

1Natural distinction of carbon and nitrogen isotopic niches in common fish species 2for diverse marine biotopes off the Yellow River estuary and adjacent sea areas

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14Abstract

15Stable isotope analysis is a universally recognized and efficient method of indicating
16trophic relationships that is widely applied in research. However, variations in natural
17isotopic abundance may lead to inaccuracies due to the effects of complex
18environmental conditions. This research compared the carbon and nitrogen isotopic
19niches of fish communities between diverse biotopes around the Yellow River estuary
20and adjacent sea areas, with the aim of revealing distinctions in stable isotopic niche
21metrics, trophic positions, and feeding preferences. Stable isotopic niche results
22indicated that the communities of estuarine habitants were compatible in most study
23biotopes, and may provide a corridor for energy and material transportation between
24Laizhou Bay and the open water. Local biocoenosis was embodied in the wider
25isotopic niche corresponding to frequent environmental changes and abiotic gradients.
26This implied that they used various food sources to adapt to the fickle environment,
27including marine-terrestrial boundaries and the estuary. Our analysis of the food
28source contribution indicated that allochthonous sources were considered major

29energy sources in estuarine areas directly affected by Yellow River-diluted water,
30while autochthonous benthic and pelagic producers dominated carbon input into the
31food web in Laizhou Bay and the open water. A significant variation in the fish $\delta^{15}\text{N}$
32characteristic was found within estuarine adjacent regions, so, together with the
33results from previous studies, we deemed the local high concentration of dissolved
34inorganic nitrogen as the original trigger of the abnormal $\delta^{15}\text{N}$ characteristic in fishes
35via a transport process along food chains. These results provide a new perspective on
36the natural distinction of carbon and nitrogen isotopic niches. The detailed data
37reported here enhance our understanding of variations in fish communities in
38estuarine ecosystems.

39**Key words:** stable isotopic niche; estuarine offshore biotope; food source; trophic
40level, fish community

41**Introduction**

42Estuarine biotopes display distinct trophic structures of biocoenosis driven by the
43supply and transformation of multiple energy sources (Underwood, 2010), while
44flowing waters sustain riverine and marine biodiversity, and make important
45contributions to global biogeochemical cycles (Palmer et al., 2019). It is widely
46believed that most adjacent marine ecosystems are strongly connected due to the
47water transference of organic matter and bioelements (Stasko et al., 2018; Sujitha et
48al., 2019). Adjacent ecosystems also provide ecological corridors for animal migration
49(Hastie et al., 2016). Numerous investigations have shown that transference of energy
50and materials occurs frequently between biotopes influenced by strong coastal
51physical and biological dynamics (Livernois et al., 2019). This implies the potential
52connectivity of trophic niches and biocoenosis structures (Palmer et al., 2019).
53However, there is relatively little research on identifying discrepancies in fish trophic
54niches caused by diverse marine biotopes around the estuary directly. These related
55studies are limited to temporal and spatial heterogeneity in food sources (McMahon et
56al., 2015) and the complexity of marine ecosystems within various biotopes
57(Christianen et al., 2017; Ramshaw et al., 2017). Investigations comparing the trophic
58relationships between diverse biotopes in estuarine ecosystems have seldom been
59performed, and this has hindered our understanding of the energy and material

60transporting mechanisms in the food webs.

61Stable isotopes record information of marine lives accumulating nutrients over
 62integrated time periods in lifecycles as opposed to a snapshot of food ingestion (Plass-
 63Johnson et al., 2013). They can be used to reconstruct the trophic structures of
 64biocoenoses in marine food webs (Parnell et al., 2010; Boecklen et al., 2011; Fry,
 652013). However, as reality is a bit more complicated, mixed models that take isotope
 66values of multiple food sources into account according to user-specified data have
 67been developed and successfully applied to food web studies to solve complex
 68interpretation processes (Jackson et al., 2011; Phillips et al., 2012). Stable isotopic
 69analysis can greatly contribute to research on fish community connectivity in marine
 70ecosystems on account of isotopic signatures corresponding to estuarine biotopes
 71(Selleslagh et al., 2015). As the trophic relationship can be concisely expounded using
 72stable isotope analysis together with advisable models, such as SIBER (Stable Isotope
 73Bayesian Ellipses in R, Jackson et al., 2011) and IsoSource (Layman et al., 2012), a
 74comparison of trophic relationships of biocoenoses between diverse biotopes,
 75including variations in trophic structures, can be further indicated.

76The Yellow River is the second longest river in China and its input routes have
 77changed over time, leading to complicated biotopes in the estuary (Xu et al., 2013).
 78There is a strong interaction between ocean and land as with many large river
 79estuaries, and the research value is significant in terms of the diverse marine food
 80webs within the Yellow River estuary. However, to date, few studies have compared
 81the trophic relationships between diverse biotopes in Yellow River estuarine
 82ecosystems. This shortage of information hinders our understanding of the energy and
 83materials transportation mechanism in local food webs and impedes the restoration
 84and conservation process in estuarine Marine Protected Areas (MPAs).

85This research aimed to compare the isotopic niches of fish communities between
 86diverse biotopes in the Yellow River estuary with the main expectation of revealing
 87distinctions in stable isotopic niche metrics, trophic positions, and feeding
 88preferences. These results provide new perspectives on trophic relationships, and they
 89provide detailed data that can enhance our understanding of the variations in fish
 90communities in estuarine ecosystems, with important implications for fishery
 91conservation and the restoration of estuarine MPAs.

92Method

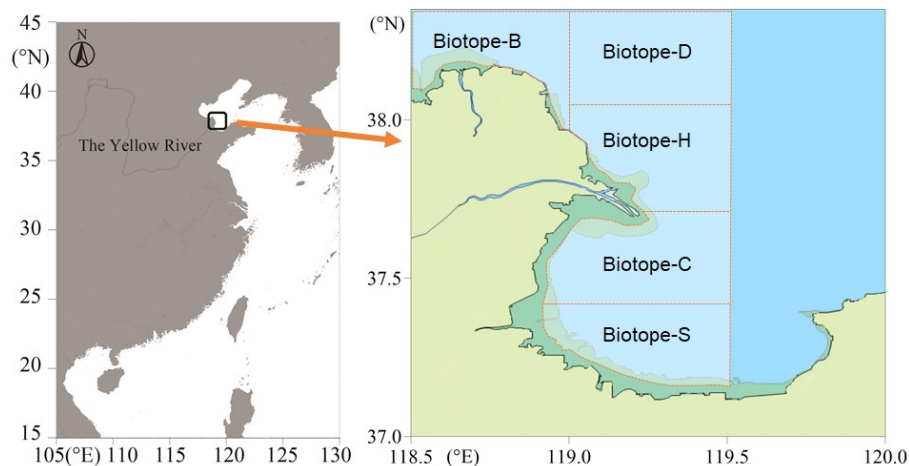
93Research area and sampling methods

94Two sampling cruises were launched in September 2017 and September 2018. Study
 95areas were the coastal sea located from 118.5°E to 119.5°E and 37.0°N to 38.4°N
 96around the Yellow River estuary (Figure 1). The location included five MPAs: Yellow
 97River Estuary MPA, Lijin MPA, Hekou MPA, Laizhou MPA, and Guangrao MPA.
 98MPAs are considered ideal experimental systems in which to obtain environmental
 99background values in areas that restrict human activities. The inherent variety of fish
 100communities in a stable isotopic niche caused by a diverse marine biotope can be
 101compared with different biotopes that exclude direct human exploitation.

