Contribution of biological effects to carbonate-system variations and the air-water CO2 flux in inner and outer bays in Japan

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Abstract

We evaluated the contribution of biological effects (photosynthesis, respiration, and decomposition) to the carbonate parameters and air-water CO₂ fluxes in Tokyo Bay, Ise Bay and Osaka Bay in Japan. The carbonate parameters were measured mainly by cargo ships travelling between Japan and other countries. We used the measurement data from three inner bays and surrounding outer bays in Japan along with reference data from previous studies for complementary analysis. We found that 1) the inner bays in this study were strong annual atmospheric CO₂ sinks, 2) the annual biological effect on the air-water CO₂ fluxes was about 5-25% of the measured CO₂ fluxes and it affected the seasonal variation of the CO₂ flux, and 3) the biological effect was largest in Tokyo Bay, and almost the same in Ise and Osaka Bays. The intensity of the biological effect corresponded mainly with nutrient concentrations, which seemed to be controlled by the wastewater treatment in urbanized areas around the bays. The CO₂ flux was also affected by the seawater residence time, salinity, and stratification. Our results suggest that labile carbon/nutrient ratio of wastewater should be a major consideration for evaluating the biological effect on the carbon cycle in urbanized inner bays, which will continue to expand globally.

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24	Key points:
25	• Inner bays in Japan are reportedly annual atmospheric CO_2 sinks but
26	comprehensive measurements are few.
27	• We quantified biological effects on carbonate parameters and CO_2 flux using an
28	empirical relationship with riverine and oceanic endmembers.
29	• The biological effect was regulated mainly by nutrient concentrations related to
30	wastewater treatment.
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36 Abstract

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38 We evaluated the contribution of biological effects (photosynthesis, respiration, and 39 decomposition) to the carbonate parameters and air-water CO₂ fluxes in Tokyo Bay, Ise 40 Bay and Osaka Bay in Japan. The carbonate parameters were measured mainly by cargo 41 ships travelling between Japan and other countries. We used the measurement data from three inner bays and surrounding outer bays in Japan along with reference data from 42 43 previous studies for complementary analysis. We found that 1) the inner bays in this study were strong annual atmospheric CO_2 sinks, 2) the annual biological effect on the 44 45 air-water CO₂ fluxes was about 5-25% of the measured CO₂ fluxes and it affected the 46 seasonal variation of the CO_2 flux, and 3) the biological effect was largest in Tokyo Bay, 47 and almost the same in Ise and Osaka Bays. The intensity of the biological effect 48 corresponded mainly with nutrient concentrations, which seemed to be controlled by the 49 wastewater treatment in urbanized areas around the bays. The CO₂ flux was also 50 affected by the seawater residence time, salinity, and stratification. Our results suggest 51 that labile carbon/nutrient ratio of wastewater should be a major consideration for 52 evaluating the biological effect on the carbon cycle in urbanized inner bays, which will 53 continue to expand globally.

54

55 Plain Language Summary

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57 We analyzed the biological effect (photosynthesis, respiration, and decomposition) on air-water CO₂ exchange in Tokyo Bay, Ise Bay and Osaka Bay in Japan using data from 58 59 cargo-ship measurements and previously published reports. We concluded that 1) bay 60 water strongly absorbs atmospheric CO₂, 2) biological effects accounted for 5–25% of 61 the evaluated CO_2 absorption and had significant effects on its seasonal variation, and 3) 62 the biological effects seemed to be mediated mainly by the carbon/nutrient ratio in 63 wastewater. This study should improve our understanding of the carbon flow in 64 urbanized coastal areas, which are expanding globally.

66 1. Introduction

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68 The ocean is one of the largest carbon reservoirs on earth, and the 69 quantification of the exchange of atmospheric carbon dioxide (CO_2) with the ocean is 70 necessary for predicting future climate change. The air-water CO₂ flux in the major 71 oceans has been studied since the late 1970s and the regional and seasonal variations 72 have been estimated (e.g., Takahashi et al., 2009; Wanninkhof et al., 2019). Meanwhile, 73 the quantification of the flux in coastal areas is still challenging because of the large 74 temporal and spatial variations. Recent studies have shown that near-shore areas are 75 sources of atmospheric CO₂ on average because of the input and of organic carbon and 76 the mineralization (Borges et al., 2005; Cai, 2011; Chen and Borges, 2009; Chen et al., 77 2013), whereas some other studies showed a local annual CO_2 sink in areas with 78 submerged autotrophic ecosystems (Kayanne et al., 1995; Tokoro et al., 2014). Marginal 79 seas (continental shelves) have been reported as atmospheric CO_2 sinks but there is still 80 uncertainty surrounding the actual estimates has been debated (Borges et al., 2005; Cai, 81 2011; Chen and Borges, 2009; Chen et al., 2013; Laruelle et al., 2014).

82 The role of inner and outer bays in atmospheric CO₂ exchange has not been 83 analyzed comprehensively despite their importance in anthropogenic activities (Chen and Borges, 2009; Chen et al., 2013). Although the near-shore area is generally 84 85 considered to be a CO₂ source region, some studies have reported that urbanized inner 86 bays in Japan are annual atmospheric CO₂ sinks (Endo et al., 2017; Fujii et al., 2013; 87 Kubo et al., 2017). In these bays, CO₂ undersaturation might result from wastewater 88 treatment (Kubo et al., 2017; Kuwae et al., 2016). The treatment process removes 89 labile carbon, yielding water with relatively less carbon than nutrients, which promotes 90 primary production in the bay water. In addition, the organic matter remaining in the 91 treated water is refractory (Kubo et al., 2015) and thus further mineralization and 92 increase in the CO₂ concertation in the bay water is suppressed. However, the effects of biological processes such as photosynthesis, respiration, and decomposition on the 93 94 carbonate parameters and the air-water CO₂ flux have not been precisely quantified.

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Here, we evaluated the biological effects of inner-bay water on temporal and 96 spatial changes in the carbonate parameters and the air–water CO_2 flux, including the 97 surrounding outer bays, in Japan. We discuss the biological effects on the flux in terms 98 of the regulating factors and the extendibility or our observations to other areas and for 99 global estimation.

102

103 2.1 Study site

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105 This study took place in Japan in Tokyo Bay, Ise Bay, and Osaka Bay, and 106 surrounding areas (Figure 1). These bays have similar topographic conditions such as a 107 southward entrance to the bay open to the Pacific Ocean and a surrounding, highly 108 urbanized coastal area. The populations of the urban areas surrounding Tokyo, Ise and 109 Osaka Bays were 37, 9 and 19 million, respectively, in 2014 (Global Metro Monitor; 110 https://www.brookings.edu/research/global-metro-monitor/). In this study, the term "bay" 111 refers to the inner and outer areas of each bay, which were analyzed together for 112 expedience.



