such statistical results by

338   themselves connote practical and informative differences can be difficult to discern. Effect sizes

339   (Omega-squared, ω^2) were calculated according to two-factor ANOVAs where independent

340   variables were investigated by season (cold-water vs. warm-water season), day/night period and

341   the interaction of season and day/night. The independent variables compared were: K_0, CO_2 flux,

342   ΔpCO_2, k_{600}, pCO_{2air}, pCO_{2water}, and wind speed. To account for temporal autocorrelation and

343   lack of independence of observations that are typical of environmental time series data, we

344   corrected for overinflation in the residual mean square used in the effect size calculations by

345   removing the autocorrelation present within residuals, leaving the white-noise component as the

346   unbiased estimate of residual variability (Cochrane-Orcutt procedure, R package "orcutt", Spada

347   et al., 2018).

348

349   **3. Results and Discussion**

350   3.1 Daily and Seasonal Cycling of pCO_2

351   Hourly averaged measurements of pCO_{2water} in the Rhode River across three years revealed

352   strong diel and seasonal cycling (Fig. 2). Mean and maximum pCO_{2water} were significantly higher

353   in warm-water vs. cold-water months (Table 1). During warm-water months (Jun–Nov) daily

354   oscillations of pCO_2 frequently transit from far above to below ambient atmospheric conditions

355   over the course of the day, only to reverse direction (from low to high) during the nighttime

356   hours (Fig. 2). During the summer, pCO_{2water} levels sometimes shifted by as much as 4500 μatm

357   in both directions during a single 24 hr period (Fig. 2). This pattern is consistent with

358   biologically driven cycling whereby very high early morning pCO_{2water} conditions are depleted

359   by net photosynthetic activity (inorganic carbon fixation) over the course of the day, but high

360   pCO_{2water} is restored by respiration in the benthos and water column at night (Song et al. 2023).

361   Comparing dissolved oxygen (DO) over the same period, similar harmonic cycling is observed,

362   but maximums and minimums of pCO_2 and DO were inversely related (Fig. S1), hallmarks of a

363   production/respiration driven system (Herrmann et al., 2020; Van Dam et al., 2019).

364

365   **Table 1.** Descriptive statistics comparing seasonality of pCO$_2$, CO$_2$ flux and associated
366   parameters in cold-water (Dec–May) and warm-water seasons (Jun–Nov).
367

Season	Time Period	Variable	Units	N	Mean	Min	Max	SD
overall	-	**CO$_2$ flux**	**mmol · m^{-2} · hr^{-1}**	**20971**	**-0.09**	**-4.89**	**11.18**	**1.823**
cold	day	CO$_2$ flux	mmol · m^{-2} · hr^{-1}	4494	-1.39	-4.89	8.26	1.134
cold	night	CO$_2$ flux	mmol · m^{-2} · hr^{-1}	5050	-1.39	-4.66	5.24	0.927
warm	day	CO$_2$ flux	mmol · m^{-2} · hr^{-1}	6007	1.18	-3.95	11.18	1.731
warm	night	CO$_2$ flux	mmol · m^{-2} · hr^{-1}	5421	0.78	-3.97	8.05	1.467
overall	-	K_0	**mmol · m^{-3} · µatm^{-1}**	**20971**	**0.04**	**0.03**	**0.07**	**0.011**
cold	day	K_0	mmol · m^{-3} · µatm^{-1}	4494	0.05	0.03	0.07	0.009
cold	night	K_0	mmol · m^{-3} · µatm^{-1}	5050	0.05	0.03	0.07	0.008
warm	day	K_0	mmol · m^{-3} · µatm^{-1}	6007	0.03	0.03	0.06	0.007
warm	night	K_0	mmol · m^{-3} · µatm^{-1}	5421	0.04	0.03	0.07	0.008
overall	-	k_{600}	**cm · hr^{-1}**	**20971**	**7.86**	**5.57**	**18.36**	**2.047**
cold	day	k_{600}	cm · hr^{-1}	4494	8.71	5.57	16.33	2.251
cold	night	k_{600}	cm · hr^{-1}	5050	7.74	5.57	18.36	2.081
warm	day	k_{600}	cm · hr^{-1}	6007	7.92	5.57	18.36	1.868
warm	night	k_{600}	cm · hr^{-1}	5421	7.20	5.57	18.36	1.751
overall	-	**ΔC**	**µatm**	**20971**	**154**	**-436**	**4750**	**645.8**
cold	day	ΔC	µatm	4494	-239	-436	1553	220.9
cold	night	ΔC	µatm	5050	-256	-434	1204	164.2
warm	day	ΔC	µatm	6007	570	-399	4750	745.5
overall	-	**pCO$_{2air}$**	**µatm**	**20971**	**437**	**387**	**500**	**20.0**
cold	day	pCO$_{2air}$	µatm	4494	430	390	497	16.0
cold	night	pCO$_{2air}$	µatm	5050	432	387	499	17.8
warm	day	pCO$_{2air}$	µatm	6007	439	390	499	20.7
warm	night	pCO$_{2air}$	µatm	5421	443	387	500	21.5
overall	-	**pCO$_{2water}$**	**µatm**	**20971**	**591**	**15**	**5182**	**651.8**
cold	day	pCO$_{2water}$	µatm	4494	191	15	1982	220.9
cold	night	pCO$_{2water}$	µatm	5050	176	17	1638	163.9
warm	day	pCO$_{2water}$	µatm	6007	1009	47	5182	752.6
warm	night	pCO$_{2water}$	µatm	5421	844	38	4855	632.2
overall	-	**wind speed**	**m · s^{-1}**	**20971**	**2.4**	**0.1**	**9.8**	**1.42**
cold	day	wind speed	m · s^{-1}	4494	3.1	0.3	8.9	1.53
cold	night	wind speed	m · s^{-1}	5050	2.4	0.3	9.1	1.45
warm	day	wind speed	m · s^{-1}	6007	2.5	0.1	9.8	1.28
warm	night	wind speed	m · s^{-1}	5421	2.0	0.1	9.1	1.23

368
369   On the seasonal timescale, pCO$_2$ was consistently lowest and DO highest during cold-water

370   months of the year (Dec–May; Fig. S1). Importantly, for both gases the temporal variability (diel

371 cycling; Fig. S2) was most constrained during cold-water months across years, strongly

372 suggesting that carbon fixation exceeds respiration for prolonged periods (weeks to months). In

373 contrast, during warm-water months (Jun–Nov), photosynthesis/carbon fixation and respiration

374 are more evenly balanced, compensating one another over 24 hr periods (i.e., respiration >

375 productivity at night and productivity > respiration during daylight hours; Fig. 2).

376

377 3.2 Air-water concentration gradient = ΔC (μatm)

378 When hourly pCO_{2water} and pCO_{2air} values (composed of 4 hourly measurements and 20

379 interpolated values per day) were plotted across the three years of observation, the diel and

380 seasonal cycles of pCO_{2water} are evident. As expected, atmospheric concentrations of CO_2

381 remained relatively constant compared with aqueous loads. When the mean raw pCO_{2air}

382 measurements (mean = 435.1, 95% CI [434.4, 435.7]) were compared with raw pCO_{2air}

383 measurements + imputed estimates (mean = 435.4, 95% CI [435.2, 435.7]) no statistical

384 difference was observed, indicating that no substantial bias was introduced by linear

385 interpolation of atmospheric measurements.

386

387 Although nearshore atmospheric CO_2 concentrations are expected to vary more than those in

388 isolated well-mixed atmosphere (e.g., at the Mona Loa Observatory), annual mean values were

389 consistent and within the published uncertainty of the K30 NDIR sensor, when compared with

390 global measurements conducted at Mona Loa (Thoning et al., 2023). Local perturbations (e.g.,

391 effects of terrestrial photosynthetic drawdown when wind is absent) were apparent in

392 measurements (Fig. 2) but there were no instances when the measured local atmospheric values

393 were suspiciously high or low for days on end, as compared with expected global mean

394 atmospheric values for the time period (i.e., 408–416 ppmv; https://www.co2.earth/annual-co2,

395 Thoning et al., 2023). This lack of sustained anomalous deviation served as additional

396 confirmation that the K30 was functioning properly and had not drifted outside its calibration

397 range. Importantly, given the extreme diel cycling and seasonal variability of the Rhode River's

398 pCO_{2water}, the absolute accuracy necessary for determining year-over-year changes in

399 atmospheric or ocean pCO_2 is not a requirement for these CO_2 flux calculations which rely on

400 relative differences between water and atmospheric measurements.

