



1       **An interrupting mechanism to prevent the formation of coastal hypoxia by winds**

2                       **Juan Yao<sup>1,2</sup>, Juying Wang<sup>3</sup>, Hongbin Liu<sup>2,4</sup>, Kedong Yin<sup>1,2\*</sup>**

3

4       <sup>1</sup>School of Marine Sciences/Guangdong Key Laboratory of Marine Resources and Coastal  
5       Engineering, Sun Yat-Sen University, Guangzhou 510275, Guangdong, China.

6       <sup>2</sup>Southern Marine Science and Engineering Guangdong Laboratory (Zhuhai), Zhuhai 519082,  
7       Guangdong, China.

8       <sup>3</sup>National Marine Environmental Monitoring Center, Ministry of Ecology and Environment,  
9       Dalian 116023, Liaoning, China.

10       <sup>4</sup>Department of Ocean Sciences, Hong Kong University of Science and Technology, Clear Water  
11       Bay, Hong Kong SAR, China.

12

13       **Corresponding author:** Kedong Yin ([yinkd@mail.sysu.edu.cn](mailto:yinkd@mail.sysu.edu.cn))

14       **Email address for each author listed:** Juan Yao ([yaoj8@mail2.sysu.edu.cn](mailto:yaoj8@mail2.sysu.edu.cn)); Juying Wang  
15       ([jywang@nmemc.org.cn](mailto:jywang@nmemc.org.cn)); Hongbin Liu ([liuhb@ust.hk](mailto:liuhb@ust.hk))

16

17       **Running head:** frequent winds interrupt the formation of hypoxia

18       **Keywords:** hypoxia; wind events; Hong Kong waters; ecosystem buffering; climate change



19 **Abstract**

20           Enrichment of nutrients is believed to lead to coastal hypoxia which have become a  
21 seasonal phenomenon over large river estuarine areas such as the Mississippi River-Northern  
22 Gulf of Mexico and Changjiang-East China Sea. There is a similar nutrient enrichment process  
23 in the Pearl River. However, hypoxia occurs only as episodic events over a relatively small area.  
24 We hypothesize that frequent wind events play the interruptive mechanism in preventing the  
25 seasonal formation of bottom hypoxia. We used 29 years time series data of dissolved oxygen  
26 (DO) and winds in the Hong Kong coastal waters to test the hypothesis. Our results show that  
27 bottom DO at 3 stations in southern waters of Hong Kong occasionally drops below the hypoxic  
28 level (2 mg/L), lasting only for less than one month in summer. Episodic hypoxia events appear  
29 to occur more frequently in recent years, but bottom DO does not show a significantly decreasing  
30 trend. The wind speed of 6 m/s appears to be a threshold, above which a wind event could  
31 destroy water column stratification and interrupt the formation of low-oxygen (DO <3 mg/L)  
32 water mass. The wind events above the threshold occur 14.3 times in June, 14.2 times in July and  
33 10.0 times in August during 1990-2018. This explains why episodic events of hypoxia hardly  
34 occur in June and July, and only occasionally in August. The frequency of such the above-  
35 threshold events appears to show a decreasing trend during 1990-2018, which coincides with an  
36 increasing occurrences of episodic hypoxia events in recent years.

37 **1. Introduction**

38           Hypoxic environments are natural existence throughout geological time and distributed in  
39 many coastal ecosystems around the world (Diaz and Rosenberg 2008). During the past half-  
40 century, however, the duration, intensity and extent of coastal hypoxia have been exacerbated by  
41 the increased nutrients input associated with human activities (Breitburg et al. 2018; Du et al.



42 2018). Typical examples of large dead zones in coastal zones around the world include Baltic  
43 Sea, northern Gulf of Mexico, northwestern shelf of the Black Sea and Changjiang-East China  
44 Sea (Bianchi et al. 2010; Capet et al. 2013; Vali et al. 2013; Zhu et al. 2016).

45 The Pearl River is the second largest river in China and the 13<sup>th</sup> largest in the world. The  
46 annual average river discharge is  $10,524 \text{ m}^3\text{s}^{-1}$  with 20% occurring during the dry season and 80%  
47 during the wet season (Yin et al. 2004). The Pearl River estuary flows into the northern part of  
48 the South China Sea. Hong Kong is part of its eastern shores (Figure 1). The Pearl River  
49 estuarine coastal waters are very dynamic driven by factors such as river discharge, oceanic  
50 waters, coastal currents and monsoons (Yin et al. 2004).

51 In recent years, the loading of anthropogenic nutrients has been increasing in the Pearl  
52 River (Hu and Li 2009; Su et al. 2017), which is comparable to Mississippi and Yangtze where  
53 hypoxia has become a seasonal phenomenon (Li et al. 2002; Rabalais et al. 1998, 2002, 2010;  
54 Zhu et al. 2011; Wang et al. 2016). The increase in nutrients in the coastal waters is usually  
55 assumed to result in hypoxia in the estuary and coastal waters. However, over the coastal scale of  
56 the Pearl River estuarine influenced waters in the South China Sea, hypoxia has only occurred as  
57 episodic events over small areas (Yin et al. 2004; Xu et al. 2010; Li et al. 2019). Recent  
58 investigations reported a new occurrence of hypoxia in the coastal waters south of Macau (Ye et  
59 al. 2013; Su et al. 2017; Lu et al. 2018; Qian et al. 2018), but the spatial scale is only a small part  
60 of the Pearl River estuarine plume which influences the large part of the Northern South China  
61 Sea.

62 Water column stratification and decomposition/oxygen consumption of organic matter in  
63 the bottom water are two critical factors that lead to hypoxia, and both favourable conditions  
64 must occur simultaneously for hypoxia to develop and persist (Diaz 2001). An event of episodic



65 hypoxia in the Pearl River coastal waters was related to the hydrodynamics and water depth  
66 (Zhou et al. 2012; Ye et al. 2013; Qian et al. 2018). In addition to the buoyancy flux induced by  
67 freshwater discharge, local wind forcing plays a regulating role in the stratification stability.  
68 Previous studies found that wind-driven vertical mixing accelerates the ventilation of water  
69 column and increases oxygen replenishment in the bottom layer in the Pearl River estuarine  
70 coastal waters (Zhou et al. 2012; Wang et al. 2015) and other regions (Wilson et al. 2008; Scully  
71 2010, 2013). For example, bottom DO in hypoxic zone was observed to increase rapidly after the  
72 passage of a typhoon (Ni et al. 2016; Su et al. 2017). However, the mixing effect of typhoons is  
73 relatively short-lived in coastal ecosystems and the enhanced freshwater discharge can re-  
74 establish stratification in only a few days (Zhou et al. 2012), which facilitates the re-formation of  
75 hypoxia (Su et al. 2017). Therefore, whether strong wind events relieve the tendency of hypoxia  
76 on a longer time scale depends partly on the frequency of wind events. Numerical modeling  
77 studies have illustrated the effects of wind stress/speed variations on coastal hypoxia (Chen et al.  
78 2015; Wei et al. 2016; Lu et al. 2018). However, the role of frequency of wind events on the  
79 formation and maintenance of hypoxia is rarely studied, partly due to lack of long time series  
80 data. In order to explain the lack of seasonal phenomenon of hypoxia over the coastal scale in the  
81 Pearl River estuarine waters, we hypothesize that frequent wind events interrupt the formation of  
82 bottom hypoxia and prevent hypoxia from becoming a persistent seasonal phenomenon in the  
83 Pearl River estuarine coast. We used 29 years (1990-2018) time series data of dissolved oxygen  
84 and winds in Hong Kong to test the hypothesis. The approach is to examine the temporal trend of  
85 dissolved oxygen and frequency of wind speeds and to determine a threshold of wind speeds  
86 above which a wind event is strong enough to interrupt the formation of hypoxia.

87



88 **2. Materials and Methods**

89 The time series data of DO and other water quality variables at three stations SM17,  
90 SM18 and SM19 during 1990-2018 are obtained from the Environmental Protection Department  
91 (EPD) which has maintained a comprehensive marine water quality monitoring programme since  
92 1986 at 86 stations (Figure 1). The marine monitoring vessel “Dr. Catherine Lam” is equipped  
93 with a CTD profiler and a computer-controlled rosette water sampler to measure salinity,  
94 temperature and dissolved oxygen in situ and collect water samples for nutrients for later  
95 analysis in the laboratory. The water samples were analyzed in the EPD’s laboratory (EPD  
96 Report 2017). The three stations SM17, SM18 and SM19 are located in the southern waters of  
97 Hong Kong in coastal area of the Northern South China Sea, with depths being 12 m, 21 m and  
98 24 m, respectively (Figure 1). They are visited monthly for sampling. They are heavily  
99 influenced by the Pearl River estuarine plume during summer. Station SM18 is located in the  
100 southern end of Lamma Channel, the north end of which receives the sewage effluents from the  
101 outfalls of the CEPT (Chemically Enhanced Primary Treatment) of Stonecutter’s Island Works.

102 The daily averaged wind speeds and prevailing wind directions at Waglan Island (N  
103 22°10’56”, E 114°18’12”, Figure 1) are obtained from the Hong Kong Observatory (HKO), and  
104 are assumed to represent the overall wind field over southern waters of Hong Kong including  
105 station SM17, SM18 and SM19. The period of 29 years (from 1990 to 2018) is divided into six  
106 groups with 5 years per group to illustrate the variations in wind speeds and the frequency of  
107 wind events in the long term.