114 Figure 1. (a) Map of Japan and the general locations of the bays and surrounding seas included in 115 this study. (b) Tokyo Bay, (c) Ise Bay, (d) Osaka Bay. The black filled circles indicate the locations 116 of the National Institute for Environmental Studies (NIES) data used in this study. The grey circles in 117 (b) are filtered NIES data from east of 140°E, which were defined as data external to Tokyo Bay. The 118 blue circles indicate additional data collected by Tokyo University of Marine Science and 119 Technology (TUMSAT) in Tokyo Bay and by Osaka City University (OCU) in Osaka Bay. The red 120 stars mark the river mouths of the main river located in the inner part of each bay and used as the 121 zero point for the distance parameter dist (Tokyo Bay, 35.65°N, 139.85°E; Ise Bay, 35.04°N, 122 136.74°E; Osaka Bay, 34.68°N, 135.41°E).

123

124 2.2 Data processing

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126 The data used for this study are measurements of water temperature, salinity 127 and fugacity of CO_2 (fCO₂) in the main urbanized inner bays and in the surrounding 128 outer bays, as observed by the National Institute for Environmental Studies (NIES), 129 Tokyo University of Marine Science and Technology (TUMSAT) and Osaka City 130 University (OCU). The other carbonate parameters-total alkalinity (TA) and 131 dissolved inorganic carbon (DIC)-were estimated using an empirical relationship 132 between TA and salinity, and the equilibrium calculation. The biological effect on 133 carbonate parameters and the air-water CO₂ flux was calculated from the difference 134 between the above estimated DIC and the value from the conservative mixing line 135 between the oceanic and riverine endmembers. We quantified the air-water CO₂ flux 136 and the magnitude of the biological effect to evaluate the contribution of the biological 137 effect in the bays to the exchange of atmospheric CO₂.

138 The water temperature, salinity, and fCO₂ in water and air were obtained from 139 the NIES database (https://soop.jp). The data are also available from the Surface Ocean 140 CO₂ Atlas (SOCAT; http://www.socat.info), which has been a public database since 141 2011 and represents an international collaboration among research institutes. The NIES 142 observations implemented as the Voluntary Observing Ship (VOS) programs by cargo 143 ships were originally for understanding the global carbon cycle but their data also 144 include inner bay measurements from 2005 to 2016 in Tokyo Bay and Ise Bay and 145 from 2011 to 2016 in Osaka Bay, where the cargo ships taking measurements have 146 anchor stations. We first extracted the data for 30-40°N and 130-145°E observed by 147 the cargo ships as the original coastal data. The details on fCO₂ measurements are 148 reported by Nakaoka et al. (2013).

149

For complementary analysis, we included some previous data as more

150 landward information than the NIES data. For Tokyo Bay, we added the data collected 151 by TUMSAT from 2007 to 2010 (Kubo et al., 2017) to the NIES data after converting pCO₂ to fCO₂ using the empirical relationship incorporating temperature (Körtzinger, 152 153 1999). Likewise, for complementary analysis, we added more landward measurement 154 data from Osaka Bay collected by OCU in spring and autumn of 2014 (Endo et al., 155 2017). The additional raw data for Osaka Bay included the water temperature, salinity, and DIC. We therefore estimated fCO₂ using the equilibrium calculation (Zeebe and 156 157 Wolf-Gladrow, 2001; we used their "recommended" coefficients for the calculation) 158 and the TA from the empirical relationship with salinity (Taguchi et al., 2009).

159 The data were filtered by the distance from the inner part of each bay. The160 distance parameter *dist* (km) was calculated as follows:

161

162
$$dist = 6370 \times \sqrt{(lat - dist0_lat)^2 + (lon - dist0_lon)^2 \times \cos\left[\left(\frac{lat + dist0_lat}{2}\right)^2\right]}$$
(1)

163

164 where *lat* and *lon* are the latitude and longitude of the measurement point in radians, 165 respectively. *dist0_lat* and *dist0_lon* are the latitude and longitude of the point 166 representing the mouth of the river with the highest flow among the rivers in the inner 167 part of each bay (Ara River in Tokyo Bay, Kiso River in Ise Bay, and Yodo River in 168 Osaka Bay; Figure 1). The data for which dist > 100 km were excluded from analysis as 169 being out-of-range of the inner and outer bays, as determined by changepoint analysis 170 using the salinity and DIC (see Text S1). In addition, the data from Tokyo Bay with 171 longitude $>140^{\circ}E$ were excluded because even though these data were within the range 172 of $dist \le 100$ km, they were from locations on the opposite side of a peninsula from the 173 inner bay (Figure 1). In total, we analyzed 18,118 data points from Tokyo Bay (16,924 174 from TUMSAT), 1926 from Ise Bay and 1067 from Osaka Bay (28 from OCU).

Because the NIES data for the inner bays were distributed uni-dimensionally along the course of the cargo ships, the spatial information for the data in this study was standardized by the distance parameter *dist* mentioned above. For evaluating seasonality, we calculated the parameter *monthlydata* for each data point using the temporal information as follows:

180

181
$$monthlydata = month + \frac{(day-1)}{365.25} \times 12$$
 (2)

182

183 where *month* and *day* are the month (1-12) and day (1-28 or 30 or 31) of the

184 measurement, respectively. Additionally, given the temporal and spatial heterogeneity 185 of the measurement data, we used natural neighbor interpolation (Sibson, 1981) to 186 interpolate a grid with values at intervals of 1 km (dist) and 0.1 (monthlydata). In order 187 for the interpolation to reflect the seasonal cycle from the minimum and maximum 188 values for *monthlydata* (on 1 January and 31 December, respectively), the interpolation 189 was performed with the data from the latter half of the year (July to December) added 190 before the start of original data and that from the first half (January to June) appended 191 to the end.

192 In order to minimize the effect of the trend in fCO_2 due to anthropogenic CO_2 193 input to the ocean, we corrected for the increase rate in each bay and in the additional 194 data, except for the OCU data because those measurements were only taken for a single 195 year. The linear increase rate for each bay was estimated from the annual average fCO₂. 196 The increase in fCO_2 in water was corrected to that in 2010 if the increase was 197 significant. The linear rates of fCO₂ increase in water were 5.83 and 3.24 μ atm yr⁻¹ in 198 Tokyo and Ise Bay, respectively, and were significant (p < 0.001), whereas that in Osaka 199 Bay was not significant (p = 0.64). Also, the additional TUMSAT data for Tokyo Bay 200 showed no significant increase (p = 0.85). Therefore, the rate of increase was corrected 201 to the base year of 2010 only for NIES data for Tokyo Bay and Ise Bay. For fCO₂ in air, we assumed an increase of 1.5 μ atm yr⁻¹ and corrected to this from the base year value. 202 203

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206

205 2.3 Calculation of the biological effect

207 Because fCO_2 in water is affected by temperature and salinity, the biological 208 effect cannot be quantified using only carbonate-system parameters. Although some 209 previous studies normalized fCO₂ using an empirical relationship between temperature 210 and fCO₂ (e.g., Takahashi et al., 2009), such a technique cannot be applied to coastal 211 carbonate analysis because the salinity varies widely and affects TA and DIC 212 non-linearly. Instead, we evaluated the biological effect using DIC, which is a 213 conservative parameter, and direct information about the biological effect from 214 processes such as photosynthesis, respiration, and decomposition (Figure 2). The DIC 215 was calculated using the carbonate equilibrium calculation (Zeebe and Wolf-Gladrow, 216 2001) with TA estimated from the empirical relationship with salinity in each bay from 217 a previous study (Taguchi et al., 2009).

