

1 **A tale of two species: disaggregating mixed historical catches of two most common skates in**
2 **the Northeast Pacific Ocean.**

3

4 Vladlena Gertseva^{1*}, Sean E. Matson², Ian G. Taylor¹

5

6 ¹Fishery Resource Analysis and Monitoring Division, Northwest Fisheries Science Center,
7 National Marine Fisheries Service, National Oceanic and Atmospheric Administration, Seattle,
8 WA 98112, USA.

9 ²Sustainable Fisheries Division, West Coast Region, National Marine Fisheries Service, National
10 Oceanic and Atmospheric Administration, Seattle, WA 98115, USA.

11

12 *Corresponding author: Vladlena Gertseva. E-mail: Vladlena.Gertseva@noaa.gov. Phone: (+1)
13 206-860-3457.

14 **Abstract**

15 Historical catch data represent a key source of information for fisheries stock assessment
16 models. Historically, commercial fishery catch statistics were estimated from the portion of catch
17 landed in ports. However, many species with relatively low economic value have not been
18 recorded on a species basis but instead as a part of an aggregate category. Reconstructing
19 component species catch from an aggregate category is a common challenge for fisheries stock
20 assessment efforts around the world. Skates are one group of species with low economic value
21 for which landed catch has not been commonly reported by species. In this paper, we present a
22 novel approach to disaggregate the historical catch of the two most abundant skate species on the
23 West Coast of the United States, longnose skate (*Caliraja rhina*) and big skate (*Beringraja*
24 *binoculata*), landed within the aggregated skate category, in ports of Washington State. We used
25 a combination of fishery-dependent and fishery-independent data sources to account for changes
26 in the spatial extent of the fishery over time, and differences in the depth distribution of these
27 two skate species. While developed to disentangle aggregate catch of longnose and big skates,
28 the approach is not limited to skates on the West Coast of the United States, but can be adapted
29 for any species which landings have been reported within an aggregated category elsewhere.

30

31 **Keywords:** historical catch estimates; fisheries stock assessment; longnose skate; big
32 skate; Northeast Pacific Ocean.

33

34 **Introduction**

35 Historical catch data are a key source of information for fisheries stock assessment
36 models (Hilborn and Walters 2003, Branch et al. 2011, King 2013). Inaccurate catch histories
37 can lead to assessment model misspecification, errors in estimation of key model parameters and,
38 in turn, inaccurate estimation of stock size and the scale of expected recruitments (Gertseva and
39 Matson 2021). Such errors can degrade the quality of assessment results that are necessary for
40 the conservation and management of fish stocks. Catch information from commercially
41 harvested species is also used to track global trends in fishing and aid to sustainable management
42 and conservation of marine resources (Hilborn et al. 2003, Branch 2008, Branch et al. 2011,
43 King et al. 2017).

44 Total catch of a fish stock consists of two components: the portion of catch that was
45 retained and subsequently landed in ports, and the portion of the catch that was discarded at sea
46 (Hilborn and Walters 2003, Haddon 2011). Historically, commercial fishery catch statistics have
47 primarily consisted of the fraction landed in port, which originates from landing receipts filled
48 out by fish dealers or dockside catch monitors. However, for many species with relatively low
49 economic value that are caught incidentally with other commercially important species, even
50 landed catch information has been limited, because their catch has not been recorded for
51 individual species, but instead as a part of an aggregate category (Dulvy and Reynolds 2002). It
52 has been shown that seemingly stable aggregated catch statistics among several species can mask
53 the decline in one or more species within the aggregate group, due to increases (sometimes
54 compensatory) in others (Dulvy et al. 2000).

55 Skates (family *Rajidae*) are the most widely distributed group of batoid fish with
56 approximately 200 species described (Last et al. 2016). They are benthic inhabitants that are
57 found in all coastal waters but are most common in cold temperatures (Ebert and Compagno
58 2007). Skates, like other elasmobranch species, also represent one of the most vulnerable group
59 of fishes because their low fecundity, slow growth and late maturation make them highly
60 susceptibility to overfishing (Dulvy and Reynolds 2002, Dulvy et al. 2014, Matson and Gertseva
61 2020). Skates are one group of species with low economic value for which landed catch
62 (landings) has not been commonly reported by species (Dulvy et al. 2000, Gertseva et al. 2019,
63 Taylor et al. 2019, Gertseva and Matson 2021). Lack of species-specific catch statistics has
64 presented a challenge for the accurate assessment of the status and sustainable management of
65 skate stocks. Therefore, progress in estimating species-specific composition of aggregate skate
66 landings is necessary to ensure long-term sustainability of these vulnerable species.

67 Multiple approaches have been used to estimate individual species' proportions to the
68 catch of an aggregated category. Many of these are based on calculating an individual species
69 proportions from a time period (usually recent) in which these are known, and applying those
70 proportions to the period with no information on species-specific catch (Gertseva 2009). This
71 approach assumes that proportions of different species within the group stay relatively stable
72 throughout time, and the recent period is representative of the earlier period with undocumented
73 species composition. However, this assumption may not be true if the recent period, with known
74 species-specific data, is different from the earlier period with aggregate data only. For example,
75 the spatial distribution of the fishery may change over time due to changes in the target stock
76 spatial distribution, bycatch avoidance behaviors, or management measures, such as spatial
77 closures (Miller et al. 2014). When species that comprise an aggregate category differ in their

78 preferred habitats and distributions, spatial changes in fishing effort result in changes in relative
79 contribution of individual species to an aggregate.

80 Longnose skate (*Caliraja rhina*) and big skate (*Beringraja binoculata*) are the most
81 abundant skate species on the West Coast of the United States in terms of biomass and
82 abundance as they represent over 90% of skate catches in the area (Gertseva 2009, Gertseva et al.
83 2019, Taylor et al. 2019). Both species are broadly distributed in the Northeast Pacific Ocean,
84 from Alaska to beyond southern Baja California (Love et al. 2021, Snytko 1987, Eschmeyer and
85 Herald 1983, Mecklenburg et al. 2002), but their depth distributions differ. Longnose skate is
86 most common at depths between 150 and 400m (Tolimieri and Levin 2006, Bizzarro 2015),
87 while big skate is mostly found on the continental shelf, shallower than 200m (Bizzarro et al.
88 2014, Farrugia et al. 2016). Both of these species are caught along with other, more-valuable
89 target species, including sablefish (*Anoplopoma fimbria*) and petrale sole (*Eopsetta jordani*), in
90 the groundfish demersal trawl fishery of the West Coast of the U.S., and the retention of both
91 species increased since mid-1990s (Gertseva et al. 2019, Taylor et al. 2019). Also, on the U.S.
92 West Coast, fishery management efforts were actively developing over the last four decades, and
93 since the early 2000s, these measures included implementation of multiple spatial conservation
94 areas, closed to fishing to help recover some of the depleted groundfish stocks. Given the
95 differences in depth distribution between these two species and changes in depth of fishing, not
96 accounting for depth specific species compositions creates a potential for a masked decline in
97 individual species within the aggregate group. Therefore, a spatially explicit approach for
98 separating skate species-specific historical landings is necessary.