401

402   Hourly air-water concentration gradient values = ΔC (μatm) were calculated and plotted across

403   the three years of study (Fig. 2). During warm months, pCO_{2water} routinely shifts from

404   supersaturated to sub-atmospheric and back again, over the course of 24 hours (e.g., between

405   >2000 μatm and <410 μatm on a single day). These large daily swings in pCO_{2water} produced

406   concomitant directional reversals of ΔC ($pCO2_{water} - pCO_{2air}$), which result in longer term

407   averaged gradients (e.g., multi-day, multi-week averages) near zero (Fig. 2). In contrast, most of

408   the time during cold-water months is spent in a state of sub-atmospheric pCO_{2water} (under-

409   saturation with respect to overlying atmosphere), resulting in ΔC values that are negative and

410   which promote movement of CO_2 from the atmosphere into the water.

411

412   3.3 Gas-specific solubility (K_0)

413   To account for the physical effects of temperature and salinity on the solubility of CO_2 in

414   estuarine water, K_0 was calculated by methods of Weiss and Price (1980). K_0 varied strongly

415   across seasons over the 3 yr observation period. The maximum annual range = 0.027 to 0.071

416   mmol $\cdot$ m^{-3} $\cdot$ μatm^{-1}; mean cold-water months = 0.051 and mean warm-water months = 0.035

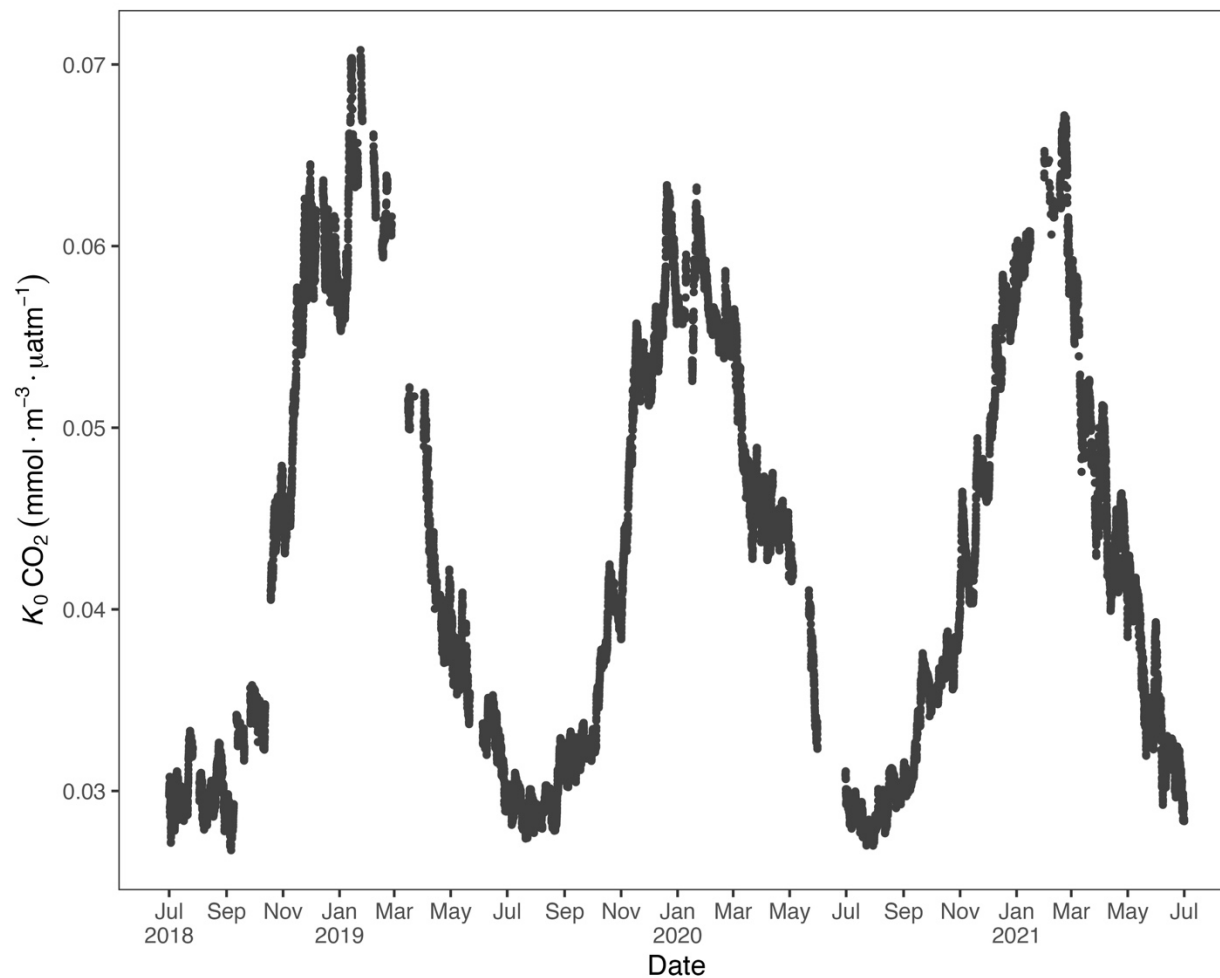
417   mmol $\cdot$ m^{-3} $\cdot$ μatm^{-1}, confirming that CO_2 was most soluble during winter and least soluble in

418   summer (Fig. 3). This is inverse to observed dissolved CO_2 values: pCO_{2water} was lowest and

419   least variable during winter and highest and most variable during summer (Fig. 2, Table 1)

420   suggesting that solubility, in and of itself, plays only a minor and non-limiting role in pCO_{2water}

421   in the Rhode River. Effect size (ω^2) estimates indicated that the greatest proportion of variability

422   in K_0 was associated with season, vs. day/night or the interaction of the two (Table 2).

423

424 **Fig. 3.** Gas-specific solubility (K_0) for CO_2 based on water temperature and salinity.
425 Units are mmol m^{-3} μatm^{-1} in the Rhode River (01 Jul 2018 to 01 Jul 2021).

426 **Table 2.** Contrast effect sizes based on two-factor ANOVA where independent variables were
427 compared by season (cold-water season = Dec–May vs warm-water season = Jun–Nov),
428 day/night period and the interaction of the two. ω^2 is a measure of effect size, estimating the
429 proportion of total variance explained by each parameter. Effect sizes were corrected for inherent
430 temporal autocorrelation using the Cochrane-Orcutt procedure (Spada et al., 2018).
431

Variable	Factor	Effect Size (ω^2)
K_0	Season	0.0300
K_0	Day/Night	0.000575
K_0	Season:Day/Night	0.0000140
CO_2 flux	Season	0.415
CO_2 flux	Day/Night	0.00295
CO_2 flux	Season:Day/Night	0.00301
ΔC	Season	0.310
ΔC	Day/Night	0.00501
ΔC	Season:Day/Night	0.00333
k_{600}	Season	0.00164
k_{600}	Day/Night	0.00269
k_{600}	Season:Day/Night	0.0000549
pCO_{2air}	Season	0.000137
pCO_{2air}	Day/Night	0.0000134
pCO_{2air}	Season:Day/Night	0.00000137
$pCO_{2\,water}$	Season	0.188
$pCO_{2\,water}$	Day/Night	0.00275
$pCO_{2\,water}$	Season:Day/Night	0.00191
wind speed	Season	0.00711
wind speed	Day/Night	0.0186
wind speed	Season:Day/Night	0.000182

432

433 3.4 Temperature/Biology ratio

434 To independently parse the magnitude of the physical versus biological forcing of pCO_{2water}, we

435 normalized values to mean water temperature and estimated the Takahashi's

436 Temperature/Biology ratio (Takahashi et al., 2002) to compare the influence of temperature and

437 biological activities on pCO_{2water}. Across the 3-year period, we found that the predominant driver

438 of pCO_{2water} in the Rhode River was biological activity, accounting for nearly 4 times more

439 forcing than the physical effects of water temperature on CO_2 solubility (Table 3). These patterns

440 demonstrate the outsized role that biological processes play in shaping pCO_{2water} in nearshore

441 marine and estuarine ecosystems (Dai et al., 2022; Van Dam et al., 2019).

442

443

444 **Table 3**. Takahashi Temperature/Biology Ratio (Eq. 5a From Takahashi et al. 2002).

