

1 **Mixed stock origin of Atlantic bluefin tuna in the U.S. rod and reel fishery (Gulf of Maine)**  
2 **and implications for fisheries management**

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17

18 **Abstract**

19 The highly migratory Atlantic bluefin tuna (*Thunnus thynnus*) has a distribution that spans the  
20 North Atlantic and two distinct spawning populations, an eastern population originating in the  
21 Mediterranean Sea and a western population originating in the Gulf of Mexico. Atlantic bluefin  
22 tuna are managed as two separate management units (east and west) in the North Atlantic,  
23 despite observed mixing that occurs across the management boundary. Characterizing the effects  
24 of stock mixing has been identified as a priority for improving the management of Atlantic  
25 bluefin tuna. Identifying the stock composition of landings from the Gulf of Maine is of  
26 particular importance, because approximately 70% of the U.S. western Atlantic total allowable  
27 catch is removed from this region annually. The aim of our research was to apply otolith  
28 chemistry techniques to characterize the origin of bluefin tuna caught in the U.S. rod and reel

29 fishery in the Gulf of Maine and to demonstrate how this information can be applied in fisheries  
30 management. Prior research established otolith stable isotope chemistry ( $\delta^{13}\text{C}$  and  $\delta^{18}\text{O}$ ) as an  
31 effective and reliable stock identification tool, and we applied this approach to determine the  
32 population of origin of bluefin tuna collected from fishery dependent sampling (recreational and  
33 commercial) in the Gulf of Maine. Results indicated that the majority of fish caught in the Gulf  
34 of Maine from 2010 to 2013 were eastern origin. We found the highest proportion of eastern  
35 origin fish were caught in 2012 and the proportion of eastern origin fish was greater in late  
36 summer to fall. Although the majority of fish in small and intermediate size classes were eastern  
37 origin, fish in the largest size class (>250 cm) were predominantly western origin. Using these  
38 data, we demonstrated an approach for integrating mixed stock composition information into  
39 fishery-specific harvest data (U.S. rod and reel catch, catch-at-age, and catch-per-unit-effort).  
40 This information can be used to monitor mixed stock composition of the fishery, partition catch  
41 to population of origin, and to inform management decisions aimed at controlling population of  
42 origin harvest.

43

44 **Keywords:** Atlantic bluefin tuna, stock composition, stock mixing, otolith chemistry, fisheries  
45 management

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## 47 **1. Introduction**

48 The Atlantic bluefin tuna (*Thunnus thynnus*) is a large, highly migratory species with a  
49 distribution that spans the North Atlantic. Initially considered to be a single panmictic stock, two  
50 distinct spawning populations, an eastern population originating in the Mediterranean Sea and a  
51 western population originating in the Gulf of Mexico, were recognized and accounted for in the  
52 stock assessment process in the 1980s through the creation of separate management units. The  
53 International Commission for the Conservation of Atlantic Tunas (ICCAT) management  
54 boundary divides east and west stocks at the 45°W meridian, and the current assessment used for  
55 management purposes assumes no mixing occurs across this boundary (Anon., 2017). In recent  
56 decades, a suite of research methods has improved our understanding of Atlantic bluefin tuna  
57 movement and stock mixing. Stock identification methods, including genetics, otolith chemistry,  
58 conventional and electronic tagging, and organochlorine data, indicate that bluefin tuna  
59 populations exhibit a high rate of natal homing, as well as a high degree of stock mixing,  
60 particularly during juvenile and sub-adult life stages (Lutcavage et al., 1999, 2001; Block et al.,  
61 2005; Rooker et al., 2008a,b; Dickhut et al., 2009; Galuardi et al., 2010; Rodríguez-Ezpeleta et  
62 al., 2019). In addition, another spawning ground was recently identified in the Slope Sea of the  
63 western Atlantic, however the origin of these spawners is currently unknown (Richardson et al.,  
64 2016). Mismatch in the scale of Atlantic bluefin tuna populations' movements and stock units  
65 has important implications to the sustainable management of this species (Kerr et al., 2017).

66 Recent stock assessments estimate significant differences in the biomass of bluefin tuna stocks,  
67 with the eastern stock estimated to be an order of magnitude greater than the western stock  
68 (Anon., 2017). Because of these relative differences in biomass, even low movement rates of  
69 eastern origin fish into western Atlantic waters can exert significant influence on the biomass and

70 stock composition of bluefin tuna in the western stock area (Secor et al., 2015; Kerr et al., 2017).  
71 The combined effect of asymmetric size of bluefin tuna populations and the assumption of no  
72 mixing in current stock assessments can result in an overly optimistic perception of western  
73 bluefin tuna biomass (Kerr et al., 2017). This influence can result in a decoupling of the stock  
74 and population view of the resource. In this case, the stock view provides insight on the  
75 availability of fish to the fishery within the stock area, regardless of origin, and the population  
76 view provides insight regarding the unit of production and sustainability. Both views are needed,  
77 but the current approach to assessment and management for Atlantic bluefin tuna assumes the  
78 stock and population view are equivalent and this can confound sustainable management.

79 Alternative approaches have been proposed to account for the impact of stock mixing on the  
80 stock assessment and management process for Atlantic bluefin tuna. Integrating the movement  
81 dynamics of bluefin tuna has been explored through fitting spatially explicit stock assessment  
82 models that incorporate tagging and stock composition data (e.g., dual zone virtual population  
83 analysis [VPA-2BOX], Porch et al., 2001; the multi-stock age structured tag integrated model  
84 [MAST], Taylor et al., 2011; the modifiable multi-stock model [M3], Anon., 2017); however,  
85 these have not been considered reliable for providing the basis of management advice to date.

86 Modeling spatial population dynamics is statistically demanding because it requires the  
87 estimation of many additional parameters to capture both seasonal and ontogenetic spatial  
88 dynamics (Punt, 2019). Furthermore, the tagging and stock composition data (based on genetics  
89 and otolith chemistry) used to fit these statistical models is limited for certain regions, fisheries,  
90 and for historical periods (Morse et al., 2018). In addition, information from different methods  
91 can provide different perspectives on mixing that integrate across different time scales. For  
92 example, natural markers, such as otoliths, and genetic markers provide distinct ecological and

93 evolutionary perspectives on population connectivity (Reis-Santos et al., 2018). Work is ongoing  
94 to determine the best approach for integration of this information within the context of ICCAT's  
95 management strategy evaluation of bluefin tuna (Anon., 2017). However, in the meantime, this  
96 information is not used to inform monitoring or management of bluefin tuna populations.

97 Analysis of otolith chemistry using stable isotopes ( $\delta^{13}\text{C}$  and  $\delta^{18}\text{O}$ ) has been established as an  
98 effective stock identification tool, providing answers to critical stock structure questions for  
99 Atlantic bluefin tuna (Rooker et al., 2008a,b; Schloesser et al., 2010; Secor et al., 2013, 2014a,  
100 2015; Rooker et al., 2014; Siskey et al., 2016). The water chemistry in the two principal nursery  
101 regions differs significantly and leaves distinct chemical signatures in the otoliths of tuna. Fish  
102 inhabiting the cooler, more saline waters of the eastern nursery (Mediterranean Sea) exhibit  
103 elevated  $\delta^{13}\text{C}$  and  $\delta^{18}\text{O}$  values compared fish from the western nursery habitat (NW Atlantic  
104 shelf; Rooker et al., 2014). By analyzing the chemical composition of otoliths, we can assign fish  
105 to their respective population of origin based on this nursery signature. Otolith chemistry of  
106 archived bluefin tuna otolith samples has been used to estimate historical and recent stock  
107 mixing levels, which depend on the region sampled within the Atlantic, as well as fish size, life-  
108 stage, and year-class (see review by Rooker and Secor 2019).

109 Stock composition analysis of fish collected in the Gulf of Mexico and Mediterranean Sea  
110 revealed that nearly all of these fish (~ 100%) originate from their respective spawning  
111 populations, indicating natal homing and little to no mixing on the spawning grounds (Rooker et  
112 al., 2008a,b, 2014). Limited stock mixing is also evident in the eastern and central Atlantic, with  
113 fish predominantly of eastern origin (Rooker et al., 2014). However, analysis of fish caught in  
114 the U.S. western Atlantic suggests extensive mixing of eastern and western origin fish within this  
115 region (Rooker et al., 2008b; Siskey et al., 2016; Anon., 2017). Stock composition of

116 commercial size (>185 cm CFL) bluefin tuna landed in the Gulf of Maine indicated this size  
117 class was predominately (95%) western origin fish in the 1990s (Rooker et al., 2008a). However,  
118 the sample size was relatively small (n=72), temporally short (1996 and 1998), and collected  
119 from a geographically restricted area within the Gulf of Maine (Ipswich Bay), and from one size  
120 class of fish (“giants”, i.e. >140 kg,  $\geq$  age 10; Rooker et al., 2008a). More recent work by Siskey  
121 et al. (2016) described changes in mixing of eastern and western origin bluefin tuna in the Gulf  
122 of Maine over recent decades, with fish being nearly entirely western origin in the 1970s and  
123 2010s, but with greater representation of eastern origin fish (41%) in the 1990s. These variable  
124 results for the Gulf of Maine suggests there is a need for higher resolution sampling in this  
125 region of dynamic stock mixing to characterize stock composition and evaluate any differences  
126 across size classes, time, and location.