99 Here, we present a novel approach to estimate the historical species-specific catch of
100 longnose and big skate landed in Washington State using historical aggregate catch data from the

101 groundfish demersal trawl fishery while accounting for changes in the spatial extent of the
102 fishery and depth differences of the two skate species. We used a combination of fishery-
103 dependent and fishery-independent data sources to estimate the contribution of individual species
104 to aggregate landings, and validated our results using the data from the most recent period for
105 which fishery landings are available by individual species.

106 This study was initiated to resolve a critical need for recent longnose skate and big skate
107 stock assessments (Gertseva et al. 2019, Taylor et al. 2019), and focused on the waters off
108 Washington State because that area was lacking species-specific estimates of skate historical
109 landings, unlike other areas along the US West Coast. Since then, we refined our approach and
110 improved the estimates, which are now ready to be used in the next stock assessments for these
111 two species. Our approach, evaluated using one specific area, can potentially be expanded to
112 other parts of the coast. Also, even though we focused on longnose and big skates on the West
113 Coast of the United States, the approach we describe is flexible, and can easily be adapted for
114 other species elsewhere in the world, for which landings have been reported within an
115 aggregated category.

116 **Methods**

117 *Data sources*

118 The groundfish fishery existed on the West Coast since late-1800s (Miller et al. 2014),
119 but the bottom trawl fishery advanced in the 1930s with the invention of balloon trawl nets (Love
120 et al. 2002), and quickly expanded along the U.S. West Coast, and to deeper waters by the late-
121 1940s (Harry and Morgan 1961, Alverson et al. 1964, Love 2002).

122 The commercial landings made by the groundfish demersal trawl fishery of the West
123 Coast of the U.S from 1981 forward are reported in the Pacific Fisheries Information Network
124 (PacFIN), which is a collaboration between Pacific States Marine Fisheries Commission,
125 National Marine Fisheries Service (NMFS), and West Coast state fishery management agencies.
126 Prior to 1980s landings of skates were minimal (most were discarded), and we limit our
127 reconstruction here to years from 1981 forward.

128 PacFIN manages information on landed catch from landings receipts (also called fish
129 tickets) and dockside samplers along the U.S. West Coast and reports landed catch by year, gear
130 type, port of landing and many other categories. Until 2010, all skate landings were reported
131 together in the ‘Unspecified Skate’ category (Fig. 1). However, since 2010, coastwide landings
132 of longnose skate have been reported separately from other skates, and in 2015 big skate was
133 also separated into a single species category (Fig.2).

134 Fish tickets rarely include information on depth of fishing. However, PacFIN also hosts
135 logbooks, which are recorded by vessel captains and contain information on the spatial
136 distribution of individual hauls within commercial trawl fishing trips. Logbook records include
137 landed catch for aggregated skates, geographic location of catch, and depth of fishing. This
138 source of data is more detailed than the fish tickets recorded by the dockside samplers and
139 processors. However, since logbook records have not been mandatory, this source (unlike fish
140 tickets) represent only a portion of the total landed catch. Logbook data for Washington State
141 skate landings in PacFIN goes back to 1987. Until 2016, all skates were reported in logbook as
142 one category of unspecified skate (even though longnose skate catch was reported separately on
143 fish tickets since 2010). From 2017 onward, logbook data include multiple skate categories (Fig.

144 2). Logbook records illustrate that depth of skate catch landed in Washington ports indeed varied
145 among years (Fig. 3).

146 To estimate the contribution of different skate species to aggregate skate fishery catches
147 by depth, we used fishery-independent data from the NMFS West Coast Groundfish Bottom
148 Trawl Survey (WCGBTS). This survey has been conducted annually since 2003, covering depths
149 between 55 and 1280 m along the U.S. West Coast between the U.S.-Canada and U.S.-Mexico
150 borders (Keller et al. 2017). The survey data contain haul-level skate catch by species,
151 geographic location, and fishing depth. We filtered the survey data to only include catches in
152 coastal waters off Washington State. We then divided skate catch into a series of depth bins, and
153 estimated proportions of longnose skate and big skate within each depth bin. The data were
154 divided into 25 fm (46 m) bins for depths up to 150 fm (274 m), and into 50 fm (91 m) bins for
155 depths of 150 fm and deeper. The finer bins were used for depths where the vast majority of big
156 and longnose skate co-occur, to better account for changes in percent contribution of these
157 species by depth within the aggregate. We used depth bins in fathoms (rather than meters), to
158 align with and better account for impacts of spatial management measures (such as spatial
159 closures) on relative species contribution to an aggregate; such spatial measures are commonly
160 applied for selected depths defined in fathoms.

161 We explored multiple binning options, and investigated the potential for including
162 latitudinal as well as seasonal components, in addition to depth. However, preliminary data
163 evaluation indicated that species proportions did not trend with latitude, when stratifying by one
164 degree latitude. The same was true among temporal strata, when dividing the survey as
165 granularly as the data would support, in this case, dividing the survey into two time periods (one
166 for spring/early summer, and two summer/early fall). Therefore, we only focused on depth-

167 specific bins, as data indicated clear differences in big skate and longnose skate distributions by
168 depth (Fig. 4).

169 From WCGBTS data, it is evident that big skate occupy shallower depths (100 fm and
170 less), while longnose skate predominantly occurs in deeper waters (Fig. 4), which is consistent to
171 what is reported in literature (Tolimieri and Levin 2006, Bizzarro et al. 2014, Bizzarro 2015,
172 Farrugia et al. 2016). There is some degree of interannual variability in proportional species
173 contribution to the aggregate category (Fig. 5), potentially due to environmental variability,
174 behavior and movements (and resultant distribution) of species at the time of sampling each year
175 or other year-specific factors, as well as sampling variability associated with the random
176 stratified survey design (Keller et al. 2017).

177 *Method overview*

178 The main goal of the study was to estimate historical catch of longnose and big skates
179 retained within the groundfish bottom trawl fishery and landed in Washington State ports in the
180 aggregate skate category for use in stock assessments of both species. Our method included three
181 main steps: 1) estimating proportions of longnose skate and big skate in combined skate catches
182 by depth and year within the WCGBTS catches, 2) estimating longnose skate and big skate catch
183 by depth and year in logbook catch data by applying survey proportions of longnose skate and
184 big skate to logbook reported catches by year, and 3) expanding longnose skate and big skate
185 catches from logbook data to the total species-specific landings by year (reported in fish tickets),
186 to account for unsubmitted logbooks. The reconstruction covered the period between 1981 with
187 the start of PacFIN data and goes through 2009 for longnose skate and through 2014 for big
188 skate, when these were removed from the aggregate category.

