Quantifying the contributions of riverine vs. oceanic nitrogen to hypoxia in the East China Sea

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Abstract. In the East China Sea, hypoxia (oxygen $\leq 62.5 \, \mathrm{mmol \, m^{-3}}$) is frequently observed off the Changjiang (or Yangtze) River estuary covering up to about $15{,}000 \, \mathrm{km^2}$. The Changjiang River is a major contributor to hypoxia formation because it discharges large amounts of freshwater and nutrients into the region. However, modelling and observational studies have suggested that intrusions of nutrient-rich oceanic water from the Kuroshio Current also contribute to hypoxia formation. The

5 relative contributions of riverine versus oceanic nutrient sources to hypoxia have not been estimated before. Here, we combine a three-dimensional, physical-biogeochemical model with an element tracing method to quantify the relative contributions of nitrogen from different riverine and oceanic sources to hypoxia formation during 2008–2013. Our results suggest that the hypoxic region north of $30 \, \mathrm{^\circ N}$ is dominated by Changjiang River inputs, with its nitrogen loads supporting 74% of oxygen consumption. South of $30 \, \mathrm{^\circ N}$, oceanic nitrogen sources become more important supporting 39% of oxygen consumption during the

10 hypoxic season, but the Changjiang River remains the main control of hypoxia formation also in this region. Model scenarios with reduced Changjiang River nitrogen loads and reduced open-ocean oxygen levels suggest that nitrogen load reductions can significantly reduce hypoxia in the East China Sea and counteract a potential future decline in oxygen supply from the open ocean into the region.

15 # 1 Introduction

In the East China Sea (ECS), hypoxic conditions (i.e. dissolved oxygen (O_2) concentrations $\leq 62.5 \, \mathrm{mmol \, m^{-3}}$) are frequently observed off the Changjiang River (or Yangtze) Estuary covering up to about $15{,}000 \, \mathrm{km^2}$ (Li et al., 2002; Zhu et al., 2017). Hypoxia was first reported in 1959 (Zhu et al., 2011), and a significant increase in its spatial extent has been observed since the 1980s (Li et al., 2011; Wang, 2009; Wang et al., 2015; Zhu et al., 2011). The Changjiang River is the fifth largest river in

20 the world in terms of freshwater (FW) discharge ($9 \times 10^{11} \, \mathrm{m^{-3} \, y^{-1}}$) and its nutrient concentrations are comparable to other strongly anthropogenically affected rivers (Liu et al., 2003). The observed increase in hypoxic area since the 1980s has been

attributed to elevated nutrient loads due to fertilizer use in the Changjiang watershed (Siswanto et al., 2008; Wu et al., 2019; Yan et al., 2003).

Various studies have suggested that oceanic nutrients also play a role in hypoxia formation in the ECS (Chi et al., 2017;
25 Li et al., 2002; Wang et al., 2018; Zhou et al., 2017b, 2018; Zhu et al., 2011). The importance of oceanic nutrient supply distinguishes hypoxia in the ECS from the otherwise comparable situation in the northern Gulf of Mexico (NGoM), where a similar spatial extent of hypoxic conditions is fueled by freshwater and anthropogenic nutrient inputs from a major river, the Mississippi (Fennel and Testa, 2019). Observations in the ECS indicate that south of $30\,°N$, intrusions of nutrient-rich water from the Kuroshio Current influence the shelf dynamics (Wang et al., 2018; Zhou et al., 2017b, 2018). These intrusions vary
30 seasonally due to the influence of the East Asian monsoon, with weak northward winds in summer and strong southward winds in winter that lead to northward and southward water mass transport, respectively (Bian et al., 2013).

The complexity of the circulation and importance of different nutrient sources for hypoxia development make this system particularly amenable to model analyses with high spatio-temporal resolution (Fan and Song, 2014; Zhao and Guo, 2011; Zhang et al., same issue; Zheng et al., 2016; Zhou et al., 2017a; Zhang et al., 2018). Fan and Song (2014) used simulated
35 salinity and nutrient distributions to show that Changjiang River inputs are transported southward in winter and northward in summer. Zhao and Guo (2011) and Zhou et al. (2017a) used model sensitivity studies to highlight the role of oceanic nutrient sources on productivity and hypoxia in the ECS, respectively. Zhang et al. (2019) combined a physical-biogeochemical model with an element tracing method (e.g. Ménesguen and Hoch, 1997) to quantify the relative contributions of different nutrient sources to primary production (PP). They found that riverine nitrogen (N) supports 56% of water column integrated PP in the
40 ECS regions shallower than $50\,m$. With organic matter degradation being the main sink of O_2 in the subsurface waters of the ECS (Li et al., 2002), this suggests that riverine N also dominates O_2 consumption. However, a quantification of the relative contributions of riverine versus oceanic nutrient sources to hypoxia has not been available until now.

By combining a high-resolution biogeochemical model with this active element tracing method expanded for the quantification of the contributions to O_2 processes (Große et al., 2017, 2019), we provide such an analysis here. We apply the element
45 tracing method to an implementation of the Regional Ocean Modeling System (ROMS; Fennel et al., 2006; Haidvogel et al., 2008) configured for the ECS (Zhang et al., same issue). This allows us to quantify the contributions of N from different riverine and oceanic sources to hypoxia formation in the ECS and analyze year-to-year and seasonal variability in the individual contributions resulting from the East Asian monsoon cycle. In addition to supplying nutrients, the intrusions of open-ocean subsurface waters are relevant to hypoxia by preconditioning O_2 concentrations in the region. Subsurface O_2 has declined
50 over the past decades in the northwest Pacific (Schmidtko et al., 2017) and is projected to continue decreasing in the future (Bopp et al., 2017), which may further exacerbate hypoxia in the region (Qian et al., 2017). We analyze the effect of reduced open-ocean O_2 concentrations on hypoxia under current and reduced N loads from the Changjiang River and compare our results to previous findings for the NGoM's hypoxic zone.