445

Year	N	$\Delta p\text{CO}_2\_\text{bio}$	$\Delta p\text{CO}_2\_\text{temp}$	T/B ratio
2018	4416	3193.0	765.8	0.240
2019	8760	3669.8	1019.6	0.278
2020	8784	2772.1	846.0	0.305
2021	4345	2356.1	507.2	0.215
Overall	26305	3701.5	926.4	0.250

446

447

3.5 Gas transfer velocity (k_{600})

Gas transfer velocity is affected by both mass transfer from molecular diffusion driven by ΔC (i.e. CO_2 gradient between water and atmosphere) and momentum transfer linked to external environmental forces that enhance turbulence at the air-water boundary layer (Ho et al., 2016; Raymond & Cole, 2001; Van Dam et al., 2019). Van Dam et al. (2019) validated the use of wind speed at 10 m above the water's surface (U_{10}) to estimate gas transfer velocities of CO_2 that were standardized to a Schmidt number of 600 (k_{600}) by comparing estimated values to k_{600} values derived directly from eddy covariance CO_2 flux measurements. Given the relative uniformity of wind speed over the Rhode River where median binned U_{10} windspeed (converted from U_7 measurements) was 2.2 m · s^{-1} regardless of time of day or season, and that maximum values rarely exceeded 10 m s^{-1} (Table 1, Fig. S1), we chose to use the most conservative estuarine-specific parameterization of k_{600} (Van Dam et al., 2019) (Eq. 2). The mean overall Rhode River k_{600} value for CO_2 (mean ± SD, 7.86 ± 2.05 cm · hr^{-1}) was of comparable magnitude to that of the New River Estuary, NC (9.37 ± 9.47 cm · hr^{-1}. However, wind speed varied far less on the Rhode River than the New River estuary and day/night explained more variability in wind speed than season. Because wind speed directly influenced the formulation of k_{600} (Eq. 2), the effect size of day/night is similarly greater than the seasonal effect on gas transfer velocity (Table 2). Nevertheless, effect sizes (ω^2) indicate that "season" explained at least 10 times more of the observed variance of $p\text{CO2water}$, $p\text{CO2air}$, air-water concentration gradient, CO2 flux, and gas-specific solubility than "day/night" or their interaction (Table 2). Given the minor freshwater input and microtidal nature of the Rhode River, we do not believe that lateral water velocity and bottom turbulence appreciably affect the gas transfer velocity of CO_2 here, although we did not investigate possible influences explicitly.

471

472   Importantly, in coastal marine and estuarine habitats, ΔC can shift as much as several thousand

473   μatm per day due to diel cycling associated with CO_2 production and depletion (Figs. 2, S2). The

474   uncertainty surrounding gas transfer velocity parameterization can represent a major source of

475   error in CO_2 flux calculations (Frankignoulle et al., 1998; Upstill-Goddard, 2006; Wanninkhof &

476   McGillis, 1999); however, small errors in k_{600} have far less effect on CO_2 flux calculations in

477   estuaries which experience pCO$_2$ swings of several thousand μatm during a single day, compared

478   with more stable conditions of the open ocean where interannual ranges of pCO$_2$ are typically far

479   less (Van Dam et al., 2019).

480

481   3.6 CO_2 flux - Seasonality and Interannual Variation

482   CO_2 flux was determined according to Eq. 3 using hourly ΔC measurements, CO_2 solubility

483   values (K_0) calculated according to temperature and salinity, and estuary-specific standardized

484   gas transfer velocities (k_{600}) of Van Dam et al. (2019). CO_2 flux was plotted across the three

485   years of observations at hourly and monthly intervals (Fig. 4a-b). As observed with pCO$_2$, CO_2

486   flux in the Rhode River was shown to be strongly seasonal. Given the similarity in windspeed

487   across seasons (Fig. S1), the effect of differential mean ΔC and variation between warm- and

488   cold-water seasons (Fig. 2, Table 1) almost certainly drives the observed seasonal differences in

489   CO_2 flux (Fig. 4). Again, the specific solubility of CO_2 is greatest at low temperatures, yet this is

490   contrary to the observed mean pCO$_{2water}$ patterns, pointing toward a biological mechanism for

491   pCO$_2$, ΔC, and ultimately, CO_2 flux. The effect size of season on CO_2 flux was two orders of

492   magnitude greater than either day/night or the season day/night by interaction (Table 2).

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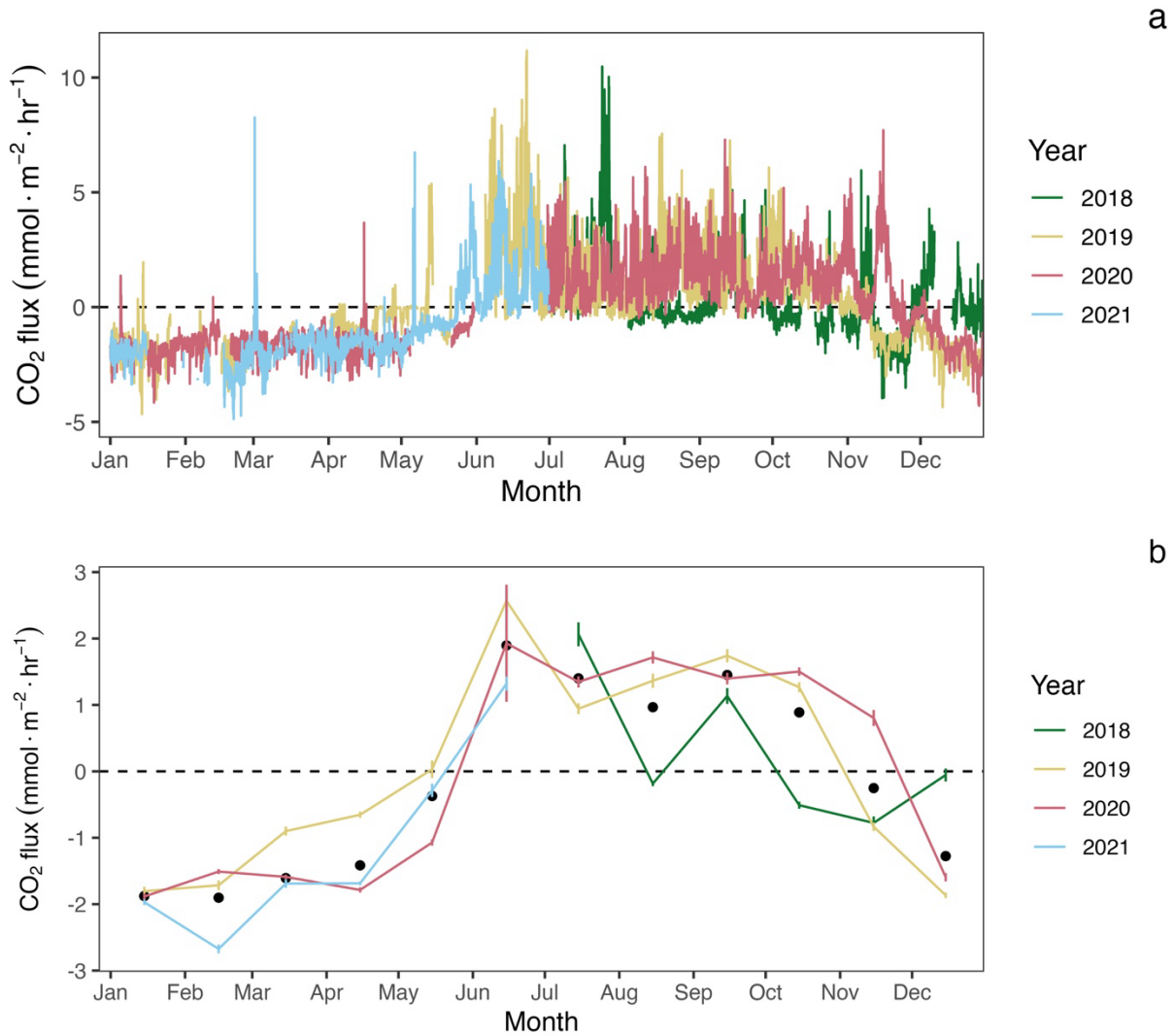
494

495   **Fig. 4**. CO_2 flux estimates by year: a. Hourly, b. Monthly average CO_2 flux estimates with 95%
496   confidence limits. Black dots in panel b indicate mean monthly fluxes across years.

