

1 **Abundance and residency patterns of common bottlenose dolphins (*Tursiops truncatus*) in**  
2 **freshwater-influenced estuaries of the northern Gulf of Mexico**

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22 **Running page head:** *Bottlenose dolphin abundance in Alabama waters*

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25 **KEYWORDS:** Capture-mark-recapture, cetacean, demographics, marine mammals, photo-  
26 identification, Robust Design models, salinity.

42 **Abstract**

- 43 1. Multiple stressors are affecting common bottlenose dolphins in the northern Gulf of  
44 Mexico, including two embayments in Alabama, Mobile Bay (MOB) and Perdido  
45 Bay (PER), where no comprehensive abundance estimates and residency data are  
46 currently available.
- 47 2. This study provides the first seasonal abundance estimates and residency patterns for  
48 bottlenose dolphins in MOB and PER and discusses the effects of seasonal entry of  
49 large volumes of freshwater on dolphin abundance.
- 50 3. In MOB, abundance estimates were larger in summer, with the highest abundance  
51 recorded in summer 2022 (1,712 dolphins, 95% CI: 1,520-1,928) and the lowest in  
52 winter 2019-2020 (518 dolphins, 95% CI: 260-1,032). The opposite pattern was  
53 found in PER, where abundances were larger in winter, with the highest abundance in  
54 winter 2021-2022 (191 dolphins, 95% CI: 157-232) and the lowest in summer 2020  
55 (100 dolphins, 95% CI: 81-122).
- 56 4. Stronger residency patterns were found in PER, with 28% (n = 52) encountered in  
57 more than three seasons compared to MOB where only 9% (n = 57) were encountered  
58 in more than three seasons, and considered resident dolphins.
- 59 5. The two studied embayments support a larger number of dolphins than previously  
60 documented and likely provide seasonally different resources, indicating high and  
61 potentially complex use of these estuaries. Combined with other data concurrently  
62 collected in this area, this study will inform conservation management and strategies  
63 for these highly impacted dolphin stocks.

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72 **1. INTRODUCTION**

73 The common bottlenose dolphin, *Tursiops truncatus* (hereafter referred as bottlenose  
74 dolphin), is the most abundant delphinid species in coastal waters of the northern Gulf of Mexico  
75 (GOMx) and the only cetacean species inhabiting bays, sounds and estuaries (BSEs) along this  
76 coastline (Garrison et al., 2021; Mullin et al., 1990). For management purposes in the northern  
77 GOMx, the species is divided into 32 distinct BSE stocks from Texas to the Florida Keys, and  
78 several of these are poorly monitored, with no recent abundance estimates available (Hayes et  
79 al., 2023). In particular, bottlenose dolphin populations inhabiting Mobile Bay and Perdido Bay  
80 (Alabama, USA) have received very little research attention. The last abundance estimate  
81 available in this area was conducted in 1993 using aerial surveys and estimated 122 dolphins in  
82 Mobile Bay only (Hayes et al., 2023). Nevertheless, estimation of species abundance is  
83 fundamental to monitoring populations through time and informing conservation assessments  
84 and strategies (Hammond et al., 2021). This population metric is particularly essential when  
85 studying apex predators that have top-down control on ecosystem structure and function and  
86 often serve as sentinels of ecosystem health (Baum & Worm, 2009; Wells, et al., 2004).  
87 Abundance assessments are also notably important in areas heavily impacted by anthropogenic  
88 activities, where multiple stressors may have negative effects on population demographics and  
89 survival (Baum & Worm, 2009; Hammond et al., 2021).

90 In the northern GOMx, multiple stressors affect bottlenose dolphin populations (e.g.,  
91 Bloodgood et al., 2023; Carmichael et al., 2012). These stressors include infectious diseases  
92 (Bloodgood et al., 2023; Cloyed et al., 2021a; Fauquier et al., 2017), harmful algal blooms (Fire  
93 et al., 2007; Litz et al., 2014; Twiner et al., 2012), negative human interactions (Carmichael et  
94 al., 2022; Collins et al., 2020; Powell et al., 2018; Samuels & Bejder, 2004), pollutant exposure  
95 (e.g., *Deepwater Horizon* oil spill; J. Balmer et al., 2018; Schwacke et al., 2014, 2017, 2022;  
96 Takeshita et al., 2017), and perennial freshwater inputs (e.g., Carmichael et al., 2012, Takeshita  
97 et al., 2021). The northern GOMx, is characterised by rapid and large seasonal volumes of cold  
98 freshwater discharge that can drastically change the marine ecosystem and the suitability of  
99 habitat for dolphins (Alexander et al., 2001; Carmichael et al., 2012). Bottlenose dolphins  
100 typically tolerate seasonal changes in their habitat (e.g., water temperature and/or salinity), but  
101 they cannot physiologically support prolonged exposure to low salinity water (< 10 parts per  
102 thousand, ppt; Bloodgood et al., 2023; Deming et al., 2020) or temperatures outside their

103 thermoneutral zone (Carmichael et al., 2012; Colbert et al., 1999; Yeates & Houser, 2008). When  
104 environmental conditions become unsuitable, dolphin health can be impacted and lead to  
105 changes in abundance and residency patterns through movements (Fazioli & Mintzer, 2020;  
106 McBride-Kebert & Toms, 2021) or direct mortalities (Deming et al., 2020; Ewing et al., 2017;  
107 Ridgway & Venn-Watson, 2010). From 1978 to 2019, environmental and human stressors have  
108 resulted in three unusual mortality events (UMEs) of bottlenose dolphins in the northcentral  
109 GOMx (Louisiana, Mississippi and Alabama) with several thousand dolphin mortalities (e.g.,  
110 Carmichael et al., 2012; Colegrove et al., 2016; Litz et al., 2014). Recent mortality events were  
111 mostly attributed to the *Deepwater Horizon* oil spill in April 2010, which was the largest marine  
112 oil spill that occurred in the United States (USA) (Beyer et al., 2016), and exposure to low  
113 salinity waters (Litz et al., 2014; NOAA NMFS, 2022). Consequently, assessing the impact of  
114 mortality and developing long-term monitoring programs will be crucial to the conservation of  
115 dolphin stocks in this region and, in particular, in Alabama waters where abundance and  
116 residency estimates are not available.

117         The goal of this study is to inform understanding of abundance and residency patterns of  
118 bottlenose dolphin populations in Alabama waters that are needed for understanding local  
119 ecosystem functions as well as management and conservation actions under the Marine Mammal  
120 Protection Act (MMPA; 50 CFR 216). To estimate abundance and define residency patterns for  
121 bottlenose dolphins in Alabama waters, we conducted dedicated seasonal capture-mark-recapture  
122 (CMR) photo-identification (photo-ID) surveys across three years (three summer and three  
123 winter seasons). Surveys were conducted in two adjacent estuaries, Mobile Bay and Perdido Bay,  
124 to address the crucial lack of information on abundance and demographic parameters for  
125 bottlenose dolphins in the region. We also discussed the relevance of established methods (i.e.,  
126 Pollock's Robust Design models and standardised CMR photo-ID surveys) in these two  
127 embayments with very different characteristics, such as embayment size, level of openness and  
128 connectedness with adjacent coastal waters, and volumes of freshwater discharge that affect  
129 salinity regimes through time.

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134 **2. METHODS**

135 2.1) Study areas

136 2.1.1) Mobile Bay (MOB)

137 The MOB estuary, Alabama (30°27'30.3" N, 87°59'44.3" W; Figure 1) is a large (1,070  
138 km<sup>2</sup>) semi-open embayment that is the sixth largest estuary in the USA, draining the fourth  
139 largest volume of freshwater discharge (mean annual discharge = 1,840 m<sup>3</sup> s<sup>-1</sup>) from watersheds  
140 contained in the USA, usually peaking in winter and spring (Alexander et al., 2001; Schroeder et  
141 al., 1990; Valentine et al., 2013). This influx of freshwater can drastically change the salinity of  
142 the water in MOB and adjacent waters (Carmichael et al., 2012; Schroeder et al., 1990), with a  
143 low salinity regime during times of high freshwater flow and a higher salinity regime during  
144 periods of lower discharge (Orlando & Klein, 1989). The bay is mostly shallow with an average  
145 water depth of 3 m but has a deeper ship channel of approximately 23 m, which runs primarily  
146 north-south and connects MOB to the intracoastal waterway and the GOMx at its southernmost  
147 extent (Schroeder et al., 1990; Valentine et al., 2013). Seasonal sea surface temperatures range  
148 on average between ~ 13 °C in winter and ~ 30 °C in summer (Orlando & Klein, 1989). The Port  
149 of Mobile, located in the northern reaches of MOB, has been the second fastest growing port in  
150 the USA during the past decade (Roberts, 2023). Expansion plans are ongoing and include  
151 deepening and widening of the Mobile Ship Channel to make it the largest harbour on the GOMx  
152 coast (USACE, 2023). As a result, MOB is utilised by a range of vessels from small recreational  
153 boats to large commercial ships that access resources in the bay and its tributaries and support  
154 significant recreational (e.g., fishing, boating) and industrial activities (e.g., metallurgy, aviation,  
155 chemicals; USACE, 2023; Valentine et al., 2013).