Calculation protocol



218 219

Figure 2. Procedure for calculating abiotic and biotic CO_2 flux. 1: Calculation of total alkalinity (TA) using the empirical relationship of Taguchi et al. (2009). 2: Carbonate equilibrium calculation (Zeebe & Wolf-Gladrow, 2001). 3: Calculation of abiotic dissolved inorganic carbon (DIC) using the interpolation of the oceanic and riverine endmembers. 4: Calculation of biotic parameters as the difference between the net parameter and the abiotic parameter. 5: Calculation using equation (4). fCO₂, CO₂ fugacity.

226

For quantifying the biological effect, we defined DIC_b as the increase or decrease in DIC through biological activities. The DIC_b was calculated as the difference between DIC estimated from the equilibrium calculation and DIC_{ab} , which was the interpolated value between the oceanic and riverine endmembers in the coordinate system of salinity and DIC (Tokoro et al., 2014). DIC_{ab} represents the DIC concentration resulting from the mixing of the endmembers without any additional

change in the bays such as through biological activity.

The endmembers and the biological effect were determined as follows. 1) the salinity and DIC of the oceanic endmember were estimated as the value outside of the respective outer bays. In this study, those values were estimated as the monthly average

of the salinity and DIC between *dist* of 90 km and 100 km. 2) We assumed that

biological activity reached a minimum (DIC_b \approx 0) in the winter period (here, the three

239 months with the lowest average water temperatures), according to the positive

240 correlation between production by seagrass and algae and water temperature (Davison,

241 1991; Lee et al., 2007; Tait & Schiel, 2013). Then the least-squares line of the salinity 242 and DIC through the oceanic endmember average value during the above winter period 243 can be used to approximate the DIC_{ab} in the winter period, and the riverine endmember 244 (DICr, μ mol kg⁻¹). That is, the former parameter is the approximation using the 245 regression line of DIC vs salinity, and the latter parameter is the regression line value

when the salinity is zero. These parameters can be estimated as follows:

247

248
$$DICr = \frac{\sum [(DIC_w - DIC_{ow})(Sal_w - Sal_{ow})]}{\sum [(Sal_w - Sal_{ow})^2]} \times (Sal_r - Sal_{ow}) + DIC_{ow}$$
(3)

249

250 Here Sal_w and DIC_w are the salinity and DIC in each bay during the three months with the lowest average water temperature. Sal_{ow} and DIC_{ow} are the mean values of the 251 252 salinity and DIC, respectively, of the oceanic endmember during these three months. 253 Sal_r is the salinity of the riverine endmember and assumed to be zero in this study. 3) 254 DIC_{ab} in each month was calculated again as the linearly interpolated value between the 255 endmembers. To evaluate the uncertainty of the estimated riverine DIC, we defined the range of DIC as $\pm 200 \ \mu mol \ kg^{-1}$ (see Text S2) and calculated the precision of the range 256 257 for the following procedures. 4) DIC_b was calculated as the difference between the 258 estimated DIC and DIC_{ab}. 5) fCO₂ was calculated from the equilibrium calculation 259 using DIC_{ab} as the fCO₂ without any biological effects (fCO_{2ab}) and fCO_{2b} (the 260 difference between fCO₂ and fCO_{2ab}) as the fCO₂ change due to the biological effects in 261 each bay.

262 The air-water CO₂ flux (F, µmol m⁻² s⁻¹ or mol m⁻² yr⁻¹) was calculated as 263 follows:

264

5
$$F = k \cdot S(fCO_{2water} - fCO_{2air})$$
(4)

where *k* is the gas transfer velocity (m s⁻¹) and calculated as described in the next paragraph. *S* is the solubility of CO_2 in water (mol m⁻³ atm⁻¹) and estimated using an empirical equation using water temperature and salinity (Weiss, 1974). fCO_{2water} and fCO_{2air} are the fugacity of CO₂ in water and air (µatm), respectively. A positive value for 271 the flux indicates a CO_2 efflux to the atmosphere, and vice versa. The air–water CO_2

- 272 flux due to abiotic factors (F_{ab}) and the contribution of biological effect to the flux (F_b)
- 273 were also calculated using fCO_{2water} calculated from fCO_{2ab} and fCO_{2b} , instead of the
- fCO_{2water} , respectively. Because the temporal and spatial information for fCO_{2air} from
- 275 TUMSAT did not completely correspond to that of fCO_{2water} measurements, the data
- were approximated by the data with the same *dist-monthlydata* grid information, which was calculated by natural neighbor interpolation using the original TUMSAT fCO_{2air} data. Because the OCU data did not include fCO_{2air} measurements, we used the average
- value of the NIES data for Osaka Bay (392.6 μatm) for the calculation.
- 280 The gas transfer velocity k (here, the units are cm hr⁻¹) was calculated as 281 follows (Wanninkhof, 2014):
- 282

283
$$k = 0.251 \times U_{10}^{2} \times (Sc/_{660})^{-0.5}$$
 (5)

284

where U_{10} (m s⁻¹) is the wind speed at the height of 10 m from the water surface. Sc is the Schmidt number, defined as the ratio of the CO₂ molecular diffusion coefficient to the dynamic viscosity of seawater; Sc of seawater can be calculated from an empirical equation using the water temperature (Jähne et al., 1987). The wind data were taken from the database of the NEDO Offshore Wind System (NeoWins;

290 http://app10.infoc.nedo.go.jp/Nedo_Webgis/top.html), which is the open database of the 291 wind simulator provided by the New Energy and Industrial Technology Development 292 Organization (NEDO) for evaluating the efficiency of offshore wind power generation 293 in Japan. The wind data were supplied as the wind rose data, monthly averaged data, 294 and averaged wind profile data, with 500-m resolution. We took the monthly averaged 295 data at 10-km intervals beginning at dist = 0 along the water course (Table S1). Because 296 the monthly averaged data were supplied as the data for 60–140 m above the water surface, while the wind profile was at 10-200-m height, we calculated the monthly 297 298 averaged data at 10-m height according to the power-law of wind profile as follows: 299

300
$$U_{10} = U_{60} \times (\frac{10}{60})^n$$

301 $n = \frac{\log(U_{p60}/U_{p10})}{\log(60/10)}$ (6)

302

303 where U_{60} (m s⁻¹) is the monthly averaged data at 60-m height. U_{p10} and U_{p60} are the