189 The approach to estimating individual skate proportions by depth within the WCGBTS
 190 catches (step one) was described above. The second and third steps are more complex and
 191 include multi-stage algorithms. To estimate species-specific skate catch from fishery logbook
 192 data, haul-specific catch of aggregated skates from logbooks was assigned to the appropriate
 193 depth bin. We then applied depth-specific proportions of longnose skate and big skate (as
 194 estimated from WCGBTS data) to each haul of total skate catch from logbooks, to obtain the
 195 species-specific catch of each skate by depth of each haul. Equation (1) below describes what
 196 was done in step 2 of the algorithm:

$$197 \quad (LBL)_{s,y} = \sum_{d=1}^{d=n} \left[(LBL)_{y,d} \cdot \frac{(SC)_{s,y,d}}{(SC)_{y,d}} \right]$$

198 Equation (1)

199 Where *LBL* is the amount of logbook landings of skates, *SC* is survey skate catch by species, *s*,
 200 year, *y*, and depth bin, *d*, with *d* ranging from one to *n*.

201 When survey data were available (Fig. 2), survey proportions of big and longnose skates
 202 within total skate catch $\left(\frac{(SC)_{s,y,d}}{(SC)_{y,d}}\right)$ were applied to catch reported in logbooks (*LBL*) by depth (*d*)
 203 and year (*y*) to account for interannual variability in depth-specific proportions of individual
 204 species. Prior to 2003 (before the survey began to operate), average proportions of longnose and
 205 big skates at depth between 2003 and 2007, were applied to depth-specific commercial logbook
 206 data. We summed the depth-specific estimates of longnose and big skates catch in trips with
 207 logbooks records into year-specific catch time series of each of these skate species ($(LBL)_{s,y}$).

208 As the third step in estimating species-specific skate landings, we expanded catch of
209 longnose and big skates reported in commercial logbooks to the level of total skate landings
210 reported via fish tickets. For this, we calculated the proportions of longnose and big skates in
211 aggregate skate catch from logbooks each year $\left(\frac{(LBL)_{s,y}}{(LBL)_y}\right)$, and applied these year-specific
212 proportions to total Washington skate landings by year $(L)_y$ (Equation 2 below).

$$213 \quad (L)_{s,y} = (L)_y \cdot \frac{(LBL)_{s,y}}{(LBL)_y} \quad \text{Equation (2)}$$

214 Where L is fish ticket landings, LBL is amount of logbook landings of skates, s in the
215 subscript stands for species and y for year.

216 For the period with logbook data available (1987 forward, Fig. 2), proportions of big and
217 longnose skates were applied by year $\left(\frac{(LBL)_{s,y}}{(LBL)_y}\right)$, again to account for interannual variability in the
218 contribution of individual species within the aggregate group. To disaggregate catch data
219 between 1981 and 1987 (when logbook data were not available), we applied the average
220 proportions of big and longnose skate calculated using the earliest five years of logbook data
221 (1987-1991) and applied those to the total skate landings in Washington from 1981 - 1987..

222 *Method validation and uncertainty*

223 To validate our method and the results, we compared our estimated landings for longnose
224 skate versus species-specific landings recorded in PacFIN, calculated based on port sampling,
225 over the period between 2010 and 2016. Landings of longnose have been reported in PacFIN
226 separately from other skate species since 2010 (Fig. 2). However, they continued to be reported
227 within the aggregate skate category in logbooks until 2016. Therefore, for the period between
228 2010 and 2016, we had longnose skate landings from PacFIN, but also had logbook data on

229 aggregate skates to make predictions using our approach. Big skate has been reported in PacFIN
230 separately from other skates since 2015, and we did not have enough overlap to compare the
231 landings between two sources (Fig. 2); therefore, our validation efforts focused on longnose
232 skate alone. The relationship between our estimated landings amounts, produced using the
233 described approach (independent variable), versus longnose skate landings records from
234 logbooks in PacFIN (dependent variable) was fitted by linear regression, and goodness of fit was
235 calculated as R^2 .

236 To account for uncertainty in using average proportion at depth, we estimated high and
237 low catch streams, by applying increased (plus two standard deviations) and decreased (minus
238 two standard deviations) proportions of longnose and big skates within each depth bin, calculated
239 from WCGBTS data.

240 **Results and Discussion**

241 Disaggregated time series of species-specific longnose and big skate landings in
242 Washington show that longnose skate dominated historical skate catch, and amount of longnose
243 skate landings on average was three times larger than that of big skate (Fig. 6). Year-specific
244 estimates of landings (Fig. 6) account for the changes in depth of catch (Fig. 3) and therefore,
245 allow for more accurate estimation of species-specific contribution to the overall skate catch. For
246 instance, we see a shift to deeper areas in the distribution of annual, aggregate skate fishery catch
247 in recent years (Fig. 3), concomitant with fishery effort shifts reported Somers et al. (2023). This
248 trend is likely to continue due to avoidance of some nearshore species off the West Coast. This
249 shift translates into the decreased catch of big skate, since this species occurs primarily in
250 shallower depths, and larger contribution of longnose skate, which dominates the deeper areas

251 Given a more limited, and shallower depth distribution of big skate, compared to widely
252 distributed longnose skate (Fig. 4), uncertainty around big skate catch results in larger intervals
253 around estimates for both species (since composition proportions are interdependent),
254 emphasizing the importance of accounting for depth. Additionally, one would especially expect
255 to see increased uncertainty around estimates in this method during periods when depth
256 distributions of fishery effort shifts, which we see in Fig. 6.

257 Depth distribution of fishing effort can change in relation to both target seeking and
258 bycatch avoidance behavior by the fleet, as well as management-related spatial closures,
259 although the exact dynamics can be difficult to determine. On the U.S. West Coast, varied and
260 sometimes intense fishery management measures have been implemented for groundfish species,
261 which depending on the species and fishery sector, may include trip limits, quotas, mesh size
262 requirements; depth, area, season, and gear restrictions; and other measures (Matson et al. 2017).
263 Since the early 2000s, multiple management measures have been implemented to recover some
264 depleted groundfish stocks. For instance, yelloweye rockfish (*Sebastes ruberrimus*) has been
265 managed under a rebuilding plan since 2002, and bycatch of this species has been constraining to
266 shelf fishery effort since. Only recently, allowable catch of yelloweye rockfish catch limits
267 started to increase, which is leading to recovering some level of fishing efforts to depths on the
268 shelf in some areas of the coast. Such management measures, whether directed to impact catch of
269 targeted or bycatch stocks, can influence spatial and temporal effort distributions (including
270 depth of fishing), and have immediate or downstream effects within a mixed stock groundfish
271 fishery, including on species composition. Not accounting for fishing depth dynamics over time
272 can lead to unanticipated correlated errors among species within the aggregate catch (Karnowski

273 et al. 2014, Gertseva 2009). The method described in the paper can account for changes in depth
274 of fishing, caused from the shift in spatial coverage of the fishery.

275 Comparison of our estimated landings of longnose skate and big skate, versus actual
276 landings informed by species-composition estimates from dockside sampling (reported in
277 PacFIN) allowed us to validate the method. For big skate, the two data sources had only a two-
278 year overlap (2015 and 2016), but for longnose skate both sources were available for the period
279 between 2010 and 2016 (Fig. 2), which enabled direct comparison and validation of the proposed
280 method. Linear regression between our estimated longnose skate landings versus actual landings
281 reported in PacFIN for that period demonstrated excellent overall fit ($R^2 = 0.795$, $p=0.0007$,
282 $RMSE=12.089$, Figure 7), indicating that our approach yields realistic and reasonably accurate
283 results overall. Discrepancies can be explained by uncertainty associated with our estimates, but
284 also potentially by limited dockside sampling of landed aggregate skate landings in some spatial
285 strata informing records of actual landed catch, which can cause some degree of uncertainty in
286 PacFIN records.