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157 2.1.2) Perdido Bay (PER)

158 The PER estuary (30°18'37.1" N, 87°30'25.0" W; Figure 1) is a small (130 km<sup>2</sup>) enclosed  
159 bay located on the Alabama-Florida border with an average water depth of approximately 2.6 m  
160 (Livingston, 2001; Xia et al., 2011). The PER estuary is composed of several smaller lagoons  
161 and bayous that are connected to the GOMx in the south through a narrow channel (Perdido  
162 Pass). This embayment receives freshwater inflow primarily from the Perdido River in the north  
163 (mean daily discharge = 62 m<sup>3</sup> s<sup>-1</sup>), and the water quality can rapidly change in response to  
164 rainfall, wind, and tides due to its small size (PERG, 1998). Seasonal sea surface temperatures

165 range on average between  $\sim 15$  °C in winter and  $\sim 30$  °C in summer (Orlando & Klein, 1989).  
166 The watershed surrounding PER is characterised by an urbanisation gradient, where areas to the  
167 north near Perdido River are largely vegetated and agriculture and residential development  
168 increases down bay, with the highest tourism and urbanisation in the southern parts of the bay  
169 (PPBEP, 2022). Human activities are primarily residential, recreational, and tourism-related,  
170 with smaller-scale commercial activities than MOB.

171

## 172 2.2) Field sampling and data collection

173 Dedicated boat-based surveys following systematic transect and contour lines were  
174 conducted in winter (December - February) and summer (June - July) from January 2020 to June  
175 2022 in the MOB and PER estuaries (Figure 1, Table 1). In each embayment, we conducted  
176 CMR photo-ID surveys following standardised methodologies specifically designed to derive  
177 dolphin abundance and population demographics (Rosel et al., 2011). A CMR photo-ID survey  
178 was typically a day at sea partially or completely following either a transect or a contour line  
179 searching for dolphins on effort.

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### 181 2.2.1) Capture-mark-recapture

182 Photo-ID data of individual dolphins were collected following Pollock's Robust Design  
183 (RD) (Pollock, 1982), with six primary sampling periods (i.e., seasons) between which dolphin  
184 populations were considered open to births, deaths, immigration, and emigration. Within primary  
185 sampling periods, the crew surveyed each embayment three times (i.e., secondary sampling  
186 occasions) in a short period of time to allow for closed population assumptions, such that no  
187 demographic changes would be expected (Kendall et al., 1995, 1997; Otis et al., 1978; Pollock,  
188 1982). Once surveys were completed three times in one embayment, the crew immediately  
189 started sampling the second embayment. By the end of the study period, all transects and contour  
190 lines were completely surveyed 18 times in each embayment, and thus a total of 18 second  
191 sampling occasions were successfully completed in each location during the course of the survey  
192 period. One secondary sampling occasion took approximately three to six days of survey effort  
193 in MOB and two to three days in PER. Under the RD models, abundance estimates and dolphin  
194 survival (S) were generated within each season (primary sampling period) by collapsing data  
195 from the secondary sampling occasions into a single instance of being encountered or not

196 encountered (Pollock, 1982). Finally, temporary emigration between seasons was estimated as  
197 the probability of being a temporary migrant ( $\gamma''$ ) between primary sampling periods, and the  
198 probability of remaining a temporary migrant ( $\gamma'$ ) between primary sampling periods (Kendall et  
199 al., 1997).

200 During surveys in MOB, two outboard engine boats (6 m and 7 m length) were usually  
201 used to simultaneously survey the embayment, following roughly parallel transect lines within  
202 the bay and the entire perimeter of the bay along the contour of the shoreline (Figure 1). In PER,  
203 due to the smaller and narrower size of the study area, surveys were conducted using only the 6  
204 m outboard engine boat, following the contour of the shoreline (Figure 1). Perimeter contour  
205 lines in each embayment were approximately ranging between 0.2 and 2.0 km distance from  
206 shore in  $\sim 2$  m of water depth. At least three crew members (maximum five) were present  
207 onboard each vessel, including a captain, one photographer, and data recorder. All crew  
208 members searched for dolphins with the naked eye and scanned an estimated area of 500 m,  $180^\circ$   
209 ahead of the research vessel. To maximize the likelihood of detecting dolphin groups, surveys  
210 were conducted under favourable weather conditions: wind speeds  $\leq$  Beaufort Sea State 4, clear  
211 visibility (no fog or haze), and a swell height of  $\leq 1$  m. All transect and contour lines were  
212 surveyed at a consistent speed of approximately 14 - 16 knots.

213 Sightings began when a dolphin group was encountered; the vessel was slowed and the  
214 dolphins were approached. A group of dolphins was defined as several dolphins usually  
215 swimming within a distance of 10 m of any other individuals (i.e., 10 m “chain” rule), moving in  
216 the same direction and engaged in similar behavioural activities (Smolker et al., 1992; Syme et  
217 al., 2022). The crew attempted to photograph the left and right sides of the dorsal fins of every  
218 dolphin in the group for individual identification (Würsig & Würsig, 1977), using a Canon EOS  
219 7D digital camera equipped with a 100 - 400 mm telephoto zoom lens (Canon USA, Inc.). For  
220 each dolphin sighting, the minimum (lowest value), maximum (highest value), and best number  
221 of dolphins, calves and neonates were estimated through a consensus decision among team  
222 members onboard. Daily effort was logged to keep track of the surveyed transects, the number of  
223 km surveyed on and off effort, and changes in weather conditions. In both embayments, the total  
224 number of km surveyed on effort (i.e., distance on effort) was calculated for each season and was  
225 used to generate the seasonal encounter rates (number of sightings per surveyed km) and relative  
226 abundances (number of individuals encountered per surveyed km).

227                    2.2.2) Photo-identification and photo analyses

228                    Each individual encountered and captured in photos was identified using natural and  
229 long-lasting nicks and notches present on the dorsal fin (Würsig & Jefferson, 1990). A  
230 standardised protocol was used to rate photographic quality and dorsal fin distinctiveness as  
231 described in Melancon et al. (2011). Briefly, overall photographic quality of dorsal fins was  
232 graded as Q1 (excellent), Q2 (good) or Q3 (poor) based on the focus and clarity, contrast, angle,  
233 dorsal fin visibility, and proportion of the fin within the photo frame. A distinctiveness rating  
234 was also attributed to each individual dorsal fin as very distinctive (D1), moderately distinctive  
235 (D2), marginally distinct (D3) or not distinctive (D4) (Melancon et al., 2011; Urian et al., 2015).  
236 Once dorsal fin photographs were sorted and individuals identified, images were stored in  
237 *FinBase* (Microsoft Corporation, Redmond, WA, USA; Adams et al., 2006). In the subsequent  
238 analyses, only excellent and good quality dorsal fin images (Q1, Q2) of very and moderately  
239 distinct individuals (D1, D2) were used to avoid misidentification while still achieving high  
240 match rates (Tyson et al., 2022). This threshold used for distinctiveness also allowed exclusion  
241 of calves and neonates from analyses because of their dependence on their mothers (i.e.,  
242 photographic captures must be independent; Pollock et al., 1990).

243                    Finally, recent development of computer vision applications using artificial intelligence  
244 algorithms have proven their performance in assisting with dorsal fin matching (Tyson et al.,  
245 2022). Therefore, we used the semi-automated dorsal fin matching software *finFindR*  
246 (Thompson et al., 2022) to compare dorsal fin images collected during surveys with the photo-ID  
247 catalogue in development in Alabama waters which contained images of individuals previously  
248 identified during the project. Dorsal fin matching was performed in the laboratory by  
249 experienced researchers and each match was verified by a second, independent researcher not  
250 involved in the original matching (Urian et al., 2015). A third opinion was sought in cases of  
251 disagreement.

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253                    2.2.3) Environmental data collection

254                    In both embayments, from summer 2020 through summer 2022, at least one boat was  
255 equipped with a Dataflow data logging system, including a Manta multiprobe sonde (© Eureka  
256 Water Probes) to continuously collect environmental data on salinity, water temperature (°C),  
257 dissolved oxygen (mg/l) and pH at the water surface (~ 50 cm depth) while underway during



258 surveys and sightings. To obtain environmental data of homogenous resolution throughout the  
259 study areas, the Dataflow system was programmed to collect data every 5 seconds while on  
260 transect and cruising at higher speeds, and every 30 seconds during focal observations of  
261 dolphins at lower speeds. During the first season of sampling (winter 2019-20), the Dataflow  
262 system was not available and salinity and water temperature were recorded manually using a  
263 handheld data sonde (YSI® Professional Series Pro30) at gridded waypoints located at 9 km  
264 intervals in MOB and 7 km in PER along the survey transects.

265

## 266 2.3) Data analyses

### 267 2.3.1) Abundance estimates

268 As a first step in calculating abundance estimates, we tested the robustness of the photo-  
269 ID catalogue and the assumption of population closure for each embayment by computing  
270 discovery curves based on the cumulative counts of distinct individuals (D1, D2) in each  
271 embayment during the study period (Wilson et al., 1999). To compute the abundance estimates,  
272 we tested 16 full-likelihood parameterization RD models with temporary emigration constraints  
273 (e.g., Markovian and random movement), with constant (.) or time-varying (*t*) survival (*S*) and  
274 recapture probability (*p*). For each of these 16 models, we set up recapture probabilities equal to  
275 capture probabilities ( $p = c$ ) because dolphin photo-ID is a non-invasive method that should not  
276 change the behaviour of dolphins within a primary sampling period (Hammond, 2010).  
277 Markovian models allow for different immigration and emigration probabilities between primary  
278 sampling periods, while random movement models require equal probabilities between  
279 immigration and emigration (Kendall et al., 1997). Preliminary analyses investigating movement  
280 patterns of dolphins between embayments found that some individuals regularly move in and out  
281 of the two embayments (Bouveroux, pers. com.; Clance, 2022), hence, the “no movement”  
282 models were not computed.