127 The goal of our research was to characterize the stock composition of bluefin tuna landed in the  
128 U.S. rod and reel fishery in the Gulf of Maine (2010-2013) using otolith chemistry techniques  
129 and to demonstrate how results could be used to inform fisheries management. The concentration  
130 of landings in the Gulf of Maine, approximately 70% of the entire U.S. bluefin tuna quota  
131 allocation, highlights the importance of understanding mixing dynamics of bluefin tuna in this  
132 region on a finer spatial and temporal scale.

## 133 **2. Material and Methods**

### 134 *2.1. Sample Selection*

135 Atlantic bluefin tuna otoliths were sampled from an archived otolith collection held at the Gulf  
136 of Maine Research Institute and collected by the University of Maine. The otoliths were  
137 dissected from tuna heads that were donated by commercial and recreational fishermen who

138 typically discard the heads when landing fish. This collection is the most comprehensive archive  
139 to date for Atlantic tuna landed in the Gulf of Maine with samples that are representative of the  
140 recreational and commercial components of the U.S. rod and reel fishery. Samples that had the  
141 most associated data (e.g., length, sex, location of capture) were preferentially selected for otolith  
142 chemistry analysis. Samples were selected across the duration of the fishing season (June –  
143 October), as well as across a spectrum of size classes to represent the fishery and account for  
144 isotopic differences by time of year or age of fish. A total of 782 otoliths were analyzed for  
145 stable isotope chemistry in this study.

## 146 *2.2. Otolith Preparation and Analysis*

147 Sagittal otoliths were removed from sampled fish, cleaned of adhering tissue, and dried. One  
148 randomly selected otolith from each pair was embedded in a fast curing high visibility epoxy  
149 (EpoHeat, Buehler Inc.). A transverse section (1.5 mm) which included the nucleus and distal  
150 knob (protuberance) of the otolith was prepared using a Buehler IsoMet 1000 Precision Saw and  
151 attached to a sample plate using SPI Crystalbond 509 thermoplastic glue with the nucleus side  
152 up. The region corresponding to the first year of growth, which was identified from  
153 measurements of transverse sections of otoliths from yearling bluefin tuna (Rooker et al., 2014),  
154 was isolated and powdered using a New Wave Research MicroMill (Fig. 1). A series of 14 drill  
155 passes at 55  $\mu\text{m}/\text{sec}$  and 55  $\mu\text{m}$  depth per pass were run over a pre-programmed drill path by a  
156 500  $\mu\text{m}$  diameter carbide drill bit until a depth of approximately 770  $\mu\text{m}$  was reached. Powder  
157 was collected with a microspatula, folded into wax paper, and sealed into plastic sample vials.  
158 Processing tools were cleaned with 95% EtOH to prevent cross contamination among specimens.

159 Otolith powder was analyzed for  $\delta^{13}\text{C}$  and  $\delta^{18}\text{O}$  using an automated carbonate preparation  
160 coupled to a gas chromatograph – isotope ratio mass spectrometer (Finnigan MAT 252; Thermo  
161 Fisher Scientific, Inc.) at the University of Arizona Environmental Isotope Laboratory. Powdered  
162 samples were reacted with dehydrated phosphoric acid under vacuum at 70°C. Analytical  
163 precision of the mass spectrometer has been measured at  $\pm 0.1\%$  for  $\delta^{18}\text{O}$  and  $\pm 0.06\%$  for  $\delta^{13}\text{C}$  (1  
164 standard deviation; Schloesser et al., 2010). Isotope ratios were calibrated based on repeated  
165 measurements of NBS-19 and NBS-18 and reported relative to the Pee Dee Belemnite (PDB)  
166 standard.  $\delta^{13}\text{C}$  values were corrected for the Suess effect:

$$167 \text{Suess Effect}_{13\text{C}} = \beta_{13\text{C}} * (\text{Year of Baseline} - (\text{Year of Capture} - \text{Age})) \quad (1)$$

168 where year of baseline (2006) was chosen as the modal year over which the baseline sample was  
169 collected (1998-2011; Siskey et al., 2016) and  $\beta_{13\text{C}}$  was calculated as the slope of  $\delta^{13}\text{C}$  data in  
170 this study plotted against year class  $\beta_{13\text{C}} = -0.0299$ .

171 The corresponding otolith of each pair was sectioned (0.8 mm thickness) along a transverse  
172 plane on the distal side of the nucleus and using a series of gritted papers polished for ageing  
173 purposes. Otolith preparation conventions and age interpretation procedures followed the  
174 standardized procedures adopted by past international efforts (Secor et al., 2014b; Busawon et  
175 al., 2014). Blind counts of annuli were conducted twice from the images using Adobe Photoshop  
176 CS2 Version 9.0. When counts differed by <2 years, the second count was accepted. When  
177 counts differed by >2 years, the image was inspected a third time along with the two previous  
178 annulus assignments.

### 179 2.3. Mixed Stock Analysis

180 Population assignment methods use stock identification information to ascertain population  
181 membership of individuals (individual assignment) or groups of individuals (mixture analysis;  
182 Manel et al., 2005). Alternative methods can support different inferences because of different  
183 assumptions. Mixture analysis estimates the proportion of individuals that each source  
184 population contributes to a mixed sample, whereas individual assignment classifies each  
185 individual to the population with the highest likelihood and estimated mixture proportions are the  
186 sum of the individual assignments (Manel et al., 2005). We applied both mixture and individual  
187 assignment methods to understand the influence on conclusions regarding mixed stock  
188 composition of bluefin tuna in the Gulf of Maine.

189 Mixture analysis, specifically a maximum likelihood mixture model, has been applied in the  
190 majority of previous studies on Atlantic bluefin tuna stock composition (Rooker et al., 2008a,b;  
191 Schloesser et al., 2010; Secor et al., 2013, 2014a, 2015; Rooker et al., 2014; Fraile et al., 2015;  
192 Siskey et al., 2016). We applied a conditional maximum likelihood mixture model (CMLE) fit  
193 using the program HISEA (PC executable of FORTRAN program) as described by Millar  
194 (1990a,b), as this is the most frequently applied mixture assignment approach for continuous  
195 characters. This procedure fits the mixture distribution based on the source population  
196 distributions of the characters and possible mixing proportions in the unknown sample. Mixture  
197 analysis (CMLE) provided mixed stock composition for all samples in terms of the mean and  
198 standard deviation of the proportion of eastern and western origin fish based on bootstrapping  
199 with 10,000 simulations. Individual assignment is increasingly being applied in otolith chemistry  
200 studies, and ICCAT has adopted this approach in compiling a database of stock of origin  
201 information on bluefin tuna across the Atlantic (Anon., 2017). The application of individual  
202 assignment requires a decision on both the statistical approach (e.g., discriminant analysis,

203 random forest, or other machine learning methods) and the appropriate threshold probability  
204 level for assignment. Results can vary based on both of these choices. Individual probabilities of  
205 belonging to the eastern or western population were assigned using the random forest technique  
206 (RF). RF is based on classification trees that are built from a random bootstrap resampling (with  
207 replacement) of the data. One of the benefits of this approach is that there are no *a priori*  
208 distributional assumptions (Mercier et al., 2011). In comparative analyses with other common  
209 statistical approaches, such as linear discriminant analysis and neural network analysis, RF was  
210 shown to be a powerful tool for classification based on otolith chemistry (Mercier et al., 2011).  
211 The RF approach provided assignment of individuals based on an associated probability. In  
212 assigning individual probabilities, we adopted a threshold probability level of 0.7 based on  
213 ICCAT's thresholds for assignment wherein the probability level for assignment to eastern or  
214 western origin was  $\geq 0.7$  and some individuals were not able to be assigned to origin (Anon.,  
215 2017). Individual assignment analysis was conducted in the R programming environment (R  
216 Development Core Team, 2018).

217 Population assignment for Atlantic bluefin tuna relied on a baseline of yearling (age 1) samples.  
218 These were collected in both principal nursery systems during the past 15 years (Rooker et al.,  
219 2014). One assumption of the mixture analysis is that the predictor variables (i.e., stable isotope  
220 ratios) are independent. Oxygen and carbon stable isotopic ratios are driven by different  
221 processes, with  $\delta^{18}\text{O}$  influenced by water temperature and salinity at the time of otolith  
222 precipitation, and  $\delta^{13}\text{C}$  ratios reflecting metabolic sources and ontogenetic changes in rates  
223 (Trueman et al., 2012). Previous investigations of this baseline have demonstrated temporal  
224 stability of  $\delta^{18}\text{O}$ , the primary marker for bluefin tuna classification (Rooker et al., 2008b; Rooker  
225 et al., 2014). Classification success based on linear discriminant function analysis of the most up-

226 to-date baseline is 90% east and 75% west (overall 83% classification accuracy; Rooker et al.,  
227 2014). Specimens of unknown origin were compared with yearling baseline samples and  
228 classified to eastern or western nurseries using both mixture classification and individual  
229 assignment approaches.

230 We characterized the stock composition of Atlantic bluefin tuna in the Gulf of Maine by year  
231 (2010-2013), month of capture (June-October), general size classes (<150, 150-199, 200-249,  
232 >250 cm curved fork length [CFL]), age classes (ages <4, 4-6, 7-9, 10-12, 13-15, 16-18,  $\geq$ 19),  
233 and by sex. In addition, we characterized stock composition by U.S. rod and reel size categories  
234 66-114 cm (small school), 115-144 cm (large school), and >177 cm (combined large medium-  
235 large school).