2 Methods

2.1 The physical-biogeochemical model

We used an implementation of ROMS (Haidvogel et al., 2008) for the ECS (Bian et al., 2013). The model covers the region from 116 °E to 134 °E and from 20 °N to 42 °N with a resolution of 1/12 ° (Fig. 1) with 30 terrain-following σ-layers.
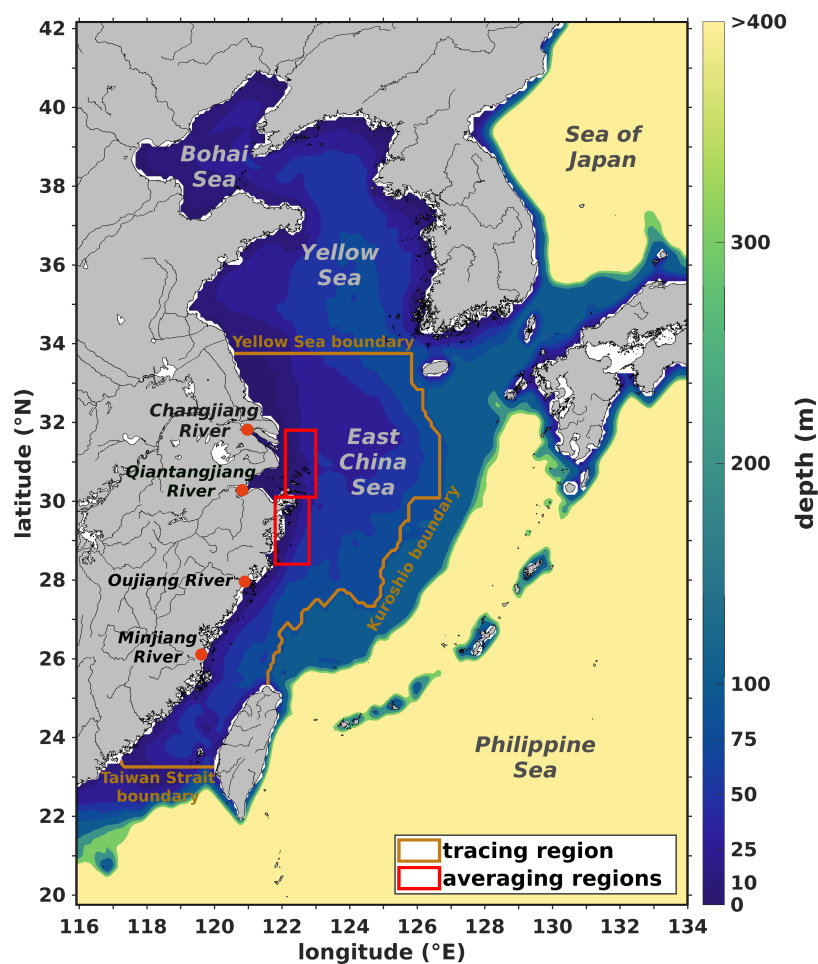
Figure 1. Model domain and bathymetry, with sub-domain used for nitrogen tracing, rivers inside the tracing region, and northern and southern regions used for time series analysis.

The biogeochemical component is based on the N-cycle model of Fennel et al. (2006, 2011) but was expanded to include phosphate Laurent et al. (2012), oxygen Fennel et al. (2013) and riverine dissolved organic matter Yu et al. (2015). The model state variables are: nitrate (NO_3^-), ammonium (NH_4^+), phosphate (PO_4^{3-}), one functional group each for phyto- and zooplankton, three groups of detritus (small, large, and riverine dissolved organic matter), and dissolved O_2. Instantaneous benthic

remineralization was applied at the sediment-water interface (Fennel et al., 2006, 2011). Light attenuation was expanded by including a term dependent on bottom depth (Zhang et al., same issue). For a complete set of model equations we refer the reader to the Appendix of Laurent et al. (2017).

65   Initial and open boundary conditions for temperature and salinity were derived from World Ocean Atlas 2013 version 2 (WOA13-v2) climatologies (Locarnini et al., 2013; Zweng et al., 2013). Temperature and salinity were nudged weekly toward the climatology using a nudging scale of 120 days. Horizontal velocities and sea surface elevation at the open boundaries are based on the SODA reanalysis (Carton and Giese, 2008). Eight tidal constituents (M_2, S_2, N_2, K_2, K_1, O_1, P_1 and Q_1) are imposed using tidal elevations, and tidal currents are derived from the global tide model of Egbert and Erofeeva (2002). Initial

70   and open boundary conditions for NO_3^-, PO_4^{3-} and O_2 are also based on WOA13-v2 (Garcia et al., 2013a, b), while small positive values are prescribed for all other biogeochemical variables. In regions deeper than 100 m and in the Yellow Sea north of 34 °N, NO_3^- concentrations are nudged toward the climatology using time constants of 7 and 10 days, respectively.

The model was run for the period 2006–2013, with the first two years used as spin-up, and forced by 6-hourly wind stress, surface heat and FW fluxes from the ECMWF ERA-Interim dataset (Dee et al., 2011). Daily FW discharge and nutrient loads

75   for 11 rivers were imposed, with the Changjiang River being by far the largest. Discharge for the Changjiang is obtained from Datong Hydrological Station (http://www.cjh.com.cn/en/). Concentrations of NO_3^-, NH_4^+ and PO_4^{3-} for the Changjiang River were obtained from the monthly Global NEWS data set (Seitzinger et al., 2005). For the other rivers, FW discharge and nutrient loads were prescribed using climatologies (Liu et al., 2009; Tong et al., 2015; Zhang, 1996). The model demonstrates good skill in representing the hydrography and biogeochemistry of the ECS (Zhang et al., same issue) and reproduces the

80   main features of the ECS circulation (see supplement Fig. S1). The simulation using this setup is hereafter referred to as the reference simulation.

In addition to the reference simulation, three scenario simulation were performed: (1) To assess the influence of a reduction in riverine N load on hypoxia, a nutrient reduction scenario was run with 50% smaller N concentrations in the Changjiang River input. All other forcing remained the same. (2) To investigate the potential effect of reduced open-ocean O_2 supply, a

85   scenario similar to the reference simulation but with reduced O_2 in the open ocean was performed where the initial and open-boundary O_2 concentrations were reduced by 20% throughout the water column in regions deeper than 200 m. (3) The reduced open-ocean O_2 scenario was repeated for reduced river N as in the N reduction scenario.