102Selection of survey biotopes mainly considered the diversity between typical estuarine
 103and gulf ecosystems, as well as subtidal zones and deeper water. Five biotopes (Figure
 1041) with unique individual characteristics were selected. All five biotopes were
 105separated by the Yellow River estuary and differentiated by intertidal and subtidal
 106zones, and represented environment characteristic data, such as salinity around the
 107Yellow River estuary, which were synthetically used to distinguish the influencing
 108scope of diluted water from the river. Each of these five biotopes had discriminative
 109marine ecosystem characteristics. Biotope-H, located at the intertidal and subtidal
 110zone of the recent Yellow River estuary, was directly affected by the abundant diluted
 111water from the Yellow River, and characterized by typical estuarine features, such as
 112abundance of terrestrial organic matters, lower salinity, and depth. Biotope-C was
 113located at the intertidal and subtidal zone of the southern branch of the Yellow River
 114estuary that was almost blocked in the 1990s. Biotope-D was located at the subtidal
 115zone facing open water and featured deeper water and higher salinity. Biotope-B was
 116located at the intertidal and subtidal zone near the ancient Yellow River estuary facing
 117open water. Biotope-S, which was located at the intertidal and subtidal zone in
 118Laizhou Bay, was closer to the mainland and influenced more by the weaker diluted
 119water from the bottom of Laizhou Bay rather than diluted water from the Yellow
 120River. Five evenly spaced sites were chosen from each biotope for comparison; thus,
 121in total, twenty-five sites were selected for collecting fish specimens for stable
 122isotopic analysis. After the summer fishing moratorium ended in September 2017 and
 123September 2018, a fishing boat trawled at 2 kn for 30 min (Choy et al., 2015) at each
 124site. Specimens were weighed and frozen at -20 °C after species identification. Three

125replicates of stable isotope analysis were carried out on each species. Zooplankton
 126specimens were collected as an assemblage of communities using a 200- μ m
 127zooplankton net that was horizontally trawled at 2 kn for 10 min. Three replicate
 128samples were taken, filtered using a pump system and 200- μ m bolting silk filters,
 129wrapped in foil, placed in sealed bags, and then stored at -20 °C until further analysis.
 130Suspended particulate organic matter (POM), which mainly contained phytoplankton
 131and organic detritus, was collected by filtering seawater through pre-combusted
 132Whatman GF/F glass fiber filters (Kohlbach et al., 2016), and subsequent methods
 133were the same with zooplankton sample collection. Microscopic photosynthetic
 134organisms living on the sediment surface were referred to as microphythobenthos,
 135mainly comprising diatoms and cyanobacteria (Christianen et al., 2017), which
 136compose the sedimental organic matter (SOM) with other organic detritus. SOM
 137specimens were collected using a clam grab bucket. The surface layers (< 5 mm) of
 138the sediment were scraped off, foil-wrapped, and then frozen at -20 °C for further
 139processing. Enteromorpha and spartina were, respectively, collected and then foil-
 140wrapped and frozen as a proxy for macroalgae and cordgrass, because of their
 141dominating position in local biotopes.

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143

144Figure 1. Research area off the Yellow River estuary and adjacent sea areas

145Sample treatment and stable isotope analysis

146In this study, local dominating fish species (average site biomass > 90%), from the
 147five research biotopes, were collected for $\delta^{13}\text{C}$ and $\delta^{15}\text{N}$ analysis. The dorsal muscle
 148tissues of fishes were used for stable isotope analysis, which reflects long-term

information about nutrient accumulation (McIntyre and Flecker, 2006). Mixed zooplankton samples composed of several species were analyzed together to gain nitrogen stable isotope data for the TL baseline (Hoen et al., 2014). Each specimen was separated into two equal quantity samples. One sample was treated with 1 mol/L hydrochloric acid to remove inorganic carbon for $\delta^{13}\text{C}$ analysis (Kanaya et al., 2007), while the non-acidified one was used directly for $\delta^{15}\text{N}$ analysis. All samples were oven dried for approximately 24 to 48 h at 70 °C until a constant weight was achieved, and then homogenized into uniform particle size powder using a triturator. After pretreatment, the main process of carbon and nitrogen isotopic analysis was performed using an isotope ratio mass spectrometer (Delta V™, Thermo Fisher, Germany). Stable isotope ratios were expressed in standard δ unit notation ($\delta^{13}\text{C}$ and $\delta^{15}\text{N}$), and defined as follows:

$$\delta^{13}\text{C}(\text{‰}) = \left(\frac{{}^{13}\text{C}/{}^{12}\text{C}_s}{{}^{13}\text{C}/{}^{12}\text{C}_{\text{VPDB}}} - 1 \right) \times 1000 \quad (1)$$

$$\delta^{15}\text{N}(\text{‰}) = \left(\frac{{}^{15}\text{N}/{}^{14}\text{N}_s}{{}^{15}\text{N}/{}^{14}\text{N}_{\text{air}}} - 1 \right) \times 1000 \quad (2)$$

where ${}^{13}\text{C}/{}^{12}\text{C}_s$ and ${}^{15}\text{N}/{}^{14}\text{N}_s$ are the ratios of heavy isotopes to light isotopes from the samples and ${}^{13}\text{C}/{}^{12}\text{C}_{\text{VPDB}}$ and ${}^{15}\text{N}/{}^{14}\text{N}_{\text{air}}$ are the Vienna Pee Dee Belemnite (VPDB) standard and atmospheric N_2 standard for ${}^{13}\text{C}$ and ${}^{15}\text{N}$, respectively (Jegliniski et al., 2013).

Data analysis

$\delta^{13}\text{C}$ and $\delta^{15}\text{N}$ were analyzed using the current most efficient procedures, including SIBER metrics (Stable Isotope Bayesian Ellipses in R version 3.6.1), SIAR package (Stable Isotope Analysis in R), IsoSource (Phillips et al., 2012, www.epa.gov/wed/pages/models.htm), and the TL model (Post, 2002; Du et al., 2020). SPSS Statistics Subscription (IBM Inc., Armonk, NY, USA) was also used to determine how similar isotopic signatures were, and to distinguish the sources of identical stable isotope characteristics.

175 Isotopic niche analysis

176 SIBER is a multivariate ellipse-based model available in an R statistical computing
 177 package, which can reformulate metrics in a Bayesian framework for direct
 178 comparison of isotopic niches across biocoenosis (Jackson et al., 2011). When
 179 comparing individual groups with each other, either within a single community or in
 180 groups of communities, the Standard Ellipse Area (SEA) was recommended by the
 181 program author. With the SIBER object created, isotope biplots could be displayed
 182 using stated functions, and some summary statistics could be calculated for each
 183 group in the dataset. In this study, a pairwise comparison of biotopes was
 184 implemented using SIBER. The stable isotopic niche areas of each group, which were
 185 determined by the SEA, represented the trophic niche of respective fish communities
 186 plotted on a $\delta^{13}\text{C}$ - $\delta^{15}\text{N}$ dot plot. The ellipse area corrected for the small sample size
 187 (SEAc) and the stable isotopic niche width of each fish community was computed for
 188 comparison.