#### 236 *2.4 Integration of Mixed Stock Information in Fisheries Data*

237 We developed a data parsing approach for integrating stock of origin information into fishery-  
238 specific harvest data for the purpose of monitoring and informing management decisions. The  
239 data parsing approach involves revising the fishery-dependent data, including catch, catch-at-  
240 age, and catch-per-unit-effort (CPUE) collected for each size category of the fishery to reflect  
241 mixed stock composition information. Thus, instead of assigning a fishery's catch series to either  
242 the eastern or western stock, as currently assumed by ICCAT assessments, total catch biomass  
243 from each fishery is proportionally assigned to eastern and western populations based on stock  
244 composition information (Kerr et al., 2017; Morse et al., 2018). This approach requires  
245 identification of samples analyzed for mixed stock composition that are representative of the  
246 timing and location of fishery operation, as well as the size classes targeted by the fishery.

247 We illustrate the application of this approach to U.S. rod and reel recreational fishery data and  
 248 commercial fishery data which is categorized by size classes: 66-114 cm (small school), 115-144  
 249 cm (large school), and >177 cm (combined large medium-large school) from the 2017 ICCAT  
 250 stock assessment (Anon., 2017). We utilized mixed stock composition from this study, as well as  
 251 a previous study (Siskey et al., 2016; note results were re-estimated using the RF approach for  
 252 consistency), to inform parsing of U.S. rod and reel fishery data. Size category-specific ( $i$ ) catch,  
 253 catch-at-age ( $a$ ), and CPUE data representing the mixed western stock ( $ws$ ) were revised by  
 254 removing the proportion of eastern origin fish caught in western fisheries to derive western  
 255 origin ( $wo$ ) data series. The revised size category-specific catch ( $C_{wo,i,y}$ ), catch-at-age ( $C_{wo,i,y,a}$ ),  
 256 and U.S. rod and reel combined catch-at-age ( $C_{wo,y,a}$ ) of western origin fish were calculated as:

$$257 \quad C_{wo,i,y,a} = C_{ws,i,y,a} (P_{i,y,a}) \quad (2)$$

$$258 \quad C_{wo,i,y} = \sum_a^{a_{max}} C_{wo,i,y,a} \quad (3)$$

$$259 \quad C_{wo,y,a} = \sum_i C_{wo,i,y,a} \quad (4)$$

260 where  $C_{ws,i,y,a}$  is the annual size category-specific catch-at-age of fish by western stock fisheries,  
 261 and  $P_{i,y,a}$  is the annual age-specific proportion of western origin fish in mixed stock catches by  
 262 the fishery. Size category-specific CPUE of western mixed stocks ( $U_{ws,i,y}$ ) were adjusted  
 263 downward to derive CPUE of western origin fish ( $U_{wo,i,y}$ ), using the proportion of western origin  
 264 fish for catch at age and weight at age ( $w_{y,a}$ ):

$$265 \quad U_{wo,i,y} = U_{ws,i,y} \frac{\sum_a^{a_{max}} P_{i,y} C_{ws,i,y,a} w_{y,a}}{\sum_a^{a_{max}} C_{ws,i,y,a} w_{y,a}} \quad (5)$$

### 266 3. Results

#### 267 3.1. Sample Characteristics

268 In total, 782 otoliths were analyzed from Atlantic bluefin tuna caught in the Gulf of Maine from  
269 2010 to 2013 (2010 = 221, 2011 = 255, 2012 = 149, 2013 = 157, Table 1). Fish were landed  
270 from June to November and descriptive location information was provided. Specific location  
271 information from a subsample of fish indicated that catches ranged from Nantucket,  
272 Massachusetts (~41°N) to Mid Coast Maine (~44°N; Fig. 2). Fish ranged in size from 84 to 305  
273 cm CFL with the majority of samples within the 150 to 199 and 200 to 249 cm CFL length bin  
274 categories (Table 2). Fish ranged in age from 1 to 26 years, with 47% between the ages of 8 and  
275 10, and birth years ranging from 1986 to 2012. Sex was determined for 479 individuals (female =  
276 175, male = 304, Table 2). Across U.S. rod and reel size categories, the majority of samples were  
277 within the largest size category (US RR 66-114 cm = 24, US RR 115-144 cm = 41, US RR >177  
278 cm = 412, Table 3).

### 279 *3.2. Alternative Approaches to Mixed Stock Analysis*

280 The classification accuracy of the mixture analysis and individual assignment methods were  
281 similar based on the baseline classification accuracy. Baseline classification accuracy of  
282 yearlings to eastern and western nurseries using mixture analysis (quadratic discriminant  
283 function analysis) was 90 and 73%, respectively (overall 83%; Rooker et al., 2014).  
284 Classification accuracy of yearlings to their respective eastern and western nurseries using the  
285 RF approach was also 83%. Between the two methods, the RF approach consistently provided  
286 lower estimates of eastern contribution compared to the mixture analysis. The choice of cutoff  
287 value of 0.7 for probability of assignment using the RF method meant that fish with eastern and  
288 western origin probabilities less than 0.7 were not classified. This resulted in differences in  
289 sample size difference between the two methods. However, when all samples were assigned  
290 through the RF approach the results varied as well.

291        *3.3. Mixed Stock Analysis*

292        Results for all samples combined indicated that the majority of individuals caught in the Gulf of  
293        Maine during 2010-2013 were of eastern origin (CMLE: 85%  $\pm$  12% and RF: 67% eastern  
294        origin; Table 1). Across the years, the mixed stock composition of bluefin tuna was dominated  
295        by eastern origin fish, with the highest percentage of eastern origin fish caught in 2010 (CMLE:  
296        89%  $\pm$  9% and RF: 73% eastern origin) and 2012 (CMLE: 93%  $\pm$  9% and RF: 73% eastern  
297        origin) with slightly lower eastern origin catches in 2013 (CMLE: 82%  $\pm$  13 and RF: 63%  
298        eastern origin) and 2011 (CMLE: 76%  $\pm$  19% and RF: 61% eastern origin; Table 1, Fig. 3).  
299        Estimated mixed stock composition by month of capture indicated that the majority of bluefin  
300        tuna caught from June to October in the Gulf of Maine were consistently eastern origin (Table 1,  
301        Fig. 3). However, the greatest proportion of eastern origin fish were caught in August to October  
302        (Table 2, Fig. 3) with slightly lower proportions caught in June to July (Table 1, Fig. 3).  
303        Estimated mixed stock composition by size class indicated that the majority of the small and  
304        intermediate size classes of bluefin tuna caught in the Gulf of Maine were estimated to be of  
305        eastern origin (<150 cm, 150-199 cm, and 200-249 cm; Table 2, Fig. 3). However, the largest  
306        size class (>250 cm) of bluefin tuna was estimated to be predominantly western origin fish or  
307        nearly equal proportions of east and west, depending on classification method (CMLE: 48%  $\pm$   
308        21%; RF: 63% western origin, Table 2, Fig. 3). Likewise, mixed stock analysis by age class  
309        indicated that younger fish were more likely to be of eastern origin ( $\leq$  age 12) compared to the  
310        oldest group of fish sampled ( $\geq$  age 13, Table 2, Fig. 3). For samples with sex information,  
311        mixed stock analysis indicated that both male and female fish were predominantly eastern origin  
312        fish, with a slightly higher proportion of female fish being eastern origin (Table 2, Fig. 3).

313        *3.4. Integration of Mixed Stock Information in Fisheries Data*

314 Integration of mixed stock information in U.S. rod and reel data required characterization of  
315 mixed stock composition by U.S. fleet size categories, rather than generic size bins (as described  
316 previously). Also, in this characterization we adopted the RF classification as this is now the  
317 prescribed approach used by ICCAT. This revealed that fish in the U.S. rod and reel size  
318 category 66-114 cm and 115-144 cm had a higher proportion of eastern origin fish (RF: 78% and  
319 79% eastern origin, respectively) compared to fish in the >177 cm size category (RF: 65%  
320 eastern origin, Table 3). The application of mixed stock information to inform fishery-specific  
321 catch, catch-at-age, and CPUE demonstrates an approach to separate mixed stock fishery-  
322 dependent data into population-of-origin data. We applied stock composition estimates from this  
323 study and Siskey et al. (2016) that aligned with the size class, timing (years of collection), and  
324 geographic location of the U.S. rod and reel fishery data. Although we do not have stock  
325 composition information that spans the full time series of the data, this exercise can provide  
326 insight into the utility of collecting data at this scale in the future. The estimated CPUE, catch,  
327 and catch-at-age of western origin fish in the U.S. rod and reel fishery (size categories: 66-114  
328 cm, 115-144 cm, and >177 cm) were considerably less than that of the western mixed stock,  
329 reflecting the dominance of eastern origin fish in the stock composition (Table 3, Figs. 4, 5, 6).  
330 Across the U.S. rod and reel size categories, the catch of >177 cm fish had the greatest  
331 contribution of western origin fish to western Atlantic fisheries (Fig. 6).