304	wind-profile data at 10-m and 60-m height, respectively (Pagon, 1935). The wind data
305	were applied to the carbonate measurement data for each of the 12 months and 10 km of
306	dist.
307	
308	2.4 Data interpolation and error range
309	
310	Because the measurement data were not homogeneous temporally or spatially,
311	the averages and deviations described in this study were calculated from the interpolated
312	data, which is a homogeneous grid of the distance (every 1 km dist) and date (every
313	0.1-monthlydata) data ($n = 100 \times 120 = 12,000$). The error values for the carbonate
314	parameters are shown in Table S2.
315	The procedure used for error estimation is described in the Supporting
316	Information (Text S3). The error propagation was complicated because of the
317	non-linearity of the carbonate equilibrium calculation, so the error was separated into
318	the random error mainly due to the TA estimation and the riverine error due to
319	uncertainty in estimating the riverine DIC.
320	
321	3. Results
322	
323	3.1 Temperature, salinity, and carbonate parameters
324	
325	The water temperature and salinity (mean \pm SD) in each bay were 19.15 \pm
326	4.59 °C and 31.42 \pm 3.39 (Tokyo Bay), 18.78 \pm 5.36 °C and 31.97 \pm 2.23 (Ise Bay), and
327	18.66 \pm 5.12 °C and 31.88 \pm 1.43 (Osaka Bay), respectively (Figure S1). fCO ₂ was
328	305.56 ± 87.80 µatm (Tokyo Bay), 294.88 ± 46.38 µatm (Ise Bay), and 300.56 ± 70.30
329	μatm (Osaka Bay) (Figure 3). The estimated TA and DIC were 2161 \pm 125 and 1872 \pm
330	129 μ mol kg ⁻¹ (Tokyo Bay), 2166 \pm 115 and 1883 \pm 120 μ mol kg ⁻¹ (Ise Bay), and 2179
331	\pm 64 and 1893 \pm 85 µmol kg ⁻¹ (Osaka Bay) (Figure S2).
332	



333

Figure 3. Temporal and spatial distributions of the fugacity of CO_2 (fCO₂) in (a) Tokyo, (b) Ise, and (c) Osaka Bays. The color indicates the interpolated 0.1-month × 1-km grid value. The gray dots show the distribution of direct measurements. The white area in (a) indicates extreme values that were excluded from the analysis for clarity. The parameter *dist* represents the distance from a zero

point in the mouth of the main river feeding the inner bay (equation [1] in the main text).

339

340 The salinity and DIC of the oceanic endmembers were calculated for each 341 month as the average values between 90 and 100 km from the river mouth reference point in each bay: 34.34 ± 0.31 and $1968 \pm 35 \mu mol kg^{-1}$ in Tokyo Bay, 34.19 ± 0.39 342 and $1978 \pm 45 \ \mu\text{mol} \ \text{kg}^{-1}$ in Ise Bay, and 33.43 ± 0.66 and $1958 \pm 45 \ \mu\text{mol} \ \text{kg}^{-1}$ in 343 Osaka Bay, respectively. These values were higher during winter and lower during 344 345 summer and were consistent with the results from an empirical equation derived for the 346 Kuroshio stream area (Ishii et al., 2011) (Figure S3). The DIC of the riverine endmember was estimated to be $1162 \pm 200 \ \mu mol \ kg^{-1}$ in Tokyo Bay, $675 \pm 200 \ \mu mol$ 347 kg^{-1} in Ise Bay, and $852 \pm 200 \mu mol kg^{-1}$ in Osaka Bay. 348 The DIC_{ab} and DIC_b were 1899 \pm 87 and -27 \pm 64 µmol kg⁻¹ in Tokyo Bay, 349 1896 ± 97 and $-13 \pm 28 \ \mu\text{mol} \ \text{kg}^{-1}$ in Ise Bay, and 1906 ± 55 and $-13 \pm 39 \ \mu\text{mol} \ \text{kg}^{-1}$ in 350 Osaka Bay, respectively (Figure 4). The estimated DIC_b showed a significant decrease 351 352 within 70 km from land in summer in all bays. Meanwhile, there was an increase in 353 Tokyo Bay within about 10 km of land from October to January. The seasonal trend of 354 photosynthetic activity (decrease in DIC_b) was almost the same among the bays. The 355 magnitude and spatial distribution of the photosynthetic activity was the greatest in 356 Tokyo Bay, where respiration and/or organic-matter decomposition (increase in DIC_b) 357 were evident mainly in autumn to winter. In the other two bays, the magnitude and spatial distribution of DIC_b were almost the same, and smaller than in Tokyo Bay. 358



359

Figure 4. Temporal and spatial distributions of abiotic dissolved inorganic carbon (DIC_{ab}) ((a): Tokyo Bay, (b): Ise Bay, (c): Osaka Bay) and biotic DIC (DIC_b) ((d): Tokyo Bay, (e): Ise Bay, (f): Osaka Bay). The colors and dots are as defined in Figure 3. The parameter *dist* represents the distance from a zero point in the mouth of the main river feeding the inner bay (equation [1] in the

main text).
3.2 Air–water CO ₂ fluxes
The air-water CO ₂ fluxes in the three bays indicated that these areas were
annual atmospheric CO ₂ sinks (–2.87 \pm 2.39, –3.20 \pm 1.52 and –2.44 \pm 1.71 mol m $^{-2}$ yr $^{-1}$
¹ in Tokyo, Ise and Osaka Bays, respectively) (Figure 5). The temporal and spatial
distribution of the air-water CO ₂ flux in Ise and Osaka Bays were similar and showed a
seasonal pattern whereas the distribution in Tokyo Bay seemed to be more consistent
with that of DIC_b than a seasonal pattern (Figure 4). The peak of CO_2 absorption was
during February to April, and CO ₂ absorption was lowest in August and September in
Ise and Osaka Bays, and in October and November in Tokyo Bay.



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Figure 5. Temporal and spatial distribution of air–water CO_2 flux in (a) Tokyo, (b) Ise, and (c) Osaka Bays. The colors and dots are as defined in Figure 3. The parameter *dist* represents the distance from a zero point in the mouth of the main river feeding the inner bay (equation [1] in the main text).

381

The fluxes without (F_{ab}) and from biological activity (F_b) were -2.14 ± 1.75 and $-0.73 \pm 2.68 \text{ mol m}^{-2} \text{ yr}^{-1}$ (Tokyo Bay), -2.82 ± 1.73 and $-0.38 \pm 1.09 \text{ mol m}^{-2} \text{ yr}^{-1}$ (Ise Bay), and -2.31 ± 1.51 and $-0.13 \pm 1.09 \text{ mol m}^{-2} \text{ yr}^{-1}$ (Osaka Bay), respectively (Figure S4). The standard deviations of the biotic fluxes (F_b) indicated that the biological effect in Ise and Osaka Bays was weaker than in Tokyo Bay. The temporal and spatial distribution of F_{ab} showed a seasonal pattern whereas the distribution of F_b seemed to be related to the distribution of DIC_b.