287 The uncertainty intervals in estimated landings (Fig. 6) encompass variation related to
288 characteristics of data used to estimate the species-specific contributions to an aggregate total.
289 However, multiple factors can contribute to species compositions of an aggregate that were not
290 accounted for here. For instance, there can be potential limitations in using data from the survey
291 conducted during only a portion of the year (from late spring to early fall) to inform fishery
292 catches that occur year around. We also assume here that species compositions in survey total
293 catch are representative of the fishery landings; while landings are only a part of fishery total
294 catch. Since both skate species have not been targeted, and discards primarily occur because of
295 lack of market (Rogers 1994), it was assumed that there is no preference in retaining one skate

296 species over the other, and both species had the same discard probability; and thus survey catch
297 composition can be used to inform species composition of landings.

298 In the recent observer data, there has been evidence for discarding of smaller individuals
299 while retaining larger ones. However, we do not have reliable information as to the criteria for
300 relative species retention preference of one species over the other, over a variety of situations
301 with both single and mixed species compositions. In some cases, fishers are said to prefer skates
302 that are not too small because of not large enough marketable (wings) body portion. In others,
303 not too large, due to difficulty handling them. Also, the applicability of the recent data in this
304 aspect to other historical periods is not known, since discard amounts and retention trends also
305 have been known to change over time with other species.

306 Another source of uncertainty in our estimates is related to reliability and
307 representativeness of logbook data, given that only part of trips were accompanied by logbook
308 records. It is reasonable to assume that available logbook data realistically represent general
309 fishing practices over the years. However, the relative amount of logbook reported catch versus
310 total PacFIN landings dropped since 2011, which could have affected accuracy of our estimates
311 during years used for method validation and partially explain discrepancies between our
312 estimates and actual landings reported in PacFIN. The reason for decrease in logbook reported
313 catch is not clear, but it coincided with implementation of an Individual Fishing Quota (IFQ)
314 system on the West Coast of the United States (Matson et al. 2017).

315 Uncertainty calculated around estimated historical landings can be directly used in
316 fisheries stock assessment process as well, to help evaluate the influence of potential alternatives
317 of high and low scenarios in longnose and big skate landings on model results through sensitivity

318 analysis. Sensitivity analysis is an excellent way to reveal and explicate to what degree the
319 assessment model output is affected by varying amounts of deviation from assumed fishery
320 landings time series. Cope and Gertseva (2020) provide a detailed overview of using sensitivity
321 analysis to evaluate structural and data uncertainty in stock assessments and to identify aspects of
322 the model that deserve further attention when quantifying uncertainty in model outputs and
323 management quantities.

324 Although focused on skates landed in Washington State, on the West Coast of the United
325 States, the method presented here is not limited to a particular species or geographic location, but
326 can be easily adapted and used for other species landed within aggregate categories around the
327 world. In our study, we only divided data into depth-specific bins, since we considered a
328 relatively limited latitudinal range of waters off Washington State, and the data did not show
329 trends in relative species occurrences by latitude. However, our approach could be easily applied
330 to a more complex spatial grid, based for instance on depth and latitude; the choice of grid would
331 depend on habitat preference of the contributing species to the aggregate category, and
332 distribution of fishing efforts. Also, seasonal bins could be applied for migrating species when
333 relative contribution of species within an aggregate varies within the year.

334 It is noteworthy that landings represent a fraction of the total fishery catches; there is also
335 discarded catch that contributes to total removals of a stock. Limited information on historical
336 discard indicate that prior to mid-1990s, the vast majority of skate catch was discarded at sea due
337 to lack of market (Rogers 1994), and only small amount of total skate catch was landed along the
338 West Coast and in Washington State ports. Discarded catch of skate in the mid-1980s, for
339 instance, accounted for more than 90% of total skate removals (Rogers 1994). However, in the
340 mid-1990s, a limited market for skate products appeared on the U.S. West Coast and retention of

341 skates in the bottom trawl fishery increased, while discards decreased to about 50% of the total
342 removals (Martin and Zorzi 1993, Bonfil 1994, Gertseva et al 2019, Gertseva and Matson 2021).
343 Currently, West Coast skates are marketed in limited amounts when they are sold fresh or fresh-
344 frozen, as well as dried or salted and dehydrated (Love et al. 2002). Gertseva and Matson (2021)
345 recently developed a method to predict total removals of bycatch species based on the catch of a
346 co-occurring targeted species; the method has already been applied in multiple stock assessments
347 of elasmobranch stocks (Gertseva et al. 2019, Taylor et al. 2019, Gertseva et al. 2021). In stock
348 assessment models however, landings and discards are commonly treated separately to account
349 for differences in length composition between retained and discarded fish and more accurately
350 estimate fishery selectivity curves. Therefore, landings represent an essential data source for
351 stock assessment and management of both targeted and non-targeted species, despite what
352 portion of the catch is discarded, and progress in reconstructing historical landings is necessary.

353 Acquisition or estimation of accurate catch time series is of utmost importance in
354 obtaining valid and reliable stock assessment results, a critical endeavor for ensuring
355 sustainability and conservation the world over, and disentangling aggregate species categories
356 presents a common problem for stock assessment. The approach described here represents
357 progress in this area, as here we developed a method enabling improvement in reconstruction of
358 species-specific landings within an aggregate category, accounting for fishing depth dynamics
359 over time. The approach presented here was already used to resolve a critical need for species-
360 specific estimates of skate historical landings in the most recent longnose skate and big skate
361 stock assessments (Gertseva et al. 2019, Taylor et al. 2019). Currently evaluated using a limited
362 area, our approach can be expanded to other parts of the coast in future assessments of these two
363 skate species.

364 It is important to continue efforts to improve our understanding of historical fishery
365 removals. In the future, approaches based on using statistical methods, such as Dirichlet
366 regression (Maier 2014) and multinomial logistic regression (Hilbe 2009) could be attempted for
367 these two species, in order to produce model-based predictions of species composition, using
368 depth as a predictor, similar as Moran et al. (2021) used latitude to predict Chinook stock
369 composition.

370 **Acknowledgements**

371 The authors wish to express their thanks to John Wallace of the Northwest Fisheries
372 Science Center, National Oceanic and Atmospheric Administration, who extracted PacFIN
373 logbook data, and to staff of the Washington Department of Fish and Wildlife for their help
374 providing details on the data collection process. We also thank Dr. James Hastie, of the
375 Northwest Fisheries Science Center, National Oceanic and Atmospheric Administration, and two
376 anonymous reviewers whose valuable suggestions help improve the manuscript.

377 **Funding sources**

378 This research did not receive any specific grant from funding agencies in the public,
379 commercial, or not-for-profit sectors.