108 **3. Results**

109 **3.1. Time series of DO at surface and bottom**

110 The time series of DO during 1990-2018 at SM17, SM18 and SM19 showed that both the  
111 surface and bottom DO fluctuated, usually being high in winter and low in summer (Figure 2). In



112 summer, the bottom DO was all low in June, July and August, and drops to the hypoxic levels  
113 occasionally in August. Hypoxic DO occurred more frequently at SM18 than SM17 and SM19,  
114 with 2, 4 and 2 times at SM17, SM18 and SM19, respectively, during 29 years. A hypoxic event  
115 has never lasted over 2 months in a year at one station and it does not usually occur at the 3  
116 stations in the same month with the exception in August 2011 when the bottom DO was 0.4  
117 mg/L and 0.9 mg/L at SM18 and SM19, respectively. These indicate that hypoxic events are  
118 episodic and have not developed a seasonal phenomenon over a coastal scale covering the 3  
119 stations. Furthermore, the bottom DO does not show any clear decreasing trend over the past  
120 decades as the linear regression over time (the red dashed line) is not significant (Figure 2).

121

### 122 3.2. The relationships between bottom DO, stratification and winds

123 In summer, the water column is usually stratified in estuarine and coastal areas due to  
124 river outflow and surface heating. In this study, we use the differences in sigma density between  
125 surface and bottom layers ( $\Delta\sigma$ ) to describe the strength of water column stratification. We use  
126 the differences in DO between surface and bottom layers ( $\Delta\text{DO}$ ) and Apparent Oxygen  
127 Utilization (AOU) to indicate DO consumption due to decomposition of organic matter in the  
128 bottom water mass. The correlation analysis (Table 1) shows that bottom DO is correlated to  $\Delta\sigma$   
129 at the 3 stations with correlation coefficient,  $r$ , being  $\sim 0.70$  at  $p < 0.01$ . The correlation coefficient  
130  $r$  between  $\Delta\text{DO}$  and  $\Delta\sigma$  is all significant at 3 Stns, reaching 0.8 at SM19. Similarly, AOU is  
131 significantly correlated to  $\Delta\sigma$  as AOU increases with  $\Delta\sigma$  increasing. These relationships  
132 indicate that the strength of water column stratification plays a regulating role in the DO  
133 variability.

134 Since it takes some time for DO to be consumed to a low level from the surface DO,  
135 wind speeds over 7 days are placed before a bottom DO value in summer during 1990-2018 at



136 SM17, SM18 and SM19 (Figure 3) to inspect visually the wind effect on the low bottom DO.  
137 Generally, bottom DO is seen to decrease during a period of low wind speeds and is elevated  
138 obviously after each episodic high wind period. For example, the bottom DO increased, reaching  
139 7.8 mg/L and 6.8 mg/L in June 1990 at SM17 and SM18, respectively, after the daily averaged  
140 wind speed blew for 7 days at 14.9, 15.2, 11.5, 8.2, 6.8, 3.8 and 9.6 m/s. When an event of  
141 hypoxia occurs, it is usually after a period of low winds. For example, bottom DO was 0.4 mg/L  
142 and 0.9 mg/L in August 2011 at SM18 and SM19, respectively, after the daily averaged wind  
143 speed blew for 7 days at 4.0, 3.8, 4.4, 3.4, 5.7, 4.5 and 3.6 m/s. This reflects the close connection  
144 between bottom DO and wind speed. The correlation analysis of bottom DO, AOU,  $\Delta DO$ ,  $\Delta\sigma$  vs.  
145 wind speeds for 1-7 preceding days before a low DO event (Table 2) shows that correlation is  
146 better between these DO related parameters and 5-7 days averaged wind speed before sampling  
147 than 1-3 days averaged wind speed before sampling. We choose  $V_7$  (7 days averaged wind speed  
148 before sampling) to represent the preceding wind speed. Bottom DO is positively correlated to  
149 wind speeds while AOU,  $\Delta DO$  and  $\Delta\sigma$  are negatively and significantly correlated to wind  
150 speeds.

151 The time series of  $V_7$  for DO in summer at SM17, SM18 and SM19 (Figure 4) does not  
152 show any decreasing trend during 1990-2018. The relationship between bottom DO,  $\Delta\sigma$  and  $V_7$   
153 during 1990-2018 at three stations (Figure 5) shows that when the wind speed is higher than 8  
154 m/s,  $\Delta\sigma$  is almost close to 0 and bottom DO is usually 5-6 mg/L. The frequency of wind speed  
155  $V_7$  vs bottom DO during 1990-2018 in summer at SM17, SM18 and SM19 (Table 3(a)) shows  
156 that occurrences of bottom DO <3 mg/L is 0 when the averaged wind speed in preceding 7 days  
157 is >8 m/s.



158            However, when wind speed is between 5 and 7 m/s, the bottom DO drops to the hypoxic  
159 level occasionally. Since hypoxic events ( $\text{DO} < 2 \text{ mg/L}$ ) occurred rarely in the Pearl River  
160 estuarine coastal waters, we take 3 mg/L as the indicator or an event of low-oxygen water mass  
161 formation. The frequency of the low-oxygen events varies at different wind speeds, being 9.4%,  
162 4.9%, 1.7% and 0 of bottom DO in summer of 29 years for  $>5$ ,  $>6$ ,  $>7$  and  $>8$  m/s in wind speed,  
163 respectively. We consider an event with a probability  $<5\%$  as a small-probability event. When  
164 wind speed is  $>6$  m/s, the water column can be largely mixed and the probability of bottom  
165 hypoxic events have a  $<5\%$  chance to occur. Therefore, wind speeds  $>6$  m/s are strong enough to  
166 interrupt the water column stratification. The accumulative frequency of the  $\Delta\sigma$ -descending  
167 group vs the ascending wind speed  $V_7$  groups during 1990-2018 in summer (Table 3(b)) shows  
168 that occurrences of  $\Delta\sigma > 15$  and the other 3 groups decrease when the wind speed increases. For  
169 example, occurrences of  $\Delta\sigma > 15$  is 0% at SM19 and  $<5\%$  at SM17 and SM18 when the averaged  
170 wind speed in preceding 7 days is  $>6$  m/s.

### 171            3.3. The frequency of wind events

172            Winds  $>6$  m/s are found to be a wind event that is strong enough to mix the water column  
173 (Figure 5). Wind events were very frequent over southern waters of Hong Kong (Figure 3). The  
174 frequency vs. wind speeds in June, July, August and September during 1990-2018 shows that  
175 wind speeds are usually between 2 m/s and 9 m/s and wind speeds  $<2$  m/s or  $>9$  m/s occur rarely  
176 (Figure 6). In June and July, wind speeds are mostly 4-7 m/s and are reduced to 2-5 m/s in  
177 August. In September, strong wind speeds occur more frequently than August. The accumulative  
178 frequency of wind speeds  $>6$  m/s is 14.3, 14.2, 10.0 and 13.0 days per month in June, July,  
179 August and September, respectively. Compared to other months, the wind condition is much  
180 weaker in August. This means that August is most vulnerable to episodic events of hypoxia.



181           The monthly averages of wind speeds for July and August decrease significantly during  
182 29 years (Figure 7). It is more apparent that there appears to be a decrease in wind speeds  $\geq 6$  m/s  
183 and an increase in wind speeds  $< 6$  m/s in August based on 5 years grouping of wind speeds  
184 during 1990-2018 while no major changes in wind speeds are apparent in June, July and  
185 September (Figure 8). This suggests that the strong wind events occur less frequently in August  
186 than other months during the last 10 years (Figure 6). Among the five groups, the frequency of  
187 wind speeds  $\geq 6$  m/s appears to show a decreasing trend, especially in June, July and August. For  
188 example, in August, the frequency of wind speeds  $\geq 6$  m/s is 12.6, 12.4, 8.4, 10.6, 7 and 9 days  
189 per month during 1990-1994, 1995-1999, 2000-2004, 2005-2009, 2010-2014 and 2015-2018,  
190 respectively. In the long period of time, this decreasing trend means that low winds may be more  
191 frequent, which potentially results in an increase in the frequency of hypoxia in summer.

#### 192 **4. Discussion**

193           There is a lack of a significantly decreasing trend in DO in the southern water of Hong  
194 Kong during 1990-2018 with occasional drops below the hypoxic level a few times in summer.  
195 Nutrients in the southern waters are non-limiting (EPD Report 2017). However, the drop has not  
196 lasted for two consecutive months at one station and has not happened at 2 stations in the same  
197 month. This demonstrates that the temporal scale of hypoxia occurring in southern waters of  
198 Hong Kong are episodic, not a seasonal phenomenon and the spatial scale is small, not even  
199 covering the two stations within 12 km. Yin et al. (2013) proposed the concept of ecosystem  
200 buffering capacity against hypoxia, which are determined by a number of drivers and processes  
201 (Yin and Harrison 2007, 2008; Harrison et al. 2008; Ho et al. 2008; Yin et al. 2010). Wind  
202 events of typhoons have been reported to mix the water column and subsequently increase  
203 bottom DO (Paerl et al. 1998, 2001; Yin and Harrison 2007; Zhou et al. 2012, 2014; Ni et al.



204 2016). This study gives evidence to testify the hypothesis that frequent strong winds interrupt the  
205 stratification and slow down the formation of hypoxia.

#### 206 4.1. The formation of low DO water mass

207 It is the residence time of the bottom layer and DO consumption rate that determine the  
208 formation of hypoxia in the bottom. The former depends on the physical processes of water  
209 advection, vertical mixing, and air-sea exchange and the latter photosynthesis, chemolithotrophic  
210 production, and respiration in the water column and sediment oxygen demand (Paerl 2006; Chen  
211 et al. 2015). When the supply of oxygen is cut off to bottom waters, usually due to stratification  
212 of the water column, and consumption of DO through respiration exceeds resupply during a  
213 sufficiently long period of time, DO will decrease, reaching the level of hypoxia if organic  
214 matter is sufficient (Diaz 2001). In many estuarine and coastal systems, excess nutrients lead to  
215 increased primary production, adding new organic matter to the coastal waters. Generally, a  
216 coastal water body receiving a large freshwater input with basic features of low physical energy  
217 (tidal, currents, or wind) is prone to hypoxia (Diaz 2001).