- 390
- 391 4. Discussion
- 392

393 The validity of our estimates of biological effects on DIC (DIC_b) and CO₂ flux 394 $(F_{\rm b})$ depended on the precision of the riverine DIC endmember because the abiotic DIC 395 (DIC_{ab}) and CO_2 flux (F_{ab}) were determined from the riverine endmember, along with 396 the oceanic endmember whose error was relatively small. Although we could not find 397 reasonable reference data for riverine DIC in the bays, the reported riverine TA data by Taguchi et al. (2009) which are 1006, 518 and 759 μ mol kg⁻¹ in Tokyo, Ise, and Osaka 398 399 Bays, respectively, support the validity of the DIC estimation because water with a 400 higher TA can contain more carbonate and bicarbonate ions as DIC under the same fCO₂ conditions. The range of estimated riverine fCO₂ in each bay also supports the 401 402 validity of our estimates of riverine DIC and its precision ($\pm 200 \ \mu mol \ kg^{-1}$) (see Text 403 S2).

404 Our results support the hypothesized mechanisms related to wastewater 405 treatment (Kubo et al., 2017; Kuwae et al., 2016). Typical wastewater treatment method 406 removes carbon in the form of sludge and CO₂ gas more efficiently than nutrients such 407 as nitrogen and phosphorus (Sedlak, 1991). Through these effluents, the balance of 408 primary production and respiration in inner bay is offset towards an excess of primary 409 production and the resultant suppression of fCO₂ increase. Second, the effluent contains 410 relatively refractory carbon, because labile organic matter has already been removed by 411 treatment (Kubo et al. 2015). Therefore, respiration and mineralization rates of effluent 412 are low, and subsequent fCO₂ increases are suppressed.

The negative average annual biotic DIC in all three of the bays indicates that
the ecosystems were net autotrophic. The magnitude of the biotic DIC seemed to be
consistent with the nutrient concentrations reported in the reference material of the
Ministry of the Environment, Japan

417 (https://www.env.go.jp/council/09water/y0917-07/ref02.pdf). In this report, the highest 418 concentrations of total nitrogen and phosphorous were in Tokyo Bay, and were almost the same in Osaka and Ise Bays (average total nitrate, 0.8, 0.3, 0.4 mg L⁻¹; average total 419 phosphate, 0.06, 0.03, 0.04 mg L^{-1} in Tokyo, Ise, and Osaka Bays, respectively). The 420 421 nutrient concentrations in the three bays were probably determined by the volume of 422 treated wastewater discharged into the bays, which is related to the size of the 423 population of the surrounding urbanized area. For example, the inflow of the treated 424 water is reported as the main component of the total freshwater inflow to Tokyo Bay 425 and thus the effect of treated water should be noticeable on the nutrient concentration in 426 the bays (Kubo et al., 2015). Although the net primary production in the bays depends 427 on several parameters such as the seawater residence time and vertical stratification, its 428 general magnitude would be similar to that of the biotic DIC because it is similarly 429 influenced by hydrographic conditions.

430 It is possible that the net primary production in Osaka Bay was underestimated 431 because Osaka Bay connects with the Seto Inland Sea at its landward end and this 432 topography results in strong tidal currents in the bay (Odamaki, 2002). Stronger tides 433 would enhance the water exchange between the bay and the surrounding areas and 434 weaken the biological effects on DIC and the air-water CO₂ flux. In addition, the area at 435 dist of 50-100 km coincides with narrow straits (Kitan Strait and Kii channel) where 436 several large rivers flow into the bay. Thus, the calculation of the endmember effect in 437 Osaka Bay might be biased compared with those for the other two bays.

438 The CO₂ fluxes found in this study indicate that the inner bays and surrounding 439 outer bays in Japan are one of the largest atmospheric CO₂ sinks among the global 440 coastal areas reviewed in previous studies (Borges et al., 2005; Chen and Borges, 2008). The overall average CO₂ flux for the bays ($-2.84 \text{ mol m}^{-2} \text{ yr}^{-1}$) is more negative than 441 the average in these previous studies for estuaries $(7.74-10.26 \text{ mol m}^{-2} \text{ yr}^{-1})$ and 442 marginal seas (continental shelves) $(-1.64 \text{ to } -1.06 \text{ mol } \text{m}^{-2} \text{ vr}^{-1})$. The absorption in the 443 444 bays was mainly based on the abiotic flux (75%, 88%, and 95% of the net CO_2 flux (F) 445 in Tokyo, Ise, and Osaka Bays, respectively). The oceanic endmember from the 446 Kuroshio stream area is the most plausible explanation for the CO₂ absorption (overall average CO_2 absorption of $-2.32 \text{ mol m}^{-2} \text{ yr}^{-1}$). That area has been reported as the one 447 of the largest CO₂ sinks in the world because of the cooling effect of cold Oyashio water 448 449 on warm Kuroshio water (Takahashi et al., 2002, 2009). We suggest that the CO₂ 450 absorption in the bays was enhanced by additional cooling due to the terrestrial effect 451 during winter (Figure S1).

452

The biotic CO₂ flux was the largest in Tokyo Bay, followed in order by Ise Bay

- and Osaka Bay. The magnitude was basically the same as that of the biotic DIC and
- 454 nutrient concentrations in each bay. The higher CO₂ absorption in Ise Bay compared to
- 455 Osaka Bay despite almost identical biotic DIC can be explained by the lower salinity
- and TA in the near-shore area in Ise Bay (Figures S1 and S2). Because lower-TA water
- 457 has less buffering effect on fCO_2 , the decrease in fCO_2 in Ise Bay was greater than that
- 458 in Osaka Bay even when the decrease in DIC was the same. Although the annual
- 459 average of the biological effect was limited (25%, 12%, and 5% of the net CO_2 flux in
- 460 Tokyo, Ise, and Osaka Bays, respectively), it affected the temporal and spatial
- 461 distribution of the air-water CO_2 flux. Temporally, the pattern of F_b was the opposite of
- 462 F_{ab} , with an influx (F_b) or efflux (F_{ab}) in summer and vice versa in winter (Figure 6).
- 463 The biological effect was the strongest in Tokyo Bay and less notable in the other bays,
- 464 as with the biotic DIC. As a result, the seasonal variation of the CO_2 flux in Tokyo Bay
- 465 was different from that in the other two bays despite having almost the same variation in
- the abiotic CO_2 flux (Figure 6). For example, the peak CO_2 influx in winter in the bays
- 467 continued into summer in Tokyo Bay, but the winter influx in Tokyo Bay was about
- 468 two-thirds that in the other bays.