189 Trophic level calculation

190 Seven common fish species were collected to calculate and compare their TLs in each
 191 biotope. TLs were determined based on the nitrogen isotopic fractionation for ^{15}N
 192 enrichment through the food chains considering the consumer ingestion and metabolic
 193 process (Caut et al., 2009, 2010), undergoing predictable changes with each
 194 successive level up the trophic ladder (Smit et al., 2005). The recognized trophic
 195 fractionation factor of $\delta^{15}\text{N}$ ($\Delta^{15}\text{N}$) was 3.4‰ between contiguous TLs (Post, 2002).
 196 TLs could be calculated using the traditional model formula, as follows:

$$197 \quad TL = TL_{base} + \frac{\delta^{15}\text{N}_c - \delta^{15}\text{N}_b}{\Delta^{15}\text{N}} \quad (3)$$

198 where TL is the consumer trophic level, TL_{base} is the baseline trophic level, $\delta^{15}\text{N}_c$ is the
 199 consumer nitrogen isotope ratio, $\delta^{15}\text{N}_b$ is the marine primary consumer nitrogen
 200 isotope ratio, and $\Delta^{15}\text{N}$ is the trophic fractionation factor. Primary consumers occupied
 201 the 2nd TL at the base of the trophic ladder, so the $\delta^{15}\text{N}$ value of zooplankton was
 202 considered the baseline in this study.

203Food source analysis

204In this study, we identified five paralic organisms as the original food sources,
 205including autochthonous primary producers (phytoplankton and microphytobenthos)
 206and allochthonous food sources (macroalgae, cordgrass, and organic matter from the
 207Yellow River (YROM)), which were the primary energy providers for local paralic
 208food webs, as the potential primary food sources of fish species, analyzed the
 209contributions of each potential food source using IsoSource and SIAR, and then drew
 210block diagrams illustrating the results with SIAR. According to Wang et al. (2018),
 211estuarine organic matter is predominately from autochthonous sources, and the
 212estimated autochthonous organic carbon is approximately 58 to 82% of total organic
 213carbon. Therefore, POM with a diameter between 20 and 200 μm was deemed
 214representative of estuarine-marine phytoplankton and the associated isotope values
 215were used in food source analysis. Surface SOM ($< 5\text{ mm}$) excluded inorganic carbon
 216and was represented by microphytobenthos in local research areas. Enteromorpha and
 217spartina were represented by macroalgae and cordgrass in the subtidal and intertidal
 218zone, respectively. Since the diluted water directly influenced Biotope-H and Biotope-
 219C more than Biotope-B, Biotope-D, and Biotope-S, YROM in Biotope-H and
 220Biotope-C was included in the food source analysis, while that in Biotope-B, Biotope-
 221D, and Biotope-S was not. The results from previous study indicated that there was no
 222significant difference between YROM and primary terrestrial vegetation in the Yellow
 223River Delta (Qu et al., 2019), so YROM was suitable for representing delta vegetation
 224in a local finite area. Consequently, five potential food sources were identified in
 225Biotope-H and Biotope-C, corresponding to a 20% average contribution for fish
 226species, while four potential food sources corresponding to a 25% average
 227contribution were identified in Biotope-B, Biotope-D, and Biotope-S. The
 228contribution of each potential carbon source to fish communities was estimated using
 229SIAR (Jackson et al., 2009 and 2011), while the contribution of each potential carbon
 230source to the seven fish species was analyzed using IsoSource (Philips et al., 2012).

231Results

232Variation in $\delta^{13}\text{C}$ and $\delta^{15}\text{N}$

233Seventeen common fish species involving 168 specimens were randomly collected
 234from the five local research biotopes using a consistent sampling method for $\delta^{13}\text{C}$ and
 235 $\delta^{15}\text{N}$ analysis (Biomass proportions are shown in Appendix Table S1). The average
 236 $\delta^{13}\text{C}$ values for each species in the five survey areas are shown in Table 1. Biotope-H
 237was located at the subtidal zone of the northern Yellow River estuary, and was
 238affected most by the abundant diluted water from the Yellow River. In our study, $\delta^{13}\text{C}$
 239values for fish in this biotope had the broadest range, from -22.17‰ to -16.94‰ with
 240an average of -19.52‰. Biotope-C was located at the subtidal zone of the southern
 241Yellow River estuary, where the effect of diluted water was weaker than the northern
 242area because of silting at the southern river mouth. The $\delta^{13}\text{C}$ values for fish here
 243ranged from -21.50‰ to -17.48‰ with an average of -19.63‰. Biotope-D, which
 244was located further north of the Yellow River estuary, approaching open water, had
 245less of a diluted water effect. The $\delta^{13}\text{C}$ values for fish here ranged from -23.30‰ to -
 24620.00‰ and the average was -21.40‰. Biotope-B, located at the edge of the intertidal
 247zone northwest of the Yellow River estuary, was strongly influenced by local diluted
 248water from river branching rather than the Yellow River mainstream. The $\delta^{13}\text{C}$ values
 249for fish here ranged from -19.80‰ to -17.87‰ and the average was -18.87‰. For
 250Biotope-S, which was located near the intertidal zone of Laizhou Bay, south of the
 251Yellow River, the most influential factor was local land input instead of diluted water
 252from the Yellow River. The $\delta^{13}\text{C}$ values for fish here ranged from -21.56‰ to -
 25318.68‰ with an average of -20.02‰.

254The average $\delta^{15}\text{N}$ values are shown in Table 2. Similar to $\delta^{13}\text{C}$ values, the $\delta^{15}\text{N}$ range
 255for fish in Biotope-H was the broadest, from 7.05‰ to 14.30‰ with an average of
 25612.08‰. The $\delta^{15}\text{N}$ for fish in Biotope-B ranged from 10.62‰ to 13.16‰ with an
 257average of 11.97‰. The $\delta^{15}\text{N}$ for fish in Biotope-D ranged from 10.08‰ to 13.12‰
 258with an average of 11.57‰. The $\delta^{15}\text{N}$ for fish in Biotope-C ranged from 10.20‰ to
 25915.28‰ with an average of 12.57‰, and the $\delta^{15}\text{N}$ for fish in Biotope-S ranged from
 26011.89‰ to 17.09‰ with an average of 14.80‰. The $\delta^{15}\text{N}$ data for Biotope-S were
 261significantly higher than all other biotopes ($P < 0.01$), while the $\delta^{15}\text{N}$ data for Biotope-
 262C were significantly higher than those for Biotope-B and Biotope-D ($P < 0.05$,
 263Appendix Table S2).

265Table 1. $\delta^{13}\text{C}$ (mean values \pm SD, n = 3) of seventeen fish species in five survey areas in 2017 and 2018