380 **References**

- 381 Alverson, D.L., Pruter, A.T., Ronholt, L.L. 1964. A study of demersal fishes and fisheries of the
382 northeastern Pacific Ocean. Institute of Fisheries, University of British Columbia.
- 383 Branch, T. A. 2008. Not all fisheries will be collapsed in 2048. *Marine Policy*, 32:38–39.
- 384 Branch, T.A., Jensen, O.P., Ricard, D., Ye, Y., Hilborn, R.A.Y., 2011. Contrasting global trends
385 in marine fishery status obtained from catches and from stock assessments. *Conservation*
386 *Biology*, 25(4): 777-786.
- 387 Bizzarro, J.J., Broms, K.M., Logsdon, M., Yoklavich, M.M., Kuhnz, L., Summers, A.P. 2014.
388 Spatial segregation in eastern North Pacific skate assemblages. *PLOS ONE*, 9: e109907.
- 389 Bizzarro, J.J. 2015. Comparative resource utilization of eastern North Pacific skates (Rajiformes:
390 Rajidae) with applications for ecosystem-based fisheries management. Ph.D.
391 Dissertation. University of Washington, School of Aquatic & Fishery Sciences. Seattle,
392 WA.
- 393 Bonfil, R. 1994. Overview of world elasmobranch fisheries. *FAO Fisheries Technical Paper No*
394 341.
- 395 Cope, J. and Gertseva, V., 2020. A new way to visualize and report structural and data
396 uncertainty in stock assessments. *Canadian Journal of Fisheries and Aquatic Sciences*,
397 77(8): 1275-1280.
- 398 Dulvy, N.K., Metcalfe, J.D., Glanville, J., Pawson, M.G. and Reynolds, J.D., 2000. Fishery
399 stability, local extinctions, and shifts in community structure in skates. *Conservation*
400 *Biology*, 14(1): 283-293.
- 401 Dulvy, N.K., Reynolds, J.D., 2002. Predicting extinction vulnerability in skates. *Conservation*
402 *Biology*, 16(2): 440-450.

403 Ebert, D.A., and Compagno, L.J. 2007. Biodiversity and systematics of skates (chondrichthyes:
404 *Rajiformes: Rajoidei*). In Biology of skates. Springer. pp. 5-18.

405 Eschmeyer, W. N. and E. S. Herald. 1983. A field guide to Pacific Coast fishes of North America
406 From the Gulf of Alaska to Baja California. Houghton Mifflin, Boston.

407 Farrugia, T.J., Goldman, K.J., Tribuzio, C., and Seitz, A.C. 2016. First use of satellite tags to
408 examine movement and habitat use of big skates *beringraja binoculata* in the Gulf of
409 Alaska. Marine Ecology Progress Series, 556: 209-221.

410 Gertseva, V. V., S. E. Matson, I. G. Taylor, J. Bizzaro, Wallace, J. R., 2019. Stock assessment of
411 the Longnose Skate (*Beringraja rhina*) in state and Federal waters off California, Oregon
412 and Washington. Pacific Fishery Management Council, Portland, OR.

413 Gertseva, V.V. and Matson, S.E., 2021. Right on target: using data from targeted stocks to
414 reconstruct removals of bycatch species, a case study of longnose skate from Northeast
415 Pacific Ocean. Fisheries Research, 236: 105841.

416 Gertseva, V.V., 2009. The population dynamics of the longnose skate, *Raja rhina*, in the
417 northeast Pacific Ocean. Fisheries Research, 95(2-3): 146-153.

418 Haddon, M. 2011. Modelling and Quantitative Methods in Fisheries. 2nd Edition. Chapman and
419 Hall/CRC.

420 Harry, G., Morgan, A.R. 1961. History of the trawl fishery, 1884-1961. Oregon Fish
421 Commission Research Briefs, 19: 5-26.

422 Hilbe, J.M., 2009. Logistic regression models. Chapman and hall/CRC.

423 Hilborn, R, Walters, C.J., 2003. Quantitative fisheries stock assessment: choice, dynamics and
424 uncertainty. Springer.

425 Hilborn, R., Branch, T.A., Magnusson, E., Minte-Vera, C.V., Scheuerell M.D., Valero J.L.,
426 2003. State of the world's fisheries. *Annual Review of Environment and Resources*,
427 28(1): 359-99.

428 Karnowski, M., Gertseva, V.V., Stephens, A. 2014. Historical Reconstruction of Oregon's
429 Commercial Fisheries Landings. 2014-02, Oregon Department of Fish and Wildlife,
430 Newport, Oregon.

431 Keller, A.A., Wallace, J.R., and Methot, R.D. 2017. The Northwest Fisheries Science Center's
432 West Coast Groundfish Bottom Trawl Survey: history, design, and description. U.S.
433 Department of Commerce, NOAA Technical Memorandum NMFS-NWFSC-136.
434 <https://doi.org/10.7289/V5/TM-NWFSC-136>.

435 King, M. 2013. *Fisheries biology, assessment and management*. John Wiley & Sons.

436 King, J.R., McFarlane, G.A., 2003. Marine fish life history strategies: applications to fishery
437 management. *Fisheries Management and Ecology*, 10(4): 249-264.

438 King, J. R., G. A. McFarlane, Gertseva, V. V., Gasper, J., Matson, S. E., Tribuzio, C. A., 2017.
439 Shark Interactions With Directed and Incidental Fisheries in the Northeast Pacific Ocean:
440 Historic and Current Encounters, and Challenges for Shark Conservation, in: Larson, S.,
441 Lowry, D. (Eds.), *Northeast Pacific Shark Biology, Research, and Conservation, Part B*.
442 Academic Press, London, pp. 9-44.

443 Love, M.S., Bizzarro, J.J., Cornthwaite, A.M., Frable, B.W. and Maslenikov, K.P., 2021.
444 Checklist of marine and estuarine fishes from the Alaska–Yukon border, Beaufort Sea, to
445 Cabo San Lucas, Mexico. *Zootaxa*, 5053(1): 1-285.

446 Love, M. S., M. Yoklavich, and L. Thorsteinson. 2002. *The rockfishes of the northeast Pacific*.
447 University of California Press, Berkeley.

448 Maier, M., 2014. DirichletReg: Dirichlet regression for compositional data in R. Research Report
449 Series / Department of Statistics and Mathematics; No. 125.

450 Martin, L., Zorzi, G.D. 1993. Status and review of the California skate fishery. In Conservation
451 biology of elasmobranchs (Ed. Branstetter, S.), p 39-52. NOAA Technical Report NMFS
452 115.

453 Matson, S.E., Taylor, I.G., Gertseva, V.V. and Dorn, M.W., 2017. Novel catch projection model
454 for a commercial groundfish catch shares fishery. *Ecological modelling*, 349, pp.51-61.

455 Matson, S. E., Gertseva, V. V., 2020. Resolving associative patterns in life history parameters
456 among marine fish stocks in the Northeast Pacific Ocean. *Journal of Sea Research*, 156:
457 101837.

458 Matson, S.E., Taylor, I.G., Gertseva, V.V. and Dorn, M.W., 2017. Novel catch projection model
459 for a commercial groundfish catch shares fishery. *Ecological Modelling*, 349: 51-61.