283 Abundance estimates for each primary sampling period were produced solely from  
284 encounter histories of distinctive animals (D1, D2). To consider the number of unmarked  
285 individuals that do not have distinct dorsal fins in the population, we divided the abundance  
286 estimates based on distinct individuals by the proportion of distinct individuals encountered  
287 during each primary sampling period  $N_{\text{total}} = N_{\text{distinct}}/\theta$ , where  $N_{\text{total}}$  is the estimated total  
288 population size,  $N_{\text{distinct}}$  is the estimate of distinct individuals derived by the models, and  $\theta$  is the

289 estimated proportion of distinct individuals encountered within each primary sampling period  
290 (Wilson et al., 1999). We followed the delta method to extrapolate the 95% Confidence Interval  
291 (95% CI) of the total abundance (Wilson et al., 1999). The Akaike Information Criterion (AIC)  
292 was used to evaluate the best fitting model, by selecting the model having the lowest AIC value  
293 (Burnham & Anderson, 2004). Models were built in R (R Development Core Team, 2023) using  
294 the package *RMark* (Laake, 2013). Finally, the mean seasonal abundances were calculated by  
295 averaging the estimated total abundance derived within winter or summer seasons for each  
296 embayment. Potential effects of seasonal mean salinity and water temperature on seasonal  
297 abundance estimates of dolphins in both study areas were tested using a General Linear Model  
298 (GLM) in R (R Development Core Team, 2023) using the package *MASS* (Venables & Ripley,  
299 2022). Environmental analyses were limited to comparisons for salinity and water temperature,  
300 including their interactions because these were the dynamic variables for which data were  
301 available for the duration of the study period in both embayments. A backward selection was  
302 used to identify the most important variables predicting seasonal abundance, starting with all  
303 variables, and systematically removing variables one at a time with a null model run last. At each  
304 step, the variable that contributes the least to the model fit was removed, and the model was  
305 refitted. Models were ranked depending on the support of each variable using the AIC and  
306 Akaike weights ( $w$ ) values (Burnham & Anderson, 2004).

307

### 308 2.3.2) Resighting and residency pattern

309 We characterised the resighting patterns of distinct individuals encountered during the  
310 study period in each embayment by determining the total number of times each individual was  
311 encountered among the 18 potential secondary sampling occasions. Residency patterns were  
312 determined as the total number of distinctive animals encountered within the six primary  
313 sampling periods that represent seasons. Individual dolphins that were encountered in more than  
314 three primary sampling periods in a given estuary were considered resident animals (adapted  
315 from Rosel et al., 2011).

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### 320 3. RESULTS

#### 321 3.1) Survey summaries

322 In MOB, a total of 85 surveys were conducted, representing 6,599 km of survey effort  
323 and resulting in 309 dolphin sightings during which 2,845 animals were encountered, including  
324 263 calves and 23 neonates. In PER, a total of 38 surveys were conducted representing 2,412 km  
325 of survey effort, with 301 dolphin sightings and 1,523 animals encountered, including 216 calves  
326 and 42 neonates. In both embayments, neonates were only observed in summer (Table 1).

327 While the total number of dolphin sightings was similar between embayments, the  
328 encounter rates differed significantly. In MOB, the encounter rates were approximately 2 to 6  
329 times lower than in PER (0.02 - 0.07 sightings km<sup>-1</sup> and 0.12 - 0.14 sightings km<sup>-1</sup>, respectively;  
330 Kruskal-Wallis  $X^2 = 8.61$ ,  $df = 1$ ,  $p$ -value = 0.003; Table 1). The relative abundances were up to  
331 4 times lower in MOB (0.15 - 0.68 dolphins km<sup>-1</sup> and 0.54 - 0.80 dolphins km<sup>-1</sup>) compared to  
332 PER, but were overall not significantly different (Kruskal-Wallis  $X^2 = 3.70$ ,  $df = 1$ ,  $p$ -value =  
333 0.05; Table 1). The greatest difference in encounter rates and relative abundances between  
334 embayments occurred during the first season of sampling in winter 2019-20. In MOB, highest  
335 encounter rates occurred in winter 2020-21 and 2021-22 (Table 1). In contrast, in PER,  
336 encounter rates were similar between seasons, with the highest encounter rate recorded in  
337 summer 2021. In both embayments, the highest relative abundance occurred during summer  
338 2022 (0.68 dolphins km<sup>-1</sup> and 0.80 dolphins km<sup>-1</sup>, respectively, for MOB and PER; Table 1).  
339 Finally, no seasonal difference was observed within embayments in dolphin encounter rates  
340 (MOB: Kruskal-Wallis  $X^2 = 0.07$ ,  $df = 1$ ,  $p$ -value = 0.796; PER: Kruskal-Wallis  $X^2 = 0.43$ ,  $df = 1$ ,  
341  $p$ -value = 0.512) and relative abundances (MOB: Kruskal-Wallis  $X^2 = 0.05$ ,  $df = 1$ ,  $p$ -value =  
342 0.827; PER: Kruskal-Wallis  $X^2 = 0.44$ ,  $df = 1$ ,  $p$ -value = 0.506).

343

#### 344 3.2) Discovery curves and abundance estimates

345 During the study period, a total of 673 distinctly marked individuals (D1, D2) were  
346 identified in MOB, and 189 distinctly marked individuals were identified in PER (Figure 2). The  
347 discovery curve, indicating the number of newly identified individuals encountered during each  
348 seasonal period, increased during the study for both MOB and PER, with the highest number of  
349 new animals documented through the first three seasons in MOB (through winter 2020-21) and  
350 only during the first season in PER (winter 2019-20) (Figure 2). After these peak periods, the

351 number of newly identified animals each season remained near 100 for MOB (ranging 79 to 112)  
352 and 20 for PER (ranging 11 to 31).

353 Among the 16 RD models we tested, random movement in MOB and Markovian  
354 movement in PER, with constant survival over time and time-varying recapture probability in  
355 both areas had the lowest corrected AIC (Tables S1 and S2). Based on encounter histories, we  
356 estimated a mean seasonal abundance of  $738 \pm 112.67$  dolphins in winter and  $1,469 \pm 152.99$   
357 dolphins in summer in MOB (Table 2). In PER, the mean seasonal abundance varied between  
358  $179 \pm 6.09$  in winter and  $130 \pm 16.39$  in summer. The abundance of bottlenose dolphins in each  
359 embayment varied with season, but in opposite directions. Depending on the seasons, the total  
360 abundance of dolphins in MOB was three to ten times higher than the abundance estimated in  
361 PER (Table 2). The lowest abundance in either embayment was found during the 2019-2020  
362 survey year (MOB winter:  $n = 518$ ; 95% CI: 260-1,032; PER summer:  $n = 100$ ; 95% CI: 81-122,  
363 Figure 3), and the highest abundances were found during the 2022 survey year (MOB summer:  $n =$   
364  $1,712$ ; 95% CI: 1,520-1,928; PER winter:  $n = 191$ ; 95% CI: 157-232; Figure 3, Table 2). In  
365 both areas, survival estimates were constant through time and estimated at  $0.93 \pm 0.02$  (95% CI =  
366  $0.89-0.95$ ) in MOB and  $0.92 \pm 0.02$  (95% CI =  $0.88-0.95$ ) in PER.

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### 368 3.2.1) Effects of environmental attributes on abundances

369 In all, we collected 127,126 environmental data points in MOB and 88,760 in PER  
370 between the winter 2019-20 and summer 2022. The mean surface salinity was generally low in  
371 both embayments during the study periods ( $< 19$  ppt; Table 3) and was significantly lower in  
372 MOB compared to PER (*one-way ANOVA*:  $F = 10.80$ ;  $df = 1$ ;  $p$ -value = 0.0082). Salinity in  
373 MOB varied significantly between seasons, ranging from  $3.53 \pm 3.51$  in summer 2021 to  $14.76 \pm$   
374  $4.79$  in winter 2021-22 (*one-way ANOVA*:  $F = 11,705$ ;  $df = 5$ ;  $p$ -value  $< 0.001$ ; Table 3; Figure  
375 S1). The mean salinity recorded in PER was also significantly different between seasonal periods  
376 (*one-way ANOVA*:  $F = 4,493$ ;  $df = 5$ ;  $p$ -value  $< 0.001$ ), ranging between  $17.04 \pm 6.12$  in winter  
377 2021-22 and  $18.39 \pm 5.67$  in winter 2020-21, but decreased to  $10.10 \pm 7.14$  in summer 2022  
378 (Table 3, Figure S1). The mean sea surface temperatures did not vary between embayments (*one-*  
379 *way ANOVA*:  $F = 0.04$ ;  $df = 1$ ;  $p$ -value = 0.85). In MOB, water temperature was significantly  
380 colder in winter compared to summer (*one-way ANOVA*:  $F = 699,509$ ;  $df = 5$ ;  $p$ -value  $< 0.001$ )  
381 and ranged from  $12.23^\circ\text{C} \pm 1.24$  in winter 2019-20 and  $31.76 \pm 1.03$  °C in summer 2022. In

382 PER, water temperature was also significantly colder in winter compared to summer (*one-way*  
383 *ANOVA*:  $F = 896,123$ ;  $df: 5$ ;  $p\text{-value} < 0.001$ ) and ranged from  $11.32 \pm 1.30$  °C in winter 2020-  
384 21 and  $29.81 \pm 1.08$ °C in summer 2022 (Figure S1). Among the models testing the effects of  
385 salinity and water temperature on dolphin abundance, only water temperature was driving the  
386 seasonal abundance within each embayment (MOB: AIC weight = 0.44, t value = 5.23, p-value <  
387 0.001; PER: AIC weight = 0.78, t value = -2.98, p-value = 0.04). In MOB, the increase in water  
388 temperature had a positive influence on the abundance of dolphins, while the opposite was  
389 observed in PER (Table S3).

390

### 391 3.3) Resighting and residency patterns

392 During the 18 possible resighting occasions, the maximum number of sightings of  
393 individual dolphins was 8 in MOB (n = 2 dolphins) and 13 in PER (n = 1 dolphin). The number  
394 of dolphins seen once was high in both areas (Figure 4A), with about 40% of the dolphins in  
395 MOB (n = 266 of 673) and 32% of the dolphins in PER (n = 61 of 189) sighted only once.  
396 Almost 40% of dolphins in MOB (n = 265) and 50% of those in PER (n = 94) were encountered  
397 at least three times (Figure 4A).