#### 218 4.2. The interruptive role of wind events on hypoxia formation

219 Many studies have demonstrated that physical processes such as estuarine circulation,  
220 tide and wind determine the residence time of bottom water and play a crucial role in the  
221 establishment, maintenance and termination of hypoxia (Simpson et al. 1981, 1990; Yin and  
222 Harrison 2007; Rabouille et al. 2008; Wang et al. 2012). Whether an estuary is stratified or  
223 mixed depends on the transformation between kinetic energy and potential energy induced by  
224 these physical factors (Simpson et al. 1981, 1990). The freshwater from the river flows above the  
225 seawater, and hence exerts a buoyancy/stratifying influence in the estuary. The tides affect the  
226 water column in two ways: tidal straining and tidal stirring (Simpson et al. 2005). Winds can



227 affect the turbulent mixing in several ways, including (1) direct mixing due to shear imposed at  
228 the surface by the wind stress, (2) generation of waves and wave breaking, and (3) modification  
229 of the plume velocity profile, and shear, through coastal set-up and/or straining of isopycnals (Li  
230 et al. 2007; Wilson et al. 2008; Wang et al. 2015; Pan and Gu 2016). Chen et al. (2015) pointed  
231 that wind speed and direction are the most important among the physical factors influencing  
232 oxygen dynamics in the Yangtze Estuary. A 10% increase in wind speed reduced the areal extent  
233 of hypoxia by 46.66%, and a 10% reduction increased the hypoxic area by 67.28% (Chen et al.  
234 2015). A previous study has shown the effect of a typhoon event on interrupting the formation of  
235 hypoxia in the Hong Kong waters (Zhou et al. 2012). In the Mississippi River-Northern Gulf of  
236 Mexico, the size of the ‘dead zone’ was found to be strongly correlated with high river  
237 discharges and strong stratification (Justic et al. 1996). The Baltic Sea with persistent  
238 stratification is prone to the occurrences of hypoxia (Conley et al. 2002; Diaz and Rosenberg  
239 2008; Lehmann et al. 2014). The lack of wind events is a favourable condition for the formation  
240 of hypoxia when organic matter supply is sufficient. Our results show that the occurrences of  
241 hypoxia are usually after a long period of low winds (Figure 3). The wind speed of 6 m/s can be  
242 considered as the threshold of a wind event, above which the stratified water column can be  
243 mixed to interrupt the formation of the bottom hypoxia in coastal waters south of Hong Kong.  
244 The examination of the monthly frequency of such wind events (>6 m/s) reveals that wind  
245 events >6 m/s occur every two or three days on average during June, July and September, which  
246 appears to be frequent enough to raise the bottom DO in southern waters of Hong Kong. The  
247 wind speed is the lowest in August, which explains why most hypoxic events at SM17, SM18  
248 and SM19 occurred in August (Figure 2). The formation process of hypoxia is interrupted, reset  
249 and starts over again after such a wind event. The consumption of bottom DO to the hypoxic



250 level will take some time as the development of phytoplankton blooms and the bacteria  
251 degradation of dissolved organic matter require a period of time, saying 7 days at least. Previous  
252 study found that phytoplankton in bottled waters took 3-4 days to reach the maximum during the  
253 incubation of the estuarine water (Yin et al. 2000, 2008) and similarly, DO consumption in  
254 bottled samples of estuarine surface waters takes 3-4 days to consume 5-7 mg/L to 2 mg/L  
255 during dark closed incubation, but DO consumption in bottled bottom water to 2 mg/L took  
256 longer time (J. Yao unpubl). Each time when the bottom hypoxia is going to be developed,  
257 strong winds slow down its formation. A stronger episodic wind event will interrupt its  
258 formation completely. The resuming processes may not be a simple recovery as estuarine coastal  
259 water masses are highly variable, which influences phytoplankton growth and its organic matter  
260 sinking to the bottom water. In addition, the consumption of DO may also be variable.  
261 Apparently, each wind event supplies oxygen to bottom layer, which resets the bottom to a  
262 higher initial DO value for consumption and hence, leads to longer formation time for the next  
263 hypoxia event to occur. This explains why seasonal and coastal scale hypoxic events have rarely  
264 occurred in Hong Kong waters despite of the large nutrient inputs.

265         Due to the southwest monsoon, the Pearl River estuarine freshwater flows across the  
266 southern waters of Hong Kong. SM17 is most affected by the river outflow. SM18 is located in  
267 the southern end of Lamma Channel. In the northern end of it, the sewage effluent outfalls of the  
268 biggest sewage treatment plant (Stonecutter's CEPT Plant) in Hong Kong are laid in the bottom.  
269 Part of the treated effluent flows through the Lamma Channel to the southern waters. Thus,  
270 SM18 is most influenced by the sewage effluent. The shallow depth of 12 m at SM17 makes  
271 wind mixing more effective at SM17. SM19 appears to be least influenced by the estuarine



272 plume and sewage effluent, and by a wind event due to its deepest depth (24 m). This explains  
273 low occurrences of hypoxia at SM19 at wind speeds >5 m/s (Table 3).

#### 274 4.3. Ecosystem buffering capacity

275 Cloern (2001) pointed out that some coastal ecosystems can accommodate an excess  
276 nutrient enrichment without showing apparent eutrophication symptoms. Yin et al. (2013)  
277 proposed that it is the ecosystem buffering that makes the Pearl River estuary “robust” to N  
278 enrichment. It is determined by physical driving forces such as monsoons, river outflow, tidal  
279 cycles and rainfall, and some of them become dominant over different temporal and spatial  
280 scales, which induce circulation, stratification and turbulent mixing. As a result, the fields of  
281 light, salinity, temperature and nutrients vary, thus influencing algal growth and DO  
282 consumption. When anthropogenic nutrients enter coastal waters, there would be a series of  
283 physical and biological processes before nutrient enrichment causes any ecological impacts. If  
284 the ecosystem buffering capacity is large enough, the input may not lead to any impacts.  
285 Inversely, algae bloom and hypoxia may occur.

286 Lacking of a seasonal hypoxia over the coastal scale in the Pearl River estuarine  
287 influenced waters suggests that the ecosystem buffering capacity plays a regulating role in  
288 controlling the production and accumulation of algal blooms and DO consumption and potential  
289 occurrence of hypoxia (Lee et al. 2006; Harrison et al. 2008). In addition to these physical  
290 controls on hypoxia, the low PO<sub>4</sub> concentrations relative to nitrogen (N:P~100:1) may limit the  
291 phytoplankton biomass production through P limitation and hence the amount of organic matter  
292 sinking to depth (Yin et al. 2004). Zooplankton grazing pressure could also be an influencing  
293 factor in limiting the phytoplankton biomass via the top down control in HK waters in summer  
294 (Ho et al. 2008). Strong solar radiation can reach the shallow bottom layer of HK waters



295 (although it might still be limiting) and support some growth of phytoplankton at depth that can  
296 release and partially replenish DO (Ho et al. 2008). In summary, hypoxia might therefore  
297 develop only when bottom DO consumption exceeds the buffering capacity maintained by all  
298 these physical and biochemical factors above.

299 The frequency of wind events (>6 m/s) appears to show a decreasing trend in summer in  
300 the long term, which may be well related to global climate change. Climate change is likely  
301 contributing to the increase in dead zones, by influencing factors such as winds, precipitation and  
302 temperature (Altieri and Gedan 2015). For example, changes in the direction and strength of  
303 seasonal wind patterns can modify hypoxic conditions by affecting circulation patterns that  
304 determine nutrient delivery and water column stratification (Conley et al. 2007; Altieri and  
305 Gedan 2015). Changes in rainfall patterns can increase discharges of freshwater and nutrients to  
306 coastal ecosystems (Diaz and Rosenberg 2008). Recently, global warming is predicted to  
307 enhance stratification, decrease oxygen solubility and accelerate respiration, thus exacerbating  
308 the oxygen depletion in nutrient-enriched coastal systems (Breitburg et al. 2018). If the weak  
309 wind condition or the tendency of decreasing wind speeds continues in the future, the occurrence  
310 of hypoxia in this system may become more frequent, and likely develops into large areas of  
311 seasonal hypoxia. This may contribute to a relatively large hypoxic zone in the south water of  
312 Macau reported recently (Su et al. 2017; Lu et al. 2018; Qian et al. 2018). This raises an alarming  
313 signal and an urgent need to fully understand the influence of climate change and how multiple  
314 factors interact to drive the dead zone dynamics.

## 315 **5. Conclusions**

316 Due to population growth and economic development in last 60 years, riverine nutrients  
317 have increased dramatically, which leads to increased organic matter production in estuarine and



318 coastal waters. However, not all estuaries or coastal waters show eutrophication symptoms such  
319 as red tides or hypoxia (Cloern 2001). Nutrients in the Pearl River have been steadily increasing  
320 in the last 4 decades, but hypoxic water mass has not developed into a seasonal phenomenon in a  
321 large scale over the plume influenced waters in the Northern South China Sea. Our study  
322 testified the hypothesis that frequent strong wind events destroy the water column stratification  
323 and interrupt the formation of hypoxia. The wind speed  $>6$  m/s can be considered to be the  
324 threshold of an interruptive wind event in Hong Kong waters. Our finding demonstrates the role  
325 winds play in the ecosystem buffering capacity against enrichment of nutrients. The finding is  
326 significant because climate change may have resulted in the decreasing trend in the frequency of  
327 wind speeds  $>6$  m/s in the recent years, which is an alarming signal for more occurrences of  
328 hypoxic events in the region. The water quality management needs to keep long-term monitoring  
329 and develop strategies for controlling and regulating the input of nutrients in coastal waters to the  
330 level that is below the threshold for triggering the hypoxia in the downstream of the estuary.