460 McEachran, J., and Miyake, T. 1990. Zoogeography and bathymetry of skates (chondrichthyes,
461 rajidae). *Elasmobranchs as living resources. Advances in biology, Ecology, Systematics*
462 *and the status of the fisheries*, pp. 305-326.

463 Mecklenburg, C. W., T. A. Mecklenburg, and L. K. Thorsteinson. 2002. *Fishes of Alaska*.
464 American Fisheries Society, Bethesda, Maryland.

465 Miller, R.R., Field, J.C., Santora, J.A., Schroeder, I.D., Huff, D.D., Key, M., Pearson, D.E. and
466 MacCall, A.D., 2014. A spatially distinct history of the development of California
467 groundfish fisheries. *PLoS One*, 9(6): e99758.

468 Moran, P., Tuttle, V.J., Bishop, S. and LaVoy, L., 2021. Compositional forecasting of Chinook
469 Salmon Evolutionarily Significant Units in bycatch for Pacific Hake fisheries. *bioRxiv*.
470 doi: <https://doi.org/10.1101/2021.11.29.470462>

471 Rogers, J.B. 1994. Assemblages of groundfish caught using commercial fishing strategies off the
472 coasts of Oregon and Washington from 1985-1987. Ph.D. Dissertation, Oregon State
473 University, Oregon.

474 Somers, K. A., C. E. Whitmire, K. E. Richerson, and V. J. Tuttle. 2023. Fishing Effort in the
475 2002–21 U.S. Pacific Coast Groundfish Fisheries. U.S. Department of Commerce,
476 NOAA Technical Memorandum NMFS-NWFSC-184.

477 Snytko, V.A. 1987. New data on the distribution of some species of fish in the North Pacific.
478 Journal of Ichthyology, 27:142-146.

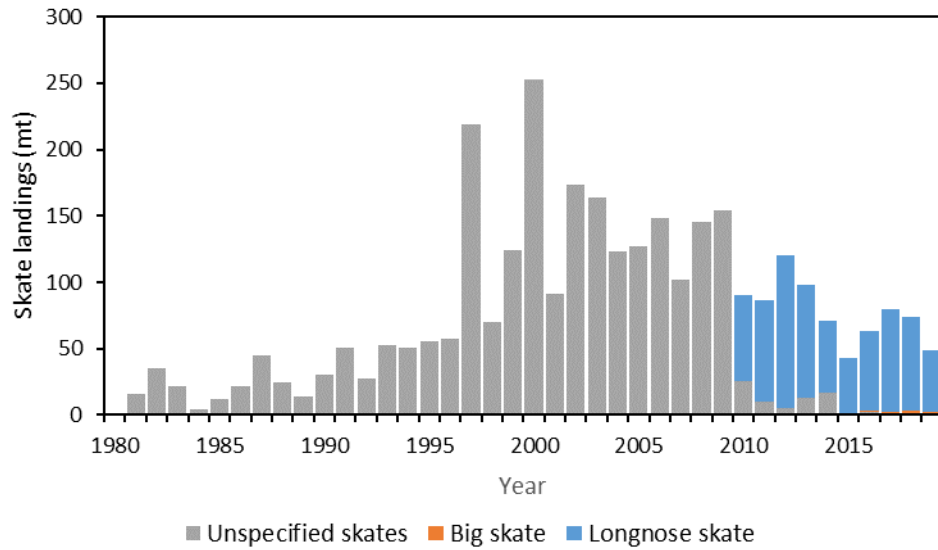
479 Taylor, I.G., Gertseva, V., Methot Jr, R.D. and Maunder, M.N., 2013. A stock–recruitment
480 relationship based on pre-recruit survival, illustrated with application to spiny dogfish
481 shark. Fisheries Research, 142: 15-21.

482 Taylor, I.G., Gertseva, V., Stephens, A., Bizzarro, J., 2019. Status of Big Skate (*Beringraja*
483 *binoculata*) Off the U.S. West Coast, 2019. Pacific Fishery Management Council,
484 Portland, OR.

485 Tolimieri, N. and Levin, P.S., 2006. Assemblage structure of eastern Pacific groundfishes on the
486 US continental slope in relation to physical and environmental variables. Transactions of
487 the American Fisheries Society, 135(2): 317-332.

488 Last, P.R., White, W.T., Carvalho, M.R. de, Séret, B., Stehmann, M.F.W and Naylor, G.J.P.
489 2016. Rays of the World. CSIRO Publishing, Melbourne.

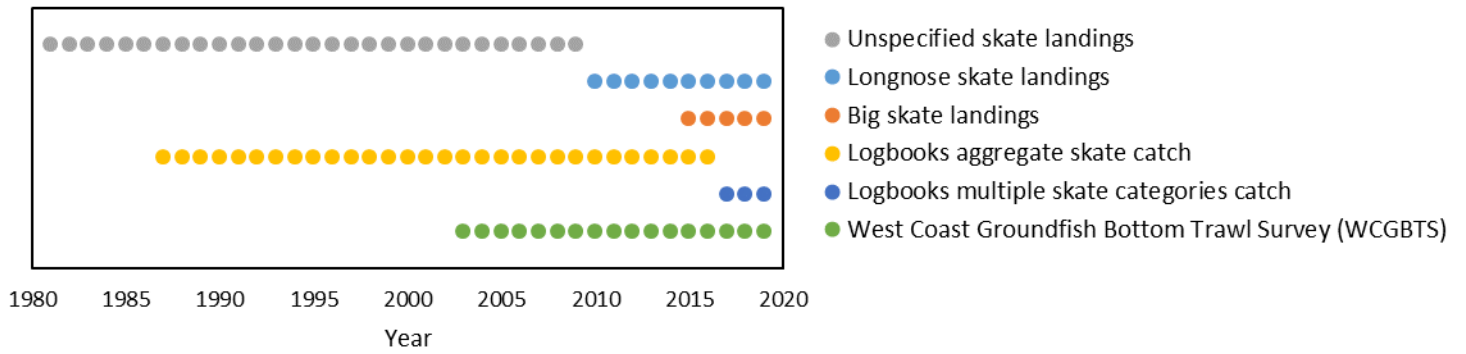
490



491

492 Figure 1. Commercial catch of skates in the U.S. West Coast groundfish demersal trawl fishery
 493 landed in Washington State ports, between 1981 and 2019, as reported in fish tickets. Longnose
 494 skate and big skate have been reported separately since 2010 and 2015, respectively.