398 The residency pattern of dolphins revealed that MOB had fewer resident animals  
399 (~ 28% of the distinct individuals) compared to PER (~ 40%) (Figure 4B). In MOB, these  
400 residents were comprised of 188 distinctive dolphins that were resighted during three seasons or  
401 more and 57 individuals (9%) that were encountered during four seasons or more. In PER,  
402 residents included 76 individuals resighted during at least three seasons and 52 dolphins (28%)  
403 resighted during four seasons or more. About 45% of the distinct individuals in MOB (n = 301)  
404 and 35% of those in PER (n = 67) were encountered in only one season. One dolphin was  
405 encountered during every season in MOB, and five dolphins were encountered in every season in  
406 PER (Figure 4B). Finally, out of a total of 862 distinctive dolphins identified in MOB (n = 673)  
407 and PER (n = 189), 15 individuals (1.74 %) were occasionally encountered in both areas during  
408 the study period.

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413 **4. DISCUSSION**

414 CMR photo-ID surveys were conducted in two embayments along the Alabama coastline,  
415 MOB and PER, providing the first estimates of bottlenose dolphin abundance and survival rates  
416 as well as a quantification of residency patterns for these two stocks. This study also provided an  
417 opportunity to assess RD modelling techniques in two embayments characterised by different  
418 physical and chemical attributes. Our study revealed that the number of dolphins inhabiting  
419 MOB is larger than PER and much larger than historically documented (1993 aerial surveys;  
420 Blaylock & Hoggard, 1994). Among the 32 recognized BSE dolphin stocks of the northern  
421 GOMx, only 11 have recent (since 2016) abundance estimates (Hayes et al., 2023). When  
422 comparing the abundances of these dolphin stocks and their respective waterbody surface area,  
423 MOB contains one of the largest populations (e.g., Hayes et al., 2023). We also provided the first  
424 abundance estimate of bottlenose dolphins in PER and revealed that it has similar (or slightly  
425 higher) abundances than similar sized bays along the northern GOMx, such as St. Andrew Bay  
426 and Sarasota Bay (e.g., B. Balmer et al., 2018; Tyson & Well, 2016). The estimated survival  
427 rates of dolphins in the study areas were similar to those reported for other bottlenose dolphin  
428 populations. Specifically, they were consistent or slightly lower than survival rates observed in  
429 Charleston, SC ( $S = 0.95$ , Speakman et al., 2010), Sarasota Bay, FL ( $S = 0.96$ , Wells & Scott,  
430 1990), and Barataria Bay, LA ( $S = 0.97$ , McDonald et al., 2017) within the Gulf of Mexico.  
431 Overall, the higher abundance estimates observed in MOB and the significant seasonal increase  
432 in abundance during summer, as well as the residency patterns of dolphins characterized in both  
433 areas, highlight a greater importance than previously expected of these Alabama embayments to  
434 BSE dolphin stocks in the northern GOMx.

435 Abundance and residency patterns of bottlenose dolphin populations are usually  
436 influenced by their habitats, where smaller populations typically have a higher degree of  
437 residency in protected and enclosed habitats (Irwin & Würsig, 2004; Shane et al., 1986; Wells et  
438 al., 1987) compared to semi-open or open habitats (Ballance, 1992; Mullin et al., 2017; Zanardo  
439 et al., 2016). Accordingly, the two studied embayments had distinctly different seasonal  
440 abundances and residency patterns. MOB, which is a large, semi-open and highly dynamic  
441 embayment with several connections to surrounding dolphin stocks, had a higher estimated  
442 abundance of dolphins but fewer resident individuals. In contrast, the smaller semi-enclosed bay  
443 of PER, with smaller connections to surrounding stocks and relatively stable marine environment

444 throughout seasons, had lower abundances and higher residency. Interestingly, the seasonal  
445 residency patterns were also opposing in these embayments, with the abundance of dolphins  
446 higher in summers in MOB and higher during winters in PER. Even with high interannual  
447 variation (e.g., 95% CI, winter 2019-20), this pattern was strongly persistent. The recruitment  
448 rates of newly identified dolphins entering these two embayments showed similar seasonal  
449 trends, with a larger number of new individuals encountered during summers in MOB and  
450 winters in PER. In MOB, the non-asymptotic discovery curve and the higher proportion of  
451 distinct animals seen once, coupled with the low proportion of animals resighted among seasons,  
452 indicates a higher percentage of transient dolphins are likely mixing with resident dolphins in  
453 this more open embayment compared to PER. Such co-occurrence between transient and resident  
454 dolphins have been previously reported in many areas along the Florida coastline (Wells et al.,  
455 1987; Quintana-Rizzo & Wells, 2001; Shane, 2004) and in San Luis Pass, Texas (Maze &  
456 Würsig, 1999). This co-occurrence observed in MOB is likely promoted by the large pass  
457 connecting with the GOMx on the southern end of the estuary, where larger number of individual  
458 transient dolphins were probably encountered, such as documented in Galveston Bay and  
459 Aransas Pass in Texas (Bräger, 1993; Weller, 1998).

460 Seasonal variation in dolphin abundance has been described in other BSE dolphin stocks  
461 along the northern GOMx as well (Balmer et al., 2008, 2019; Fazioli et al., 2006; Mullin et al.,  
462 2017; Toms, 2019). These variations were related to non-resident visitors from neighbouring  
463 systems, including the adjacent Northern Coastal Stock, which occasionally or seasonally utilises  
464 inshore embayments (B. Balmer et al., 2018; Conn et al., 2011; Fazioli et al., 2006; Irvine et al.,  
465 1981). The large increase of abundance in the MOB estuary during summer is likely associated  
466 with temporary immigration of dolphins from adjacent waters moving into the area, rather than  
467 by unmarked individuals (i.e., subadults and calves) becoming marked throughout the study  
468 period. MOB has close connections with neighbouring dolphin stocks (Hayes et al., 2023), such  
469 as the Mississippi Sound/Lake Borgne/Bay Boudreau Stock (3,469 dolphins, 95% CI: 3,113 -  
470 3,725; Mullin et al., 2017), and the Northern Coastal Stock (11,543 dolphins, CV = 0.19;  
471 Garrison et al., 2021). Preliminary results from intensive biopsy sampling surveys conducted  
472 concurrently with this study in the eastern part of Mississippi Sound, MOB, PER and in the  
473 GOMx off the coast of Alabama, found evidence of movements among these areas (Bouveroux,  
474 pers. com; Clance, 2022). Additionally, movement patterns of tagged dolphins in the eastern part

475 of MSS revealed that some individuals travel widely throughout the southwestern portion of  
476 MOB, where it also connects with the coastal water of the GOMx (Cloyed et al., 2021b). These  
477 combined results indicate connectivity among embayments and potential for frequent  
478 interactions among dolphin stocks in this region.

479         Seasonal movements into embayments have been previously described in eastern regions  
480 of the northern GOMx (e.g., St. Andrew Bay and St. Joseph Bay: Balmer et al., 2008, 2018;  
481 Bouveroux et al., 2014), but the specific drivers of these movements are not well understood.  
482 During this study, water temperature was the major environmental parameter related to seasonal  
483 abundances of dolphins in Alabama waters. Because the seasonal abundance patterns were  
484 opposite relative to temperature between embayments, however, it is likely that temperature  
485 alone is not the primary driver but may be a covariate or proxy for another as yet unidentified  
486 factor or combination of factors affecting dolphin movements in the area. A common theory is  
487 that seasonal variation in dolphin abundance could follow seasonal prey movements and  
488 variation in local resource availability. We do not have embayment-specific data on seasonal  
489 variation in prey abundance to test this theory in our study embayments. In MOB, however, the  
490 shrimp fishery is one of the most important commercial fisheries and most active in summer  
491 (GULF, 2021; Loesch, 1976). Bottlenose dolphins are often associated with shrimp boats and  
492 other fishing vessels, and they take advantage of the lower energy cost of feeding activities from  
493 depredation in trawling gears or discarded prey items (Genov et al., 2019; Greenman & Mcfee,  
494 2023; Lorenz, 2015; Shane et al., 1986). Shrimp are found in the stomachs of dolphins stranded  
495 in MOB (Russell pers. com., Alabama Marine Mammal Stranding Network), and fishery-  
496 associated feeding behaviours were frequently recorded during our surveys in MOB (authors  
497 pers. obs.). In contrast, shrimping is limited year-round in PER due to tighter restrictions on  
498 shrimping activities in Florida waters, which make up about half of PER. Therefore, animals  
499 from the Northern Coastal Stock and MSS might benefit from the seasonal increase of available  
500 resources and move into MOB during summer, which can partially explain differences in  
501 abundance observed between the two study areas during this time.