## 332 **4. Discussion**

### 333 *4.1. Mixed Stock Analysis of U.S. Fishery Landings*

334 Identification of the origin of Atlantic bluefin tuna landed in the Gulf of Maine provides critical  
335 information on stock mixing in the primary U.S. commercial fishing grounds. The number of  
336 otolith samples we analyzed from recently collected fish allows for robust estimates of stock

337 composition for the U.S. rod and reel fishery and permits estimation of stock composition over  
338 time of capture (year and month), sex, size and age classes, and by fishery size categories.  
339 Population assignment based on otolith chemistry provides origin information that enables us to  
340 track the population origin of landings, in addition to the stock view of Atlantic bluefin tuna  
341 landings which is routinely tracked by fishery managers. This information is critical to  
342 understanding both how many fish are available to the fishery (stock view) and the status of  
343 bluefin tuna populations, or units of production (population view).

344 Otolith chemistry indicated a high degree of mixing of the eastern and western origin bluefin  
345 tuna in the Gulf of Maine, with eastern fish typically dominating the composition of the catch.  
346 The prevalence of eastern origin fish caught by western fisheries represents a revised perception  
347 of the composition of fish in the U.S. rod and reel fishery relative to the current assumption  
348 underlying ICCAT's assessment and management of the bluefin tuna (i.e., no stock mixing). The  
349 representation of eastern fish in the U.S. rod and reel fishery catches relates to both the  
350 movement rate of eastern origin fish into the Gulf of Maine and relative differences in the  
351 biomass of eastern and western origin fish. The most recent ICCAT stock assessment (Anon.,  
352 2017) supports a significant increase in eastern stock biomass. There is also evidence of a  
353 moderate increase in western stock biomass (Anon., 2017). However, it is unclear if this increase  
354 reflects increased western origin biomass or increased mixed stock biomass augmented by  
355 migrating eastern origin fish. In addition, substantial shifts in the spatial distribution of Atlantic  
356 bluefin tuna have been documented in response to changing ocean conditions and prey  
357 distributions (e.g., Golet et al., 2013; Fromentin et al., 2014; MacKenzie et al., 2014; Druon et  
358 al., 2016; Faillettaz et al., 2019). Increased productivity of eastern bluefin tuna and changing  
359 movement rates in relation to changing ocean conditions are likely drivers of the current high

360 prevalence of eastern origin fish in western fisheries. Recent evidence supports an alternative  
361 spawning ground for Atlantic bluefin tuna in the Slope Sea region of the western Atlantic with  
362 spawning occurring at different times of year (June-August) than in the Gulf of Mexico  
363 (Richardson et al., 2016). The effect of this potential third group of spawning fish on  
364 assumptions for baseline information for otolith chemistry is unknown, but the collection of  
365 samples from this area is a high priority among bluefin tuna researchers.

366 It is important to note that stock mixing is a dynamic process and trends across years support this  
367 interpretation, with a change on the order of 10-20% in the mixed stock composition between  
368 2010 and 2013. Furthermore, we find patterns across months with an increased proportion of  
369 eastern origin fish present in the fall months (September to October) as compared to early in the  
370 summer. This temporal trend could be indicative of spatial and temporal differences in habitat  
371 use in the Gulf of Maine by eastern and western origin fish. There were also apparent  
372 ontogenetic trends in stock composition, with the smallest size and age classes dominated by  
373 eastern origin fish and the largest composed predominantly of western origin fish. This pattern  
374 supports previous studies that have shown high rates of transoceanic migration of eastern origin  
375 fish at younger sizes (Rooker et al., 2008b; Dickhut et al., 2009). Changes in stock composition  
376 do not appear to be sex-specific, as there was no trend between sexes with near equal  
377 representation of eastern and western origin fish for males and females. Differences across  
378 grouping factors highlight the need to understand the details of where and when fish were  
379 collected and the demographic characteristic of the fish (e.g., length and age) to more accurately  
380 interpret and understand mixed stock composition information as these factors can influence the  
381 mixed stock composition in both subtle (e.g., month of capture) or significant ways (e.g., size  
382 based differences).

383 Another important finding of this study is the differences in stock composition results based on  
384 assignment method (individual and mixture analysis). We observed that while the overall trend  
385 between approaches was similar across grouping factors (i.e., time of capture, sex, size, and age  
386 classes), the absolute values differed (Fig. 3). In general, mixture analysis is expected to offer  
387 more accurate estimates of stock composition than summing individual assignments to determine  
388 the mixture proportions as uncertainty in the classification of individuals may lead to estimation  
389 bias (Manel et al., 2005). Furthermore, the approach of summing individual assignments ignores  
390 the uncertainty associated with each individual assignment. ICCAT has adopted an individual  
391 assignment approach in compiling a database of stock of origin information on bluefin tuna  
392 across the Atlantic as this offers an advantage when sharing assignment data and allows for  
393 individuals to be easily regrouped to address new questions (Anon., 2017). However, when  
394 integrating data in this manner, it is important to standardize the assignment methodology and  
395 model decisions, such as threshold values for assignment. Recognizing the influence of  
396 assignment method is critical to resolving similarities across studies and synthesizing data for a  
397 broader picture of mixed stock composition that can be applied in fisheries management.

398 Comparison of our results with previous results from U.S. and Canadian fisheries revealed some  
399 alignment and differences with other studies. Previous work by Siskey et al. (2016) indicated that  
400 ~95% of bluefin tuna in the Gulf of Maine in 2011 were of western origin. In contrast, our  
401 analysis using the same baseline and assignment method (mixture analysis based on CMLE)  
402 indicate that  $24\% \pm 19\%$  of fish were of eastern origin during the same year. These large  
403 differences may relate to subtleties in the spatial, temporal, or demographic scale of sampling  
404 between these two projects. Our results align with stock composition results from U.S. harvested  
405 Atlantic bluefin tuna sampled in 2015 and landed in Maryland and New York (Barnett et al.,

406 2017). Stock contribution estimates for all 2015 samples combined indicated that the majority  
407 were eastern origin fish (72%). Similarly, Barnett et al. (2017) found decreased mixing with  
408 increasing size with an eastern contribution of 79%, 64%, 50%, and 37% respectively for school  
409 (69-117 cm CFL), large school (119-147 cm CFL), medium (150-203 cm CFL), and giant ( $\geq 206$   
410 cm CFL) size bluefin tuna. An increased presence of eastern origin fish has also been observed  
411 recently in Canadian fisheries, where the fishery had traditionally been composed of western  
412 origin fish nearly exclusively (Busawon et al., 2015) and has now shifted to be predominantly  
413 eastern origin fish (Hanke et al., 2017; Busawon *personal communication*).

#### 414 *Integration of mixed stock data into fisheries management*

415 Understanding patterns in mixed stock composition can provide a means of monitoring and  
416 potentially managing fishing pressure at the population level, in addition to the stock level. The  
417 application of stock composition results to revise catch, catch-at-age, and CPUE time series can  
418 enable us to more closely track fishing pressure on western origin fish, as well as the age/size  
419 structure and relative abundance of this population. One of the most relevant findings to fisheries  
420 conservation and management from our study was the size-based differences in mixed stock  
421 composition. Increased representation and dominance of western origin fish in the largest size  
422 class provides an opportunity for more precise regulation of harvest of western origin fish based  
423 on size restrictions (e.g., slot limits) or size-based quotas for this component of the fishery.

424 Our stock composition information for the U.S. rod and reel fishery complements ongoing  
425 population assignment work for other fisheries in the eastern and western stock areas and  
426 contributes to ICCAT's aim of synthesizing this information to get a more comprehensive view  
427 of stock mixing across fisheries and areas (Anon., 2017). There is ongoing work to fit models to

428 mixed stock composition based on otolith chemistry and genetics (Carruthers and Butterworth,  
429 2019). A key challenge of this effort is the lack of historical data and that all bluefin tuna  
430 fisheries (e.g., longline, purse seine, bait boat) are not well sampled and characterized. If this  
431 type of population-level information were available across fisheries, it could be integrated into a  
432 stock assessment model to estimate population status (Kerr et al., 2017; Cadrin et al., 2018;  
433 Morse et al., 2018). Operational use of otolith chemistry to inform an assessment of bluefin tuna  
434 that includes stock mixing could transform the accuracy of bluefin tuna stock assessment and  
435 effectiveness of management (Kerr et al., 2017; Cadrin et al., 2018; Morse et al., 2018). More  
436 immediately, this information can be used outside of the stock assessment to monitor mixed  
437 stock composition of the fishery in near real time and to inform management decisions aimed at  
438 controlling population of origin harvest.

## 439 **5. Conclusions**

440 Application of otolith chemistry techniques to characterize the origin of bluefin tuna caught in  
441 the U.S. rod and reel fishery in the Gulf of Maine revealed the majority of fish caught from 2010  
442 to 2013 originated from the Mediterranean Sea (eastern origin), not the Gulf of Mexico (western  
443 origin). While fish in small and intermediate size classes were primarily eastern origin, fish in  
444 the largest size class (>250 cm) were predominantly western origin. It is important to recognize  
445 the influence that alternative approaches to assignment (individual and mixture) can have on  
446 results. Such uncertainties should be taken into consideration in ongoing stock assessment and  
447 management strategy evaluation analysis. Integrating mixed stock composition information into  
448 fishery-specific harvest data (catch, catch-at-age, and catch-per-unit-effort) could be used to  
449 inform monitoring and management objectives for the fishery.

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600 **Figure Captions**

601 **Fig 1.** *Left panel:* Transverse section of bluefin tuna otolith. Solid lines represent the micromill  
602 drill path to be powered. Small dashed lines depict the targeted region of the otolith representing  
603 year one growth. *Right panel:* Stable isotope values ( $\delta^{13}\text{C}$  and  $\delta^{18}\text{O}$ ) of Atlantic bluefin tuna from  
604 the Gulf of Maine (gray circles) compared to eastern (n=150) and western (n=115) baseline  
605 samples based on yearling bluefin tuna (Rooker et al., 2014).