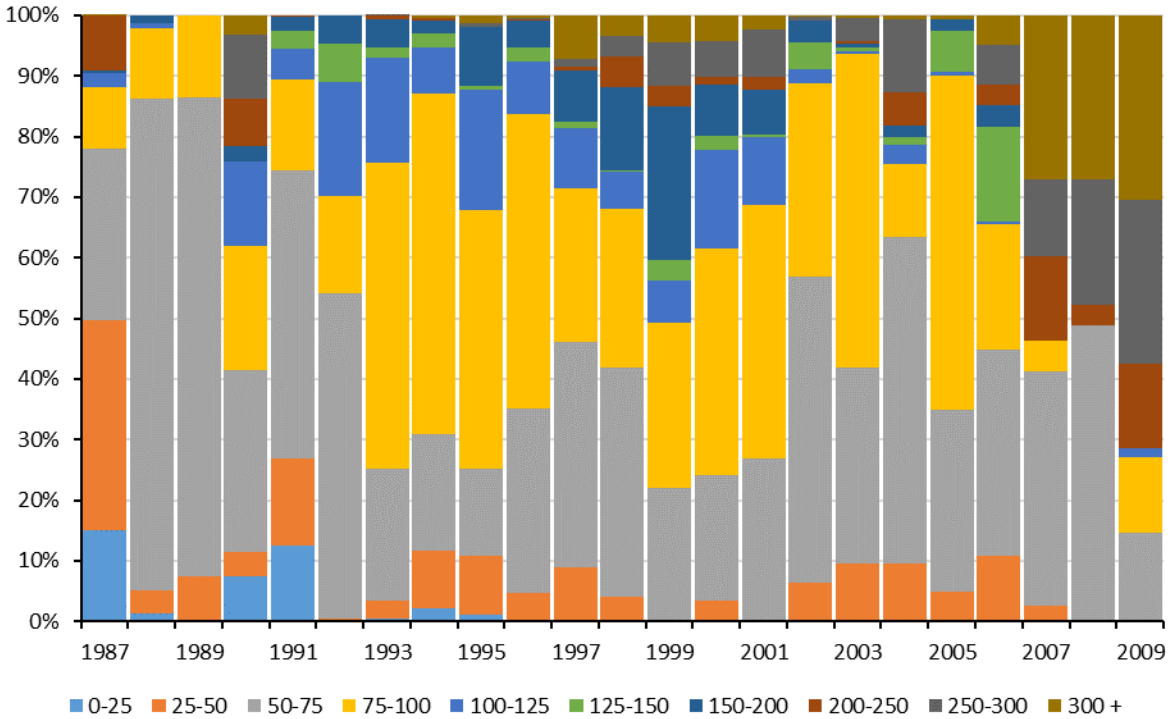
495



496

497 Figure 2. Summary of data sources used in the analysis by year.

498

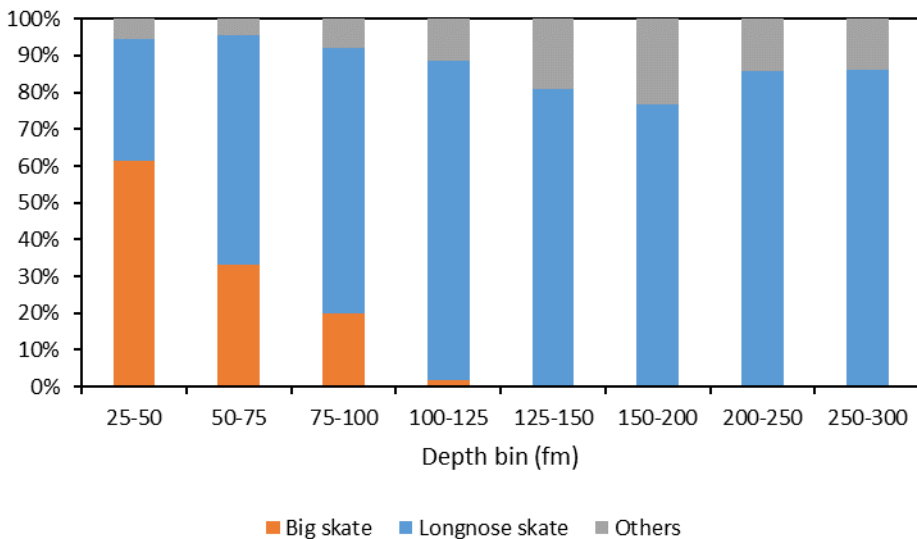


499

500 Figure 3. Annual aggregate skate catch landed in Washington State ports, by depth bin (in fm),
 501 informed by logbook data. Depth bins are sorted, from shallow at the bottom, growing deeper
 502 toward the top of the plot.

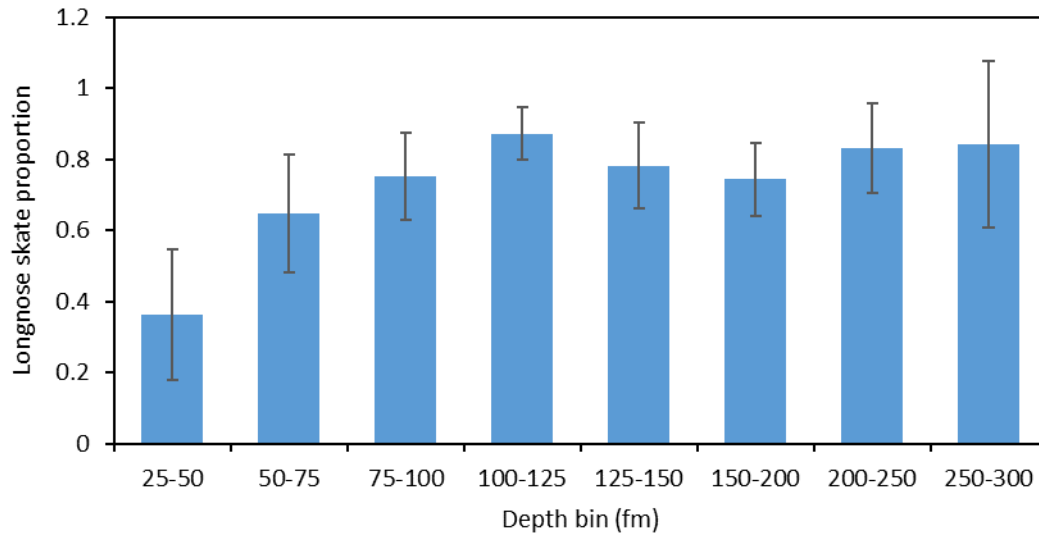
503

504

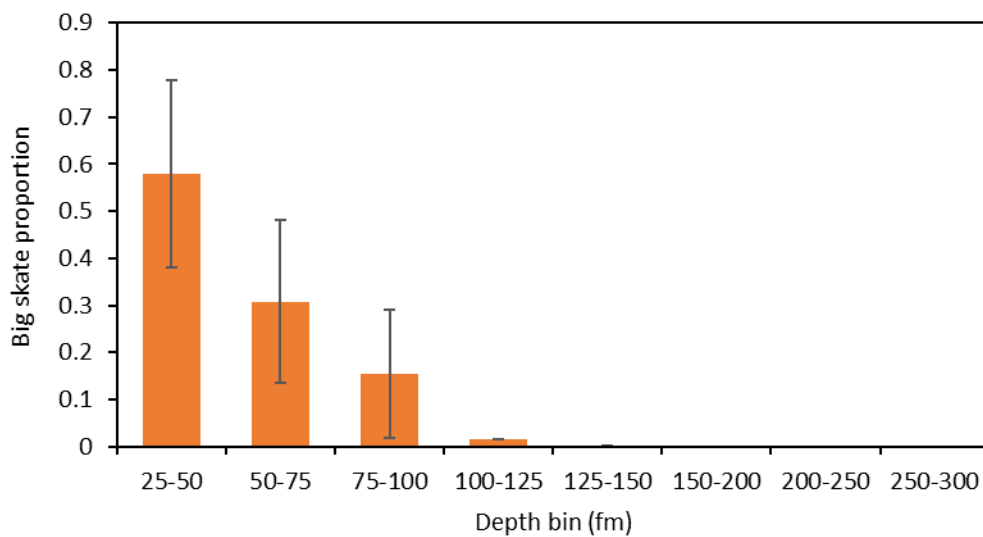


505

506 Figure 4. Average percent contribution of individual skate species to the aggregate skate catch by
507 depth bins during the West Coast Groundfish Bottom Trawl Survey, years 2003 – 2019. Other
508 skates primarily consist of sandpaper skate (*Bathyraja kincaidii*) and rougtail skate (*Bathyraja*
509 *trachura*).