331 **References**

- 332 Altieri, A. H., and Gedan K. B.: Climate change and dead zones. *GCB Bioenergy*. 21(4): 1395-  
333 1406. <https://doi.org/10.1111/gcb.12754>, 2015.
- 334 Bianchi, T. S., Dimarco S. F., Cowan J. H., Hetland R. D., Chapman P., Day J. W., and Allison  
335 M. A.: The science of hypoxia in the northern gulf of Mexico: a review. *Sci. Total Environ.*  
336 408(7): 1471-1484. <https://doi.org/10.1016/j.scitotenv.2009.11.047>, 2010.
- 337 Breitburg, D., Levin L. A., Oschlies, A., Gregoire, M., Chavez, F. P., Conley, D. J., Garcon, V.,  
338 et al.: Declining oxygen in the global ocean and coastal waters. *Science*. 359(6371): 46.  
339 <https://doi.org/10.1126/science.aam7240>, 2018.
- 340 Capet, A., Beckers J. M., and Grégoire M.: Drivers, mechanisms and long-term variability of  
341 seasonal hypoxia on the Black Sea Northwestern Shelf & Ndash; is there any recovery after  
342 eutrophication?. *Biogeosciences*. 10(6): 3943-3962. [https://doi.org/10.5194/bg-10-3943-](https://doi.org/10.5194/bg-10-3943-2013)  
343 2013, 2013.
- 344 Chen, X., Shen Z., Li Y., and Yang Y.: Physical controls of hypoxia in waters adjacent to the  
345 Yangtze estuary: a numerical modeling study. *Mar. Pollut. Bull.* 97(1-2): 349-364.  
346 <https://doi.org/10.1016/j.marpolbul.2015.05.067>, 2015.
- 347 Cloern, J. E.: Our evolving conceptual model of the coastal eutrophication problem. *Mar. Ecol.*  
348 *Prog. Ser.* 210: 223-253. <https://doi.org/10.3354/meps210223>, 2001.
- 349 Conley, D. J., Humborg C., Rahm L., Savchuk O. P., and Wulff F.: Hypoxia in the baltic sea and  
350 basin-scale changes in phosphorus biogeochemistry. *Environ. Sci. Technol.* 36(24): 5315-  
351 5320. <https://doi.org/10.1021/es025763w>, 2002.



- 352 Conley, D. J., Carstensen J., Aertebjerg G., Christensen P. B., Dalsgaard T., Hansen J. L. S., and  
353 Josefson A. B.: Long-term changes and impacts of hypoxia in Danish coastal waters.  
354 Ecological Applications, 17(sp5): S165-S184. <https://doi.org/10.1890/05-0766.1>, 2007.
- 355 Diaz, R. J.: Overview of hypoxia around the world. J. Environ. Qual. 30(2): 275-281.  
356 <https://doi.org/10.2134/jeq2001.302275x>, 2001.
- 357 Diaz, R. J., and Rosenberg R.: Spreading dead zones and consequences for marine ecosystems.  
358 Science, 321(5891): 926-929. <https://doi.org/10.1126/science.1156401>, 2008.
- 359 Du, J., Shen J., Park K., Wang Y. P., and Yu X.: Worsened physical condition due to climate  
360 change contributes to the increasing hypoxia in Chesapeake Bay. Sci. Total Environ. 630:  
361 707-717. <https://doi.org/10.1016/j.scitotenv.2018.02.265>, 2018.
- 362 Harrison, P. J., Yin K., Lee J. H. W., Gan J., and Liu H.: Physical-biological coupling in the  
363 Pearl River Estuary. Cont. Shelf Res. 28(12): 1405-1415.  
364 <https://doi.org/10.1016/j.csr.2007.02.011>, 2008.
- 365 Ho, A. Y. T., Xu J., Yin K., Yuan X., He L., Jiang Y., Lee J. H. W., Anderson D. M., Harrison P.  
366 J.: Seasonal and spatial dynamics of nutrients and phytoplankton biomass in Victoria  
367 Harbour and its vicinity before and after sewage abatement. Mar. Pollut. Bull. 57(6-12):  
368 313-324. <https://doi.org/10.1016/j.marpolbul.2008.04.035>, 2008.
- 369 Hu, J., and Li S.: Modeling the mass fluxes and transformations of nutrients in the Pearl River  
370 Delta, China. Journal of Marine Systems, 78(1): 146-167.  
371 <https://doi.org/10.1016/j.jmarsys.2009.05.001>, 2009.
- 372 Justić, D., Rabalais N. N., and Turner R. E.: Effects of climate change on hypoxia in coastal  
373 waters: a doubled CO<sub>2</sub> scenario for the Northern Gulf of Mexico. Limnol. Oceanogr. 41(5):  
374 992-1003. <https://doi.org/10.4319/lo.1996.41.5.0992>, 1996.



- 375 Lee, J. H. W., Harrison P. J., Kuang C., and Yin K.: Eutrophication dynamics in Hong Kong  
376 coastal waters: physical and biological interactions. The environment in Asian Pacific  
377 Harbors. Springer, Netherlands, 187-206. [https://doi.org/10.1007/1-4020-3655-8\\_13](https://doi.org/10.1007/1-4020-3655-8_13), 2006.
- 378 Lehmann, A., Hinrichsen H. H., Getzlaff K., and Myrberg K.: Quantifying the heterogeneity of  
379 hypoxic and anoxic areas in the Baltic Sea by a simplified coupled hydrodynamic-oxygen  
380 consumption model approach. *Journal of Marine Systems*. 134(6): 20-28. <https://doi.org/10.1016/j.jmarsys.2014.02.012>, 2014.
- 382 Li, D., Zhang J., Huang D., Wu Y., and Liang J.: Oxygen depletion off the Changjiang (Yangtze  
383 River) Estuary. *Science in China*, 45(12): 1137-1146. <https://doi.org/10.3969/j.issn.1674-7313.2002.12.008>, 2002.
- 385 Li, M., Zhong L., Boicourt W. C., Zhang S., and Zhang D. L.: Hurricane-induced destratification  
386 and restratification in a partially-mixed estuary. *J. Mar. Res.* 65(65): 169-192(24).  
387 <https://doi.org/10.1357/002224007780882550>, 2007.
- 388 Li, X., Lu C., Zhang Y., Zhao H., Wang J., Liu H., Yin K.: Low dissolved oxygen in the Pearl  
389 River estuary in summer: Long-term spatio-temporal patterns, trends, and regulating factors.  
390 *Mar. Pollut. Bull.* <https://doi.org/10.1016/j.marpolbul.2019.110814>, 2019.
- 391 Lu, Z., Gan J., Dai M., Liu H., and Zhao X.: Joint effects of extrinsic biophysical fluxes and  
392 intrinsic hydrodynamics on the formation of hypoxia west off the Pearl River Estuary. *J.*  
393 *Geophys. Res.: Oceans*. 123(9). <https://doi.org/10.1029/2018JC014199>, 2018.
- 394 Monitoring Group, Water Policy and Planning Group, Environmental Protection Department,  
395 Hong Kong Special Administrative Region. *Marine Water Quality in Hong Kong in 2017*,  
396 2017.



- 397 Ni, X., Huang D., Zeng D., Zhang T., Li H., and Chen J.: The impact of wind mixing on the  
398 variation of bottom dissolved oxygen off the Changjiang Estuary during summer. *Journal of*  
399 *Marine Systems*. 154: 122-130. <https://doi.org/10.1016/j.jmarsys.2014.11.010>, 2016.
- 400 Paerl, H. W.: Assessing and managing nutrient-enhanced eutrophication in estuarine and coastal  
401 waters: interactive effects of human and climatic perturbations. *Ecol. Eng.* 26(1): 40-54.  
402 <https://doi.org/10.1016/j.ecoleng.2005.09.006>, 2006.
- 403 Paerl, H. W., Pinckney J. L., Fear J. M., and Peierls B. L.: Ecosystem responses to internal and  
404 watershed organic matter loading: consequences for hypoxia in the eutrophying Neuse  
405 River Estuary, North Carolina, USA. *Mar. Ecol. Prog. Ser.* 166(8): 17-25.  
406 <https://doi.org/10.3354/meps166017>, 1998.
- 407 Paerl H. W., Bales J. D., Ausley L. W.: Ecosystem impacts of three sequential hurricanes  
408 (Dennis, Floyd, and Irene) on the United States' largest lagoonal estuary, Pamlico Sound,  
409 NC. *Proceedings of the National Academy of Sciences of the United States of America*,  
410 98(10): 5655-5660, 2001.
- 411 Pan, J., and Gu Y.: Cruise observation and numerical modeling of turbulent mixing in the Pearl  
412 River Estuary in summer. *Cont. Shelf Res.* 120: 122-138.  
413 <https://doi.org/10.1016/j.csr.2016.03.019>, 2016.
- 414 Qian, W., and Gan J., Liu J., He B., and Dai M.: Current status of emerging hypoxia in a  
415 eutrophic estuary: The lower reach of the Pearl River Estuary, China. *Estuarine Coastal*  
416 *Shelf Sci.* 205: 58-67. <https://doi.org/10.1016/j.ecss.2018.03.004>, 2018.
- 417 Rabalais, N. N., Turner R. E., Justić D.: Charaterization of hypoxia: Topic 1 Report for the  
418 Integrated Assessment of Hypoxia in the Gulf of Mexico. NOAA Coastal Ocean Program.  
419 Decision Analysis Series No.15, 167pp, 1998.