No.	Species	B: $\delta^{13}\text{C}$ (‰)	D: $\delta^{13}\text{C}$ (‰)	H: $\delta^{13}\text{C}$ (‰)	C: $\delta^{13}\text{C}$ (‰)	S: $\delta^{13}\text{C}$ (‰)
1	<i>Argyrosomus argentatus</i>	-18.02 \pm 0.03	-20.70 \pm 0.78	-19.31 \pm 0.32	-18.27 \pm 0.72	-20.87 \pm 0.24
2	<i>Konosirus punctatus</i>	-18.70 \pm 0.06	-21.33 \pm 0.64	-19.90 \pm 0.13	-19.36 \pm 0.06	-21.09 \pm 0.03
3	<i>Cynoglossus semilaevis</i>	-19.01 \pm 0.30	-20.60 \pm 0.78	-19.02 \pm 0.38	-19.71 \pm 0.41	-19.38 \pm 0.26
4	<i>Thryssa kammalensis</i>	-19.19 \pm 0.26	-20.80 \pm 0.70	-22.01 \pm 0.24	-20.88 \pm 0.55	-20.34 \pm 1.06
5	<i>Amblychaeturichthys hexanema</i>	-19.70 \pm 0.07	-20.57 \pm 0.81	-20.42 \pm 0.43	-19.52 \pm 0.70	-19.11 \pm 0.00
6	<i>Sardinella zunasi</i>	-19.14 \pm 0.12	-21.77 \pm 0.90	-20.20 \pm 0.30	-19.69 \pm 0.54	-20.95 \pm 0.32
7	<i>Platycephalus indicus</i>	-18.49 \pm 0.07	-22.97 \pm 0.21	-19.20 \pm 0.09	-19.66 \pm 0.07	-19.11 \pm 0.71
8	<i>Synechogobius hasta</i>	-18.35 \pm 0.31	-20.87 \pm 0.72	-20.14 \pm 0.34	-19.25 \pm 0.23	—
9	<i>Triaenopogon barbatus</i>	-19.38 \pm 0.50	-21.27 \pm 0.67	—	—	—
10	<i>Thryssa mystax</i>	-19.29 \pm 0.51	—	-20.16 \pm 0.82	-20.15 \pm 0.29	—
11	<i>Cynoglossus joyneri</i>	-18.28 \pm 0.36	—	-17.60 \pm 0.81	—	—
12	<i>Enedrias fangi</i>	—	-23.10 \pm 0.20	—	—	—
13	<i>Sillago japonica</i>	—	—	-18.99 \pm 0.48	—	—
14	<i>Eupleurogrammus muticus</i>	—	—	-19.58 \pm 0.04	—	—
15	<i>Odontamblyopus rubicundus</i>	—	—	-18.28 \pm 0.01	-19.79 \pm 0.01	—
16	<i>Setipinna tenuifilis</i>	—	—	-18.44 \pm 0.37	-19.72 \pm 1.27	-19.39 \pm 0.22
17	<i>Pampus echinogaster</i>	—	—	-19.61 \pm 0.30	—	-19.93 \pm 0.19

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Table 2. $\delta^{15}\text{N}$ (mean values \pm SD, n = 3) of seventeen fish species in five survey areas in 2017 and 2018

No.	Species	B: $\delta^{15}\text{N}$ (‰)	D: $\delta^{15}\text{N}$ (‰)	H: $\delta^{15}\text{N}$ (‰)	C: $\delta^{15}\text{N}$ (‰)	S: $\delta^{15}\text{N}$ (‰)
1	<i>Argyrosomus argentatus</i>	12.96 \pm 0.04	11.15 \pm 0.82	12.70 \pm 1.21	12.58 \pm 0.57	15.36 \pm 0.77
2	<i>Konosirus punctatus</i>	11.24 \pm 0.22	11.76 \pm 0.89	10.97 \pm 0.53	10.38 \pm 0.16	11.93 \pm 0.07
3	<i>Cynoglossus semilaevis</i>	11.31 \pm 0.60	10.84 \pm 0.90	12.16 \pm 0.48	12.36 \pm 0.30	13.58 \pm 0.06
4	<i>Thryssa kammalensis</i>	11.80 \pm 0.08	11.27 \pm 0.85	12.57 \pm 0.06	13.66 \pm 0.31	14.94 \pm 0.14
5	<i>Amblychaeturichthys hexanema</i>	11.62 \pm 0.23	10.75 \pm 0.94	12.18 \pm 0.10	12.39 \pm 0.25	15.75 \pm 0.17
6	<i>Sardinella zunasi</i>	13.01 \pm 0.03	11.91 \pm 0.84	7.89 \pm 0.80	11.46 \pm 0.97	15.89 \pm 0.09
7	<i>Platycephalus indicus</i>	12.30 \pm 0.58	12.07 \pm 0.56	13.48 \pm 0.24	13.57 \pm 0.30	14.09 \pm 0.03
8	<i>Synechogobius hasta</i>	10.98 \pm 0.07	11.47 \pm 0.51	13.52 \pm 0.01	12.70 \pm 0.58	—
9	<i>Triaenopogon barbatus</i>	11.91 \pm 0.06	11.53 \pm 0.82	—	—	—
10	<i>Thryssa mystax</i>	11.52 \pm 0.26	—	11.52 \pm 0.17	11.95 \pm 0.25	—
11	<i>Cynoglossus joyneri</i>	13.03 \pm 0.12	—	13.52 \pm 0.45	—	—
12	<i>Enedrias fangi</i>	—	12.99 \pm 0.11	—	—	—
13	<i>Sillago japonica</i>	—	—	9.79 \pm 0.56	—	—
14	<i>Eupleurogrammus muticus</i>	—	—	11.82 \pm 0.83	—	—
15	<i>Odontamblyopus rubicundus</i>	—	—	12.40 \pm 0.67	12.26 \pm 0.03	—
16	<i>Setipinna tenuifilis</i>	—	—	13.85 \pm 0.41	14.90 \pm 0.41	14.69 \pm 0.16
17	<i>Pampus echinogaster</i>	—	—	12.81 \pm 0.42	—	16.94 \pm 0.15

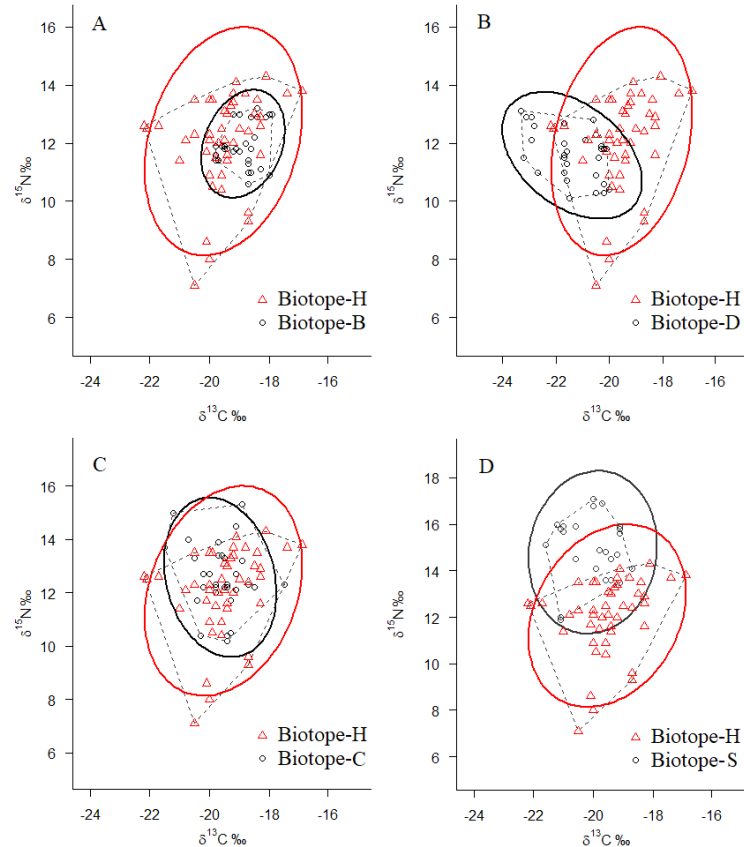
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276 Comparing stable isotopic niches using SIBER