497

498   Among years, pCO_{2water} and CO_2 flux largely repeat themselves, with dissolved CO_2 becoming

499   consistently sub-atmospheric and CO_2 flux going negative (gas exchange from atmosphere to

500   water) between December and May and abruptly transitioning to much higher maximum, yet

501   variable pCO_{2water} values with net positive CO_2 fluxes from June through November (Figs. 2 and

502   4). Monthly averaged CO_2 fluxes are consistent among years (Fig. 4b), with net positive CO_2

503   fluxes (heterotrophic conditions) between June and November and negative (autotrophic) fluxes

504   dominating when water temperatures are cold, between December and May. Despite the overall

505   similarities in seasonal CO_2 flux, inter-annual patterns can vary considerably. When hourly CO_2

506   flux values were averaged for the year, the Rhode River in 2019 was shown to have a net

507   positive flux but a net negative flux in 2020. When scaled for the year, 2019 outgassed CO_2 from

508   the water to the atmosphere at a rate of 2215.08 mmol $\cdot$ m^{-2} $\cdot$ yr^{-1} (95% CI = 1816.88, 2613.29).

509   The annual net flux rate in 2020 was negative (i.e. CO_2 moved from the atmosphere into the

510   river) at a rate of -1361.31 mmol $\cdot$ m^{-2} $\cdot$ yr^{-1} (95% CI = -1723.60, -999.01).

511

512   At shorter time scales, such as comparing the same week of the year among years, we sometimes

513   observed vast differences in the magnitude and direction of CO_2 flux (Fig. S3), signaling

514   differences in seasonal conditions between years. Transient events can also result in deviations

515   from otherwise typical CO_2 flux conditions. For example, the period from July 2018 to Jan 2019

516   deviated from other years as CO_2 flux was more erratic, with intermittent periods of negative and

517   positive CO_2 flux extending later into the winter season than in other years. When water

518   temperatures are compared among years, 2018 was shown to be more inconsistent, with more

519   pronounced temperature shifts and reversals than in 2019 or 2020 (Fig. S1). Salinities remained

520   relatively low for the latter half of 2018 into early 2019, reflecting wetter conditions (Fig. S1).

521   There were also two rapid salinity declines (>4 ppt reductions) in July and October 2018, likely

522   associated with strong precipitation events. These events were both followed by immediate

523   spikes in chlorophyll-*a* concentration to levels exceeding 200 μg $\cdot$ L^{-1}, indicative of

524   phytoplankton bloom conditions. From 2018 to 2021, chlorophyll-*a* levels of this magnitude and

525   greater were generally confined to cold-water months (Dec–May; Fig. S1 Erratic water

526   temperature and salinity are also reflected in more variable gas-specific solubility (K_0) for CO_2 in

527   2018 than later years (Fig. 3).

528

529   Gallegos et al. (1992) documented predictable phytoplankton blooms associated with freshets in

530   the Rhode River, when nutrient-rich freshwater inundates the estuary, not from point and non-

531   point sources within the local Rhode River watershed, but instead from the enormous watershed

532   that feeds the Susquehanna River, the primary source of freshwater input into the Chesapeake

533   above the Potomac as well as >50% of the entire Bay's freshwater (U.S. Geological Survey,

534   2023). Unlike river dominated estuaries, in the Rhode River estuary, volumetric influxes from

535   the Chesapeake Bay end member far exceed freshwater input from the Muddy Creek and

536   secondary tributaries. In the Rhode River, phytoplankton blooms result in the temporary

537   depletion of pCO$_{2water}$, followed by a spike, as phytoplankton senesce and organic carbon is

538   decomposed/re-mineralized back into inorganic carbon. Episodic, short-lived occurrences like

539   these demonstrate how immediate small scale biological forcing, can be coupled with, and

540   catalyzed by, distant large-scale weather and hydrological events. These in turn can influence

541   pCO$_2$ flux variations within seasons and among years (Fig. 3 and S3; and Chen et al., 2020).

542

543   Overall, except for wind speed, the effect sizes for the other six measured or calculated variables

544   were shown to be greatest for season vs. day/night or the interaction of season x day/night, and in

545   all cases the season effect was greater by at least 1 order of magnitude (Table 2). Seasonality has

546   10 to 1000 times more explanatory power than other variables investigated as estimated by ω^2

547   (Table 2).

548

549   3.7 Shifting Net Ecosystem Production

550   To better understand how the net ecosystem production (NEP) of the Rhode River shifts

551   throughout the year, where positive NEP indicates the river is storing carbon (autotrophic state)

552   and negative NEP indicates it is releasing carbon to the atmosphere (heterotrophic state), we

553   calculated hourly CO$_2$ flux values, averaged them by day (i.e. 24 hr period) and plotted each in

554   relation to the ΔC = 0 reference. Each day of the 3 yr study was categorized as either net

555   heterotrophic (CO$_2$ flux from water to atmosphere) or net autotrophic (CO$_2$ flux from atmosphere

556   to water). Each day was then further identified as either purely heterotrophic (all 24 hours were

557   heterotrophic), purely autotrophic, or mixed (some hours were heterotrophic and some were

558   autotrophic but resulting in a net autotrophic or net heterotrophic state for the day) (Fig. 5). From

559   July 2018 to July 2021, most 24 hr periods were categorized as pure autotrophic (444/920 =

560   48%), while 25% (229/990) were purely heterotrophic, and the remainder of mixed trophic status

561   (17% net heterotrophic and 10% net autotrophic; Fig. 5).

562

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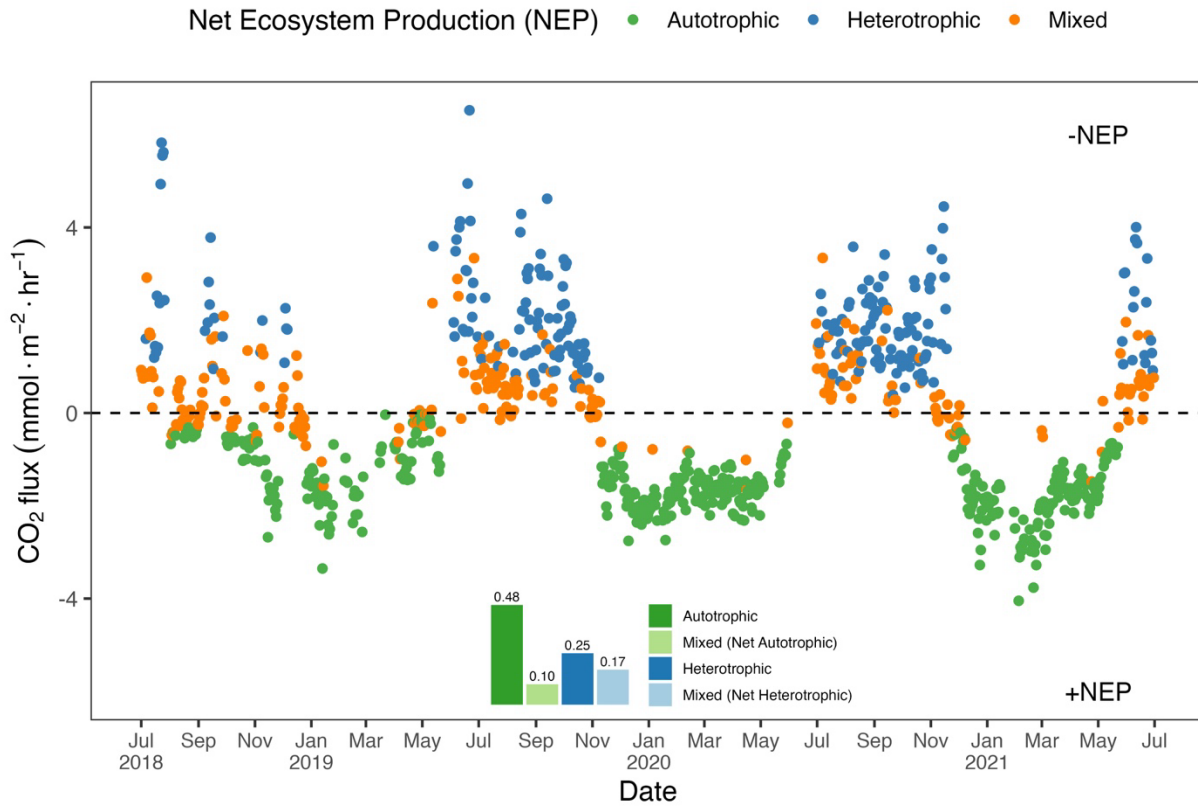

564

565   **Fig. 5**. Daily mean CO_2 flux estimates (CO_2 gradient is $CO_{2water} - CO_{2air}$). Green dots indicate
566   days when all 24 hourly flux measurements were negative (autotrophic with +NEP); blue dots
567   indicate days on which all 24 hourly flux measurements were positive (heterotrophic with -NEP).
568   Orange dots indicate that hourly fluxes were both negative and positive, and the position of the
569   orange dot below or above the zero line indicates whether the day was net autotrophic or net
570   heterotrophic. Insert describes the proportion of days in each category indicating that during 58%
571   (0.48 + 0.10) of days across three years of observation, the Rhode River was a CO_2 sink.