- 420 Rabalais, N. N., Turner R. E., and Wiseman W. J.: Gulf of Mexico hypoxia, a.k.a. \"the dead  
421 zone\". *Annu. Rev. Ecol. Syst.* 33: 235-263. <https://doi.org/10.2307/3069262>, 2002.
- 422 Rabalais, N. N., Díaz R. J., Levin L. A., Turner R. E., and Zhang J.: Dynamics and distribution  
423 of natural and human-caused hypoxia. *Biogeosciences*. 7(2): 585-619.  
424 <https://doi.org/10.5194/bgd-6-9359-2009>, 2010.
- 425 Rabouille, C., Conley D. J., Dai M., Cai W., and Mckee B.: Comparison of hypoxia among four  
426 river-dominated ocean margins: the Changjiang (Yangtze), Mississippi, Pearl, and Rhône  
427 rivers. *Cont. Shelf Res.* 28(12): 1527-1537. <https://doi.org/10.1016/j.csr.2008.01.020>, 2008.
- 428 Scully, M. E.: Wind modulation of dissolved oxygen in Chesapeake Bay. *Estuaries Coasts*. 33(5):  
429 1164-1175, <https://doi.org/10.1007/s12237-010-9319-9>, 2010.
- 430 Scully, M. E.: Physical controls on hypoxia in Chesapeake Bay: a numerical modeling study. *J.*  
431 *Geophys. Res.: Oceans*. 118(3): 1239-1256. <https://doi.org/10.1002/jgrc.20138>, 2013.
- 432 Simpson, J. H., and Bowers D. B.: Models of stratification and frontal movement in shelf seas.  
433 *Deep-Sea Res., Part A*. 28(7): 727-738. [https://doi.org/10.1016/0198-0149\(81\)90132-1](https://doi.org/10.1016/0198-0149(81)90132-1),  
434 1981.
- 435 Simpson, J. H., Brown J., Matthews J., and Allen G.: Tidal straining, density currents, and  
436 stirring in the control of estuarine stratification. *Estuaries*. 13(2): 125-132. <https://doi.org/10.2307/1351581>, 1990.
- 437
- 438 Simpson, J. H., Williams E., Brasseur L. H., and Brubaker J. M.: The impact of tidal straining on  
439 the cycle of turbulence in a partially stratified estuary. *Cont. Shelf Res.* 25(1): 51-64.  
440 <https://doi.org/10.1016/j.csr.2004.08.003>, 2005.
- 441 Su, J., Dai M., He B., Wang L., Gan J., Guo X., Zhao H., and Yu F.: Tracing the origin of the  
442 oxygen-consuming organic matter in the hypoxic zone in a large eutrophic estuary: the



- 443 lower reach of the Pearl River Estuary, China. *Biogeosciences Discussions*, 14(18): 1-24.  
444 <https://doi.org/10.5194/bg-2017-43>, 2017.
- 445 Väli, G., Meier H.E.M., and Elken J.: Simulated halocline variability in the Baltic Sea and its  
446 impact on hypoxia during 1961-2007. *J. Geophys. Res.: Oceans*. 118(12): 6982-7000.  
447 <https://doi.org/10.1002/2013JC009192>, 2013.
- 448 Wang, H., Dai M., Liu J., Kao S., Zhang C., Cai W., Wang G., et al.: Eutrophication-driven  
449 hypoxia in the East China Sea off the Changjiang Estuary. *Environ. Sci. Technol.* 50: 2255-  
450 2263. <https://doi.org/10.1021/acs.est.5b06211>, 2016.
- 451 Wang, J. F., Macdonald D. G., Orton P. M., Cole K., and Lan J.: The effect of discharge, tides,  
452 and wind on lift-off turbulence. *Estuaries Coasts*, 38(6): 2117-2131.  
453 <https://doi.org/10.1007/s12237-015-9958-y>, 2015.
- 454 Wang, B., Wei Q., Chen J., and Xie L.: Annual cycle of hypoxia off the changjiang (yangtze  
455 river) estuary. *Mar. Environ. Res.* 77: 1-5. <https://doi.org/10.1016/j.marenvres.2011.12.007>,  
456 2012.
- 457 Wei, X., Zhan H., Ni P., and Cai S.: A model study of the effects of river discharges and winds  
458 on hypoxia in summer in the Pearl River Estuary. *Mar. Pollut. Bull.* 113(1-2): 414-427.  
459 <https://doi.org/10.1016/j.marpolbul.2016.10.042>, 2016.
- 460 Wilson, R. E., Swanson R. L., and Crowley H. A.: Perspectives on long-term variations in  
461 hypoxic conditions in western long island sound. *J. Geophys. Res.* 113(C12): C12011.  
462 <https://doi.org/10.1029/2007jc004693>, 2008.
- 463 Xu, J., Yin K., Liu H., Lee J. H. W., Anderson D. M., Ho A. Y. T., and Harrison P. J.: A  
464 comparison of eutrophication impacts in two harbours in Hong Kong with different



- 465 hydrodynamics. *Journal of Marine Systems*, 83(3-4): 276-286. [https://doi.org/](https://doi.org/10.1016/j.jmarsys.2010.04.002)  
466 10.1016/j.jmarsys.2010.04.002, 2010.
- 467 Ye, F., Huang X., Shi Z., and Liu Q.: Distribution Characteristics of Dissolved Oxygen and Its  
468 Affecting Factors in the Pearl River Estuary During the Summer of the Extremely Drought  
469 Hydrological Year 2011. *China Environ. Sci.* 34(5): 1707-1714, 2013.
- 470 Yin, K., and Harrison P. J.: Influence of the Pearl River estuary and vertical mixing in Victoria  
471 Harbor on water quality in relation to eutrophication impacts in Hong Kong waters. *Mar.*  
472 *Pollut. Bull.* 54(6): 646-656. <https://doi.org/10.1016/j.marpolbul.2007.03.001>, 2007.
- 473 Yin, K., and Harrison P. J.: Nitrogen over enrichment in subtropical Pearl River estuarine coastal  
474 waters: possible causes and consequences. *Cont. Shelf Res.* 28(12): 1435-1442.  
475 <https://doi.org/10.1016/j.csr.2007.07.010>, 2008.
- 476 Yin, K., Qian P., Chen J. C., Hsieh P. H. D., and Harrison P. J.: Dynamics of nutrients and  
477 phytoplankton biomass in the Pearl River estuary and adjacent waters of Hong Kong during  
478 summer: preliminary evidence for phosphorus and silicon limitation. *Mar. Ecol. Prog.*  
479 194(3): 295-305. <https://doi.org/10.3354/meps194295>, 2000.
- 480 Yin, K., Lin Z., and Ke Z.: Temporal and spatial distribution of dissolved oxygen in the Pearl  
481 River estuary and adjacent coastal waters. *Cont. Shelf Res.* 24(16): 1935-1948.  
482 <https://doi.org/10.1016/j.csr.2004.06.017>, 2004.
- 483 Yin, K., Xu J., and Harrison P.J.: A Comparison of eutrophication processes in three Chinese  
484 subtropical semi-enclosed embayments with different buffering capacities. *Coastal Lagoons:*  
485 *Critical Habitats of Environmental Change*, 372-398. [https://doi.org/](https://doi.org/10.1201/EBK1420088304-c15)  
486 10.1201/EBK1420088304-c15, 2010.



487 Zhou, W., Yin K., Harrison P. J., and Lee J. H. W.: The influence of late summer typhoons and  
488 high river discharge on water quality in Hong Kong waters. *Estuarine Coastal Shelf Sci.*  
489 111(4): 35-47. <https://doi.org/10.1016/j.ecss.2012.06.004>, 2012.

490 Zhou, W., Yuan X., Long A., Huang H., and Yue W.: Different hydrodynamic processes  
491 regulated on water quality (nutrients, dissolved oxygen, and phytoplankton biomass) in  
492 three contrasting waters of Hong Kong. *Environ. Monit. Assess.* 186(3): 1705-1718.  
493 <https://doi.org/10.1007/s10661-013-3487-6>, 2014.

494 Zhu, Z., Zhang J., Wu Y., Zhang Y., Lin J., and Liu S. M.: Hypoxia off the Changjiang (Yangtze  
495 River) Estuary: oxygen depletion and organic matter decomposition. *Mar. Chem.* 125(1-4):  
496 108-116. <https://doi.org/10.1016/j.marchem.2011.03.005>, 2011.

497 Zhu, J., Zhu Z., Lin J., Wu H., and Zhang J.: Distribution of hypoxia and pycnocline off the  
498 Changjiang Estuary, China. *Journal of Marine Systems*, 154(Part A): 28-40. <https://doi.org/10.1016/j.jmarsys.2015.05.002>, 2016.

500

#### 501 **Acknowledgments**

502 This study is part of NSFC-GD Joint Scheme Project (U1701247), SML99147-42080013,  
503 NSFC grant (91328203) and NMEMC Key Laboratory for Ecological Environment in Coastal  
504 Area (Grant 201819). We acknowledge the Hong Kong EPD and HKO for permitting us to use  
505 their water quality monitoring data and weather data for this study. H. Liu acknowledge the  
506 support of Hong Kong Research Grants Council (T21/602/16 and N\_HKUST609/15).

#### 507 **Data availability**

508 The wind speeds data used in this manuscript is open to the public and can be  
509 downloaded from HKO website. The time series of water quality monitoring data is provided by  
510 Hong Kong EPD and will be available after this manuscript is published.