502         Because our study was concurrently conducted in two adjacent areas under otherwise  
503 similar conditions, we were also able to further test relationships to environmental attributes. The  
504 large volumes of cold and freshwater discharge that seasonally drain into MOB can dominate the  
505 system during certain years (e.g., Carmichael et al., 2012; NOAA NMFS, 2022). During this



506 study salinity at the surface in both embayments did not appear to influence the abundance of  
507 dolphins. It is important to note, however, that the salinity during this study was recorded at the  
508 surface and not in the water column. MOB can have a strong vertical salinity stratification, with  
509 twice the surface values at the bottom, especially in the low-wind conditions required during our  
510 surveys (Kim & Park, 2012). Therefore, it is likely that our data do not fully capture the effects  
511 of salinity on dolphin abundance; for example, animals may have benefited from higher salinity  
512 water at depth as a refuge, particularly in the deep MOB ship channel. During winter 2019-20,  
513 however, almost all of MOB had an extremely low surface salinity regime, which combined with  
514 colder winter temperatures (the coldest documented during this study) could explain the low  
515 number of sightings and animals encountered, as well as their greater southern distribution  
516 (Bouveroux et al., in prep.). As described in other areas, these poor ecological conditions for  
517 dolphins likely promoted temporary dolphin movements out of the study area (Fazioli &  
518 Mintzer, 2020; McBride-Keibert & Toms, 2021). In Barataria Bay (Louisiana), a salinity  
519 threshold of ~ 8 ppt was identified as a critical area of preferred dolphin habitat, and animals  
520 were rarely seen in salinity waters below that isohaline (Hornsby et al., 2017). Dolphins can  
521 tolerate low salinity waters (< 10 ppt) for short periods of time (Takeshita et al., 2021), but  
522 longer or chronic exposure can dramatically impact dolphins' health (from generalised skin  
523 pallor to ulcerative lesions, corneal oedema, electrolyte imbalance) and contribute to mortality  
524 (Deming et al., 2020; Ewing et al., 2017). From February to November 2019, between Louisiana  
525 and the Florida Panhandle, 337 bottlenose dolphins died stranded on beaches, and an UME was  
526 declared in this region. Most of the investigated carcasses exhibited skin lesions consistent with  
527 freshwater exposure; 10% of these strandings occurred in Alabama waters (NOAA NMFS,  
528 2022). Therefore, the combined effect of low salinity water, low water temperatures, and higher  
529 mortality rates in the region likely resulted in the lowest abundance of dolphins recorded in  
530 MOB in winter 2019-20. In contrast, temporal fluctuations of salinity were less pronounced and  
531 mostly limited to the north easternmost part of PER. The stability of dolphin habitat in PER may  
532 have contributed to the smaller seasonal variation in abundance and higher residency patterns for  
533 this population. Long-term monitoring and higher resolution of environmental sampling will help  
534 better resolve the drivers of seasonal habitat use and residency patterns for these dolphin stocks.

535 Comparing dolphin abundance and residency patterns on a regional scale, across  
536 embayments with different environmental and geomorphic features, presents methodological and

537 analytical challenges. To compare results between embayments and potentially reveal seasonal  
538 patterns of dolphin abundance along the northern GOMx, it is essential to follow the same or  
539 very similar methodologies among sites. Most of the research conducted in this region used  
540 CMR photo-ID surveys and RD modelling methods (Balmer et al., 2008, 2019; Bouveroux et al.,  
541 2014; Litz et al., 2019; Tyson et al., 2011) and this study followed that approach. This study  
542 achieved high capture probabilities  $> 0.10$ , which indicates limited bias in the estimates (Otis et  
543 al., 1978). There was one exception to these capture probabilities, the MOB sampling period  
544 during winter 2019-20 (probabilities 0.05-0.10), when environmental conditions were poor.  
545 Refinements to develop more complex models that include spatial information on detections to  
546 estimate abundance, survival, and recruitment through time and consider individual  
547 heterogeneity in capture probabilities and mortality, would be an important step for future  
548 research. Such methodologies will help refine abundance and survival estimates, particularly for  
549 large and open areas with dynamic habitats. Despite these challenges, the approach used for this  
550 study appears robust.

551         This study provides much needed baseline data on abundance and residency patterns for  
552 bottlenose dolphins in Alabama waters. The two studied embayments support a large number of  
553 dolphins, in particular MOB, and likely provide seasonally different resources, indicating high  
554 and potentially complex use of these estuaries. Although survival rates were sufficient for  
555 sustainable population growth, but lower than estimated in other BSE stocks (e.g., McDonald et  
556 al., 2017; Speakman et al., 2010; Wells & Scott, 1990), these two populations should be closely  
557 monitored because the region is subject to high human activities and environmental variation.  
558 Further research is ongoing in both embayments to characterise dolphin distribution and habitat  
559 selection and define dolphin hotspots. Biopsy samples that were collected after each season of  
560 CMR surveys are being used to analyse genetic stock structure, assess dolphin health through  
561 contaminant levels and investigate dolphin diet from stable isotope analyses. These combined  
562 data will provide comprehensive information to complement ongoing photo-ID and abundance  
563 surveys in these embayments and others in support of effective conservation plans for bottlenose  
564 dolphins throughout their range.

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580

581 **6. DATA AVAILABILITY STATEMENT**

582 The datasets presented in this study can be found online at Dauphin Island Sea Lab data center:  
583 <https://doi.org/10.57778/fsqw-pj98>

584

585 **7. ETHICS AND PERMIT APPROVAL STATEMENT**

586 This study was conducted under National Marine Fisheries Service (NMFS) Scientific  
587 Research Permit No. 23772 issued to Dauphin Island Sea Lab (DISL), and No. 21938 issued to  
588 SEFSC-NOAA. Research protocols used were approved by the Institutional Animal Care and  
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590

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595

596 **9. CONFLICT OF INTEREST DISCLOSURE**

597 The authors declare they are not aware of any conflict of interest.

598

599 **10. REFERENCES**

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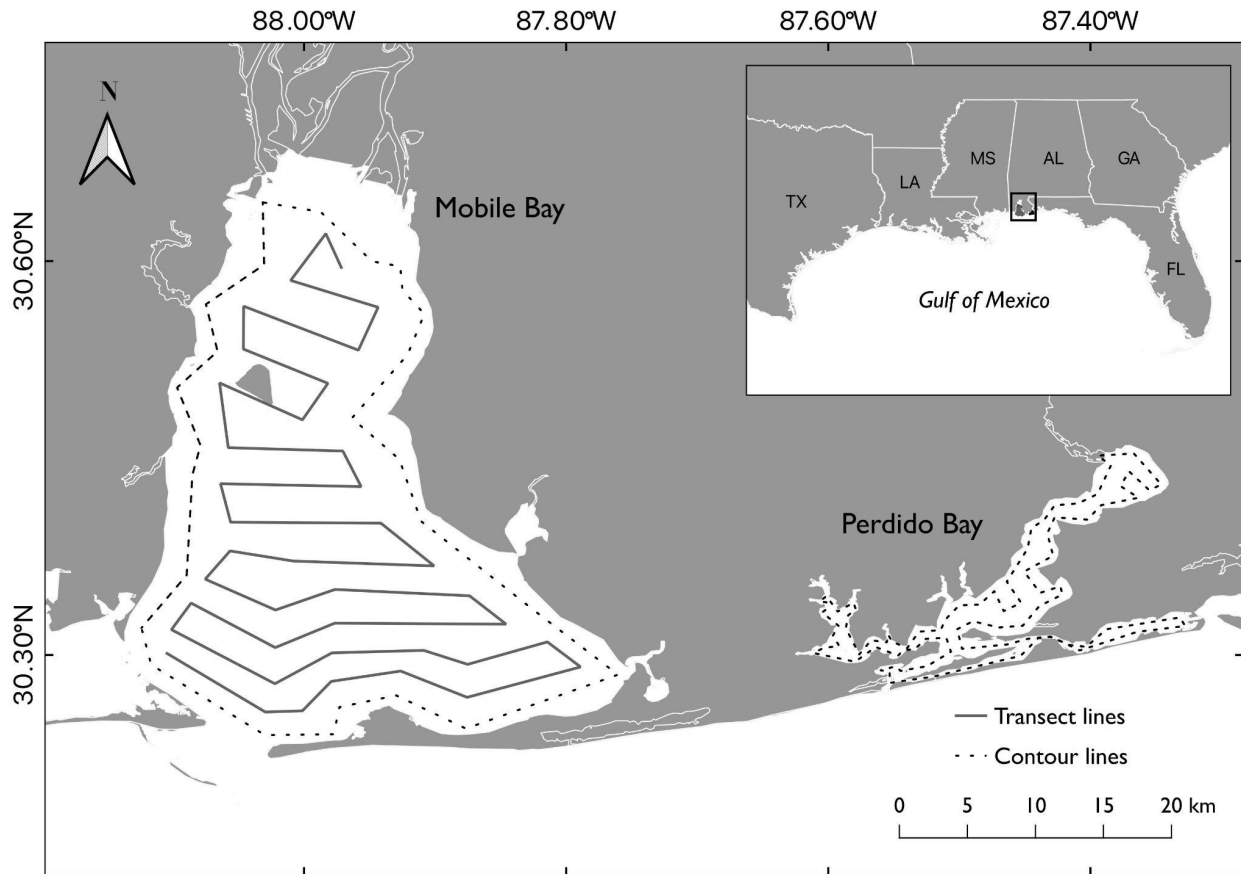
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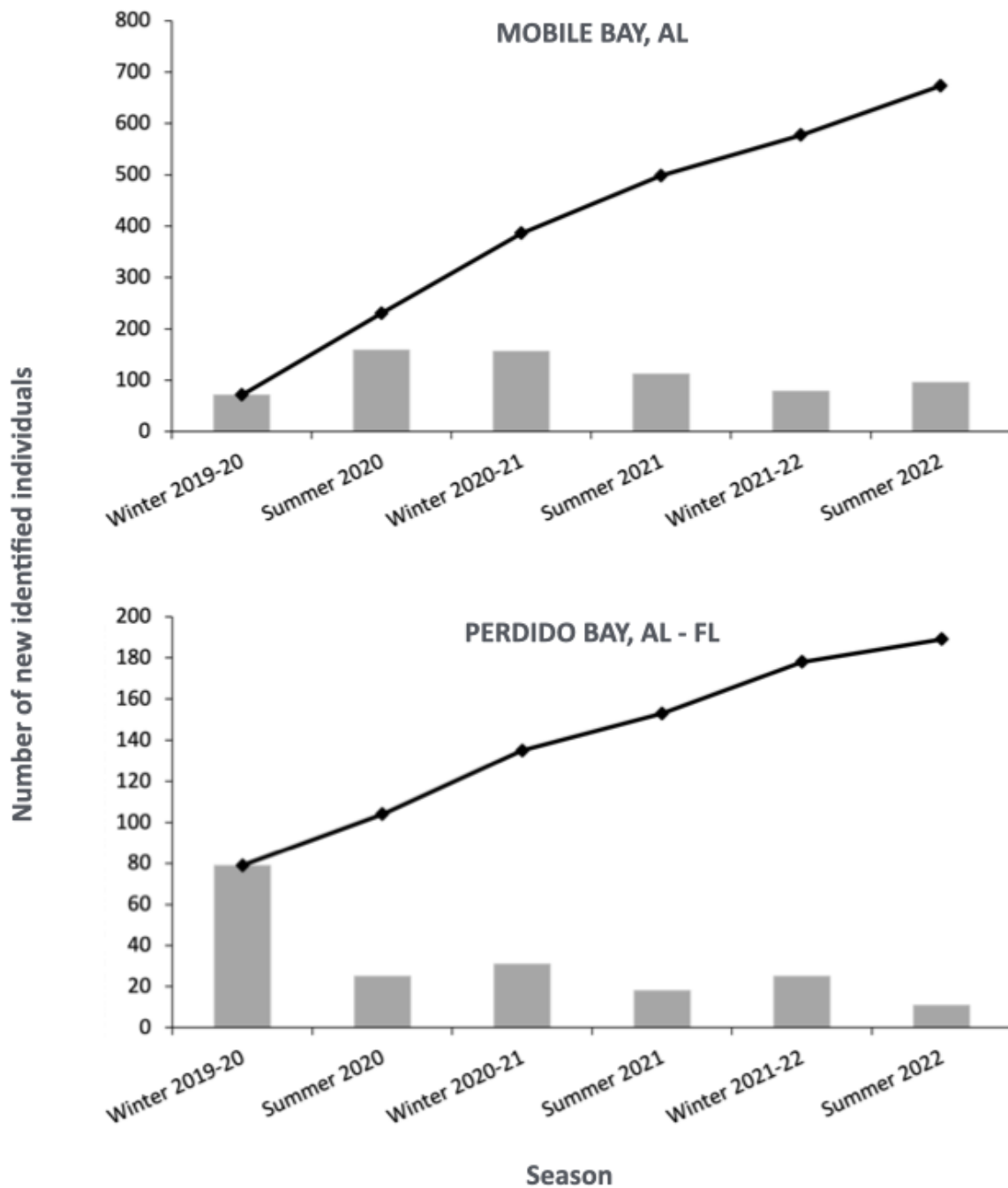
**Figures**