2.2 Passive freshwater and active nitrogen tracing

Passive dye tracers were used to track FW inputs from the Changjiang River similar to previous ROMS applications (Große

90   et al., 2019; Hetland and Zhang, 2017; Rutherford and Fennel, 2018; Zhang et al., 2010, 2012). The rate of change of the concentration of a passive dye tracer from the i-th source (C_p^i) is described as:

$$\frac{\partial C_p^i}{\partial t} = \nabla \cdot \left(\overline{\overline{D}} \nabla \right) C_p^i - \nabla \cdot \left(C_p^i \boldsymbol{v} \right) + S_{C_p^i}. \tag{1}$$

Here, $\overline{\overline{D}}$ is the second-order diffusion tensor (or diffusivity), $\boldsymbol{v}$ is the velocity vector, and $S_{C_p^i}$ represents the external dye tracer sources (i.e. riverine FW discharge). The dye tracer was initialized with zero in the entire model domain. Changjiang FW

95   discharge had a dye tracer concentration of 1 and the dye tracer was used to analyze the influence of FW from the Changjiang
River on stratification (i.e. potential energy anomaly; Simpson, 1981).

In addition, we applied an active element tracing method (Große et al., 2017; Ménesguen et al., 2006; Radtke et al., 2012)
to quantify the contributions of different N sources to O_2 consumption. For this purpose, each model variable containing N is
subdivided into fractions from the different source regions or rivers. The rate of change of a specific source fraction is described

100   as:

$$\frac{\partial C_X^i}{\partial t} = \nabla \cdot \left(\overline{\overline{D}} \nabla C_X \right) \cdot \frac{C_X^i}{C_X} - \nabla \cdot (C_X \boldsymbol{v}) \cdot \frac{C_X^i}{C_X} + R_{C_X} \cdot \frac{C_{X_{con}}^i}{C_{X_{con}}}. \tag{2}$$

C_X and C_X^i represent the concentrations of state variable X (e.g. NO_3^-) and its labeled fraction from the i-th source (e.g.
NO_3^- from the Changjiang River), respectively. R_{C_X} describes internal and external sources and sinks of X. According to
Eq. (2), source-specific processes are calculated as the product of the processes acting on each state variable (e.g. NH_4^+ uptake

105   during nitrification) and relative concentration of its source-specific fraction. The index 'con' in the source/sink term implies
that the relative concentration of the variable consumed by a process is used; e.g. source-specific nitrification is calculated
based on the relative concentration of NH_4^+. For each traced source, one set of equations of the form of Eq. (2) is applied using
the post-processing software of Große et al. (2017), which has been adapted for ROMS (Große et al., 2019). N tracing is only
applied in the region without NO_3^- nudging (see 'tracing region' in Fig. 1). Inside this region, we simultaneously traced N from

110   five different sources: the Changjiang River, three other smaller rivers (Minjiang, Hanjiang and Oujiang Rivers; grouped into
one source), the Taiwan Strait (at the southern boundary of the tracing region), the Kuroshio Current (at the eastern boundary),
and the Yellow Sea (at the northern boundary; see Fig. 1). Tracers that leave the tracing region cannot reenter. This tracer setup
is similar to that of Zhang et al. (2019), with the difference that we separate the Changjiang River from smaller rivers and that
we do not account for atmospheric nutrient deposition.

115   We applied the N tracing to the reference simulation. To spin up the tracing, we first re-ran year 2006 three times. For the
first iteration, all N mass already in the system was attributed to the small rivers. Subsequent iterations were initialized from the
final distribution of the previous year, followed by 2007 as an additional spin-up year. This procedure ensures that the different
N mass fractions have reached a dynamic steady state and produced realistic distributions at the beginning of the analysis
period 2008–2013.

120   ## 3   Results

3.1   Changjiang freshwater discharge and nitrogen concentrations

The Changjiang River is the main source of FW and nutrients in the ECS. Daily time series of its FW discharge and total
nitrogen (TN) concentrations are shown in Fig. 2. The FW discharge has a distinct seasonal cycle (highest in summer, lowest
in winter) due to the monsoon season with high precipitation in summer (Wang et al., 2008), and significant year-to-year

125   variability. The TN concentrations also show high year-to-year variability.
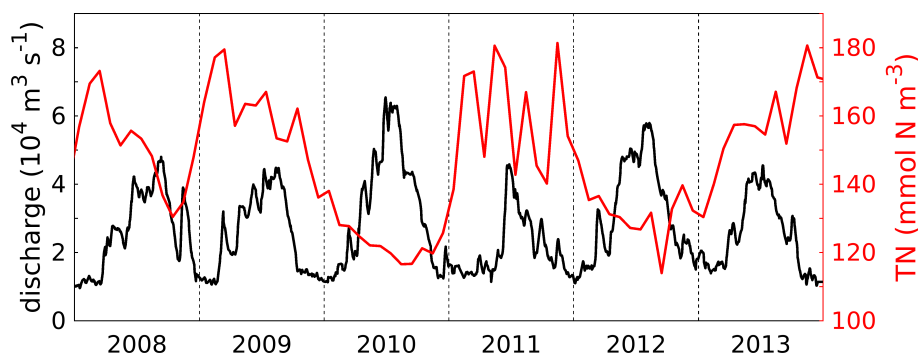
Figure 2. Daily time series of freshwater discharge (black, left y-axis) from and total nitrogen (TN) concentration in the Changjiang River (red, right y-axis).

3.2 Spatial patterns in oxygen consumption

We focus our analysis of O_2 dynamics on gross O_2 consumption (GOC), i.e. the sum of sediment O_2 consumption (SOC) and water column respiration (WR; incl. nitrification), and integrate over the 12 deepest pelagic layers (analogous to Zhang et al., same issue). This is reasonable as Zhang et al. show that both SOC and WR are relevant O_2 sinks in the hypoxic bottom

130   boundary layer of the ECS. Hypoxia in the ECS is most pronounced between July and November; we hence focus most of our analyses on this time period. First, we analyze the general spatial patterns in GOC and the relative contributions from riverine and oceanic N sources.