469

470 **Figure 6**. Temporal (left) and spatial (right) variations in the air–water CO_2 flux (*F*) ((a), (d)), abiotic 471 flux (F_{ab}) ((b), (e)) and biotic flux (F_b) ((c), (f)). Note that the unit is different from Figure 5. The 472 error bars were estimated by using equation (S4). The parameter *dist* represents the distance from a 473 zero point in the mouth of the main river feeding the inner bay (equation [1] in the main text).

474

The difference in the spatial distribution of F_b among the bays was less noticeable than that in the temporal distribution because of the offset of the influx in summer and efflux in winter in Tokyo Bay (Figure 6). The magnitude of the influx peak at *dist* of about 20 km was almost the same among the bays whereas the efflux at *dist* > 50 km was the largest in Osaka Bay. This might be caused by the decomposition of the organic matter produced by the photosynthesis at about 20 km *dist*.

481 Meanwhile, there was limited CO₂ released to atmosphere in the near-shore 482 area ($dist \le 10$ km) although fCO₂ of the riverine endmember was estimated at more 483 than 2000–5000 µatm in the case of the intermediate riverine DIC (Figure 5). As for 484 Tokyo Bay, there was a tendency toward an efflux in the near-shore area both in F_{ab} and 485 $F_{\rm b}$ (Figure 6), corresponding to the decrease in the DIC_{ab} in summer and the increase in DIC_b in winter (Figures 4 and 5). Because the distribution of the decrease in DIC_{ab} was 486 487 consistent with the salinity distribution (Figures 4 and S1), an increase in riverine flow 488 from precipitation might cause a CO₂ release as an abiotic factor. On the other hand, the 489 increase in DIC_b was consistent with the increase in pCO₂ observed in a previous study 490 (Kubo et al., 2017) in which the increase was due to the weakening of stratification in 491 Tokyo Bay due to the cooling of surface water, a decrease in precipitation, and the 492 weakening of the seasonal southward wind. These factors probably contributed to 493 bringing high-fCO₂ water and resuspended organic sediments from the bottom to the 494 surface, resulting in the CO₂ efflux to the atmosphere. Although we could not perform a 495 similar analysis for the other two bays because of a lack of measurement data, we would 496 expect a similar, considerable efflux tendency because the other bays share the same 497 hydrographic and climate conditions.

498 The comparison between Tokyo Bay and other two bays suggests that the 499 enhanced urbanization results in increased biotic CO₂ absorption. Because the 500 development of coastal areas will likely continue for several decades, the biotic 501 absorption of CO₂ in the bays is expected to be a mitigating factor for future climate change. We assumed that the area within 100 km of the global coastline $(6.2 \times 10^7 \text{ km}^2)$ 502 503 could absorb atmospheric CO₂ additionally at the same rate as the biotic flux in Tokyo Bay (0.73 mol m^{-2} yr⁻¹), and roughly estimated the potential for additional biotic 504 absorption to be 0.054 Pg yr⁻¹. This is on the same order of magnitude as the estimated 505 506 estuarine CO₂ efflux (e.g., Chen et al., 2013). The estimation of worldwide abiotic CO₂ 507 flux is difficult because the oceanic and riverine carbonate parameters differ at each 508 location. In addition, the carbon export from urbanized areas is observed to increase 509 along with the development (Barnes and Raymond, 2009; Lopes et al., 2020; Wang et

al., 2017). This increase in carbon export could mitigate the increase in biotic CO₂
absorption.

512 For accurate estimates of the global CO₂ absorption resulting from coastal 513 urbanization, it is necessary to obtain more carbonate measurements from a variety of 514 areas. For example, data from measurements at the lagoons in Ivory Coast and 515 Guanabara Bay in Brazil suggested that strong CO₂ absorption and release are mixing in 516 the tropical urbanized bay where the rate of wastewater treatment is low (Cotovicz et al., 517 2015; Koné et al., 2009). Both of these studies indicated that eutrophication and 518 stratification were the factors regulating the undersaturation of CO₂. Meanwhile, studies 519 in the Chesapeake Bay reported that both the absorption and release of CO₂ were 520 observed by numerical simulation model and abundant pH measurement, respectively 521 (Herrmann et al., 2020; St-Laurent et al., 2020). Because the main bottleneck for further 522 study is the difficulty of obtaining comprehensive measurements in coastal areas, the 523 development of novel methods like the above Chesapeake Bay studies will facilitate 524 filling gaps in the temporal and spatial distributions of available data. The review study 525 of such measurements is expected to understand the quantitative relationship between 526 the coastal urbanization and the carbon cycle.

527 528

529 5. Conclusions

530

531 In this study we clarified the temporal and spatial variations of carbonate 532 parameters in three bays in Japan from riverine water to the outer bays within a range of 533 100 km. Our results are the first to fill the gap between studies of nearshore areas and 534 the marginal seas. We found a notable atmospheric CO₂ influx due to biological activity 535 in the areas about 20 km from land in summer, accounting for 25%, 12%, and 5% of the 536 measured net CO₂ flux in Tokyo, Ise, and Osaka Bays, respectively. In addition, the 537 biological effect in the highly developed Tokyo Bay significantly affected the seasonal 538 variation of air-water CO₂ flux. The potential for atmospheric CO₂ absorption was 539 mainly regulated by the water in the surrounding marginal seas. The biological effect 540 seemed to be associated with the nutrient concentrations, which are related to the 541 volumes of treated wastewater entering the bays. In addition, the degree to which the 542 water area is enclosed likely influences the biological effect. Conditions that are more 543 closed or open would enhance or mitigate the biological effect on the air-water CO₂ 544 flux, respectively.

545

This study should contribute to future investigations into the carbon cycle in

546 urbanized coastal areas, which will likely continue to expand for the next several

547 decades. However, more detailed investigations in inner and outer bays are required for

548 more precise evaluation of their contribution to the global carbon cycle. To expand the 549 results of this study to the global scale will require further measurements in bays in a

- 550 variety of regions.
- 551
- 552

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Supporting Information for

[Contribution of biological effects to carbonate-system variations and the airwater CO₂ flux in inner and outer bays in Japan]

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Introduction

This supporting information contains: (1) a detailed description of the changepoint analysis for determining the upper limit of the distance parameter *dist* which is defined by Eq. [3]. (2) a detailed description of the uncertainty of riverine endmember dissolved inorganic carbon (DIC) (3) a detailed description of the random error and the error associated with the uncertainty of the riverine endmember DIC.

Text S1. Changepoint analysis

The upper limit of the distance parameter *dist* was determined using changepoint analysis of the salinity and DIC, which are conservative parameters and should be unaffected by water temperature and salinity. The analysis focused on the rapid change of the variance of these parameters with *dist*, given the seasonal variability, and used the following equation (Lavielle, 2004):

$$J_n = m \log(\partial_{1:m}^2) + (n - m + 1) \log(\partial_{m+1:n}^2)$$
(S1)

where *n* is the number of the salinity or DIC binned into 1-km *dist* intervals and *m* is the location of the change point $(1 \le m \le n)$. $\partial^2_{i:j}$ is the variance of the data between *i* and *j* $(1 \le i \le j \le n)$. The change point *m* is determined in order to minimize the parameter *J_n*.