606 **Fig. 2.** Illustration of known spawning habitat (dark grey) of western and eastern origin Atlantic  
607 bluefin tuna and recently discovered slope sea spawning area (light gray). Note the region of  
608 capture of fish from the U.S. Rod and Reel fishery (hatched area) and the ICCAT management  
609 boundary (dashed horizontal line).

610 **Fig. 3.** Mixed stock composition of all Atlantic bluefin tuna sampled and stock composition by  
611 (a) year of capture (2010-2013), (b) month of capture, (c) size class (cm CFL), (d) sex, and (E)  
612 age class. Values shown are the proportion of eastern origin fish in the sample as estimated by  
613 two methods: conditional maximum likelihood (light gray; error bars indicate standard deviation)  
614 and by individual population assignment using a random forest approach (dark gray).

615 **Fig. 4.** Catch-per-unit-effort of Atlantic bluefin tuna from mixed stock (solid line) U.S. Rod and Reel  
616 fishery (size categories 66-114 cm [a], 115-144 cm [b], and >177 cm [c]) in the western Atlantic.  
617 Stock of origin is determined for two time periods (1996-2002 and 2010-2015) for which stock  
618 composition data exist for these fishery size categories (dashed lines).

619 **Fig.5:** Catch-at-age of bluefin tuna from mixed-stock U.S. rod and reel fishery in the Gulf of  
620 Maine (white circles) and of western origin Atlantic bluefin tuna (black circles). Stock of origin

621 is determined for two time periods (1996-2002 and 2010-2015) for which stock composition data  
622 exist for this fishery.

623 **Fig. 6:** Catch-at-age of bluefin tuna from mixed-stock U.S. rod and reel fishery size categories  
624 (size categories 66-114 cm [a], 115-144 cm [b], and >177 cm [c]) in the Gulf of Maine (white  
625 circles) and of western origin Atlantic bluefin tuna (black circles). Stock of origin is determined  
626 for two time periods (1996-2002 and 2010-2015) for which stock composition data exist for this  
627 fishery.

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Table 1. Mixed stock composition of all Atlantic bluefin tuna sampled and stock composition by year of capture (2010-2013) and by month (June- Oct.) as estimated by two assignment methods (conditional maximum likelihood estimator [CMLE] and random forest). Values for CMLE shown are the mean and standard deviation (in parentheses) of the proportion of eastern and western origin fish in the sample.

Time	Population	n	Conditional Maximum Likelihood	n	Random Forest (Prob. > 0.7)
All samples	East	782	0.85 (0.12)	491	0.67
	West		0.15 (0.12)		0.33
2010	East	221	0.89 (0.09)	146	0.73
	West		0.11 (0.09)		0.27
2011	East	255	0.76 (0.19)	134	0.61
	West		0.24 (0.19)		0.39
2012	East	149	0.93 (0.09)	101	0.73
	West		0.07 (0.09)		0.27
2013	East	157	0.82 (0.13)	110	0.63
	West		0.18 (0.13)		0.37
June	East	107	0.84 (0.14)	65	0.66
	West		0.16 (0.14)		0.34
July	East	166	0.73 (0.18)	101	0.57
	West		0.27 (0.18)		0.43
August	East	229	0.87 (0.11)	148	0.72
	West		0.13 (0.11)		0.28
September	East	205	0.90 (0.10)	135	0.71
	West		0.10 (0.10)		0.29
October	East	67	0.87 (0.14)	37	0.70
	West		0.13 (0.14)		0.30

Table 2. Mixed stock composition of all Atlantic bluefin tuna sampled and stock composition by size class (curve fork length, cm), age-class, and sex as estimated by two assignment methods (conditional maximum likelihood estimator [CMLE] and random forest). Values for CMLE shown are the mean and standard deviation (in parentheses) of the proportion of eastern and western origin fish in the sample.

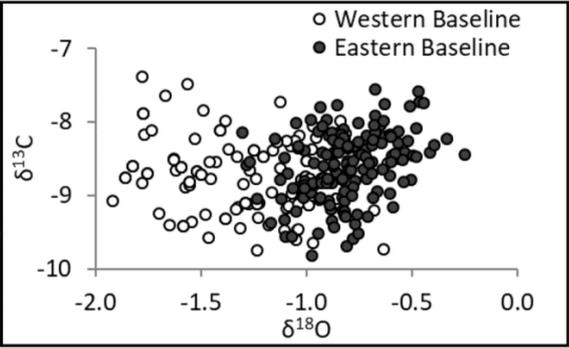
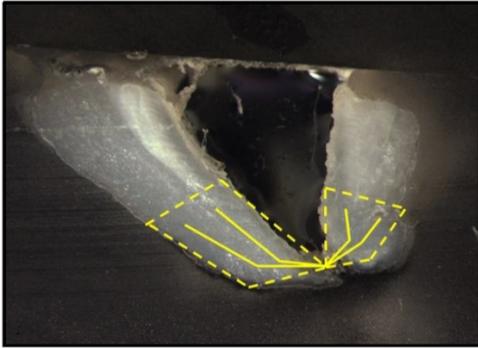
Size/Age/Sex	Origin	n	Conditional Maximum Likelihood	n	Random Forest (Prob. > 0.7)
< 150 cm	East	81	0.96 (0.06)	69	0.80
	West		0.04 (0.06)		0.20
150-199	East	233	0.86 (0.13)	143	0.66
	West		0.14 (0.13)		0.34
200-249	East	325	0.92 (0.10)	201	0.76
	West		0.08 (0.10)		0.24
>250 cm	East	143	0.52 (0.21)	78	0.37
	West		0.48 (0.21)		0.63
Age <4	East	27	0.94 (0.08)	23	0.83
	West		0.06 (0.08)		0.17
Age 4-6	East	85	0.97 (0.05)	74	0.82
	West		0.03 (0.05)		0.18
Age 7-9	East	321	0.85 (0.13)	191	0.65
	West		0.15 (0.13)		0.35
Age 10-12	East	207	0.94 (0.09)	125	0.78
	West		0.06 (0.09)		0.22
Age 13-15	East	51	0.58 (0.21)	32	0.44
	West		0.42 (0.21)		0.56
Age 16-18	East	46	0.29 (0.23)	22	0.27
	West		0.71 (0.23)		0.73
Age ≥19	East	45	0.51 (0.22)	24	0.38
	West		0.49 (0.22)		0.63
Female	East	175	0.90 (0.12)	115	0.69
	West		0.10 (0.12)		0.31
Male	East	304	0.84 (0.13)	191	0.65
	West		0.16 (0.13)		0.35

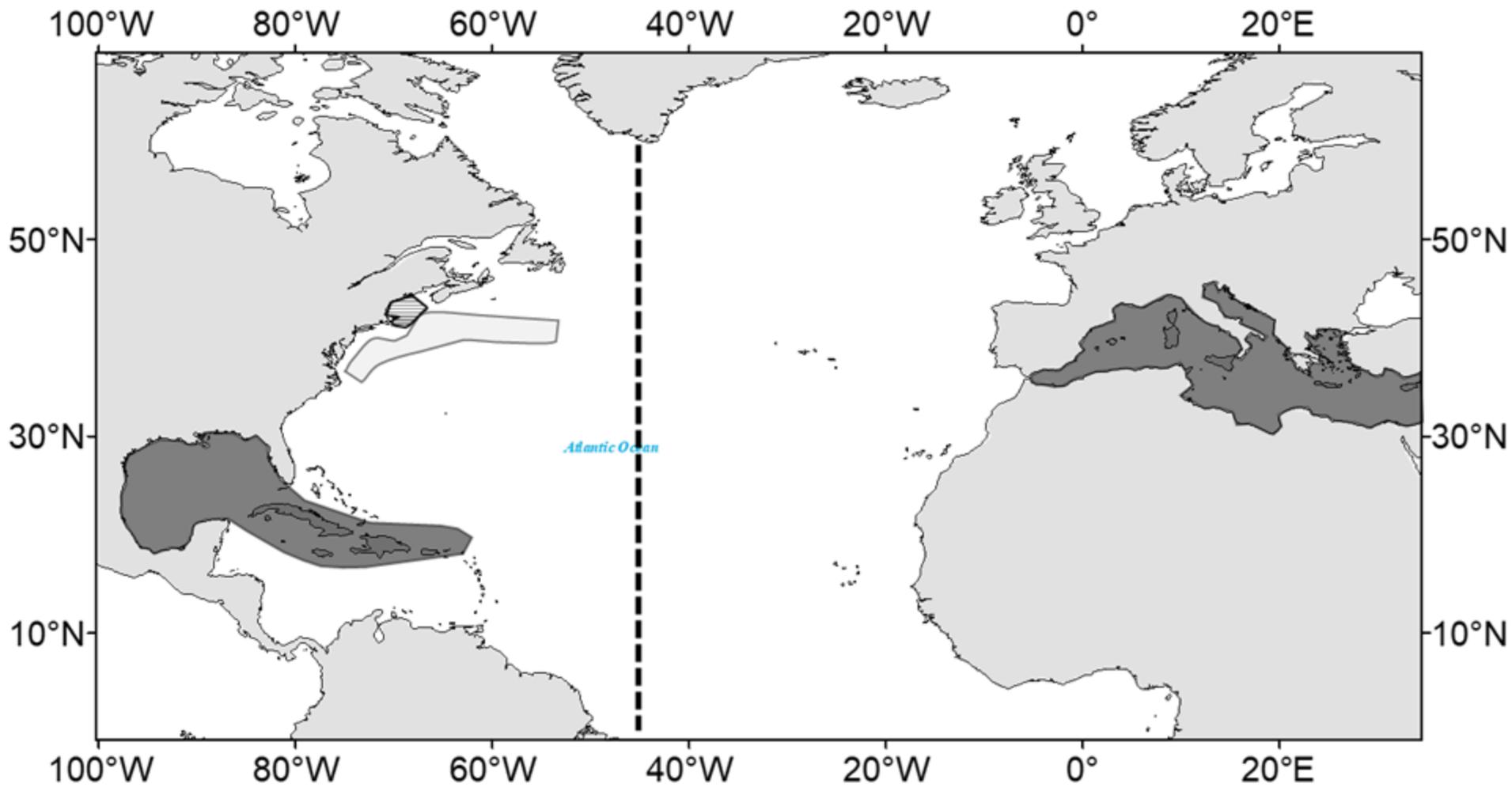
Table 3. Estimates of stock composition for the U.S. Rod and Reel fishery across size categories applied to revise fishery dependent data (US rod and reel: 66-114 cm, 115-144 cm, and >177 cm, ICCAT 2018), including catch per unit effort, catch at age, and partial catch at age. Note, representative size class information was not available from the 1990s for US Rod and Reel 115-144 cm category .