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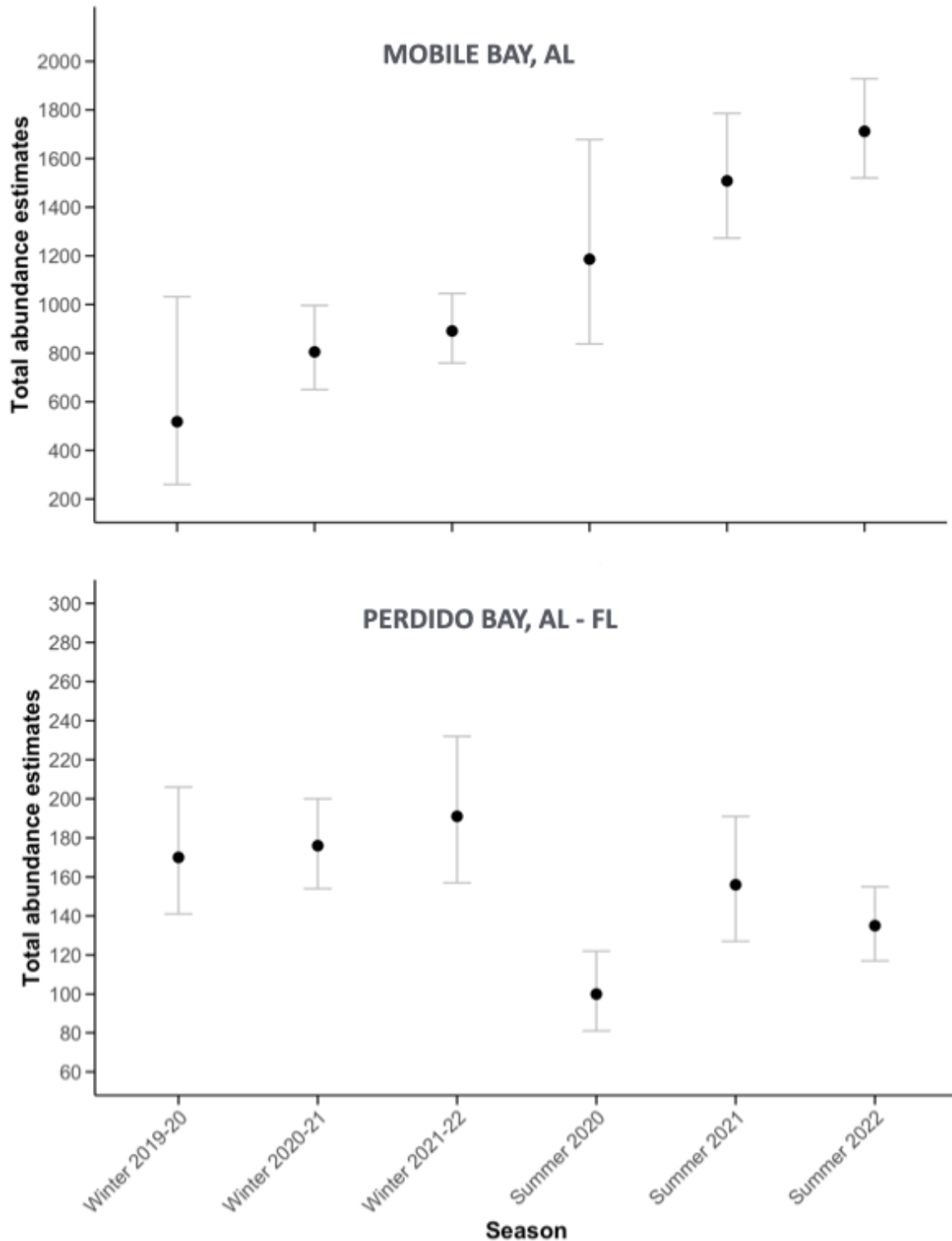
**Figure 1.** Two adjacent study areas in Alabama waters, Mobile Bay (left) and Perdido Bay (right), that were surveyed to collect mark-recapture photo-identification data on common bottlenose dolphins (*Tursiops truncatus*) following either transect lines (solid black lines) and/or contour lines (dashed black lines).





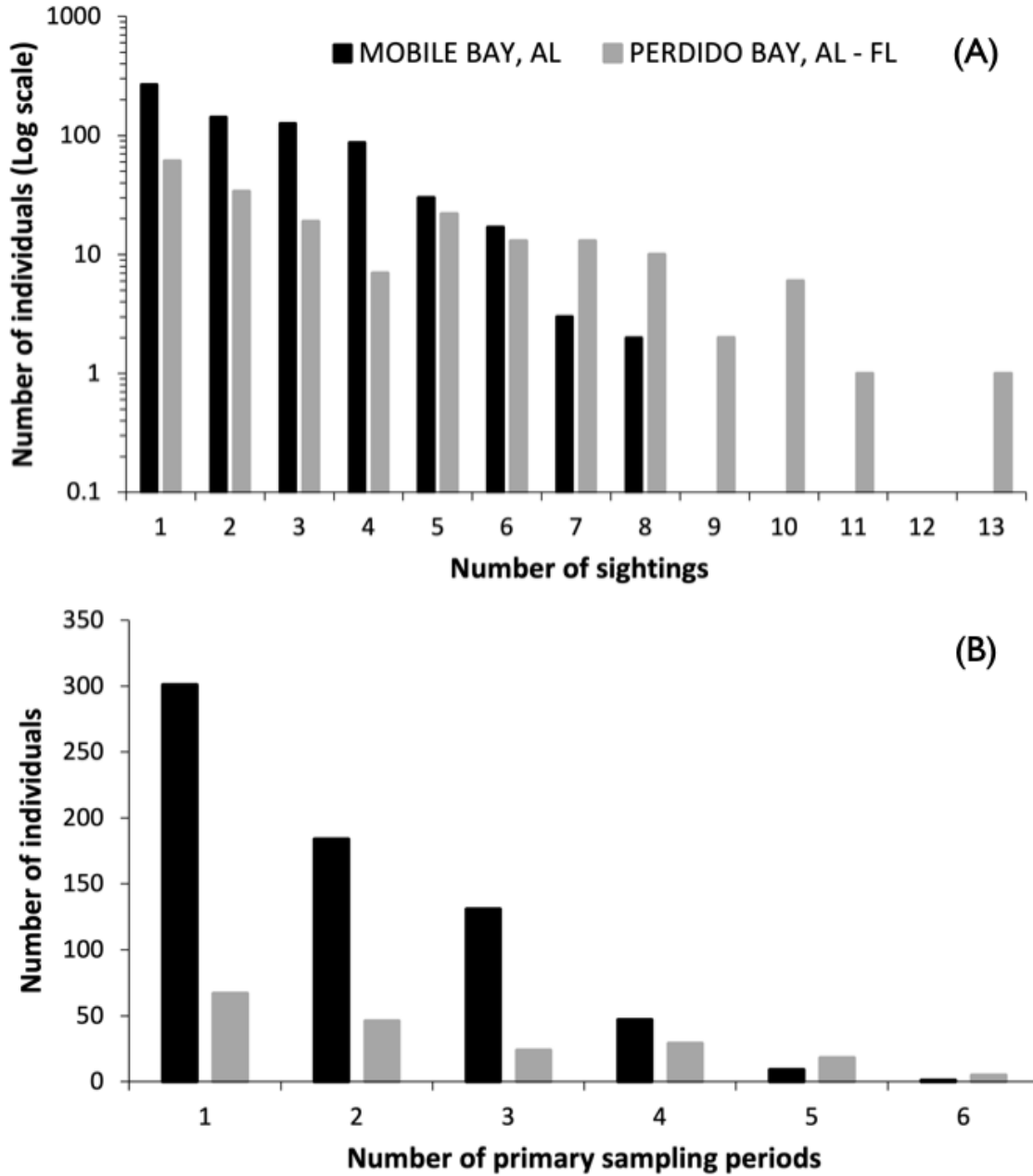
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**Figure 2.** The number of newly identified common bottlenose dolphins in each survey season (primary sampling occasions) in Mobile Bay (N = 673) and Perdido Bay (N = 189), and the resulting discovery curve showing the cumulative number of newly identified distinct individual dolphins (D1, D2), during the study period (winter 2019-20 to summer 2022).