277 Biotope-H was most influenced by diluted water from the Yellow River, so it was
 278 chosen as the object of comparison with the other biotopes (B, D, C, and S), avoiding
 279 too complex and mixed-up dots with their SEA in a single plot using SIBER (Figure
 280 2802). The niche width of $\delta^{13}\text{C}$ and $\delta^{15}\text{N}$ for fish is shown in Table 3. In Biotope-H, the
 281 niche width of $\delta^{13}\text{C}$ and $\delta^{15}\text{N}$ was 7.24 and 5.24, respectively, and both had the widest
 282 niche of all selected biotopes (Table 3). Niche widths in Biotope-B were the narrowest
 283 at 1.93 for $\delta^{13}\text{C}$ and 2.53 for $\delta^{15}\text{N}$. Accordingly, Biotope-H had the highest SEAc of
 284 5.38 followed by Biotope-S (4.10) and Biotope-C (2.98), while Biotope-B had the
 285 lowest SEAc of 1.36. The isotopic niche area of Biotope-H contained Biotope-B
 286 (Figure 2A), while similarly Biotope-H almost included Biotope-C except for one dot
 287 (Figure 2C, $\delta^{13}\text{C}$ of -21.18, $\delta^{15}\text{N}$ of 14.96). The SEAc of Biotope-D was 2.83, and its
 288 $\delta^{13}\text{C}$ value was significantly lower than that of Biotope-H (t-test, $P < 0.01$, $n = 33$)
 289 (Figure 2B). The SEAc of Biotope-S was 4.10, and its $\delta^{15}\text{N}$ was significantly higher
 290 than that of Biotope-H (t-test, $P < 0.01$, $n = 27$) (Figure 2D).

291



292

Figure 2. A dot plot of $\delta^{13}\text{C}$ - $\delta^{15}\text{N}$ with the standard ellipse area (SEAc) illustrating comparisons between Biotope-H and other biotopes (Biotope-B, Biotope-D, Biotope-C, and Biotope-S corresponding to inset A, B, C and D, respectively) using the SIBER package. (Biotope-H was drawn as red triangles, and the other biotopes were drawn as black circles)

298

Table 3. Sampling number (n), niche width ($\delta^{13}\text{C}$ and $\delta^{15}\text{N}$) and SIBER analysis results including the total area (TA), standard Bayesian Ellipse Area (SEA), and ellipse area corrected for small sample size (SEAc) in five different biotopes.

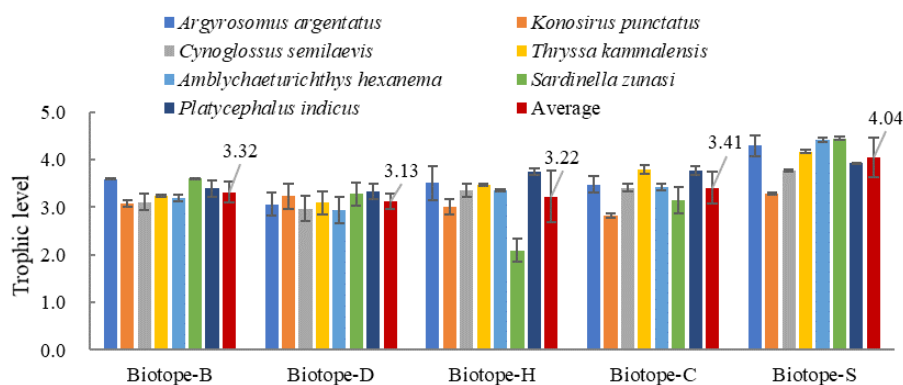
Biotope s	n	$\delta^{13}\text{C}$ niche width	$\delta^{15}\text{N}$ niche width	TA	SEA	SEAc
B	3 3	1.93	2.53	3.68	1.32	1.36
D	3 0	3.30	3.04	7.38	2.73	2.83
H	4 5	5.24	7.24	20.27	5.26	5.38
C	3 3	4.02	5.08	13.36	2.89	2.98
S	2 7	2.88	5.21	9.35	3.94	4.10

303

304Trophic levels

As seven common fish species (*Argyrosomus argentatus*, *Amblychaeturichthys hexanema*, *Cynoglossus semilaevis*, *Thryssa kammalensis*, *Konosirus punctatus*, *Platycephalus indicus*, and *Sardinella zunasi*) appeared in all five biotopes, they were chosen for TL and food source comparisons in this study. The TLs of these seven species in each biotope were calculated using a unique baseline. Figure 3 shows the average TL of the seven representative fishes in each biotope. Agreeing with $\delta^{15}\text{N}$ data, the highest average TL of 4.0 was found for Biotope-S, while the lowest average of 3.1 was found for Biotope-D. The TL of Biotope-S was significantly higher than any other biotope ($P < 0.01$, Appendix Table S3). Biotope-H had the highest standard deviation (0.5), while Biotope-D had the lowest standard deviation (0.2). For single species, the highest TL was 4.5 (*Sardinella zunasi*) in Biotope-S, while the lowest TL was 2.1 (also *Sardinella zunasi*) in Biotope-H.

317



318

319 Figure 3. TLs of the seven common fish species (average labeled) in the five biotopes

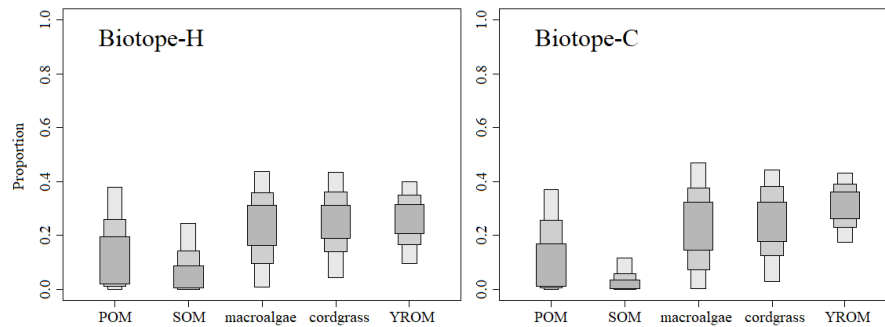
320 Food source analysis

321 As shown in Figure 4, the contribution of the five potential food sources showed a
 322 similar distribution tendency in Biotope-H and Biotope-C (Appendix Table S4).
 323 Allochthonous food sources (macroalgae, cordgrass, and YROM) showed higher
 324 proportional contributions to local fish communities than autochthonous sources
 325 (POM and SOM) (Figure 4). The 95% confidence intervals (CIs) were 0.01 to 0.44 for
 326 macroalgae, 0.04 to 0.43 for cordgrass, and 0.10 to 0.40 for YROM, and 0 to 0.37 for
 327 POM and 0 to 0.24 for SOM in Biotope-H. The 95% CIs were 0 to 0.47 for
 328 macroalgae, 0.03 to 0.44 for cordgrass, and 0.17 to 0.43 for YROM, and 0 to 0.37 for
 329 POM and 0 to 0.12 for SOM in Biotope-C. The 95% CIs for macroalgae showed the
 330 widest distribution, which was 0.43 in Biotope-H and 0.47 in Biotope-C, respectively.
 331 YROM demonstrated less variation in the confidence interval, which was 0.30 in
 332 Biotope-H and 0.16 in Biotope-C. Conversely, SOM showed the lowest contribution,
 333 with a 95% CI of 0 to 0.24 in Biotope-H and 0 to 0.12 in Biotope-C, respectively. For
 334 single species, both macroalgae and cordgrass contributed relatively more to
 335 *Argyrosomus argentatus* (mean = 0.33 and 0.30 in Biotope-H; 0.32 and 0.33 in
 336 Biotope-C), while YROM contributed a 0.37 mean proportion for *Thryssa*
 337 *kammalensis* in Biotope-H. More detailed food source contribution data for fish
 338 species are shown in Appendix Table S4 and S5.