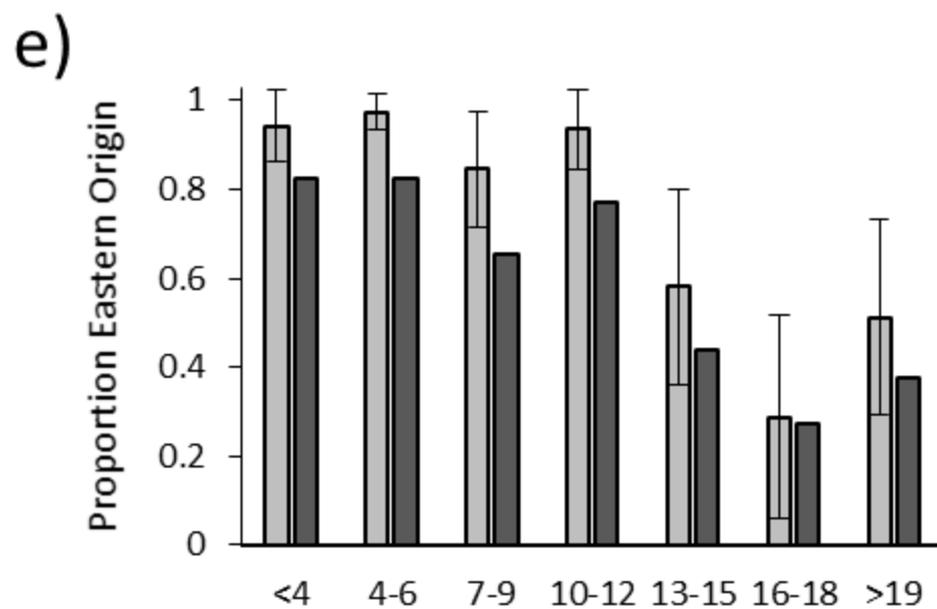
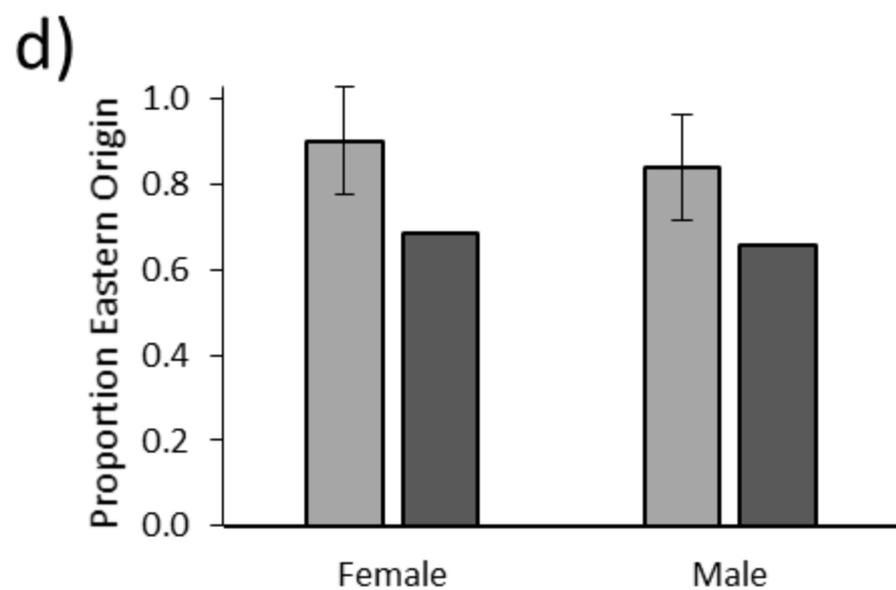
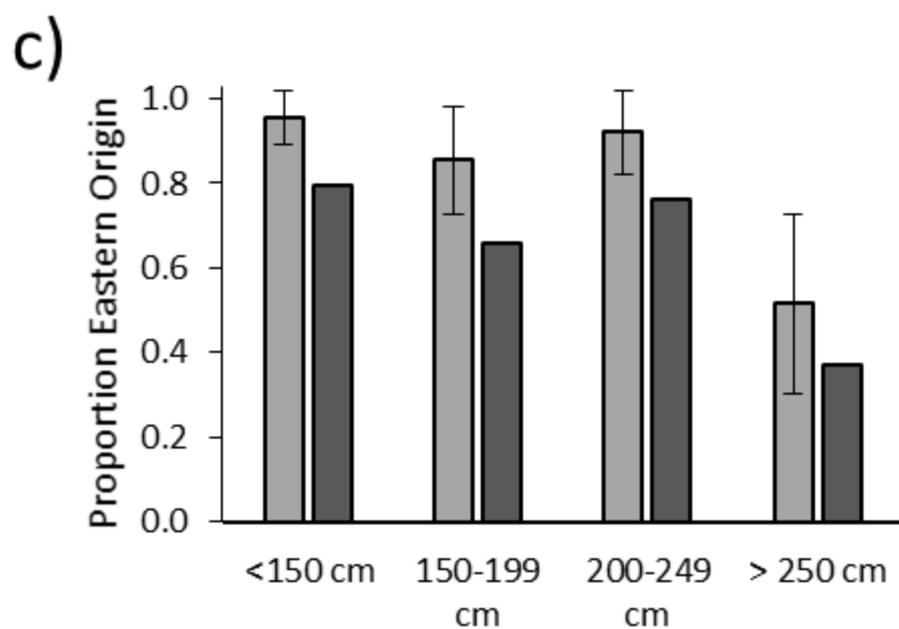
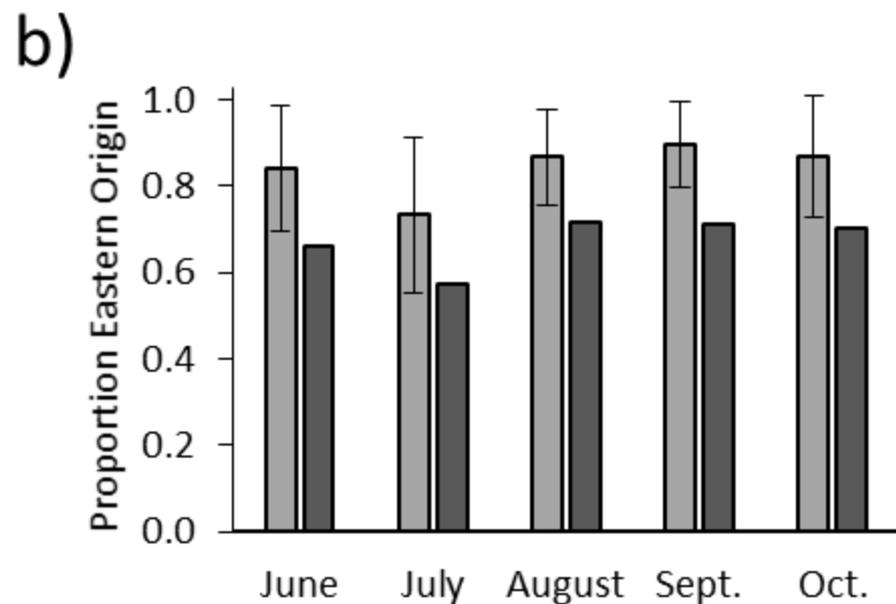
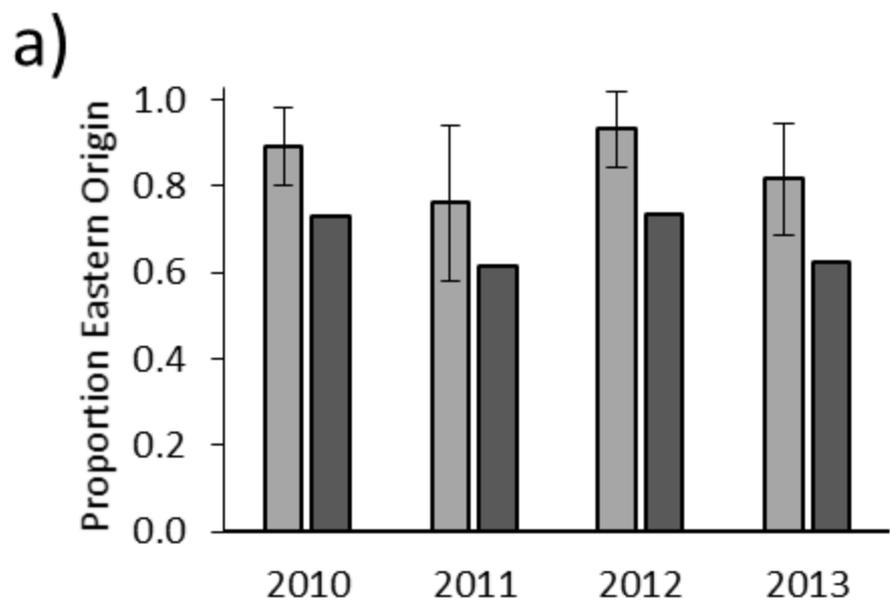
Size Category	Time Frame	N	Proportion Western	Proportion Eastern	Source
US RR 66-114 cm	2010-2013	24	0.21	0.79	This study
	1996-2002	24	0.21	0.79	Siskey et al., 2016
US RR 115-144 cm	2010-2013	41	0.22	0.78	This study
	1996-2002	NA	NA	NA	NA
US RR >177 cm	2010-2013	412	0.35	0.65	This study
	1996-2002	52	0.54	0.46	Siskey et al., 2016