Figures 3a and b show maps of GOC and its relative riverine contribution averaged from July to November over the years 2008 to 2013, respectively. GOC is highest (up to $70 \, \mathrm{mmol} \, O_2 \, \mathrm{m}^{-2} \, \mathrm{d}^{-1}$) in the northern part ($30 \,^\circ$N to $31 \,^\circ$N) of the region

135   typically affected by hypoxia (indicated by the green line). Farther north, GOC is still high (50 to 60 $\mathrm{mmol} \, O_2 \, \mathrm{m}^{-2} \, \mathrm{d}^{-1}$), but decreases significantly in the offshore and southward directions with the strongest gradient roughly between $29 \,^\circ$, $122.5 \,^\circ$E and $31 \,^\circ$N, $123 \,^\circ$E. The riverine contribution to GOC is highest ($>95\%$) in the coastal regions between $30 \,^\circ$N and $32 \,^\circ$N and steadily decreases offshore and southward. South of $32 \,^\circ$N the strongest gradient in the riverine contribution corresponds to the maximum gradient in GOC.

140   Based on the hypoxic area locations simulated in our model and the spatial patterns in GOC and its riverine contribution, we defined two distinct analysis regions (black boxes in Fig. 3): a northern region where GOC is supported mainly by riverine N ($>70\%$) and a southern region where the riverine contribution declines strongly from 75% to 20% in southeastward direction. These regions are used to quantify how the relative contributions of riverine (Changjiang River and smaller rivers) and oceanic N sources (Kuroshio, Taiwan Strait and Yellow Sea) to GOC differ regionally.
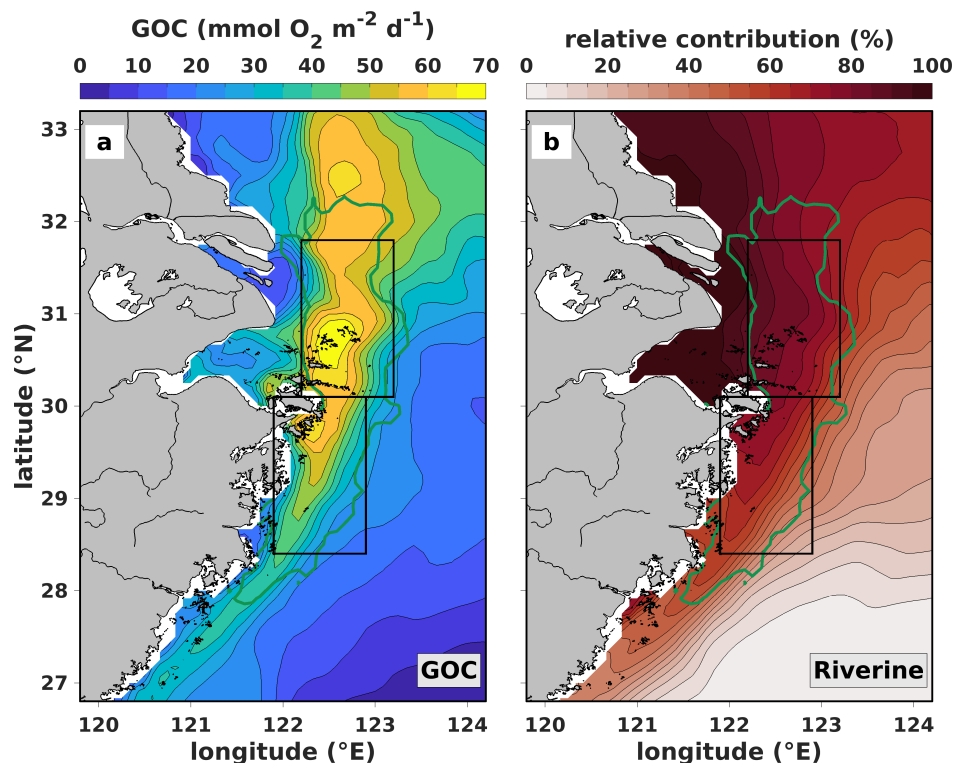
Figure 3. (a) Gross O_2 consumption (GOC) and (b) relative contribution supported by nitrogen from rivers (Changjiang and other rivers) averaged from July to November 2008–2013. Green line: area affected by hypoxia. Boxes: analysis regions used in sections 3.3 and 3.4.

145   3.3   Year-to-year variability in source-specific oxygen consumption

In order to provide insight in the relative importance of the different N sources for hypoxia formation in the northern and southern hypoxic regions, Fig. 4 shows time series of average source-specific GOC and total hypoxic area from July to November for 2008 to 2013 in both regions. Here, we define total hypoxic area as the area experiencing hypoxia at any time during July to November. The corresponding values of total GOC and the relative contributions of the different sources are provided in Table

150   S1.

In the northern region (Fig. 4a), riverine N sources (Changjiang and other rivers) account for $78.0 \pm 5.9\%$ of GOC averaged over 2008–2013, while oceanic sources support only $22.0 \pm 5.9\%$. The Changjiang River constitutes the largest contribution ranging between 63.3% in 2013 and 78.2% in 2010, which are also the years of the smallest and largest hypoxic areas in this time series, respectively. This indicates that the Changjiang River is the main control on O_2 consumption in the northern region.

155   In the southern region (Fig. 4b), total GOC is about 24% lower than in the northern region, and the riverine contribution is also lower ($61.6 \pm 5.7\%$). The average Changjiang contribution is $56.9 \pm 5.1\%$, while the contributions from the Kuroshio and
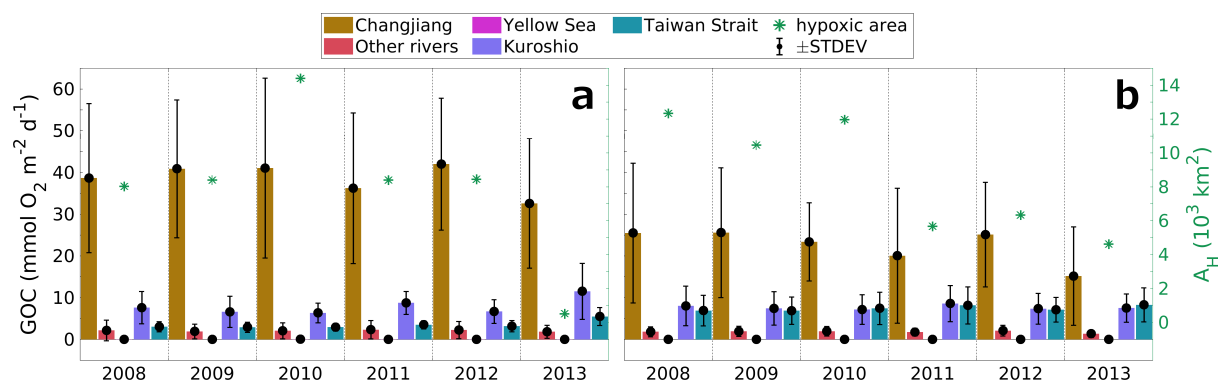
Figure 4. Average source-specific gross O_2 consumption (GOC) and total hypoxic area during July–November of 2008 to 2013 in the (a) northern and (b) southern regions (see Fig. 1). Same legend for both panels.