572

573   Altogether, the Rhode River was net autotrophic for (535 of 920 days = 58%) and net

574   heterotrophic for 42% (385 days) across three years. When CO_2 flux is integrated over all three

575   years, the Rhode River is shown to have near neutral NEP (Fig. 6). The effect size of season is

576   two orders of magnitude greater than either that of day/night or season:day/night interaction

577   (Table 2). Mean CO_2 flux values highlight the obvious correlation between season and NEP;

578   error bars (± 1 SD) reveal the importance of diel cycling where the magnitude and directionality

579   of day/night flux variability is approximately equal to the overall variability accrued across all

580   three years (Fig. 6). Although CO_2 flux is less variable and more autotrophic during cold-water

581   months than warm-months in the Rhode River, the range of possible values that occur across

582   night and day, regardless of season, must be taken into consideration to minimize incidental

583   sampling bias (Figs. 2 and 6).

584

585   A multi-year investigation of CO_2 flux in the main stem of Chesapeake Bay by Chen et al.

586   (2020) combined several bay-wide cruises that were distributed across seasons to collect discrete

587   and underway pCO_2 data for CO_2 flux calculations. They concluded that the low salinity upper

588   bay, which receives large volumes of freshwater from the Susquehanna River, was net

589   heterotrophic; the mesohaline middle bay was net autotrophic, and the polyhaline lower bay was

590   near carbon neutral. Chen et al. (2020) characterized Chesapeake Bay, on the whole, as a weak

591   source of CO_2 to the atmosphere (net flux = 0.73 mol $\cdot$ m^{-2} $\cdot$ yr^{-1}) but suggested that during wet

592   years, it may function as weak sink of CO_2. Herrmann et al. (2020) also concluded that the

593   Chesapeake Bay was a weak source of CO_2 to the atmosphere based on calculated pCO_2 values

594   from long term pH and alkalinity measurements (net flux = 1.2 mol $\cdot$ m^{-2} $\cdot$ yr^{-1}mol). Brodeur and

595   colleagues (2019) examined dissolved inorganic carbon (DIC) and total alkalinity along the

596   mainstem of the Chesapeake Bay across the year in 2016 and concluded that DIC increases from

597   north to south and from surface waters to depth, but that seasonal riverine input and biological

598   cycling were significantly important, concluding that the Bay as a whole was a sink for CO_2.

599

600   When our annual mean pCO_2 values were compared with the Chen et al. (2020) survey, the

601   Rhode River was shown to be higher on average and more variable than the mesohaline main

602   stem of the Bay (591 ± 652 vs. 416 ± 167 µatm), including a substantially greater measured

603   range (min = 15, max = 5182 µatm vs. 103 and 1033 µatm). These results suggest that water in

604   the shallow and well mixed Rhode River, and DIC in particular, undergo more acute biological

605   transformation than in the mesohaline main stem of Chesapeake Bay. Chen et al. (2020) point to

606   a variety of factors that affect pCO_2 and CO_2 flux in the main stem bay, including temperature,

607   depth, stratification, and freshwater input volume, some of which may attenuate biological

608   forcing. Interannual variability was demonstrated in both the Rhode River (some years were net

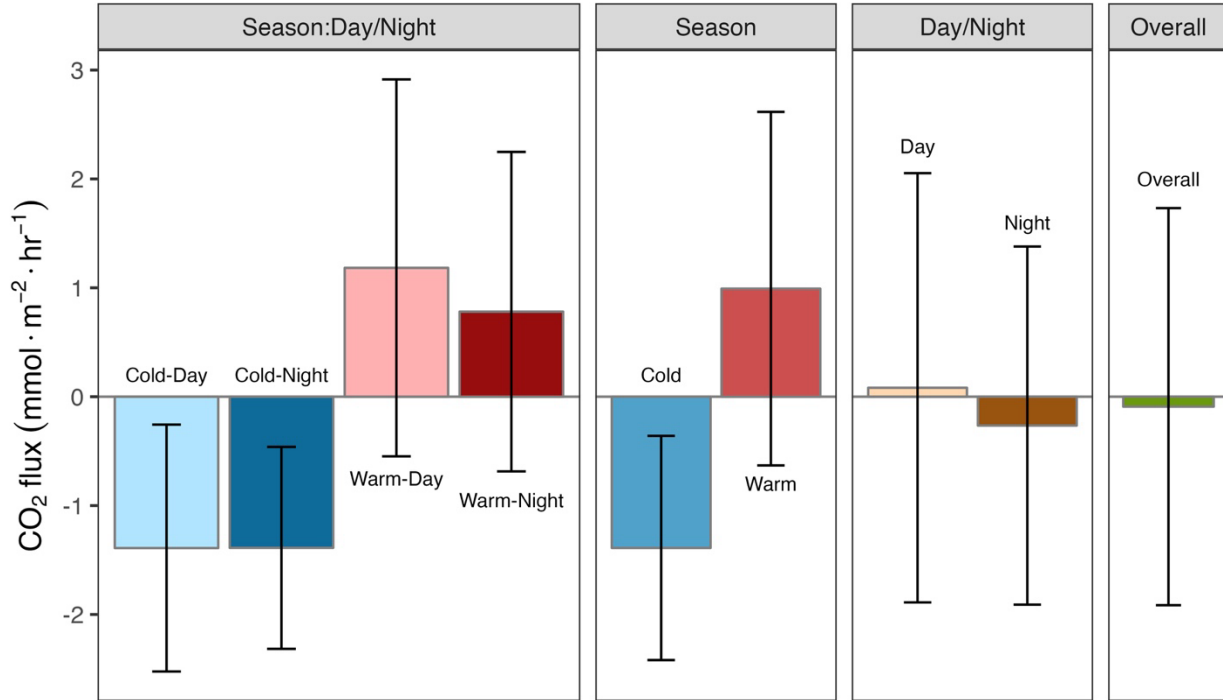
609   autotrophic and others heterotrophic, Figs. 4 and 5) and in the mesohaline main stem of the Bay;

610   however, we attribute interannual variability in pCO_2 and CO_2 flux primarily to variation in

611   water temperature that in turn drives biological activity. We conclude that seasonal variations the

612   Rhode River (and likely similar rivers in the mesohaline portion of the Chesapeake) are

613   significant and predictable, closely associated with water temperature, and that temperature

614   mediates NEP biologically rather than by changes to the solubility of CO_2.

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616

617   **Fig. 6.** Mean CO_2 flux $\pm$ 1 SD (mmol $\cdot$ m^{-2} $\cdot$ hr^{-1}) plotted by day/night cycling, cold-water/warm-
618   water season, season by day/night interaction, and overall CO_2 flux across three years of
619   observation.

620

621   In the Rhode River, we find that CO_2 flux reverses itself daily for part of the year (June–

622   November) yielding some days that are characterized as a net sink (net autotrophic) and others

623   that are a net source (net heterotrophic). From December to May, diel cycling is minimal and the

624   river is almost exclusively a net sink, autotrophic both day and night. Finally, although CO_2 flux

625   is pronounced but variable across seasons, the net CO_2 flux of the Rhode River on an annual

626   basis is near neutral.

627

628

629

630

631   3.8 Lateral transport

632   Tidal cycling has been shown to liberate and laterally transport DOC from brackish marshes to

633   adjacent estuaries (Cai, 2011; Herrmann, 2015) and therefore is of great importance to carbon

634   cycling and budgets of wetlands and estuaries (Najjar et al., 2020). DOC outwelling from the

635   Kirkpatrick Marsh (hereafter KPM), a 0.21 km^2 tidal marsh located approximately 1 km up

636   estuary from our primary study site at the SERC Dock (Fig. 1), into the Rhode River has been

637   measured and modeled extensively in recent years (Clark et al., 2020; Menendez et al., 2022;

638   Tzortziou et al., 2011; Tzortziou et al., 2008). These studies indicate that the KPM is responsible

639   for a large portion of overall DOC input to the Rhode River, as well as significant export from

640   the river to the mainstem of Chesapeake Bay. Model generation and validation by Clark et al.

641   (2020) indicate that up to 13.1% of the total DOC input to the Rhode River originates in the

642   KPM. Another important source (53% of total) is DOC derived from phytoplankton and is

643   therefore labile and readily biodegraded and remineralized into DIC. Furthermore, large

644   quantities of other, semi-labile forms of DOC are exported from the KPM, which are themselves

645   subject to photochemical and biodegradation and remineralization (Clark et al., 2020).