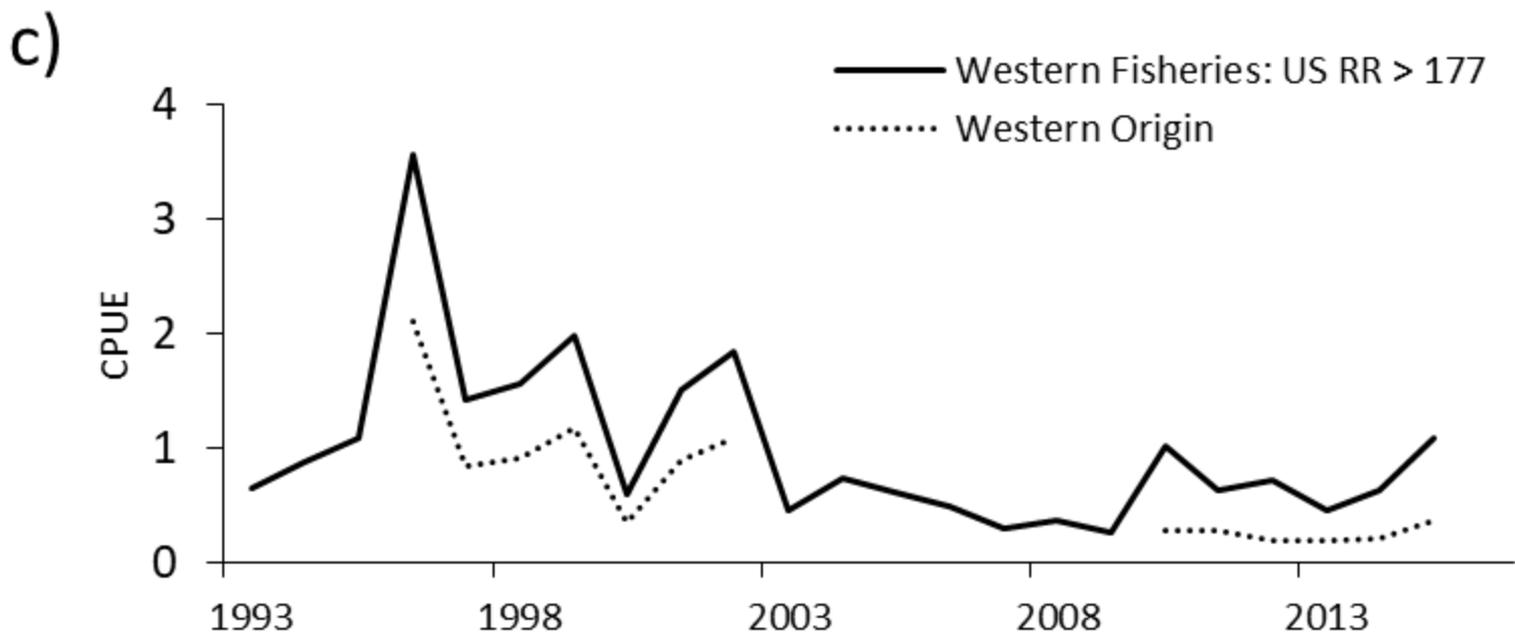
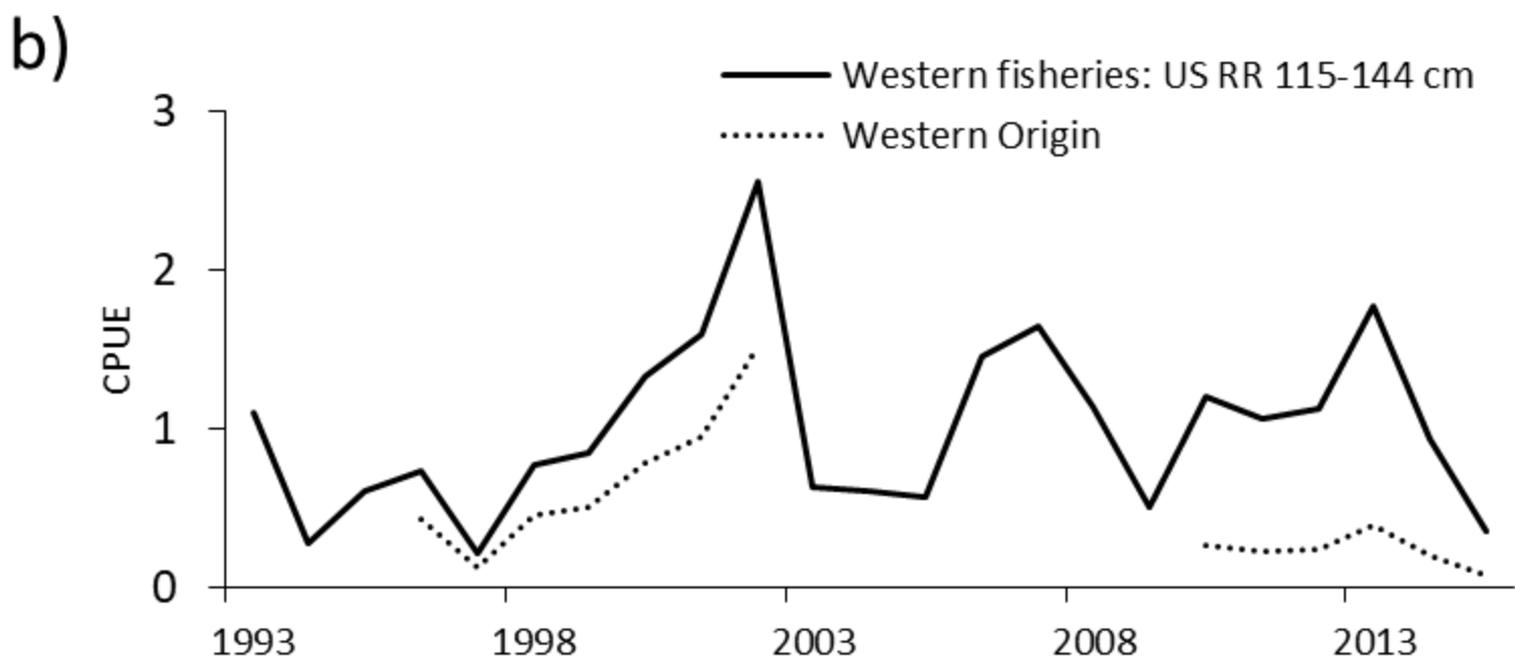
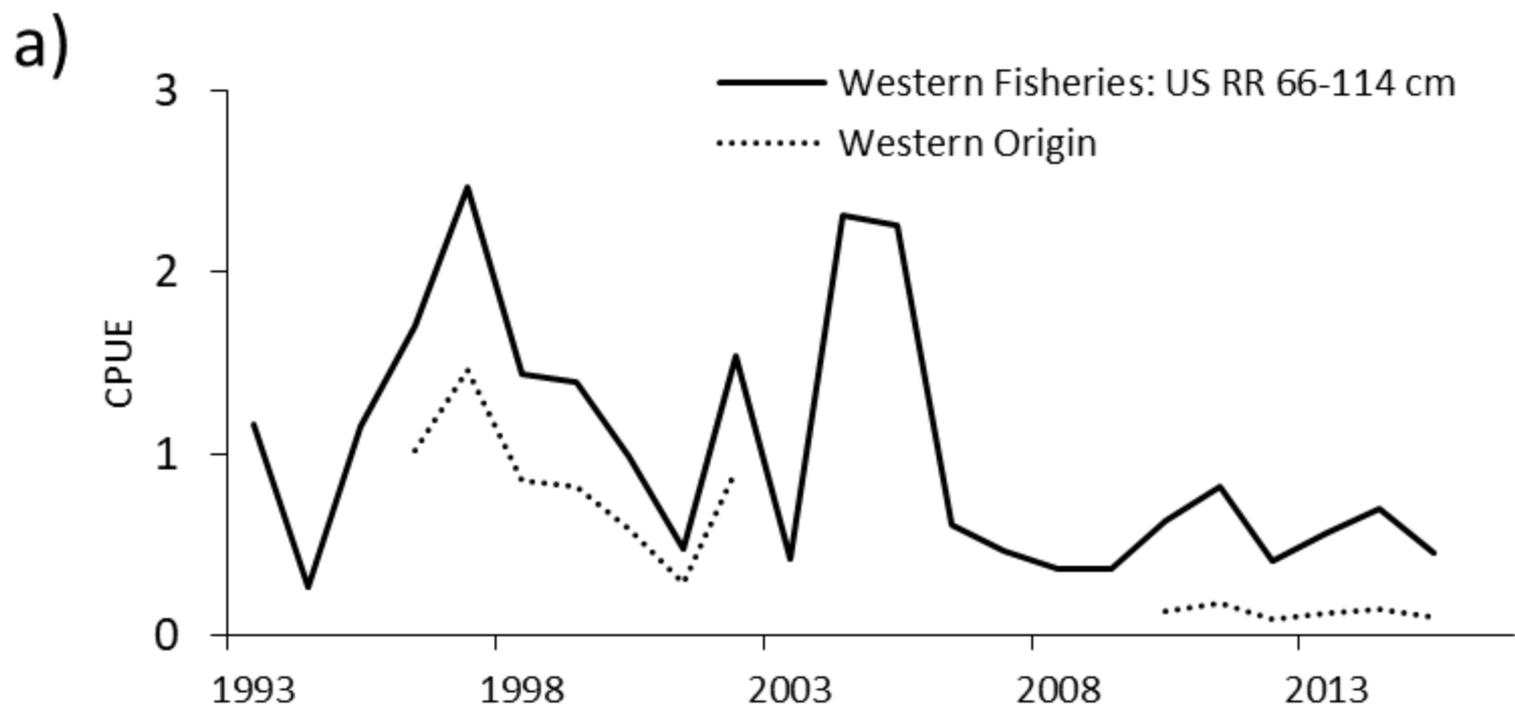
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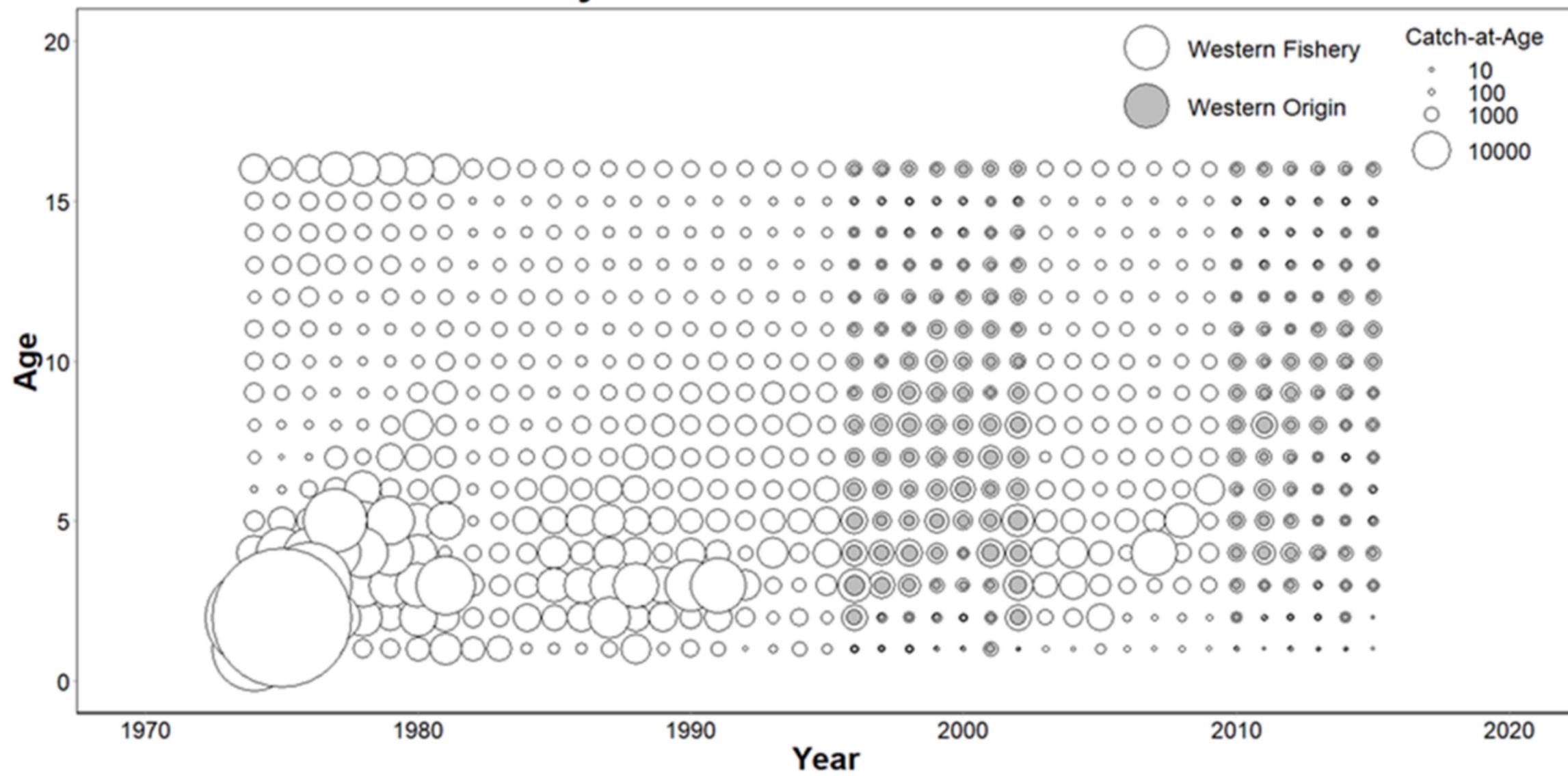