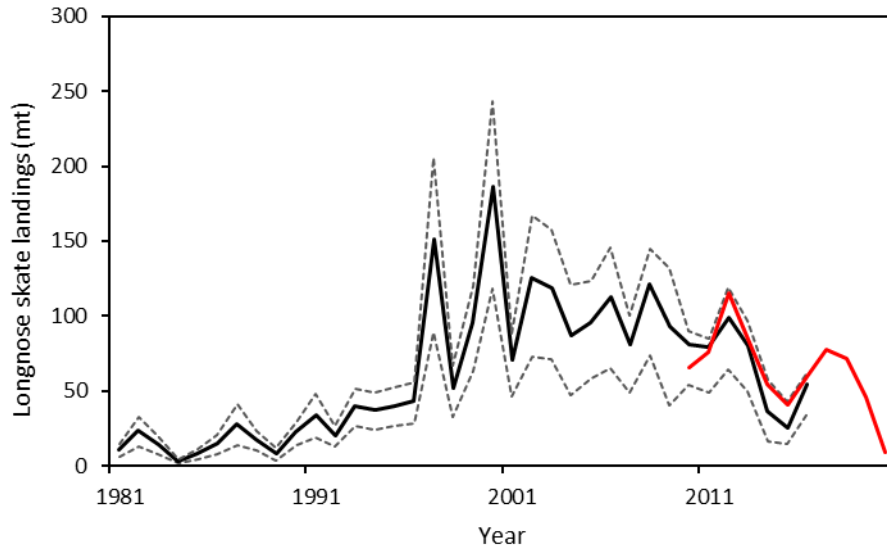
510 A)



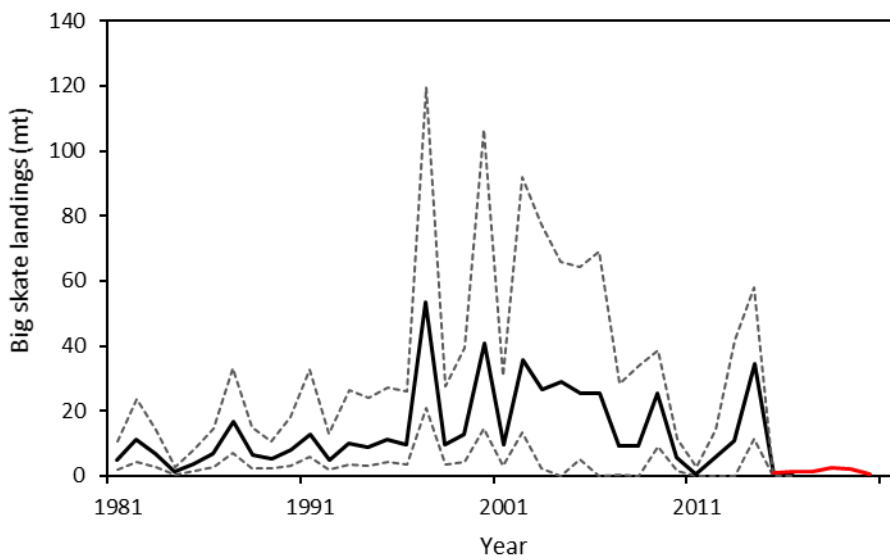
511

512 B)

513 Figure 5. Mean proportion of longnose skate (A) and big skate (B) by depth bin in the West
 514 Coast Groundfish Bottom Trawl Survey (WCGBTS), 2003 – 2019. Error bars represent one
 515 standard deviation above and below the mean (calculated across all years).



516 A)

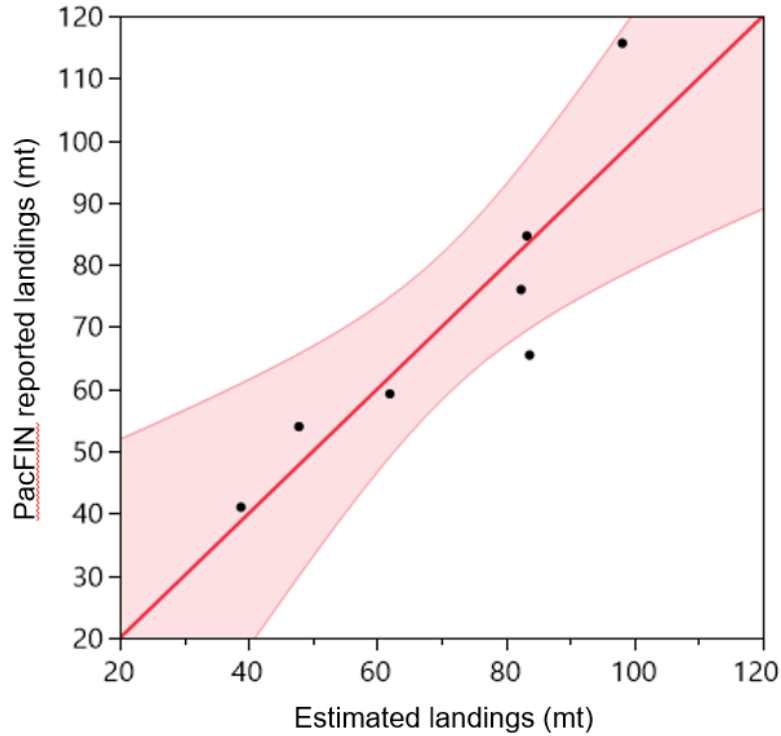


517

518 B)

519 Figure 6. Estimated landings of longnose skate (A) and big skate (B) based on combination of
 520 survey and logbook data (black lines). Dashed lines show uncertainty intervals around the
 521 reconstructed landings for each species, calculated from applying ± 2 standard deviations of
 522 depth specific proportions for each species within WCGBTS data. Red lines are landings

523 reported in PacFIN, calculated based on species composition port sampling (longnose skate
524 landings reported separately from other skate from 2010 forward and big skate from 2015
525 forward).



526

527 Figure 7. Comparison of fit between estimated longnose skate landings, versus longnose skate
 528 landings reported in PacFIN for 2010-2016 ($R^2=0.795$, $p=0.0007$). Colored area around the line
 529 represented 95% confidence region. Prior to 2010, longnose skate landings were reported in
 530 PacFIN as part of skate aggregate; after 2016, longnose skate data was reported separately from
 531 other skates in logbooks.