646   Importantly, each of these DOC streams provides a potential source of DIC, including pCO$_2$, to

647   the Rhode River.

648

649   Dissolved inorganic carbon generated in brackish tidal wetlands is also outwelled directly into

650   estuaries (e.g., Cai et al., 2000; Chu et al., 2018; Song et al., 2023). Recent work by Song et al.

651   (2023) demonstrates that pCO$_2$ in a salt marsh tidal creek in Waquoit Bay, MA was regulated by

652   both tide height (inversely) and the day/night cycle, with nighttime low tides resulting in the

653   highest pCO$_2$ values, signaling a strong local effect from respiration and photosynthesis in

654   combination with tidal outwelling.

655

656   In the Rhode River watershed pCO$_2$ was measured continuously in the single tidal creek that

657   drains the KPM using the same methods as at our primary study location. We observed that the

658   KPM tidal creek pCO$_2$ follows the tidal cycle exclusively, yet outside the mouth of the tidal

659   creek, in the estuary proper, day/night cycling overwhelms this marsh tidal signal. Simultaneous

660   pCO$_2$ measurements from the SERC dock follow a strict day/night cycle (Fig. S4). However,

661   while peak levels of dissolved CO$_2$ in the Kirkpatrick Marsh creek occur at low tide and can

662 reach values nearly 20 times greater than highs at the SERC dock (Fig. S4) there is no obvious

663 evidence of this tidal DIC input at the dock site. Remineralization of DOC exported from the

664 KPM, as well as DOC originating in other locations within the watershed are important sources

665 of DIC in the river, but given the relative volumes of these sources to that of the much larger

666 estuary, as well as the physical distance (~1 km) from SERC dock, these input signals should be

667 expected to be lagged and damped inside the estuary and not tightly coupled with tidal cycles.

668 Instead, pCO_2 exported from the KPM is expected to undergo significant dilution effects, be

669 partially off-gassed to the atmosphere, and be metabolized via photosynthesis, reducing its

670 influence on downstream sites. These findings suggest that despite periodic extreme pCO_2 in

671 KPM tidal creek (>30,000 ppmv), the overall mass of CO_2 export is not sufficient to have

672 measurable effects on the deeper, well-mixed portions of the Rhode River.

673

674 Thus, although land – sea interfaces and outwelling of DOC and DIC are important in estuaries

675 and coastal ecosystems, the relative sizes of wetlands and adjacent water bodies and the overall

676 volume of water moving between the two are also important factors. In eutrophic estuaries like

677 the Rhode River, biological forcing can rapidly assimilate DIC and degrade and mineralize labile

678 forms of DOC, as evidenced by extensive diel cycling in these systems (e.g., Brodeur et al. 2019;

679 Song et al. 2023, and the present study.) The much larger and complex Chesapeake Bay

680 generally follows seasonal changes in pCO_2 and CO_2 flux, but these appear to be most

681 predictable in the upper oligohaline portion and the polyhaline region of the bay near the mouth,

682 where freshwater and oceanic end-member effects are most pronounced (Brodeur et al. 2019;

683 Chen et al., 2020). The central mesohaline part of Chesapeake Bay comprises numerous discrete

684 and unique watersheds and subestuaries/rivers, each of which exchanges water with the bay.

685 Elucidating spatial and temporal patterns of pCO_2 and CO_2 flux are vital for understanding each

686 one's role as an atmospheric source or sink, but also could provide better insight into how each

687 may be influenced by global increases in atmospheric CO_2 (i.e., acidification and its influences

688 on estuarine metabolism, and the local biota, fisheries, and habitats each support.) Collectively,

689 these and other subestuaries will have cumulative effects on the overall water quality of

690 Chesapeake Bay, including cycling of DOC and DIC, which in turn affect pCO_2 and CO_2 flux.

691

692

693   **4. Conclusion**

694   The notion that estuaries are predominantly heterotrophic systems that invariably outgas more

695   CO_2 to the atmosphere than they absorb has been a long-held view (Abril et al., 2000; Borges et

696   al., 2004; Cai, 2011; Cai et al., 2000; Chen, 2013; Frankignoulle et al., 1998, Gattuso et al.,

697   1998). However, more recently investigators have realized that physical and hydrological

698   characteristics, geographical location, size, and biological and biogeochemical activities may

699   individually, or together, influence CO_2 flux in estuaries and therefore contributions to

700   atmospheric chemistry (Brodeur et al. 2019; Caffrey, 2004; Chen et al., 2013, 2020; Herrmann et

701   al., 2020). As indicated in this study and others, the role that biological processes play in

702   estuaries to either fix CO_2 (autotrophy) or liberate CO_2 (heterotrophy) are extensive, complex,

703   and can be quite variable over space and time (Brodeur et al. 2019; Chen et al., 2020; Herrmann

704   et al., 2020; Rosentreter et al., 2021). High frequency automated measurements revealed strong

705   seasonal contrasts in dissolved CO_2 content and CO_2 flux between water and atmosphere of the

706   Rhode River, a shallow mesohaline reach of the Chesapeake Bay. Importantly, only through high

707   frequency, multi-year measurements could diel and seasonal cycling be fully discerned. Indeed,

708   inadequate sampling can induce bias (e.g., upscaling from a small number of daytime samples

709   taken during warm-water months can skew apparent patterns; Laruelle et al., 2017; Van Dam et

710   al., 2019.). The timing and frequency of measurements are critical and have potential for strong

711   and misleading biases if sampling is insufficient. In contrast, cold-water months coincide with

712   long periods (weeks to months) of continuous sub-atmospheric sink conditions for CO_2. Using

713   these measurements, we estimated the direction and magnitude of CO_2 flux in hourly, daily, and

714   annual terms. In the Rhode River CO_2 flux reverses itself daily for part of the year (June through

715   November) yielding some days that are characterized as net sink (net autotrophic and NEP > 0)

716   and others that are net source (net heterotrophic and NEP < 0). From December to May diel

717   cycling is minimal, and the river is almost exclusively a CO_2 sink with +NEP both day and night.

718   Although CO_2 flux is pronounced but variable across seasons, the net CO_2 flux of the Rhode

719   River on an annual basis is near carbon neutral, although some years are net heterotrophic and

720   others net autotrophic.

721

722   High frequency sampling of $p$$CO_2$, although typically confined spatially, is one approach to

723   understanding fundamental aspects of estuarine metabolic states and CO_2 flux that may

724 otherwise go undetected (Song et al., 2023). To address the spatial complexity of estuarine,

725 nearshore, and inland waters, more observation locations are required. As with any

726 environmental or ecological question, careful sampling design is critical to balance efficiency

727 and statistical power.

728

729 As the largest and arguably most complex estuary in the United States, the Chesapeake Bay is

730 the subject of extensive ecosystem management efforts and ranks among the most studied and

731 monitored estuaries in the world (Boesch and Goldman 2009). Yet, information on CO_2 and

732 GHG fluxes continue to be limited (Brodeur 2019; Chen et al., 2020; Herrmann et al., 2020).

733 Given the extensive coordinated monitoring programs that either make real-time water quality

734 measurements and/or maintain routine water sampling schedules (e.g., Maryland DNR "Eyes on

735 the Bay" program) in this region, existing water quality observation assets and sampling

736 programs could be leveraged to more fully characterize and quantify CO_2 and other GHG

737 dynamics and flux in the Bay and elsewhere (see Saba et al. 2019). For example, coordinated

738 deployment of additional automated sampling devices (e.g., robust air-water equilibrators and

739 traditional atmospheric gas sensors) in key locations would enable estimates of CO_2 flux, and if

740 combined with pH, DIC, or total alkalinity measurements, carbonate chemistry calculations as

741 well. Importantly, such installations need not be permanent. Instead, a small group of

742 instruments could be systematically deployed across an existing observation network, co-located

743 with other water quality instruments using a stratified sampling approach to capture spatial

744 variability. For example, a set of shifting two to four week deployments during summer and

745 winter months could yield sufficient data to advance our understanding of Chesapeake Bay-wide

746 CO_2 flux significantly in a single year. Such information would complement underway transects

747 that are vital, but which tend to underestimate temporal variability in any given location. In the

748 case of dissolved GHGs, liquid-air equilibration techniques are being used to measure multiple

749 GHGs (Call et al. 2015; Hartmann 2018; Gülzow et al. 2011; Miller et al. 2019; Xiao et al.

750 2020).