The change points for the salinity and DIC were 71 and 66 km in Tokyo Bay, 64 and 62 km in Ise Bay, and 33 and 48 km in Osaka Bay, respectively. In other words, the salinity and DIC in the bays at a distance more than 71 km from the head of the bay were relatively constant compared to the inner areas of the bays. We defined the range of the combined inner and outer bays in this study to be 90 km, to provide a conservative range, and the water between 90 km and 100 km was defined as the oceanic endmember. Data from more than 100 km from land was filtered out and not used in this study.

Text S2. Range of riverine endmember DIC

The range of riverine endmember DIC was estimated using the fugacity of CO_2 (fCO₂). Although riverine fCO₂ is variable because of the low buffering capacity, the

upper range should be limited because high water fCO_2 results in a strong CO_2 efflux to the atmosphere, which reduces the fCO_2 in water with low buffering capacity. Given previous information about riverine fCO_2 (Aufdenkamp et al., 2011; Chen and Borges, 2009), we set the range of fCO_2 as 100 to 10,000 µatm. From this, the respective upper and lower limits of riverine DIC were 1397.5 and 982.3 µmol kg⁻¹ in Tokyo Bay (1161.8 µmol kg⁻¹ from Eq. [3]), 909.8 and 513.3 µmol kg⁻¹ in Ise Bay (675.3 µmol kg⁻¹), and 1150.7 and 746.1 µmol kg⁻¹ in Osaka Bay (852.3 µmol kg⁻¹). Note that the equilibrium calculation uses the dissociation constant of carbonic acid provided by Millero (1979) because this constant is optimized for the calculation in freshwater. According to these estimates, we set ±200 µmol kg⁻¹ as the range of riverine endmember DIC, which satisfies the fCO₂ range.

Text S3. Error estimation

The calculation errors in this study were caused by 1) the measurement error of salinity and fCO₂, 2) error in total alkalinity (TA) as estimated from the empirical equation, 3) errors in the CO₂ flux calculation from the Eq [5] for the gas transfer velocity, and 4) the error from the imprecision in the range of the riverine endmember DIC. The propagation of error was calculated according to the procedure for estimating the abiotic CO₂ flux (F_{ab}) and the biotic CO₂ flux (F_b) (Figure 2). The error from the first three factors above resulted in the random error of the propagated parameters ("random error") whereas the error from the last factor was estimated as the range of the parameters for cases with the upper and lower values for the range of the riverine DIC ("riverine error"). Because the random and riverine errors became large for the data from near-shore areas, we applied a constant random error to the averaged or binned parameters in order to avoid temporal and spatial inhomogeneity. The results of applying the constant random error and the mean of the riverine error are shown in Table S2.

First, the measurement errors of the National Institute for Environmental Studies (NIES) salinity and DIC were estimated as 0.005 and 0.5 µatm, respectively (Nakaoka et al., 2013). The errors for the corresponding Tokyo University of Marine Science and Technology (TUMSAT) data were about 0.3 (1%) and 0.4 µatm, respectively (Kubo et al., 2017), and about 0.03 (0.05 mS cm⁻¹) for Osaka City University (OCU) salinity. Note that OCU did not directly measure fCO₂. For ease of calculation, the fCO₂ error was assumed to be a uniform 0.5 µatm in this study.

The TA error from the empirical equation was defined as the standard deviation of the residuals of the empirical equation (34, 21 and 35 μ mol kg⁻¹ for Tokyo Bay, Ise Bay and Osaka Bay, respectively) (Taguchi et al., 2009). Because these values were larger than the TA error caused by the salinity measurement error (1.8, 11, and 1.3 μ mol kg⁻¹ for the data from NIES, TUMSAT and OCU, respectively), the random error of TA was determined only from the error of the empirical equation in this study.

The random error of DIC cannot be estimated by simple propagation because the carbonate equilibrium is calculated non-linearly. Therefore, the error was calculated using the asymptotic slope as follows:

$$Er_{-Ob} = \left\{ \left[\left(\frac{\partial Ob}{\partial Ex_1} \right) \cdot Er_{-}Ex_1 \right]^2 + \left[\left(\frac{\partial Ob}{\partial Ex_2} \right) \cdot Er_{-}Ex_2 \right]^2 \right\}^{0.5}$$
(S2)

where Ob and Ex are the objective and explanatory variables, respectively. Er_{Ob} and Er_{Ex1} and Er_{Ex2} are the error of Ob and Ex, respectively. Here, Ob, Ex_1 and Ex_2 are the DIC, fCO₂ and TA, respectively. The partial differential of Ob and Ex (slope of the asymptote) was calculated using the water temperature, salinity, TA and DIC for each sampling point and date.

The random error of DIC_{ab} was calculated according to equation (3):

$$Er_{-DICab} = \left\{ \left(\frac{Sal}{Sal_o} Er_{-DICo} \right)^2 + \left[\frac{(Sal_o - Sal)}{Sal_o} Er_{-DICr} \right]^2 \right\}^{0.5}$$
(S3)

where Sal and Sal_o are the salinity of the data and the oceanic endmember for the month, respectively. Er_{DICo} and Er_{DICr} are the errors of oceanic and riverine endmember DICs, respectively. The former was calculated as the propagation of the DIC error used for the calculation of DIC_o in each month and equals the DIC error divided by the root square of the number of data. The latter was set to zero because the error was estimated separately as described below. Indeed, the random error of DIC_{ab} was relatively small (about 1 µmol kg⁻¹) compared to the error of DIC.

The calculation using equation (S2) was also applied for the calculation of fCO_{2ab} error. In this case, Ob, Ex_1 and Ex_2 in equation (S2) are fCO_{2ab} , DIC_{ab} and TA, respectively. In addition, the error of CO_2 fluxes (F, F_{ab} , F_b) was calculated using the error of fCO_2 (fCO_2 , fCO_{2ab} , fCO_{2b}) and the error of the gas transfer velocity (20%; Wanninkhof, 2014). The error of the solubility was neglected in this study because the effect was small (<1%) compared with the errors estimated above.

The riverine DIC error was calculated as the range of the results using the upper (+200 μ mol kg⁻¹) and lower (-200 μ mol kg⁻¹) values for the range of riverine DIC in each bay. Because the random error of the abiotic and biotic parameters depends on each parameter itself, the random error was estimated for both cases with upper and lower riverine DIC.