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 964 **Figure 3.** Seasonal abundance estimates of common bottlenose dolphins in in Mobile Bay and Perdido Bay, derived  
 965 from Pollock's robust design full-likelihood  $p$  and  $c$  approach for each season (primary sampling period). Best fitting  
 966 model for Mobile Bay (MOB) =  $\{S(.) p(t) \text{ random emigration}\}$ , best fitting model for Perdido Bay (PER) =  $\{S(.) p(t)$   
 967  $\text{Markovian emigration}\}$ . The estimates include unmarked individuals (N total), and the vertical grey bars represent  
 968 the 95% confidence interval for each estimate.

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**Figure 4.** (A) Number of distinct individual common bottlenose dolphins (expressed in log scale) encountered during the study period (winter 2019-20 to summer 2022); (B) number of distinct individuals encountered per number of survey seasons (i.e., number of primary sampling periods) in Mobile Bay and Perdido Bay.

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**Tables**

**Table 1.** Photo-identification survey effort for each study embayment during the primary sampling periods (seasons) between winter 2019-20 and summer 2022 showing distances on effort, encounter rates (sightings km<sup>-1</sup>), relative abundance (dolphins km<sup>-1</sup>), number of surveys and sightings, as well as number of encountered dolphins, calves, and neonates of common bottlenose dolphins. MOB = Mobile Bay, PER = Perdido Bay.

<b>Study area</b>	<b>Season</b>	<b>Distance or effort (km)</b>	<b>Encounter rates</b>	<b>Relative abundance</b>	<b>No. of surveys</b>	<b>No. of sightings</b>	<b>No. of dolphins</b>	<b>No. of calves</b>	<b>No. of neonates</b>
<b>MOB</b>	Winter 2019-20	1,086.4	0.02	0.15	13	26	162	20	0
	Winter 2020-21	1,096.3	0.06	0.47	13	70	519	48	0
	Winter 2021-22	1,047.2	0.07	0.55	14	74	580	74	0
	Summer 2020	1,111.5	0.03	0.36	13	37	404	46	0
	Summer 2021	1,148.8	0.04	0.38	14	41	431	36	9
	Summer 2022	1,109.2	0.05	0.68	18	61	749	39	14
<b>Sub-total</b>		<b>6,599.4</b>	<b>-</b>	<b>-</b>	<b>85</b>	<b>309</b>	<b>2,845</b>	<b>263</b>	<b>23</b>
<b>PER</b>	Winter 2019-20	386.4	0.12	0.62	7	48	238	38	0
	Winter 2020-21	394.2	0.13	0.75	6	52	295	44	0
	Winter 2021-22	417.2	0.12	0.56	6	50	232	25	0
	Summer 2020	407.8	0.12	0.54	7	47	221	41	3
	Summer 2021	414.7	0.14	0.54	6	56	224	26	12
	Summer 2022	391.8	0.12	0.80	6	48	313	42	27
<b>Sub-total</b>		<b>2,412.1</b>	<b>-</b>	<b>-</b>	<b>38</b>	<b>301</b>	<b>1,523</b>	<b>216</b>	<b>42</b>
<b>TOTAL</b>	<b>-</b>	<b>9,011.5</b>	<b>-</b>	<b>-</b>	<b>123</b>	<b>610</b>	<b>4,368</b>	<b>479</b>	<b>65</b>

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990 **Table 2.** Abundance estimates (N distinct) of common bottlenose dolphins for Mobile Bay, derived from Pollock's  
 991 robust design full-likelihood  $p$  and  $c$  approach for each season (primary sampling period). Best fitting model for  
 992 MOB =  $\{S(.) p(t) \text{ random emigration}\}$ , best fitting model for PER =  $\{S(.) p(t) \text{ Markovian emigration}\}$ . SE = standard  
 993 errors, CI = 95% confidence interval. The table gives the proportion of distinct individuals (D1, D2), the total  
 994 abundance estimate including unmarked individuals (N total), and the overall mean abundance in Winter and  
 995 Summer.

Primary sampling period	N distinct	SE	95% CI (N distinct)	Proportion of distinct IDs	N total	95% CI (N total)	Mean abundance ± SE
<b>MOB - <math>\{S(.) p(t) \text{ random emigration}\}</math></b>							
Winter 2019-20	306	110.91	168 - 637	0.59	518	260 - 1,032	738 ± 112.67
Winter 2020-21	483	52.63	399 - 608	0.60	805	650 - 996	
Winter 2021-22	535	43.67	461 - 634	0.60	891	759 - 1,045	
Summer 2020	546	97.38	399 - 790	0.46	1,186	838 - 1,678	1,469 ± 152.99
Summer 2021	663	57.41	565 - 791	0.44	1,508	1,273 - 1,786	
Summer 2022	582	35.36	522 - 662	0.34	1,712	1,520 - 1,928	
<b>PER - <math>\{S(.) p(t) \text{ Markovian emigration}\}</math></b>							
Winter 2019-20	109	10.72	95 - 139	0.64	170	141 - 206	179 ± 6.09
Winter 2020-21	109	7.36	99 - 129	0.62	176	154 - 200	
Winter 2021-22	149	14.99	126 - 187	0.78	191	157 - 232	
Summer 2020	83	8.64	71 - 107	0.83	100	81 - 122	130 ± 16.39
Summer 2021	106	10.95	90 - 135	0.68	156	127 - 191	
Summer 2022	108	7.63	97 - 128	0.80	135	117 - 155	

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1007 **Table 3.** Mean and range (min - max) seasonal variation of the surface salinity ( $\pm$  SD) and water temperature ( $\pm$  SD)  
 1008 for Mobile Bay (MOB) and Perdido Bay (PER) collected on effort during photo-identification surveys between  
 1009 winter 2019-20 and summer 2022.

Study area	Season	Salinity (ppt)	Min - Max (ppt)	Water temp. ( $^{\circ}$ C)	Min - Max ( $^{\circ}$ C)
<b>MOB</b>	Winter 2019-20	3.72 $\pm$ 3.19	0.10 - 13.10	12.23 $\pm$ 1.24	8.40 - 16.00
	Winter 2020-21	11.51 $\pm$ 4.84	0.14 - 25.56	15.07 $\pm$ 1.56	12.30 - 20.55
	Winter 2021-22	14.76 $\pm$ 4.79	2.15 - 31.00	17.22 $\pm$ 1.03	13.92 - 20.14
	Summer 2020	12.01 $\pm$ 6.29	0.08 - 31.36	30.30 $\pm$ 1.09	23.11 - 33.55
	Summer 2021	3.53 $\pm$ 3.51	0.06 - 20.65	29.52 $\pm$ 1.46	24.12 - 34.67
	Summer 2022	8.89 $\pm$ 3.58	0.51 - 24.25	31.76 $\pm$ 1.03	27.74 - 34.71
<b>PER</b>	Winter 2019-20	18.12 $\pm$ 5.78	4.30 - 28.00	16.75 $\pm$ 1.96	13.80 - 21.20
	Winter 2020-21	18.39 $\pm$ 5.67	0.10 - 30.31	11.32 $\pm$ 1.30	6.42 - 19.93
	Winter 2021-22	17.04 $\pm$ 6.12	0.18 - 30.16	14.33 $\pm$ 1.27	9.82 - 18.39
	Summer 2020	18.11 $\pm$ 7.07	3.17 - 34.13	29.81 $\pm$ 0.89	27.79 - 32.88
	Summer 2021	17.99 $\pm$ 6.51	2.23 - 31.22	28.31 $\pm$ 0.80	24.82 - 32.53
	Summer 2022	10.10 $\pm$ 7.14	0.05 - 28.08	29.81 $\pm$ 1.08	25.63 - 34.37

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**Supplementary Material**

**Table S1:** Model selection details of Robust Design candidate models for Mobile Bay testing for Markovian and random temporary emigration ( $\gamma''$ ,  $\gamma'$ ) with combined estimated parameters for survival (S) and capture (p) probabilities. Models are ranked by lowest corrected Akaike Information Criterion (AICc). Delta AICc is the difference between a model and the best ranked model. No. parameters = number of estimated parameters.