751

752 Understanding the GHG dynamics in estuaries is a vital component to generating accurate global

753 budgets (Maher & Eyre, 2012) as well as informing where emerging carbon capture

754 technologies, including nature-based solutions, might be best located (Bradshaw & Dance, 2005;

755   Sun et al., 2021). In the case of estuaries, there have been extensive global losses of seagrasses

756   due to habitat degradation, pollution, and disease (Waycott et al. 2009). In addition to many

757   other ecosystem service benefits, restoration of seagrass and submerged aquatic vegetation has

758   the potential to restore and enhance natural carbon sequestration (i.e. blue carbon; Kennedy et al.

759   2022; Macreadie et al. 2022; Unsworth et al. 2022). In Virginia, U.S.A., Oreska et al. (2020)

760   demonstrated how the functional benefits of a restored seagrass meadow habitat can be

761   quantified ecologically in terms of their ability to sequester carbon and affect GHG fluxes

762   between the estuary and atmosphere. Uniquely, these investigators then monetized the costs and

763   benefits of habitat restoration and function as CO_2 offset credits, as part of a GHG budget, and

764   demonstrated how such approaches can be used to incentivize habitat restoration (Oreska et al.

765   (2020).

766

767   Increasing the completeness and utility of global GHG budgets, as they relate to human activities

768   and ecosystem functions, are necessary steps toward combating global climate change.

769   Measurement of GHGs at high spatial and temporal resolution using economical, automated

770   measurement solutions can increase our understanding of GHG dynamics at small ecologically

771   significant scales, as well as at the larger ecosystem level of an estuary.

772

773   **Data Availability**

774   Hourly means of pCO$_2$ and associated environmental data used in the analyses are available at

775   the Smithsonian Figshare repository https://doi.org/10.25573/serc.22491655 via under Creative

776   Commons license CC BY-NC 4.0.

777

778   **Author Contributions**

779   AWM contributed to project Conceptualization, Funding acquisition, Investigation,

780   Methodology, Project Administration, Resources, Supervision and Writing – Original Draft.

781   JRM contributed to Data Curation, Formal Analysis, Software and Visualization. ACR

782   contributed to Data Curation, Investigation, Methodology and Project Administration. MSM

783   contributed to Conceptualization, Supervision and Visualization. KJK contributed to

784   Conceptualization, Data Curation, Software, Validation. All authors contributed to Writing –

785   review and editing.