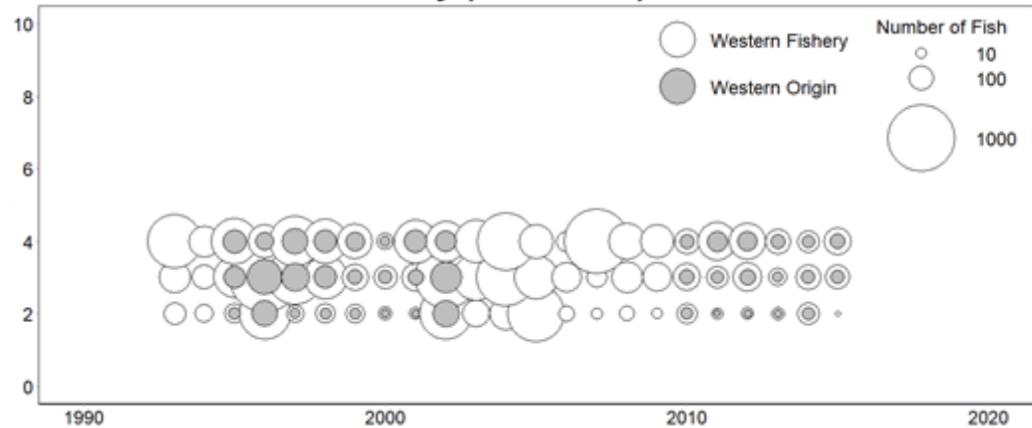




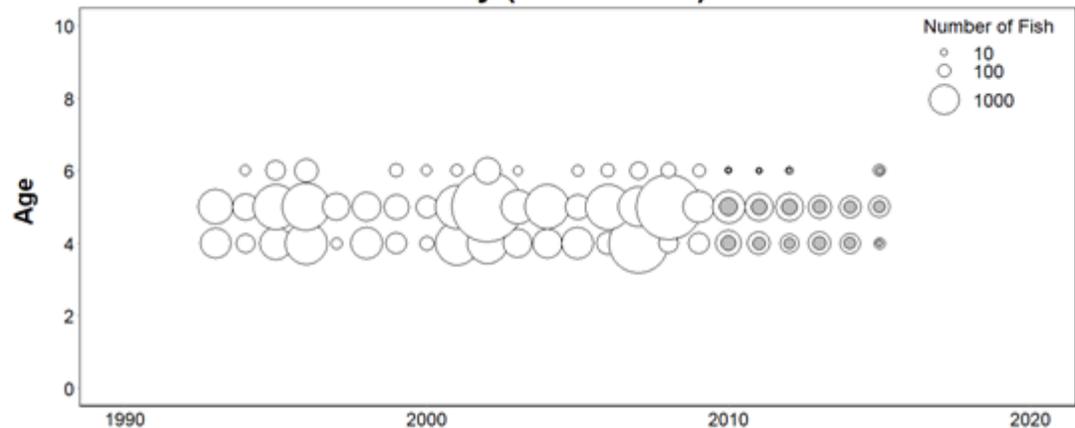
# U.S. Rod and Reel Fishery



**U.S. Rod and Reel Fishery (66-114 cm)**



**U.S. Rod and Reel Fishery (115-144 cm)**



**U.S. Rod and Reel Fishery (>177 cm)**