511 **Author contribution**

512 The contributions made by each of the authors are as follows. Juan Yao analyses the long  
513 time series data and writes the original manuscript. Juying Wang and Hongbin Liu review the  
514 manuscript and give valuable and helpful comments. Kedong Yin provides guidance on the  
515 conceptualization of the scientific story and makes revision of the manuscript.

516 **Competing interests**

517 The authors declare no conflict of interest.

518 **Figure legends**

519 **Figure 1.** Map of the Pearl River estuary and Hong Kong waters and Waglan Island showing the  
520 selected EPD water quality monitoring stations.

521 **Figure 2.** The time series of surface and bottom DO during 1990-2018 at SM17, SM18 and  
522 SM19 (the red dashed line indicates that the linear regression is not significant).

523 **Figure 3(a).** Time series of wind speed and bottom DO during 1990-2004 in summer at SM17,  
524 SM18 and SM19 (the red dashed line denotes the level of DO=2 mg/L).

525 **Figure 3(b).** Time series of wind speed and bottom DO during 2005-2018 in summer at SM17,  
526 SM18 and SM19 (the red dashed line denotes the level of DO=2 mg/L).

527 **Figure 4.** Time series of averaged wind speed  $V_7$  before sampling during 1990-2018 in summer  
528 at SM17, SM18 and SM19. There is no significant trend by linear regression.

529 **Figure 5.** The relationship between bottom DO,  $\Delta\sigma$  and  $V_7$  during 1990-2018 at SM17, SM18  
530 and SM19.

531 **Figure 6.** The frequency of grouped wind speeds during 1990-2018 in summer (left panel) and  
532 accumulative frequency of grouped wind speeds (accumulated from the largest wind speed group  
533 to the smallest one) (right panel).



534 **Figure 7.** The monthly averages of wind speeds for June, July, August and September,  
535 respectively, over 29 years (the red and blue solid lines denote the significant regression).

536 **Figure 8.** The averaged monthly frequency of wind speeds  $<6$  and  $\geq 6$  m/s at Waglan Island in  
537 summer (the red dashed line denotes the frequency of wind speeds  $\geq 6$  m/s over 29 years).



538 Table 1. The Correlation Coefficient,  $r$ , between bottom DO, AOU,  $\Delta$ DO and  $\Delta\sigma$ .

| Variables                      | SM17 (n=287) | SM18 (n=292) | SM19 (n=292) |
|--------------------------------|--------------|--------------|--------------|
| DO vs. $\Delta\sigma$          | -0.71 **     | -0.69 **     | -0.68 **     |
| AOU vs. $\Delta\sigma$         | 0.69 **      | 0.67 **      | 0.67 **      |
| $\Delta$ DO vs. $\Delta\sigma$ | 0.75 **      | 0.77 **      | 0.80 **      |

539 *Note.* n is the total number of samples, and the asterisk \*\* indicates the significant level of  $p < 0.01$ .



540 Table 2. The Correlation Coefficient,  $r$ , between bottom DO, AOU,  $\Delta$ DO,  $\Delta\sigma$  and wind speed in summer.

|                | SM17 (n=94)   |                |                 |                | SM18 (n=97)    |                 |                 |                | SM19 (n=97)    |                 |                 |                |
|----------------|---------------|----------------|-----------------|----------------|----------------|-----------------|-----------------|----------------|----------------|-----------------|-----------------|----------------|
|                | DO            | AOU            | $\Delta$ DO     | $\Delta\sigma$ | DO             | AOU             | $\Delta$ DO     | $\Delta\sigma$ | DO             | AOU             | $\Delta$ DO     | $\Delta\sigma$ |
| V <sub>7</sub> | <u>0.23</u> * | -0.23*         | -0.27**         | 0.07           | <u>0.47</u> ** | <u>-0.49</u> ** | -0.36**         | -0.14          | 0.46**         | -0.48**         | -0.36**         | -0.19          |
| V <sub>6</sub> | <u>0.23</u> * | <u>-0.24</u> * | -0.33**         | 0.001          | <u>0.47</u> ** | <u>-0.49</u> ** | -0.40**         | -0.17          | 0.47**         | -0.49**         | -0.40**         | -0.23*         |
| V <sub>5</sub> | <u>0.23</u> * | -0.23*         | <u>-0.36</u> ** | -0.08          | 0.46**         | -0.48**         | <u>-0.43</u> ** | -0.21*         | <u>0.48</u> ** | <u>-0.50</u> ** | -0.47**         | -0.26*         |
| V <sub>4</sub> | 0.20          | -0.21*         | -0.33**         | -0.15          | 0.43**         | -0.46**         | <u>-0.43</u> ** | -0.21*         | <u>0.48</u> ** | <u>-0.50</u> ** | <u>-0.49</u> ** | -0.23*         |
| V <sub>3</sub> | 0.18          | -0.19          | -0.27**         | -0.19          | 0.41**         | -0.43**         | -0.40**         | -0.21*         | 0.41**         | -0.44**         | -0.44**         | -0.22*         |
| V <sub>2</sub> | 0.13          | -0.14          | -0.20           | -0.17          | 0.32**         | -0.34**         | -0.32**         | -0.18          | 0.28**         | -0.31**         | -0.35**         | -0.17          |
| V <sub>1</sub> | 0.09          | -0.11          | -0.15           | -0.16          | 0.16           | -0.18           | -0.16           | -0.09          | 0.09           | -0.12           | -0.21*          | -0.08          |

541 *Note.* The V<sub>i</sub> means i days averaged wind speed before sampling, and n is the total number of samples, the asterisk \* or \*\* indicates the significant level  
 542 of  $p < 0.05$  or  $p < 0.01$ , the underline labels the maximum of each column.



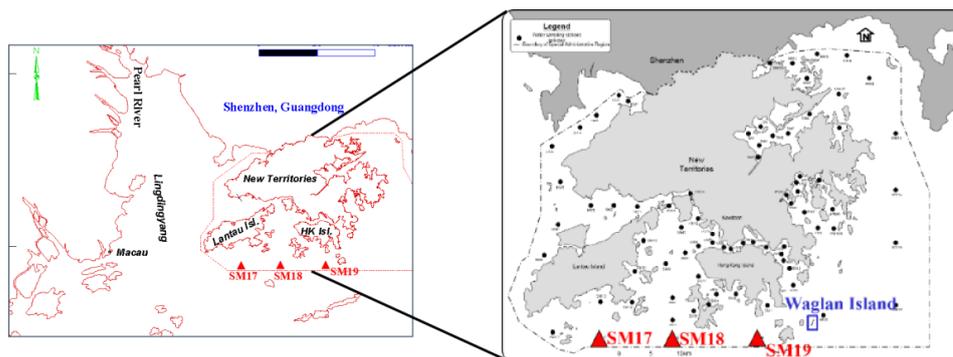
543 Table 3(a). The Frequency (%) of bottom DO at different  $V_7$  Wind Speeds during 1990-2018 in summer.

|         | (mg/L)          | $\geq 5$ m/s | $\geq 6$ m/s | $\geq 7$ m/s | $\geq 8$ m/s |
|---------|-----------------|--------------|--------------|--------------|--------------|
| SM17    | $3 < DO \leq 4$ | 16.0         | 7.5          | 4.3          | 1.1          |
|         | $2 < DO \leq 3$ | 8.5          | 5.3          | 1.1          | 0.0          |
|         | $DO \leq 2$     | 2.1          | 0.0          | 0.0          | 0.0          |
| SM18    | $3 < DO \leq 4$ | 13.4         | 8.3          | 2.1          | 1.0          |
|         | $2 < DO \leq 3$ | 10.3         | 5.2          | 3.1          | 0.0          |
|         | $DO \leq 2$     | 2.1          | 1.0          | 0.0          | 0.0          |
| SM19    | $3 < DO \leq 4$ | 21.7         | 11.3         | 3.1          | 0.0          |
|         | $2 < DO \leq 3$ | 4.1          | 2.1          | 0.0          | 0.0          |
|         | $DO \leq 2$     | 1.0          | 1.0          | 1.0          | 0.0          |
| Average | $DO \leq 3$     | 9.4          | 4.9          | 1.7          | 0.0          |



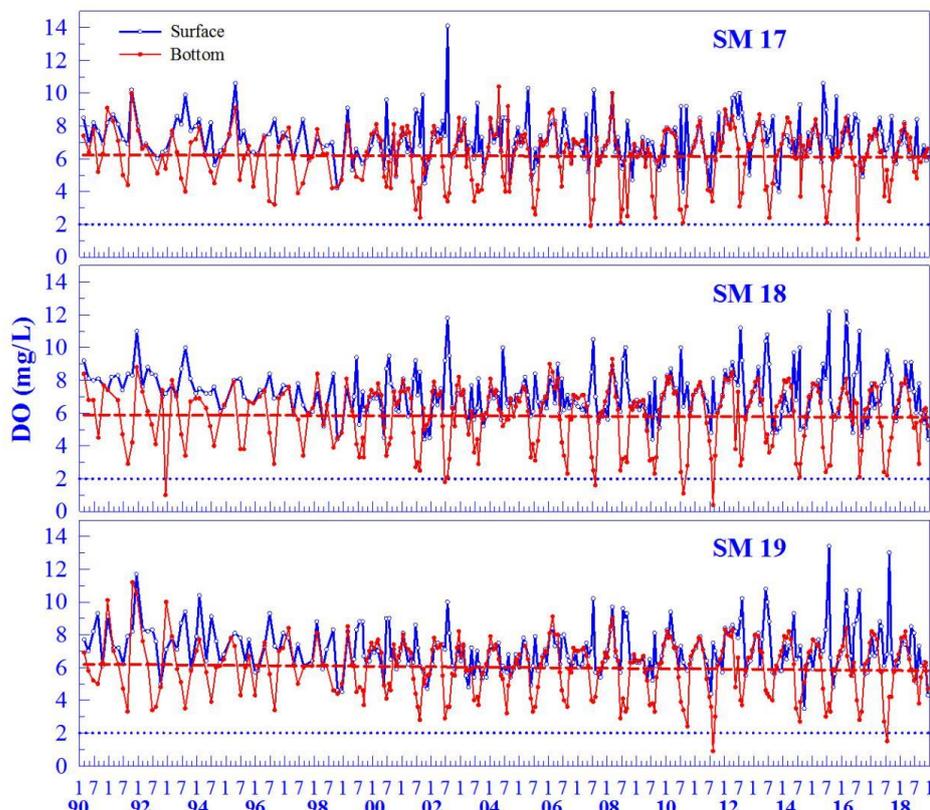
544 Table 3(b). The Accumulative Frequency (%) of  $\Delta\sigma$  in the 4 descending groups vs 4 ascending groupings of  
 545  $V_7$  Wind Speeds during 1990-2018 in summer (June-August). Group  $\geq 5$  m/s includes the other 3 groups,  
 546 group  $\geq 6$  m/s includes the other 2 groups and so on.