339 In Biotope-D, which was the farthest offshore, POM showed a significantly higher
 340 food source contribution, and its 95% CI was higher (0.26 to 0.79) than any other
 341 potential food sources (Figure 5, Appendix Table S4), and for single species, the
 342 highest mean contribution of POM was to *Konosirus punctatus* (0.76) in Biotope-D
 343 (Appendix Table S5). The 95% CI for SOM was also higher (0.17 to 0.53) than for
 344 macroalgae and cordgrass, which showed extremely low 95% CIs of 0 to 0.18 and 0

345to 0.11, respectively. In Biotope-S, POM and SOM were also considered to be the
 346primary food sources based on contribution results, which accounted for 0 to 0.61, but
 347with high distribution indeterminacy (0.61), and 0.16 to 0.54 for 95% CIs,
 348respectively. In Biotope-B, the contributions of each food source were in relative
 349equilibrium, while for single species, cordgrass showed a 0.48 contribution to
 350*Argyrosomus argentatus* (Appendix Table S5).

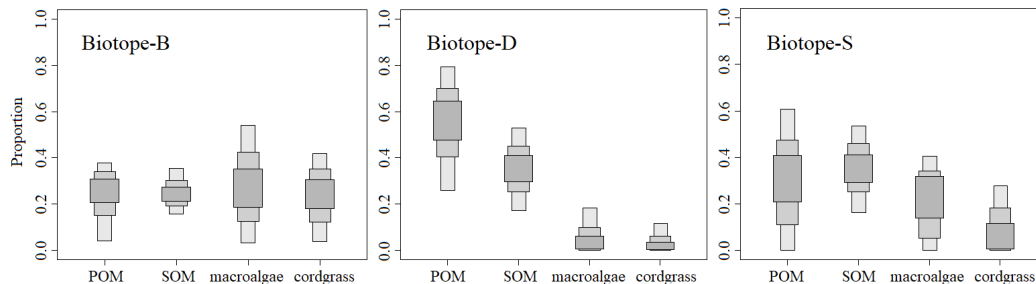
351



352

353Figure 4. Relative contribution of the five potential food sources to the diet of fish
 354communities in Biotope-H and Biotope-C using SIAR. Grey shaded areas represent
 35595%, 75%, and 50% confidence intervals. (Food source containing POM = suspended
 356particulate organic matter, SOM = sedimental organic matter, macroalgae, cordgrass,
 357and YROM = Yellow River organic matter.)

358



359

360Figure 5. Relative contribution of the five potential food sources to the diet of fish
 361communities in Biotope-B, Biotope-D, and Biotope-S using SIAR. Grey shaded areas
 362represent 95%, 75%, and 50% confidence intervals. (Food source containing POM =
 363suspended particulate organic matter, SOM = sedimental organic matter, macroalgae
 364and cordgrass.)

365 Discussion

366 Trophic niche variation

367 Increasing numbers of studies have shown that coastal and estuarine ecological
 368 connectivity plays an essential role in ecosystem conservation and restoration (Du et
 369 al., 2015). The stable isotopic niche results in this study indicated that the SEAc
 370 (5.38) and total area (TA, 20.27) of estuarine Biotope-H covered the majority of the
 371 other research areas (Figure 2, Table 3). In general, the results showed the
 372 compatibility of communities among Yellow River estuarine habitants, and that the
 373 area may be a corridor for energy and material transportation between Laizhou Bay
 374 and the open water. Further, the results highlight the importance of terrestrial-marine
 375 linkages for interpreting energy flow in estuarine ecosystems (Wai et al., 2011). Local
 376 biocoenosis used various food sources to adapt to the fickle environment, including
 377 marine-terrestrial boundaries and the estuary, and this was embodied in the wider
 378 isotopic niche corresponding to frequent environmental changes and abiotic gradients
 379 (Lange et al., 2018). These marine-terrestrial linkages served as feeding and nursery
 380 grounds for fish biocoenosis, where even top predators like shark species gained
 381 trophic subsidies along food chains from land (Wai et al., 2012). A comparative
 382 analysis of stable isotopic niches was useful for detecting patterns in trophic structure
 383 and identifying differences or similarities in trophic organization related to
 384 environmental conditions (Abrantes et al., 2014). It is evident that changes to the TLs
 385 and food sources of fish communities in terms of $\delta^{13}\text{C}$ and $\delta^{15}\text{N}$ will have an effect on
 386 conservation.

387 The $\delta^{13}\text{C}$ data in our study indicated that all fish species were mainly corresponding to
 388 benthic diatoms, macroalgae, and estuarine-marine phytoplankton (Cloern et al.,
 389 2002), which occupy normal marine isotopic niches (Newsome et al., 2007). Beyond
 390 macroalgae, our analysis of the food source contribution indicated that YROM and
 391 cordgrass were major allochthonous energy sources in estuarine areas directly affected
 392 by Yellow River-diluted water, while local autochthonous primary producers
 393 (phytoplankton and microphytobenthos) demonstrated a low contribution in those
 394 specific areas (Figure 4). Phytoplankton produce new particles that drive the
 395 biological carbon pump, contributing to the global carbon cycle in the ocean, which
 396 plays a disproportionately important role in the global climate on a range of time
 397 scales (Bolaños et al., 2020; Moreau et al., 2020). However, it is susceptible to

environmental conditions relative to other primary producers, especially the variable environment of estuaries. The bloom and extinction of phytoplankton is driven by physical, chemical, and biological seasonality (Bolaños et al., 2020). Hydrology and dissolved nutrients have been widely identified as the main drivers of phytoplankton dynamics in estuarine ecosystems (Tao et al., 2020). Our investigation indicated that a high concentration of chlorophyll-a, a representative of phytoplankton (Moreau et al., 2020), had not shown up in estuarine areas with a direct diluted water influence (Appendix 1). This is consistent with the results of Ding et al. (2020). Microphytobenthic primary producers also demonstrated a low contribution, probably due to their dependency on light (Haro et al., 2019) and encounter with high-suspended solids (Wang et al., 2017). On the other hand, besides macroalgae, allochthonous energy sources were identified as the main food sources supporting the estuarine food web. In a previous study, most of the riverine organic carbon originated from delta vegetation debris (Phragmites, Suaeda, and Tamarisk) in particulate form (Wang et al., 2018). Suspended particulate matter acts as the main carrier of organic matter, providing energy to the estuarine food web from upstream carrying, which plays an important role in the conditioning of productivity and ecosystem functions in estuaries (Li et al., 2020). As a representative of cordgrass in the intertidal zone, *Spartina* provided a considerable food contribution proportion for estuarine fish communities in this study. However, it was recognized as the main invasive plant in the Yellow River estuarine area, so its contribution to the local estuarine food web is still controversial (Chen et al., 2020). *Spartina alterniflora* was first introduced to the coastal wetlands of China from the United States in 1979 for the purpose of ecological restoration. From 1985 to 2015, it continued to spread across the coast of mainland China as a typical invasive species (Meng et al., 2020).