Taiwan Strait account for $19.5 \pm 2.2\%$ and $18.9 \pm 3.2\%$, respectively. Clearly, N sources other than from the rivers play a role in the southern region.

Interestingly, larger hypoxic areas tend to coincide with high contributions from the Changjiang River and small oceanic contributions (2008–2010), while the tends to be Changjiang is less important in years of small hypoxic areas (2011 and 2013). This suggests that the water mass distribution is an important factor for controlling the extent of hypoxia in the southern region, with a higher Changjiang contribution supporting larger hypoxic areas.

The water mass distribution in the southern region is strongly influenced by large-scale wind patterns, i.e. the East Asian monsoon, and variations in the wind field may affect both seasonal and year-to-year variability in GOC and thus hypoxia. Next, we analyze the seasonality of source-specific GOC and hypoxic area in relation to the large-scale winds in the southern region in years of small and large hypoxic areas.

3.4 Seasonal cycle of oxygen consumption and hypoxia in the southern region

In the southern region, the largest and smallest hypoxic areas are simulated in 2008 and 2013, respectively. Figure 5 presents monthly time series of source-specific GOC and total hypoxic area for both years. In addition, it shows the 6-year monthly average of the meridional wind speed 10 m above sea level (v_{10}) and the corresponding anomalies for both years.

In general, the Changjiang's GOC fraction tends to be high in spring (March and April) and in late summer (September). This seasonal pattern is explained by shifts in the dominant wind direction, where the transition from southward to northward winds from March to July results in a northward transport of both the Changjiang River plume and Kuroshio and Taiwan Strait waters increasing the oceanic contribution to GOC. In contrast, the reversal from northward to southward winds in September results in a southward movement of river-influenced water masses. Year-to-year differences in hypoxia, e.g. in June or September, result from differences in the Changjiang contribution. The large Changjiang contribution to GOC in September 2008 (Fig. 5a) and in June 2013 (Fig. 5b) are followed by large hypoxic areas, while hypoxia almost vanishes in August 2013 when the
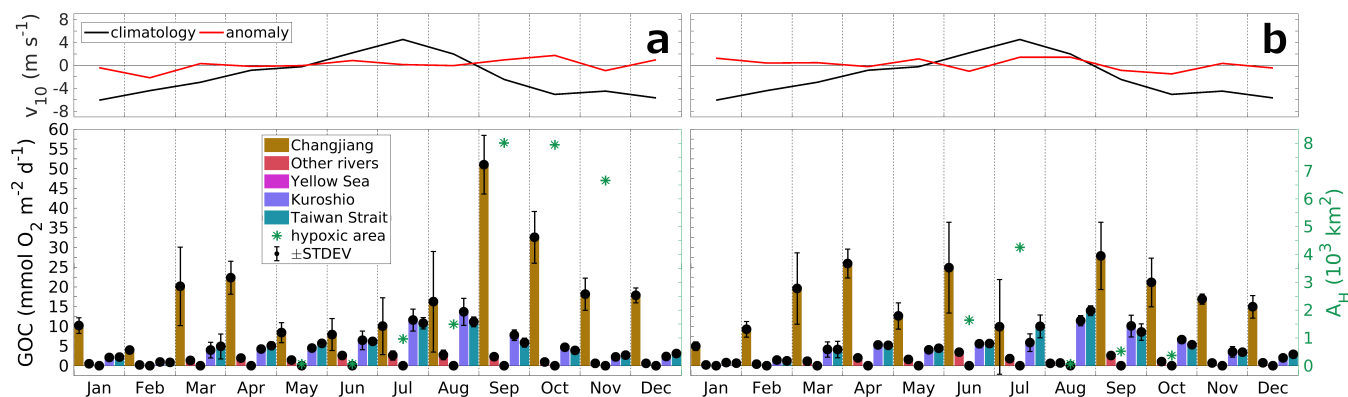
Figure 5. Monthly time series of source-specific contributions to GOC and total hypoxic area (A_H) in the southern region (see Fig. 1) in (a) 2008 (year of largest A_H) and (b) 2013 (smallest A_H), and anomaly of northward wind speed 10 m above sea level (v_{10}) relative to 2008–2013 ('climatology') averaged over the ECS (25–33 °N, 119–125 °E). Same legend and axes for both panels.

Changjiang contribution diminishes. These increases (decreases) in the Changjiang contribution also coincide with significant increases (decreases) in FW thickness and thus stratification (see Fig. S2).

180   The significant year-to-year differences in both the Changjiang contribution to GOC and thus the hypoxic area can be related to year-to-year variability in the wind field and also in FW discharge. The anomalously weak southward winds in September/October of 2008 allowed for a longer presence of Changjiang FW and nutrients in the region stimulating organic matter production and stabilizing vertical density stratification (see Fig. S2). The higher FW discharge in 2008 compared to 2013 (see Fig. 2) also supported stronger stratification. Consequently, the Changjiang contribution to GOC and thus hypoxic

185 area remained large through October and only dropped in November. In June 2013, the anomalously weak northward winds allowed Changjiang water to be transported into the region. This caused an increase in GOC and – supported by less wind-induced mixing – stratification, and thus hypoxia. The opposite occurred in July/August of 2013, when anomalously strong northward winds pushed the Changjiang River water northward resulting in a decrease in GOC, stratification and thus hypoxia.

    The results from sections 3.3 and 3.4 highlight the importance of Changjiang River nutrients for hypoxia formation in both

190 the northern and the southern regions and suggest that N load reductions in the Changjiang River would mitigate hypoxia also in the southern region. This raises the question how a potential future decline in open-ocean O_2 concentrations would affect hypoxia in the ECS, which is analyzed next.