|      | ( $\text{kg/m}^3$ )         | $\geq 5$ m/s | $\geq 6$ m/s | $\geq 7$ m/s | $\geq 8$ m/s |
|------|-----------------------------|--------------|--------------|--------------|--------------|
|      | $\Delta\sigma > 15$         | 3.2          | 3.2          | 0.0          | 0.0          |
| SM17 | $10 < \Delta\sigma \leq 15$ | 14.9         | 10.6         | 6.4          | 1.1          |
|      | $5 < \Delta\sigma \leq 10$  | 29.8         | 18.1         | 7.4          | 5.3          |
|      | $\Delta\sigma \leq 5$       | 25.5         | 16.0         | 9.6          | 4.3          |
|      | $\Delta\sigma > 15$         | 4.1          | 4.1          | 2.1          | 1.0          |
| SM18 | $10 < \Delta\sigma \leq 15$ | 13.4         | 9.3          | 4.1          | 1.0          |
|      | $5 < \Delta\sigma \leq 10$  | 17.5         | 10.3         | 5.2          | 2.1          |
|      | $\Delta\sigma \leq 5$       | 24.7         | 16.5         | 13.4         | 5.2          |
|      | $\Delta\sigma > 15$         | 1.0          | 0.0          | 0.0          | 0.0          |
| SM19 | $10 < \Delta\sigma \leq 15$ | 12.4         | 11.3         | 3.1          | 1.0          |
|      | $5 < \Delta\sigma \leq 10$  | 18.6         | 9.3          | 4.1          | 0.0          |
|      | $\Delta\sigma \leq 5$       | 33.0         | 23.7         | 17.5         | 8.2          |

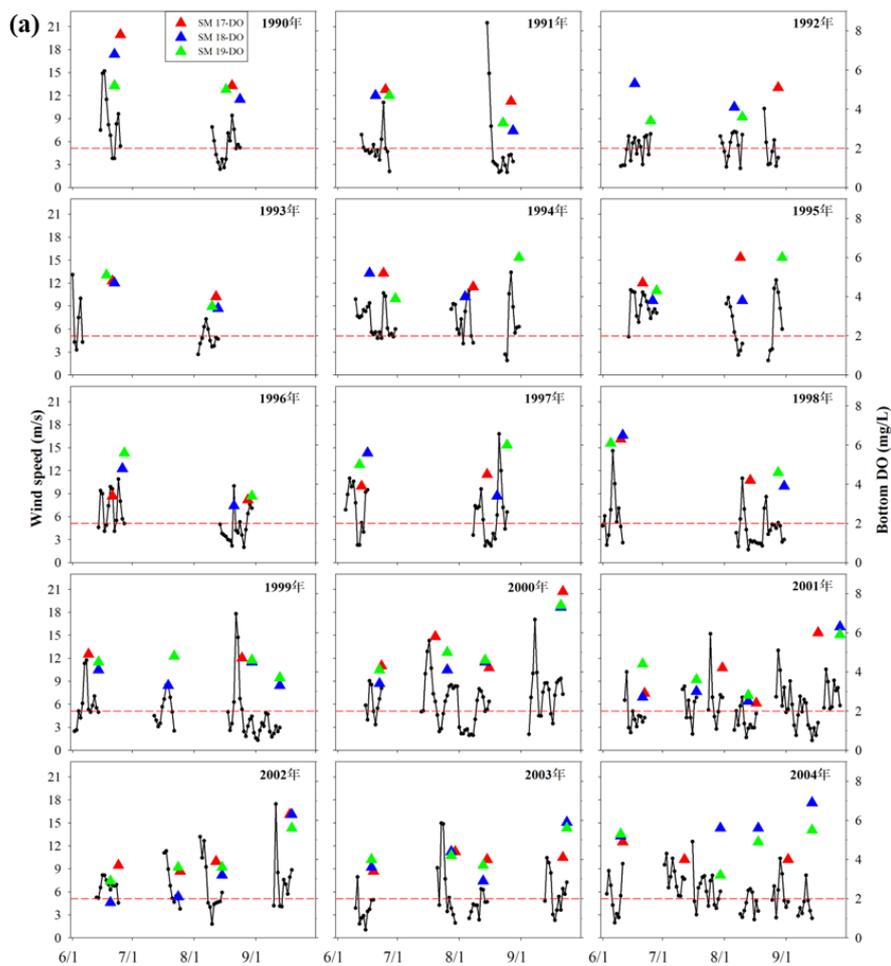


547  
548 Figure 1. Map of the Pearl River estuary and Hong Kong waters and Waglan Island showing the selected EPD  
549 water quality monitoring stations.

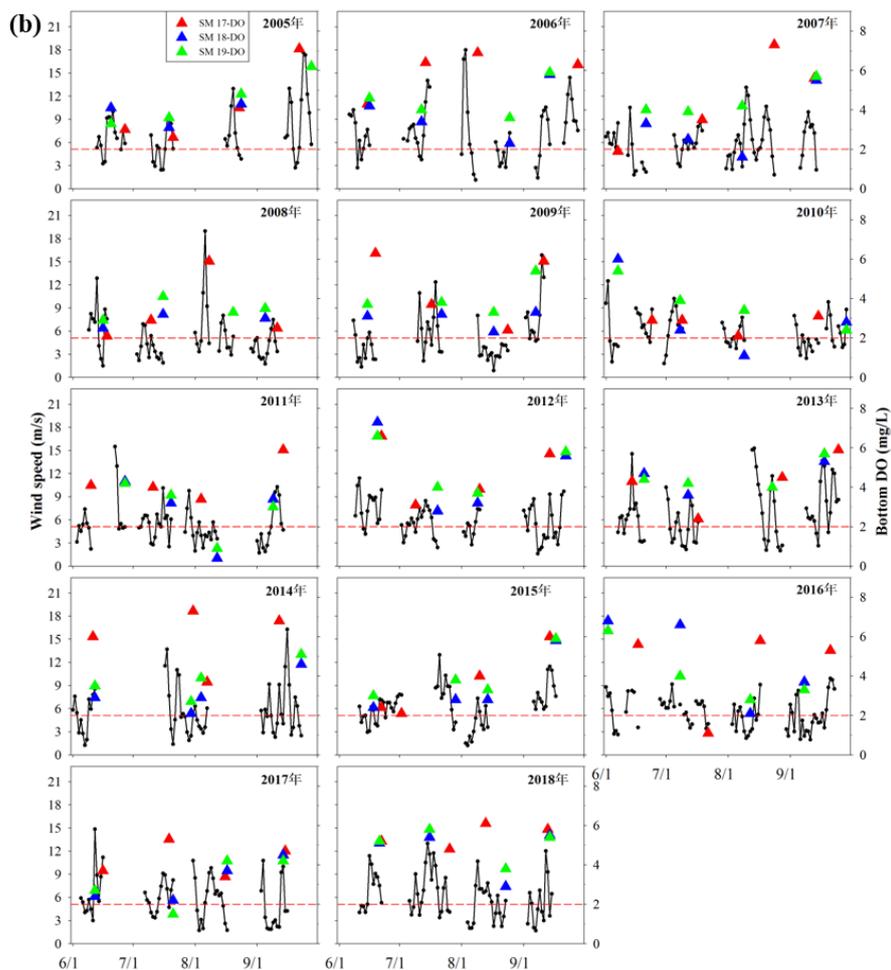
550



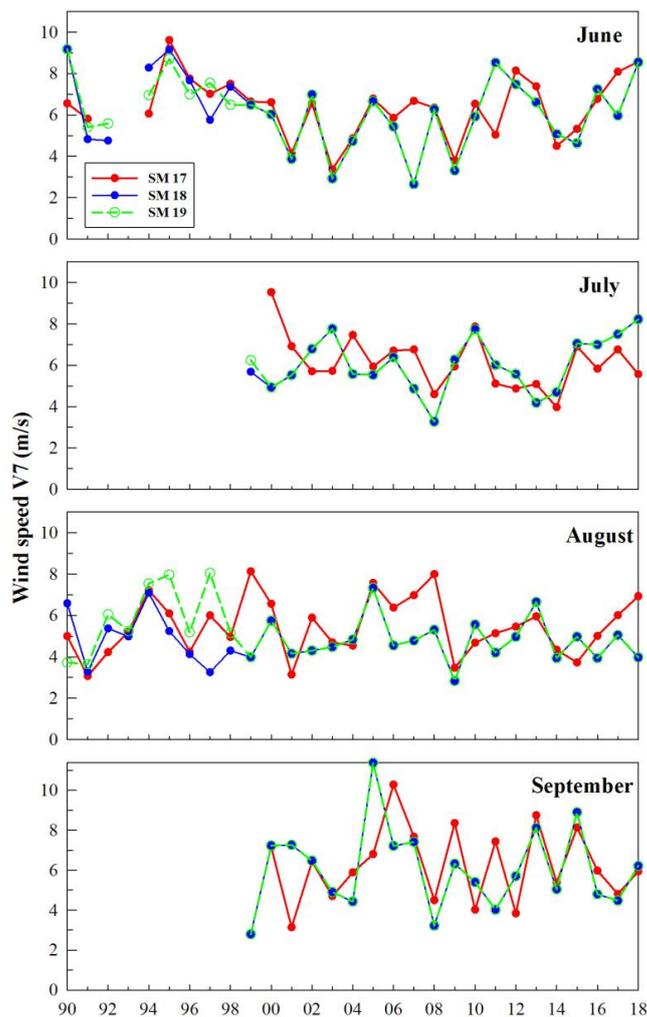
551  
552 Figure 2. The time series of surface and bottom DO during 1990-2018 at SM17, SM18 and SM19 (the red dashed  
553 line indicates that the linear regression is not significant).