786

787 **Competing Interests**

788 The corresponding author has declared that none of the authors has any competing interests.

789

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795

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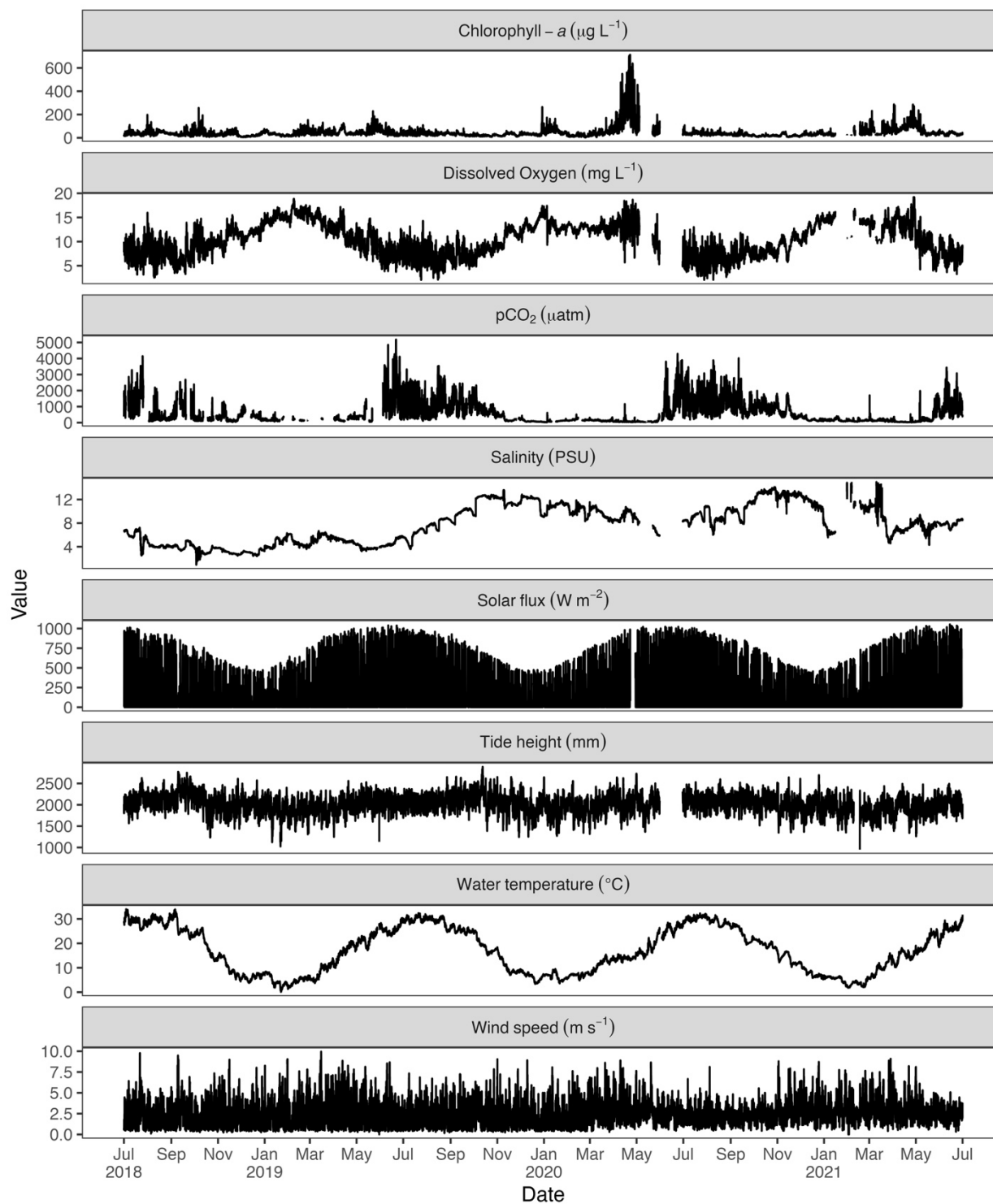
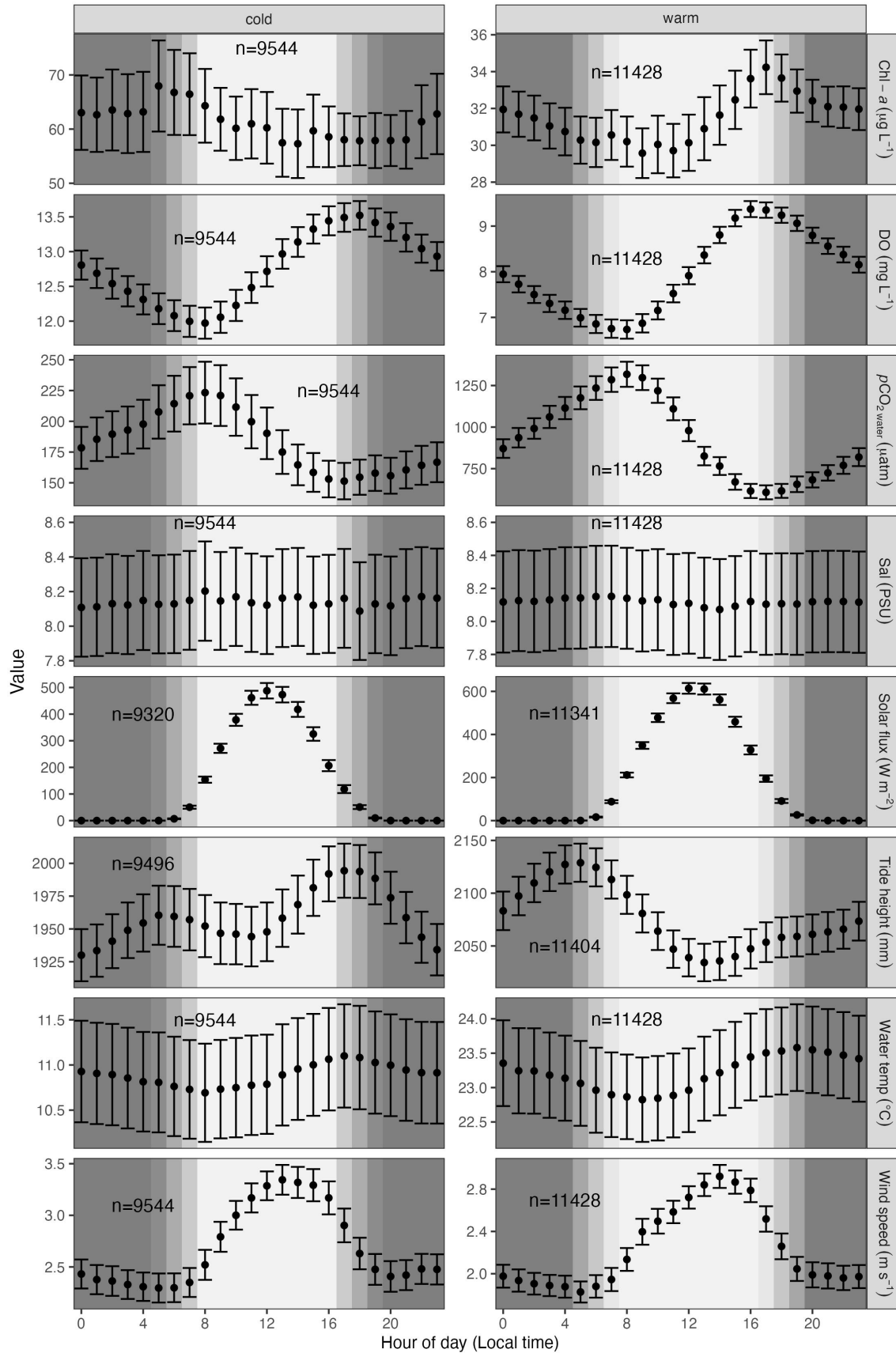
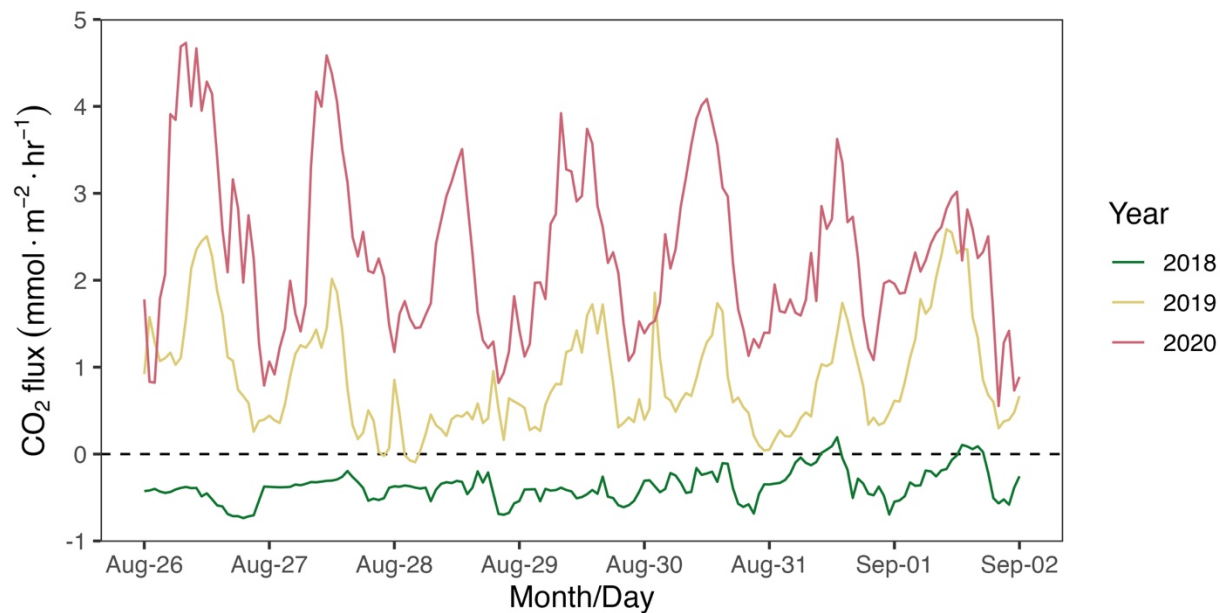
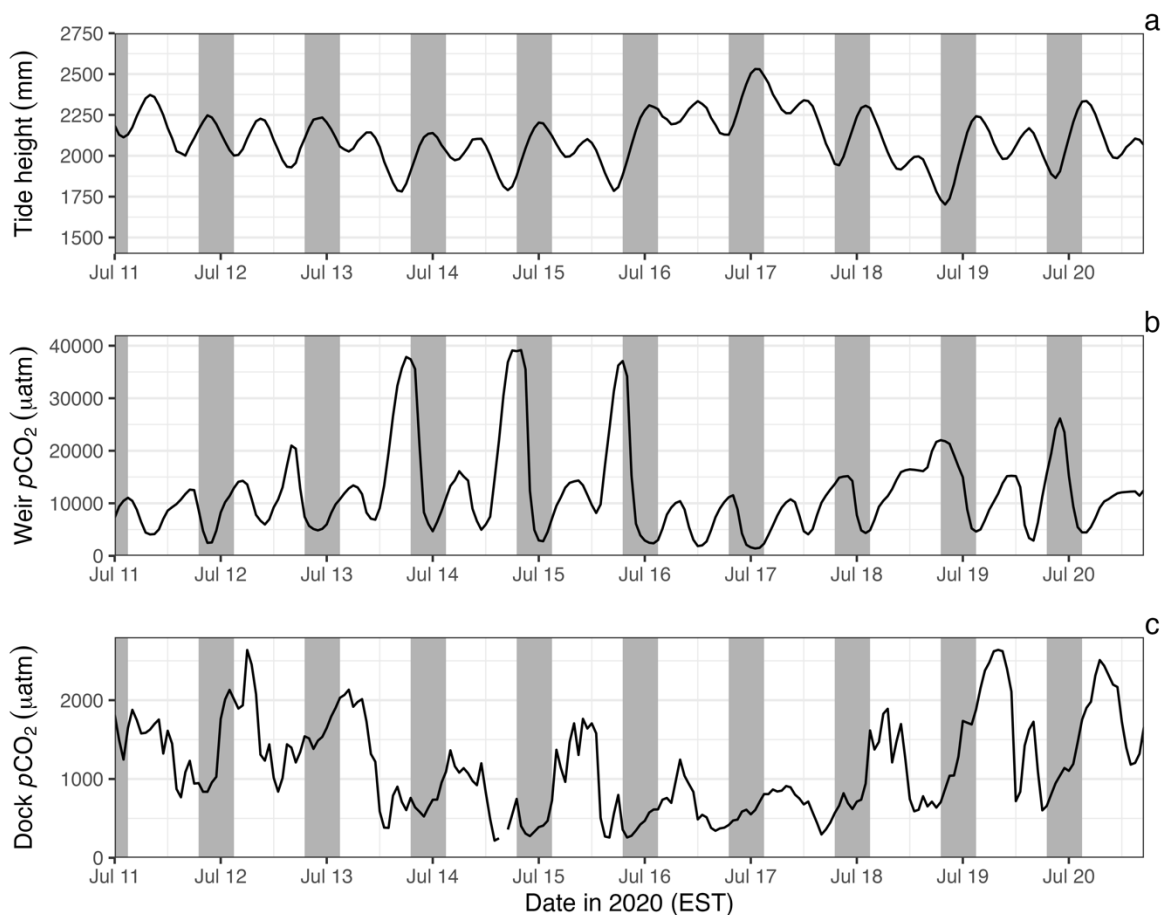
1051 **Supplemental**

1052
1053 **Fig. S1**. Plot of all raw values from environmental variables for the same time period as CO_2 flux
1054 (July 2018–July 2021).

1056   **Fig. S2**. Average hourly values (95% CI) of environmental variables across 24 hours of the day
1057   (July 2018–July 2021) in cold and warm seasons. Light/dark background indicates day/night
1058   conditions.
1059
1060

1061
1062   **Fig. S3.** Hourly CO_2 flux estimates for the week of August 26 to September 2 where CO_2 flux
1063   status differs among years.
1064

1065
1066 **Fig. S4.** Simultaneous pCO$_2$ measurements (1 hr intervals) from SERC dock (panel c) and the
1067 mouth of the single tidal creek that drains the Kirkpatrick Marsh (panel b) (11–20 Jul 2020)
1068 indicate that dissolved CO$_2$ varies at the dock according to a day/night cycle while CO$_2$ in the
1069 marsh tidal creek rises and falls inversely with tide height (panel a), indicating outwelling of
1070 marsh derived CO$_2$ (e.g., root respiration, pore and groundwater).