3.5   Potential future changes in hypoxia

The open-ocean water masses travelling to the hypoxic region are not only relevant as N sources but may also act as sources

195 of low-O_2 water, thus preconditioning the region for hypoxia. Consequently, a future decline in subsurface open-ocean O_2 concentrations may exacerbate hypoxia in the ECS. To investigate how reduced open-ocean O_2 concentrations and reduced Changjiang River N loads would affect hypoxia, we performed the scenario simulations described in section 2.1. Figure 6

shows the cumulative hypoxic exposure in the bottom layer from July to November averaged over 2008–2013 for the reference
simulation (Fig. 6a), the nutrient reduction scenario with 50% lower Changjiang River N loads (Fig. 6b), the scenario with

200 20% lower open-ocean O_2 concentrations (Fig. 6c) and the combined scenario with reduced N load and lower open-ocean O_2
(Fig. 6d). Table 1 presents the corresponding changes in different hypoxia metrics between the reference case and the scenarios.

In the reference case (Fig. 6a), hypoxia occurs between 28–32 °N and from the coast to about 123 °E, on average affecting
$19.3 \pm 8.1 \times 10^3 \, \mathrm{km}^2$. Maximum hypoxic exposure is 28 days in the central parts of the hypoxic region. Hypoxia is still present
in the simulation with 50% lower Changjiang River N loads (Fig. 6b), but its areal extent and hypoxic exposure are significantly

205 reduced. Areas affected by continuous hypoxia of more than one and two weeks are reduced by >60%. Hypoxic area increases
by about 50% under reduced open-ocean O_2 concentrations (Fig. 6c) and regions affected by continuous hypoxia for more
than one and two weeks expand by 86% and 118%, respectively (Table 1).

Under reduced open-ocean O_2 and reduced Changjiang River N load (Fig. 6d), simulated changes relative to the reference
are still negative, although comparably small (see Table 1). This suggests that N load reductions constitute a potent means

210 to mitigate hypoxia under present conditions and to counteract deterioration of O_2 conditions in ECS due to open-ocean
deoxygenation.
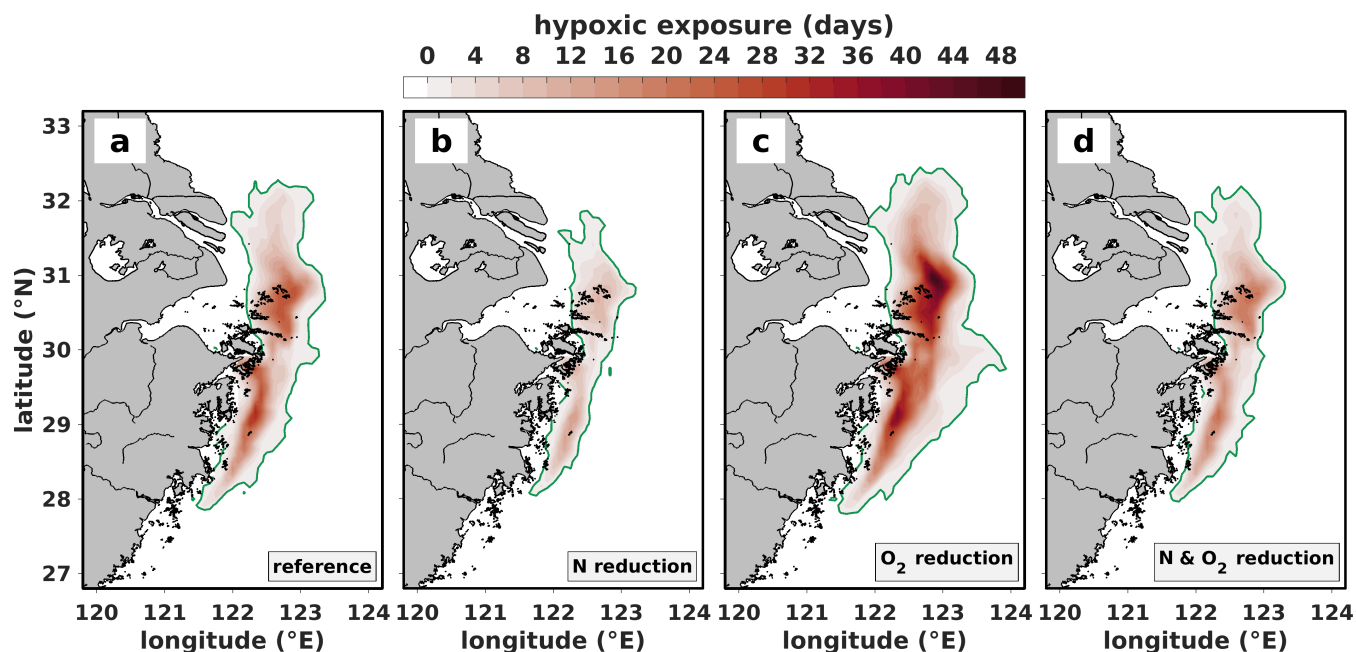
Figure 6. Average cumulative hypoxia during 2008–2013 derived from simulated bottom O_2 concentrations for (a) the reference simulation,
(b) the nitrogen reduction scenario, (c) the reference simulation with reduced oceanic O_2 concentrations, and (d) the nitrogen reduction
scenario with reduced oceanic O_2 concentrations.

Table 1. Hypoxia metrics for the four different cases presented in Fig. 6. A_H: hypoxic area; $A_{H,1w}$/$A_{H,2w}$: areas with continuous hypoxia for one and two weeks, respectively; A_A: anoxic area (all in 10^3 km^2). Values for the reference simulation are total areas. Other values are changes in area (Δ) relative to the reference simulation.