Areas away from the Yellow River estuary show different food contribution characteristics compared with the estuarine area. As indicated by the food source contribution results, autochthonous benthic and pelagic producers (microphytobenthos and phytoplankton) dominated carbon input into the food web in Biotope-S and Biotope-D, which conformed to the normal characteristics of an intertidal ecosystem like the Wadden Sea (Christianen et al., 2017). Microphytobenthos form extensive biofilms on the sediment surface conducive to its stabilization. They are not easily disturbed and thus provide a more stable food source for local consumers (Hart and Lovvorn, 2003; Miyatake et al., 2014). In contrast, phytoplankton are more vulnerable to influence from environmental conditions (Armbrecht et al., 2015), while providing

an unstable food source according to the more discrete confidence interval of contribution (Figure 5).

Majorization of the trophic model baseline

The $\delta^{15}\text{N}$ data indicated that the fish species in our study system belonged to TL 2.1 to 4.5 using a unique baseline (Figure 3), covered a TL distance of 2.4, which differed from the general trophic pattern of fish communities in Chinese coastal waters, such as 3.0 to 4.1 for Changjiang Estuary at the junction of the Yellow Sea and the East China Sea (Chang et al., 2014), 3.1 to 3.6 in the coastal water of the Yellow Sea (Feng et al., 2014), and 2.9 to 3.9 at the junction of the East China Sea and South China Sea (Du et al., 2015). Their trophic niche was also wider than that in the western Mediterranean (2.9 to 4.0, Valls et al., 2014), but lower than that in the Gulf of Maine (3.7 to 5.2, Scharf et al., 2019). From the above comparisons, the variation in TL in our research area (2.4) was much wider than in other similar coastal areas. Intriguingly, this variation was reflected not only in single species, but also significantly in the biotopes according to the results of this study (Appendix Table S3). Our results demonstrated that the average fish TL was 3.3 in Biotope-B, 3.1 in Biotope-B, 3.2 in Biotope-H, 3.4 in Biotope-C, and 4.0 in Biotope-S, giving the trend of $S > C > B > H > D$. This tendency showed that the average TL was significantly higher in Laizhou Bay than any other biotope and decreased from the near shore biotope to the far shore, which meant significant variation in our original $\delta^{15}\text{N}$ data. It was an abnormal phenomenon that similar species had such a significantly different $\delta^{15}\text{N}$ characteristic between connected biotopes.

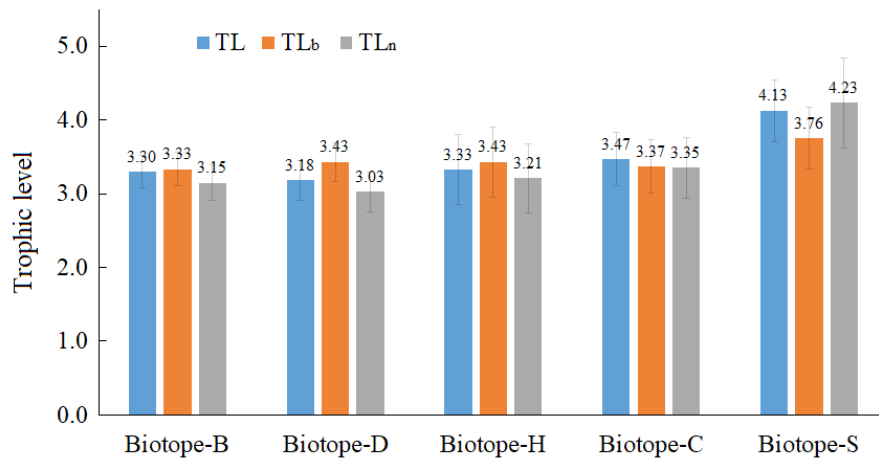
$\delta^{15}\text{N}$ can also be used to indicate scenopoetic dimensions, such as marine-terrestrial (Lange et al., 2018) and eutrophication (Goeddy et al., 2016). Generally, the difference in habitat conditions is mainly related to the scenopoetic dimensions where a high $\delta^{15}\text{N}$ value indicate a marine characteristic while a low value indicate a terrestrial characteristic; a high value also indicates a eutrophic area while a low value indicates a pristine area (Newsome et al., 2007). The environmental condition of Laizhou Bay is closer to a mainland and it belongs more to a terrestrial rather than a marine characteristic with its lower $\delta^{15}\text{N}$ value, and thus the results in our study did not obey the marine-terrestrial pattern. We also suspected that local aquaculture activities might lead to a high $\delta^{15}\text{N}$ value from wild marine lives by releasing organic bait based on similar conditions in the aquaculture water in Jiaozhou Bay, off the coast of China (Feng et al., 2014). Therefore, a field survey was conducted in May

4672020 and scallop culture was identified as the main local aquaculture with no release
468of anthropogenic organic bait. After excluding the above conditions, we needed a
469more reasonable theory to explain the abnormal phenomenon in our study area.

470After the above analysis, we turned our attention to probing into the $\delta^{15}\text{N}$ variation
471based on the food source contribution. Whether in a high $\delta^{15}\text{N}$ (Biotope-S) or low $\delta^{15}\text{N}$
472(Biotope-D) area, the autochthonous food source demonstrated a relatively high
473contribution to the local fish communities, which implied that it is probably due to the
474influence of primary producers at the base of the food web (Oakes et al., 2010).
475However, primary producers seldom directly provide energy to high-TL predators
476(Warne et al., 2010), so it does not adequately explain the $\delta^{15}\text{N}$ variation in different
477biotopes. Therefore, in May 2020, we reanalyzed the zooplankton, which was
478considered a mediator of energy-transfer from primary producers to high-TL predators
479and tended to respond to variations in food source $\delta^{15}\text{N}$ values (Schmidt et al., 2003),
480and then recalculated the TL of fishes using zooplankton $\delta^{15}\text{N}$ values as the baseline in
481each biotope (Figure 6). Compared with the unique baseline, the differences in fish
482TLs were smaller between each biotope (Figure 6 TL_b). The fish TL in Biotope-C was
483no longer significantly higher than that in Biotope-B and Biotope-D. Though the fish
484TL in Biotope-S was still significantly higher than that in the other biotopes, the gaps
485decreased from 0.83 to 0.42 with Biotope-B, from 0.95 to 0.69 with Biotope-D, from
4860.80 to 0.33 with Biotope-H, and from 0.65 to 0.39 with Biotope-C (Appendix Table
487S6). However, the bias caused by abnormally high $\delta^{15}\text{N}$ still existed in the optimized
488results. This method demonstrated a method for solving the disparity generated from
489 $\delta^{15}\text{N}$ data in spatial distribution, which implied a tendency that the $\delta^{15}\text{N}$ variation
490originated in a more fundamental part of the estuarine food web, such as nitrogen
491cycling dynamics (Hetherington et al., 2017).

492To exclude possible error introduced during the calculating process, we also used
493Hussey's equation to recalculate the TL (Hussey et al., 2014; Reum et al., 2015,
494method in Appendix 2). Based on the new equation, there were some changes in the
495TL results, but they still retained the same trend in terms of the average biotope TL
496(Figure 6, TL_n).