Because the random and riverine errors become large for data with low salinity, the weighted averaging of the different errors for analyzing the temporal and spatial variation should underestimate the contribution of the biological activities in near-shore areas. Therefore, we estimated a constant random error for each bay to avoid temporal and spatial inhomogeneity. In this study, the random error was interpolated to the 1-km dist \times 0.1-monthlydata grid by the natural neighbor method, and the third quartile (75th percentile) was defined as the representative constant error for each bay. The error of an interpolated parameter was estimated as follows:

$$Er_{int_u} = \left(\frac{Er_{const_u}^{2} + Er_{const}^{2}}{nd^{0.5}}\right)^{0.5} + \left[\left(\overline{P_{u}} - \overline{P}\right)^{2}\right]^{0.5}$$
$$Er_{int_l} = \left(\frac{Er_{const}^{2} + Er_{const_l}^{2}}{nd^{0.5}}\right)^{0.5} + \left[\left(\overline{P} - \overline{P_{l}}\right)^{2}\right]^{0.5}$$
(S4)

where Er_{int_u} and Er_{int_1} are the errors of the interpolated parameter for the cases with the upper and lower riverine DIC, respectively. The upper riverine DIC results in a positive and negative range for abiotic and biotic parameters, respectively, and vice versa. Er_{const} , $\text{Er}_{\text{const}_u}$, and $\text{Er}_{\text{const}_1}$ are the constant errors in cases with the intermediate (the result of equation [3]), upper (+200 µmol kg⁻¹) and lower (-200 µmol kg⁻¹) riverine DIC. *P*, *P*_u and *P*₁ are the interpolated parameters for the same cases of riverine DIC. The overbar indicates averaging of binning of the parameter. The term "nd" is the number of data for the averaging or binning under the assumption that the measurement data were distributed homogeneously; nd equals the number of data for each bay divided by the dist–monthlydata grid (100 × 120) number used for the averaging. For example, if the monthly-binned average is calculated for Tokyo Bay, nd = 18,118/(12,000/12).



Figure S1. Temporal and spatial distributions of water temperature ((a): Tokyo Bay, (b): Ise Bay, (c): Osaka Bay) and salinity ((d): Tokyo Bay, (e): Ise Bay, (f): Osaka Bay). The color indicates the interpolated 0.1-month \times 1-km grid value. The gray dots show the distribution of direct measurements. The parameter *dist* represents the distance from a zero point in the mouth of the main river feeding the

inner bay (equation [1] in the main text).



Figure S2. Temporal and spatial distributions of TA ((a): Tokyo Bay, (b): Ise Bay, (c): Osaka Bay) and DIC ((d): Tokyo Bay, (e): Ise Bay, (f): Osaka Bay). The colors and dots are as defined in Figure S1. The parameter *dist* represents the distance from a zero point in the mouth of the main river feeding the inner bay (equation [1] in the main text).



Figure S3. The relationships between salinity and DIC in (a) Tokyo, (b) Ise, and (c) Osaka Bays. The black dots are the interpolated values between the oceanic and riverine endmembers and defined as DIC_{ab} for each month. The graph in (d) shows a comparison between the DIC predicted using the empirical equation from a previous study for the value in the Kuroshio stream area (Ishii et al., 2011) and the estimated DIC of oceanic endmembers in this study. The black line in this graph shows y = x.



Figure S4. Temporal and spatial distributions of the abiotic (F_{ab}) ((a): Tokyo Bay, (b): Ise Bay, (c): Osaka Bay) and biotic (F_b) ((d): Tokyo Bay, (e): Ise Bay, (f): Osaka Bay) air–sea CO₂ flux. The colors and dots are as defined in Figure S1. The parameter *dist* represents the distance from a zero point in the mouth of the main river feeding the inner bay (equation [1] in the main text).

	Tokyo Bay		Ise Bay		Osaka Bay	
Dist ^a (km)	Lat (°N)	Long (°E)	Lat (°N)	Long (°E)	Lat (°N)	Long (°E)
0	35.65	139.85	35.04	136.74	34.68	135.41
10	35.565	139.81875	34.970	136.81250	34.585	135.31875
20	35.440	139.76250	34.890	136.76250	34.570	135.18125
30	35.365	139.77500	34.815	136.72500	34.475	135.06875
40	35.270	139.76875	34.700	136.80000	34.345	135.01250
50	35.185	139.78125	34.615	136.90000	34.245	134.97500
60	35.095	139.74375	34.545	137.02500	34.130	134.94375
70	35.005	139.71250	34.445	137.06250	34.030	134.91250
80	34.915	139.71250	34.430	137.27500	33.905	134.92500
90	34.820	139.74375	34.425	137.45625	33.805	134.93750
100	34.735	139.74375	34.525	137.72500	33.705	134.95000

Table S1 The locations for the wind data provided by the NEDO Offshore Wind System database (NeoWins; http://app10.infoc.nedo.go.jp/Nedo_Webgis/top.html)

^a*Dist* is the distance from a zero point in each bay located in the mouth of the primary river at the head of the bay (equation [1] in the main text).

Table S2 The constant random (*Rand*) and riverine (*Riv*) errors in each bay for the interpolated data. The riverine error is the value of the overall average for each bay. '+200' or '200' in 'Error type' low indicates the random error or change of the parameter for the case with the upper or lower riverine DIC, respectively. *The error for DIC_{ab} was omitted because of the small (<4 µmol kg⁻¹) random error and because it has the same riverine error as DIC_b . **The error for fCO_{2ab} was omitted because it was almost the same as that of fCO_{2b} .

	Error type	Tokyo Bay	Ise Bay	Osaka Bay
fCO ₂ (µatm)	Rand	0.5	0.5	0.5
TA (µmol kg ⁻¹)	Rand	34	21	35
DIC (µmol kg ⁻¹)	Rand	28	18	29
$\text{DIC}_{b} \left(\mu \text{mol } \text{kg}^{-1} \right)^{*}$	Rand	28	18	29
	<i>Riv</i> + 200	-16	-12	-8
	<i>Riv</i> – 200	+16	+12	+8
$fCO_{2b} (\mu atm)^{**}$	Rand	52	28	51
	<i>Rand</i> + 200	64	31	51
	Rand - 200	47	27	51
	<i>Riv</i> + 200	-37	-16	-10
	<i>Riv</i> – 200	+24	+13	+9
$F (\mathrm{mol}\;\mathrm{m}^{-2}\;\mathrm{yr}^{-1})$	Rand	0.92	0.85	0.77
$F_{ab} \pmod{\mathrm{m}^{-2} \mathrm{yr}^{-1}}$	Rand	0.89	0.93	0.88
	<i>Rand</i> + 200	0.80	0.85	0.96
	Rand - 200	0.92	1.02	0.029
	<i>Riv</i> + 200	+1.70	+0.64	+0.37
	<i>Riv</i> – 200	-0.90	-0.51	-0.32
$F_b (\mathrm{mol} \;\mathrm{m}^{-2} \;\mathrm{yr}^{-1})$	Rand	1.30	1.26	1.19
	<i>Rand</i> + 200	1.25	1.21	1.12
	Rand - 200	1.42	1.33	1.25
	<i>Riv</i> + 200	-1.70	-0.64	-0.37
	<i>Riv</i> – 200	+0.90	+0.51	+0.32

References

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