Metric	2008	2009	2010	2011	2012	2013	mean $\pm$ stdev
				Reference			
A_H	23.3	20.8	31.0	16.3	18.2	6.4	19.3 ± 8.1
$A_{H,1w}$	11.2	11.5	18.7	2.8	5.2	0.1	8.2 ± 6.8
$A_{H,2w}$	4.0	5.7	11.8	0.6	0.9	0.0	3.8 ± 4.5
A_A	1.8	2.2	3.5	0.1	0.6	0.0	1.4 ± 1.4
				N reduction			
ΔA_H	-11.2	-8.3	-12.4	-14.9	-13.8	-4.9	-10.9 ± 3.8
$\Delta A_{H,1w}$	-6.9	-7.4	-11.7	-2.8	-4.7	-0.1	-5.6 ± 4.0
$\Delta A_{H,2w}$	-3.0	-4.4	-8.8	-0.6	-0.9	0.0	-2.9 ± 3.3
ΔA_A	-1.3	-1.9	-3.5	-0.1	-0.6	0.0	-1.4 ± 1.3
				O_2 reduction			
ΔA_H	13.6	10.7	9.3	10.6	11.4	9.3	10.8 ± 1.6
$\Delta A_{H,1w}$	10.4	6.6	7.5	6.5	9.3	2.3	7.1 ± 5.1
$\Delta A_{H,2w}$	6.8	6.8	7.0	3.0	3.1	0.0	4.5 ± 2.9
ΔA_A	0.5	0.5	2.8	0.1	0.3	0.0	0.7 ± 1.1
				N and O_2 reduction			
ΔA_H	-2.3	-2.5	-3.9	-11.8	-5.8	-3.8	-5.0 ± 3.6
$\Delta A_{H,1w}$	-2.1	-2.5	-4.4	-2.8	-3.2	-0.1	-2.5 ± 1.4
$\Delta A_{H,2w}$	-1.1	-1.6	-4.1	-0.6	0.0	0.0	-1.4 ± 1.4
ΔA_A	-0.7	-1.5	-3.2	-0.1	0.0	0.0	-1.0 ± 1.2

4   Discussion

4.1   Hypoxia under current environmental conditions

Previous observational studies of the hypoxic region off the Changjiang River Estuary have concluded that the regions north
and south of 30 °N differ with respect to the nutrient sources contributing to hypoxia formation (Chi et al., 2017; Li et al., 2002;
Zhu et al., 2011). The northern region is considered to be dominated by Changjiang River inputs, while oceanic nutrient sources
are thought to be important in the southern region. Here, we present the first quantification of the relative contributions of the
different nutrient sources to hypoxia formation. Our results show that the riverine contribution to GOC during the hypoxia

season (July to November) during 2008–2013 steadily decreases from the northwest to the southeast in the region typically
220 affected by hypoxia (see Fig. 3). The riverine contribution to GOC dominates the northern region, supporting $78.0 \pm 5.9\%$ of
GOC ($73.9 \pm 5.4\%$ attributed to Changjiang River), while oceanic N supports only $22.0 \pm 5.9\%$ ($15.2 \pm 3.7\%$ from Kuroshio;
see Fig. 4a and Table S1) confirming that Changjiang River nutrient loads control O_2 consumption and thus hypoxia formation
in the northern region.

When considering July–November averages in the southern region, the relative contribution of riverine N to GOC is
225 $61.6 \pm 5.7\%$ ($56.9 \pm 5.1\%$ from Changjiang River), while the oceanic contribution is $38.4 \pm 5.7\%$, nearly twice the oceanic
contribution in the northern region, with roughly equal contributions from Kuroshio and Taiwan Strait (see Fig. 4b and Table
S1). However, analysis of the seasonal cycles of the source-specific contributions to GOC reveals that hypoxia expands when-
ever the Changjiang contribution to GOC increases (see Fig. 5 and Table S2). This demonstrates that the Changjiang River also
is the major factor for hypoxia development in the southern region.

230 High Changjiang contributions to GOC correspond to high FW thicknesses and thus strong stratification (see Fig. S2), the
other essential factor for hypoxia formation and maintenance than biological O_2 consumption. In contrast, the hypoxic area
is smaller during periods with high oceanic contributions as a result of reduced GOC and weaker stratification (see Figs. 5
and S2). Nevertheless, the relatively low subsurface O_2 concentrations in the oceanic water masses precondition the southern
region for hypoxia formation.

235 Our analyses of the seasonal cycles of GOC and stratification in relation to the large-scale meridional winds further show
that the East Asian monsoon and its year-to-year variability result in seasonal and year-to-year variability in the contributions
to GOC, in stratification, and thus in hypoxia in the southern region (see Fig. 5). During winter monsoon (September–April),
the southward winds transport the Changjiang River plume towards the southern region, where it arrives in August/September
supporting the formation of hypoxia. The opposite occurs during summer monsoon (May–August), when the northward winds
240 transport oceanic water masses into the southern region and push the river plume northward out of the southern region.

With respect to year-to-year variability, our analysis further shows that anomalies in the meridional winds during summer
and during the transition from summer to winter monsoon (August–October) significantly affect the water mass distribution,
and thus GOC and hypoxia on subseasonal scales, especially south of $30\,°N$ (see Fig. 5 and Table S2). Particularly weak
southward winds in September/October of 2008 resulted in a longer presence of the Changjiang River plume in the southern
245 region, increasing stratification and GOC, which in turn caused the formation of the largest hypoxic area in that region during
2008–2013. The opposite occurred in July/August of 2013, when particularly strong northward winds pushed the Changjiang
River plume north. This resulted in a drop in GOC and stratification, and in the vanishing of hypoxia, which had started
forming in June through early July as anomalously weak northward winds allowed a southward transport of the Changjiang
River plume. This is in agreement with Zhang et al. (2018) who simulated similar almost immediate responses of hypoxia to
250 changes in the wind field as a result of the redistribution of Changjiang FW causing changes in stratification. Our study expands
on their work by additionally demonstrating the importance of the redistribution of Changjiang nutrients for GOC.

These short-term changes in GOC, stratification and hypoxic area in response to variations in the large-scale winds illustrate
the complexity of the region with respect to atmosphere–ocean interaction and their effect on hypoxia. They further highlight

the necessity of a high spatio-temporal resolution of both model approaches and observations to understand the causes of

255 hypoxia formation and maintenance in the ECS under current environmental conditions.

In general, our results are in agreement with Zhang et al. (2019) who found that riverine N supports 56% of water column integrated PP (8-year average) in the ECS regions shallower than 50 m, which are most susceptible to hypoxia (see Fig. 3). This confirms the high contribution of Changjiang N to GOC found for our analysis regions. Similarly, the comparably low oceanic contributions, especially in the northern region, are in line with their results. Zhang et al. (2019) also found a relatively

260 high contribution (22%) of atmospheric N to PP suggesting that atmospheric N deposition should be considered in future biogeochemical modeling studies of the ECS.