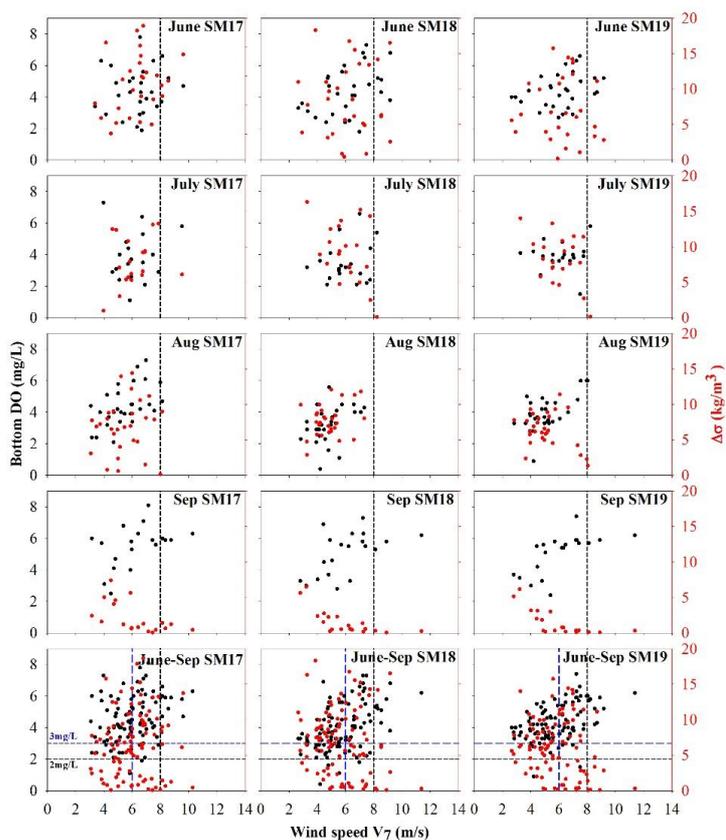
554 Figure 3(a). Time series of wind speed and bottom DO during 1990-2004 in summer at SM17, SM18 and SM19 (the  
555 red dashed line denotes the level of DO=2 mg/L).  
556



557  
558 Figure 3(b). Time series of wind speed and bottom DO during 2005-2018 in summer at SM17, SM18 and SM19 (the  
559 red dashed line denotes the level of DO=2 mg/L).

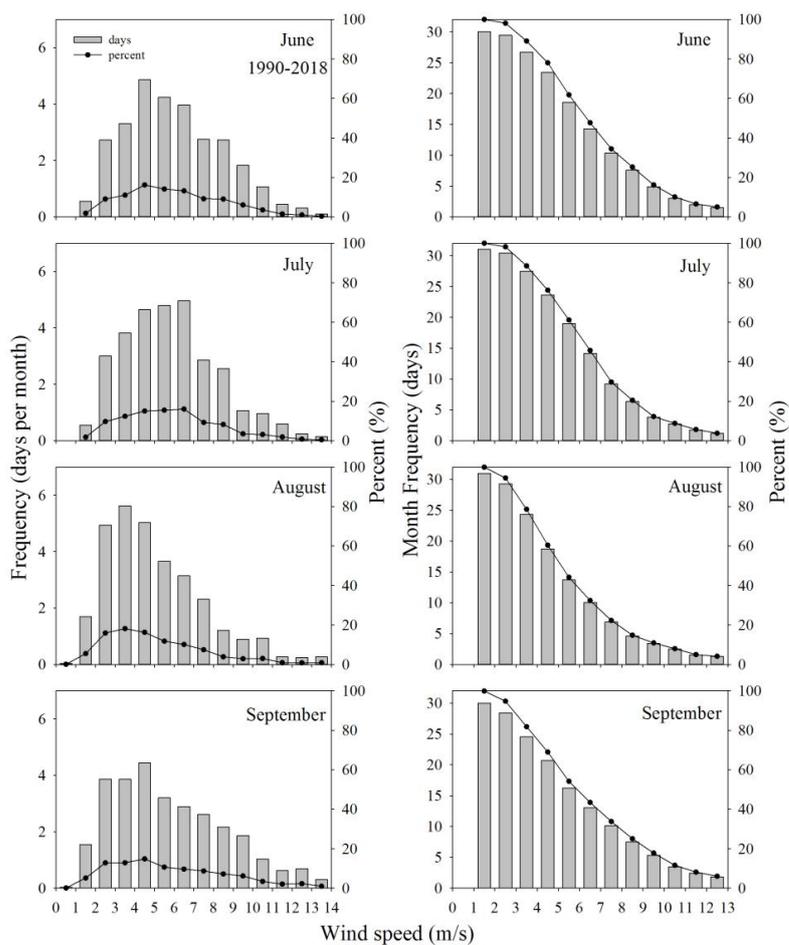


560  
561 Figure 4. Time series of averaged wind speed  $V_7$  before sampling during 1990-2018 in summer at SM17, SM18 and  
562 SM19. There is no significant trend by linear regression.

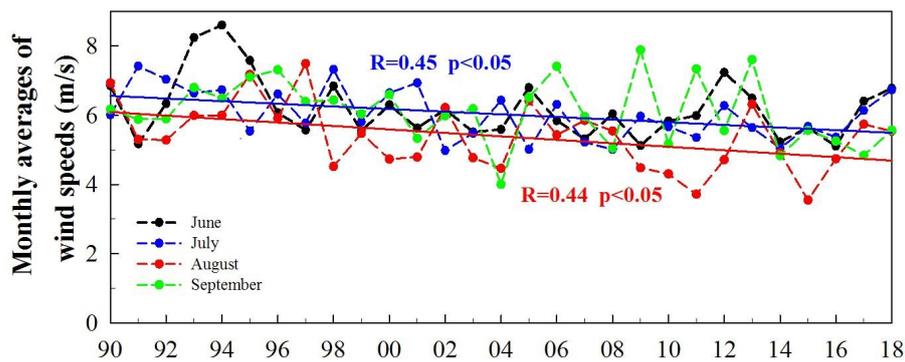


563  
564

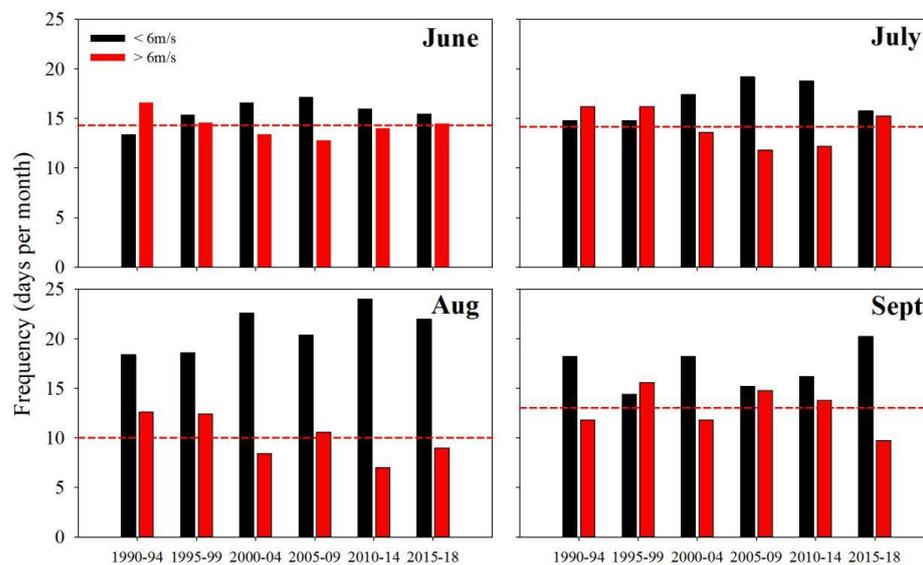
Figure 5. The relationship between bottom DO,  $\Delta\sigma$  and  $V_7$  during 1990-2018 at SM17, SM18 and SM19.



565  
566 Figure 6. The frequency of grouped wind speeds during 1990-2018 in summer (left panel) and accumulative  
567 frequency of grouped wind speeds (accumulated from the largest wind speed group to the smallest one) (right panel)



568  
 569 Figure 7. The monthly averages of wind speeds for June, July, August and September, respectively, over 29 years  
 570 (the red and blue solid lines denote the significant regression).  
 571



572  
 573 Figure 8. The averaged monthly frequency of wind speeds <6 and ≥6 m/s at Waglan Island in summer (the red  
 574 dashed line denotes the frequency of wind speeds ≥6 m/s over 29 years).