Model	No. parameters	AICc	$\Delta$ AICc
S(.) $\gamma'' = \gamma'$ p(time)	29	-5675.301	0.000000
S(t) $\gamma'' = \gamma'$ p(time)	33	-5672.253	3.047545
S(.) $\gamma'' \neq \gamma'$ p(time)	32	-5672.156	3.144824
S(t) $\gamma'' \neq \gamma'$ p(time)	36	-5666.624	8.677216
S(.) $\gamma'' = \gamma'$ p(1sess)	17	-5569.585	105.715433
S(t) $\gamma'' = \gamma'$ p(1sess)	21	-5566.859	108.441557
S(.) $\gamma'' \neq \gamma'$ p(1sess)	20	-5561.964	113.336829
S(t) $\gamma'' \neq \gamma'$ p(1sess)	24	-5561.088	114.212995
S(t) $\gamma'' = \gamma'$ p(2sess)	18	-5548.320	126.981172
S(.) $\gamma'' \neq \gamma'$ p(2sess)	17	-5547.540	127.760933
S(.) $\gamma'' = \gamma'$ p(2sess)	14	-5544.576	130.725117
S(t) $\gamma'' \neq \gamma'$ p(2sess)	21	-5544.470	130.831057
S(t) $\gamma'' = \gamma'$ p(.)	16	-5535.447	139.853347
S(.) $\gamma'' \neq \gamma'$ p(.)	15	-5534.768	140.533010
S(.) $\gamma'' = \gamma'$ p(.)	12	-5531.676	143.624543
S(t) $\gamma'' \neq \gamma'$ p(.)	19	-5529.678	145.622369

S denotes survival; p probability of capture and recapture ( $p = c$ );  
 $\gamma'' = \gamma'$  random emigration model;  $\gamma'' \neq \gamma'$  Markovian emigration model;  
 (.) denotes constancy over time; (t) vary by time;  
 (1sess) vary by primary sessions (e.g., season); (2sess) vary by secondary sessions but constant between primary sessions.

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**Table S2:** Model selection details of Robust Design candidate models for Perdido Bay, testing for Markovian and random temporary emigration ( $\gamma''$ ,  $\gamma'$ ) with combined estimated parameters for survival (S) and capture (p) probabilities. Models are ranked by lowest corrected Akaike Information Criterion (AICc). Delta AICc is the difference between a model and the best ranked model. No. parameters = number of estimated parameters.

Model	No. parameters	AICc	$\Delta$ AICc
S(.) $\gamma'' \neq \gamma'$ p(time)	32	-601.3892	0.000000
S(t) $\gamma'' \neq \gamma'$ p(time)	36	-599.7946	1.594558
S(t) $\gamma'' = \gamma'$ p(time)	33	-597.6384	3.750829
S(.) $\gamma'' = \gamma'$ p(time)	29	-590.2232	11.166009
S(.) $\gamma'' \neq \gamma'$ p(1sess)	20	-587.8069	13.582285
S(t) $\gamma'' \neq \gamma'$ p(1sess)	24	-586.6826	14.706555
S(t) $\gamma'' = \gamma'$ p(1sess)	21	-584.1481	17.241069
S(.) $\gamma'' \neq \gamma'$ p(.)	15	-583.3657	18.023545
S(t) $\gamma'' \neq \gamma'$ p(.)	19	-580.5405	20.848701
S(.) $\gamma'' \neq \gamma'$ p(2sess)	17	-579.8266	21.562595
S(t) $\gamma'' \neq \gamma'$ p(2sess)	21	-576.9509	24.438339
S(t) $\gamma'' = \gamma'$ p(.)	16	-576.4426	24.946575
S(.) $\gamma'' = \gamma'$ p(1sess)	17	-576.3465	25.042695
S(.) $\gamma'' = \gamma'$ p(.)	12	-573.5119	27.877291
S(t) $\gamma'' = \gamma'$ p(2sess)	18	-572.8887	28.500501
S(.) $\gamma'' = \gamma'$ p(2sess)	14	-570.0076	31.381615

S denotes survival; p probability of capture and recapture ( $p = c$ );  
 $\gamma'' = \gamma'$  random emigration model;  $\gamma'' \neq \gamma'$  Markovian emigration model;  
 (.) denotes constancy over time; (t) vary by time;  
 (1sess) vary by primary sessions (e.g., season); (2sess) vary by secondary sessions but constant between primary sessions.

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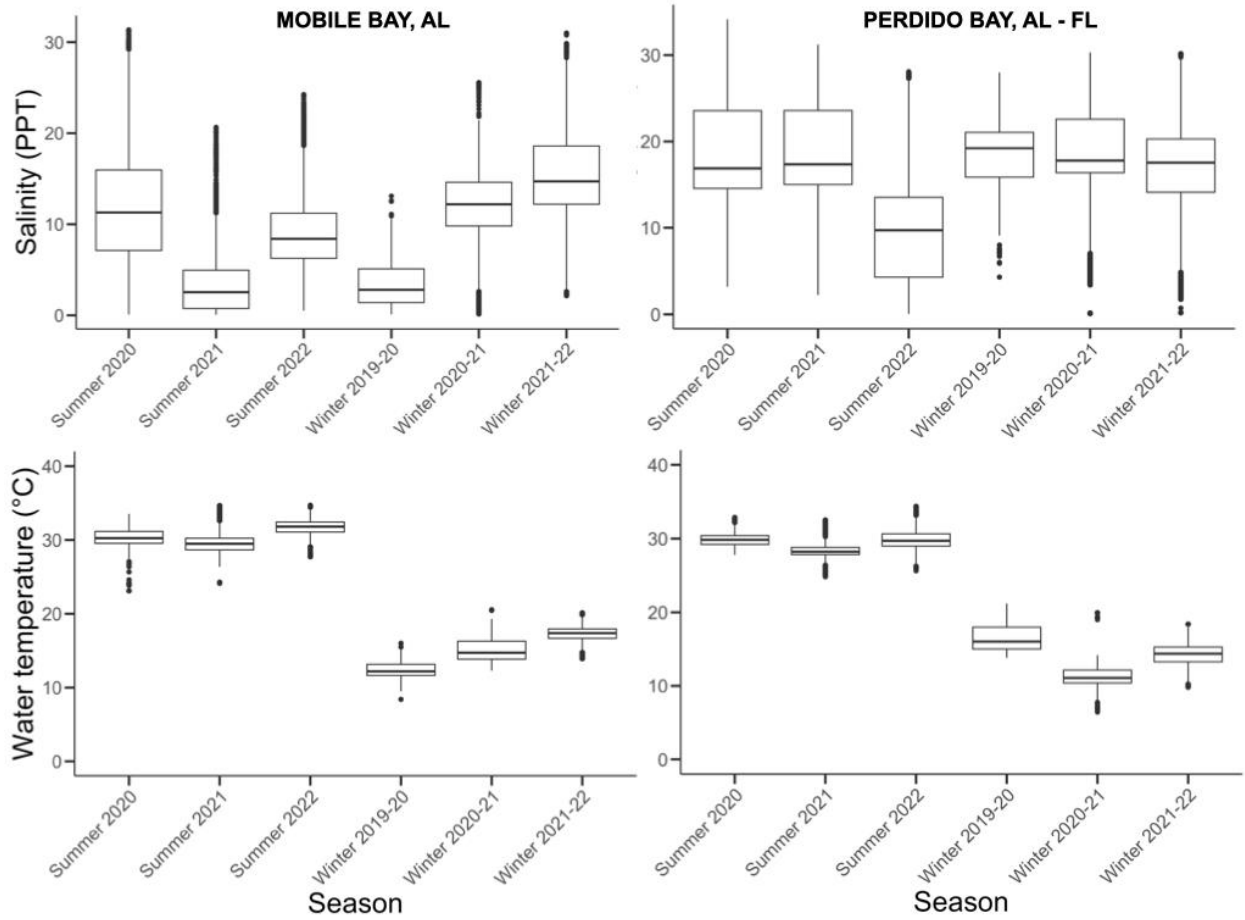


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**Table S3.** Statistics for the environmental parameters of models for Mobile Bay (MOB) and Perdido Bay (PER). SE = standard error, t = t-values, p = p-values, AIC = Akaike information criterion, and  $\Delta$ AIC = delta Akaike information criterion, w = Akaike weights.

	<b>Parameters</b>	<b>Estimate</b>	<b>SE</b>	<b>t</b>	<b>p</b>	<b>AIC</b>	<b><math>\Delta</math>AIC</b>	<b>w</b>	
<b>MOB</b>	(Intercept)	-490.46	500.36	-0.98	0.4303				
	Salinity	63.80	54.37	1.17	0.3615	83.11	0.19	0.39	
	Temp	74.68	22.41	3.33	0.0795				
	Salinity:Temp	-3.33	2.53	-1.32	0.3190				
	(Intercept)	48.53	320.11	0.15	0.8890				
	Salinity	-3.67	20.09	-0.18	0.8670	84.85	1.93	0.17	
	Temp	47.97	10.57	4.54	0.0200				
	(Intercept)	12.87	221.06	0.06	0.9564	82.92	0.00	0.44	
	Temp	48.07	9.19	5.23	0.0064				
		<b>Parameters</b>	<b>Estimate</b>	<b>SE</b>	<b>t</b>	<b>p</b>	<b>AIC</b>	<b><math>\Delta</math>AIC</b>	<b>w</b>
	<b>PER</b>	(Intercept)	768.84	837.27	0.92	0.4550			
		Salinity	-30.24	46.73	-0.65	0.5840	59.35	2.53	0.22
Temp		-20.79	28.26	-0.74	0.5380				
Salinity:Temp		0.97	1.58	0.61	0.6020				
(Intercept)		257.79	76.41	3.37	0.0433				
Salinity		-1.67	3.52	-0.47	0.6683	84.85	28.04	0.00	
Temp		-3.48	1.33	-2.61	0.0798				
(Intercept)		224.06	24.75	9.05	0.0008	56.82	0.00	0.78	
Temp		-3.20	1.07	-2.98	0.0408				

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**Figure S1:** Seasonal variation of salinity (top panels) and water temperature (bottom panels) recorded by the Dataflow continuous data logger (© Eureka Water Probes) during sampling in Mobile Bay (n = 127,126) and Perdido Bay (n = 88,760), including the median (black horizontal lines), lower and upper quartile and outlier (black circles) values.