4.2 Mitigation of hypoxia in the present and future

Our analysis of the changes in hypoxic area and hypoxic exposure under reduced Changjiang River N loads (see Fig. 6 and Table 1) underlines the high potential of riverine N load reductions to mitigate hypoxia. The simulated average reduction

265 in hypoxic area by 56.5% is almost proportional to the N load reduction of 50% but varies significantly from year to year (39.9% in 2009 to 91.4% in 2011). Comprehensive further studies are needed to assess the effect of Changjiang River nutrient reductions on hypoxia. In this context, it is important to consider all processes relevant for hypoxia formation. Zhou et al. (2017b) conducted a scenario with 50% lower Changjiang River nutrient loads and found a reduction in hypoxic area by only about 20%. However, they did not consider SOC, a major sink of O_2 in the ECS (Zhang et al., 2017), which may explain this

270 discrepancy.

Analysis of the scenario where open-ocean O_2 concentrations were 20% lower than at present shows that the reduced lateral supply due to declining open-ocean O_2 concentrations (Bopp et al., 2017) may significantly exacerbate hypoxia in the ECS, indicated by the expansion of both hypoxic and anoxic areas by about 50% (see Fig. 6 and Table 1). However, we further found that a 50% reduction of the Changjiang River N loads more than compensates for this increase in hypoxia.

275 The 20% reduction of open-ocean O_2 concentrations that was applied here corresponds to changes in subsurface O_2 of about 30–40 $mmol\,O_2\,m^{-3}$ projected by earth system models (Bopp et al., 2017), but does not consider other climate change effects such as local changes in water temperature or stratification. For instance, an increase in temperature would reduce O_2 solubility, which would worsen O_2 conditions further as shown, e.g. for the NGoM (Laurent et al., 2018). Therefore, more comprehensive studies on climate change impacts on hypoxia in the ECS are required.

280 ### 4.3 Comparison with the northern Gulf of Mexico

In the following, we compare the hypoxic zones of the ECS and the NGoM with respect to their sizes and timing, based on observations, and the main sources of O_2 consumption, based on the results of this study and an analogous nutrient tracing study for the NGoM (Große et al., 2019). The key messages of this comparison are summarized in Table 2.

The spatial extent of observed hypoxia in the ECS is similar to that in the NGoM, affecting areas on the order of 15,000 km^2.

285 In the NGoM, hypoxia is frequently observed from May to September when seasonal stratification is strongest due to high riverine freshwater inputs. In the ECS, the hypoxic season starts about one month later in June, with hypoxia forming directly off

Table 2. Comparison of hypoxia in the northern Gulf of Mexico (NGoM) and the East China Sea (ECS).

Characteristic	NGoM	ECS
Hypoxic area (km^2)	15,000 [1]	15,000 [2]
Hypoxic season	May–September [1]	June–November [3]
GOC controls (seasonal average)	riverine $\gg$ oceanic	$>30\,°N$: riverine $\gg$ oceanic $<30\,°N$: riverine $\approx$ oceanic
Main riverine source(s)	Mississippi & Atchafalaya	Changjiang
Controls of year-to-year variability	Mississippi & wind field	Changjiang & wind field

[1] Rabalais et al. (2002); [2] Li et al. (2002), Zhu et al. (2011, 2017); [3] Wang et al. (2012)

the Changjiang River Estuary. Hypoxia can be observed until October/November, though the hypoxic zone moves southward over summer.

In the NGoM, riverine nutrients from the Mississippi and Atchafalaya Rivers dominate O_2 consumption. In the ECS, only in the region north of $30\,°N$ is O_2 consumption dominated by riverine nutrient inputs that almost exclusively originate from the Changjiang River. South of $30\,°N$, the oceanic contribution to O_2 consumption is comparable to that of riverine nutrients. However, driven by year-to-year variability in the East Asian monsoon, the Changjiang River appears to be the main control of hypoxia formation also in the southern region. Similarly, the Mississippi River in concert with the local wind field seems to be the main control of year-to-year variability in hypoxic area in the NGoM. There, episodic upwelling-favorable winds can result in an offshore transport of Mississippi FW and nutrients leading to a reduction in stratification and O_2 consumption and thus hypoxia (Feng et al., 2014).

5   Conclusions

To our knowledge, this is the first study quantifying the relative contributions of individual riverine and oceanic nutrient sources on hypoxia formation in ECS using active element tracing. Therefore, it constitutes an important milestone towards the quantification of the contributions of riverine vs. oceanic nutrient sources to hypoxia formation in the ECS.

Our results suggest that N from the Changjiang River is the dominant driver of O_2 consumption ($73.9 \pm 5.4\%$) north of $30\,°N$ under present-day environmental conditions (2008–2013). Contrary to observational insights and despite high contributions of N from Kuroshio and Taiwan Strait to O_2 consumption on seasonal time scales ($19.5 \pm 2.2\%$ and $18.9 \pm 3.2\%$, respectively), the Changjiang River waters are also the main factor for hypoxia formation south of $30\,°N$.

Our analysis highlights the importance of considering subseasonal time scales for understanding the controls of hypoxia formation and its year-to-year variability in this region. The East Asian monsoon and its associated change in large-scale wind patterns control water mass distribution and thus O_2 consumption and stratification. Year-to-year variability in the intensity

of the winds can lead to significant differences in the amount of Changjiang River water in the southern region explaining year-to-year variability in hypoxia.

310 Reductions in the Changjiang River nutrient loads appear to have a high potential for mitigating hypoxia and to counteract the likely future decline of open-ocean O_2 supplied to the region. However, more comprehensive studies of both the effect of riverine nutrient load reductions and climate change effects are required.

Code and data availability. A modified version of the ROMS source code that facilitates writing of model diagnostics needed by the element tracing software (ETRAC) is available at https://github.com/FabianGrosse/ROMS_3.7_for_ETRAC.

315 ETRAC is available via https://github.com/FabianGrosse/ETRAC. Due to the large data size, ROMS and ETRAC output is freely available upon request from the corresponding author (fabian.grosse@dal.ca).

Competing interests. The authors declare that they have no conflict of interest.

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