497



498

499Figure 6. Average TLs of fishes in each biotope using Post's (TL) and Hussey's (TL_n)
500methods, and revised TLs using a different baseline value ($\delta^{15}\text{N}$ of primary
501consumers) Post's method (TL_b)

502

503The inorganic nitrogen assimilation process is a key driver from primary producers in
504the marine nitrogen cycle (Hetherington et al., 2017). Due to the high and stable
505contribution for fishes, SOM was considered a primary food source in Biotope-S. We
506further investigated its $\delta^{15}\text{N}$ distribution (Figure 7) and found a strong link between
507the high $\delta^{15}\text{N}$ of fish and the distribution of dissolved inorganic nitrogen (DIN)
508(Appendix 3). Nitrogen-fixing microorganisms, such as nitrospinae, were also
509significantly enriched in ^{15}N under conditions of a high inorganic nitrogen
510concentration (Kitzinger et al., 2020). Therefore, the high $\delta^{15}\text{N}$ value of SOM was
511most probably caused by microorganisms and other primary producers assimilating
512high-concentration DIN, which was likely the reason for the eutrophic pattern of $\delta^{15}\text{N}$
513(Newsome et al., 2007). The major primary consumers, zooplankton, tended to
514respond to this variation (Schmidt et al., 2003). This characteristic would translate to
515high-TL consumers like fish communities via marine food chains leading to biases in
516the statistical process (Auerswald et al., 2010; Layman et al., 2012).

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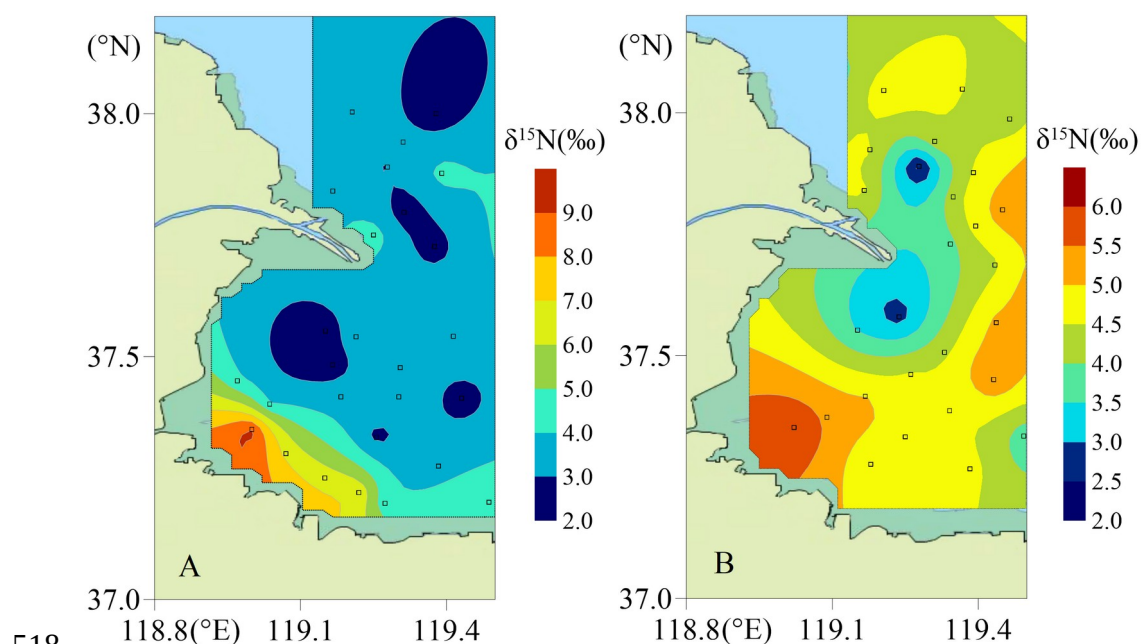


Figure 7. The distribution of $\delta^{15}\text{N}$ in organic sediment in August 2017 (A) and May 2020 (B).

Hetherington et al. (2017) used linear mixed effects models (LMEs) to verify that source amino acid $\delta^{15}\text{N}$ values in marine lives were related to nitrate concentrations, which pertained to fluctuations in biogeochemical cycling at the base of the food web. Therefore, if we intend to solve this issue fundamentally, the baseline should be optimized through comprehensive and integrated analysis in further research. This conclusion will also promote our next step research with more specific verification to explain the integrated transfer process. Although this conclusion connected a series of transfer processes from high concentration DIN to primary producers, to primary consumers, and then to higher consumers like fishes, which were so complicated and contained too many intermediate processes, it is the most likely reason to explain this abnormal phenomenon based on a previous theory (Teichberge et al., 2010; Kitlinger et al., 2020) combined with our results, which suggested highly connected food web loops.

Many previous studies found that their application was further complicated by potential shifts in baseline $\delta^{15}\text{N}$ for many specific ecological processes, such as migration of marine nektons like bluefin tuna and swordfish (Schartup et al., 2019), significant taxonomic variation in the composition of primary producers at the base of the food webs (Ramshaw et al., 2017), and supply way of DIN sources (Kitlinger et al., 2020). The $\delta^{15}\text{N}$ characteristic of primary producers may vary by as much as 10‰ over a spatial and temporal scale (McMahon et al., 2015). Therefore, identifying an

appropriate baseline requires not only considering migratory predators but also paying more attention to the local primary producers that can determine the baseline of marine trophic structures more directly.

If our conclusion is right that the local high $\delta^{15}\text{N}$ of fishes originated from the high concentration of DIN in Biotope-S, a trophic model relying on the $\delta^{15}\text{N}$ characteristic would be further complicated by potential shifts in the baseline due to variations in the $\delta^{15}\text{N}$ characteristic in primary producers (Gutiérrez-Rodríguez et al., 2014). Regional $\delta^{15}\text{N}$ diversity in primary producers should be considered not only between broad oceans on a large spatial scale (Schmidt et al., 2003; Hetherington et al., 2017), but also among adjacent coastal waters with high DIN variation, such as the estuary. Though the research areas in our study were not significantly isolated, the bias of TLs in fish communities still emerged between biotopes, which indicates the diversity of matter and energy flows (Palmer et al., 2019).

Conclusion

Stable isotopic niche results indicated that estuarine inhabitants showed the compatibility of the communities of most study biotopes, which may provide a corridor for energy and material transportation between Laizhou Bay and the open water. YROM and cordgrass were considered the major allochthonous energy sources in estuarine areas directly affected by Yellow River-diluted water, while phytoplankton and microphytobenthos demonstrated a low contribution as local autochthonous primary producers. Areas away from the Yellow River estuary showed different food contribution characteristics compared with estuarine areas. As indicated by the food source contribution results, autochthonous benthic and pelagic producers (microphytobenthos and phytoplankton) dominated carbon input into food webs. Our results showed that the significant variation in the fish $\delta^{15}\text{N}$ characteristic presented within estuarine adjacent regions (less than 2 degrees latitude), led to significant variation in TLs in the same fish species, using a unique baseline. Although the research areas in our study were not significantly isolated, the bias of TLs in fish communities still emerged between biotopes. This indicates the diversity of matter and energy flows. Regional $\delta^{15}\text{N}$ diversity in primary producers should be considered not only between broad oceans on a large spatial scale, but also among adjacent coastal waters with high DIN variation. These results offer a new perspective on trophic relationships, and provide the first detailed data for enhancing our understanding of the variations among fish communities in estuarine ecosystems.

576 Acknowledgement and funding

577 We thank Xijie Yin's research team of Third Institute of Oceanography, MNR for
 578 testing stable isotopic data. This work was supported by National Natural Science
 579 Foundation of China (NSFC, Grant Nos. 41706140 and 41606140), China Ocean
 580 Mineral Resources R&D Association Project (Grant Nos. DY135-E2-1-02 and
 581 DY135-E2-2-05) and Open Fund of Key Laboratory of Ocean Ecological Monitoring
 582 and Restoration Technologies, MNR (No: 202004).

583 Data availability statement

584 Supporting information is available at online version of the manuscript.

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