

Inequality in the economic impacts from climate shocks in fisheries: the case of harmful algal blooms

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1 **Abstract**

2 Climate impacts disproportionately affect people that are most vulnerable and least able
3 to adapt. The extent to which these equity impacts extend to fishing communities in the
4 developed world is a question that has received surprisingly little attention. Here we
5 explore the distributional impacts of a climate shock in one of the largest and most
6 valuable fisheries on the West Coast of the United States. Specifically, we examine
7 whether a series of two harmful algal blooms (HABs), occurring during the 2014-2016
8 Northeast Pacific Marine Heatwave, differentially affected small and large vessels in the
9 commercial California Dungeness crab fishery. The HAB events were managed with
10 localized fishery closures in response to elevated levels of the HAB toxin, domoic acid,
11 in crab tissue. We find evidence that large vessels had a greater ability to mitigate losses
12 from the HAB events. Thus, the proportion of total revenue going to small-vessel
13 operators and the proportion of small-vessel participation in the fishery fell in response to
14 the HAB events in several California fishing ports. Our results, therefore, offer empirical
15 evidence that climate impacts on fishing communities are not uniform and offer insights
16 into potential alternative adaptation strategies for different ocean user groups.

17
18 **Keywords:** Climate change, Dungeness crab, fisheries, harmful algae, harmful algal
19 blooms, fishery management, economic impacts, equity.

20

21 **1. Introduction**

22 The societal impacts of climate change are likely to be disproportionately borne by
23 populations that are most vulnerable and least able to adapt (Mendelsohn, Dinar, &
24 Williams, 2006; Mendelsohn, Emanuel, Chonabayashi, & Bakkensen, 2012; Park,

25 Hallegatte, Bangalore, & Sandhoefner, 2015; Winsemius et al., 2015; Hallegatte &
26 Rozenberg, 2017). Inequities in the distribution of climate impacts across a population,
27 hereafter referred to as the “distributional impacts” of climate variability and change, are
28 potentially compounded for vulnerable populations due to a greater dependence on
29 natural resources (Paavola & Adger, 2002; Adger, Huq, Brown, Conway, & Hulme,
30 2003) and a lack of social, financial, and other resources to cope with and adapt to
31 environmental change (Kates, 2000; Cutter, Boruff, & Shirley, 2003; Hallegatte &
32 Rozenberg, 2017).

33 Fishers and fishing communities are closely connected to the environment and are
34 expected to be impacted by climate change in numerous ways (see Hollowed et al. (2013)
35 for a review). Key impacts of climate change on fisheries include shifts in the geographic
36 distribution of fish species that will disrupt fisheries in regions that become uninhabitable
37 for certain fish species (A. L. Perry, Low, Ellis, & Reynolds, 2005; Pinsky, Worm,
38 Fogarty, Sarmiento, & Levin, 2013; Morley et al., 2018) and changes in fisheries
39 productivity (e.g. Britten, Dowd, & Worm, 2016) which will alter biological reference
40 points used in management, e.g. maximum sustainable yield (Cheung et al., 2010).

41 Climate change is also expected to directly affect fisheries via its impacts on seafood
42 safety through, for example, climate-driven changes in the frequency, magnitude, and
43 geographic scope of harmful algal blooms (HABs) (Hallegraeff, 1993, 2010; Moore et
44 al., 2008; Backer & Moore, 2010; Marques, Nunes, Moore, & Strom, 2010; Berdalet et
45 al., 2016).

46 HABs occur when colonies of algae proliferate and cause damage to the
47 environment and/or produce toxic effects on people and other marine and aquatic life

48 (Erdner et al., 2008). Many HABs produce toxins that accumulate in shellfish, and while
49 they may not jeopardize shellfish survival, predators such as marine mammals and
50 humans may suffer illness or death if contaminated shellfish are consumed. These HAB
51 events are managed with fishery closures that are effective at preventing human exposure
52 to HAB toxins and protecting human health, but can cause severe economic and
53 sociocultural disruption to fishery-dependent communities (Bauer, 2006; Ritzman et al.,
54 2018).

55 Understanding the economic and distributional implications of climate change in
56 fisheries and fishing communities is critical as fisheries managers balance multiple, and
57 often competing, goals including achieving biological sustainability and promoting
58 economic efficiency and equity (Charles, 1992; Rosenberg, Fogarty, Sissenwine,
59 Beddington, & Shepherd, 1993; Kailin Kroetz, Sanchirico, & Lew, 2015). Equity impacts
60 depend on both climate-driven changes to fisheries and individual-, fleet-, and
61 community-level adaptive responses to these changes. For example, the economic
62 impacts of species range shifts and productivity changes depend on the ability of
63 individual fishers to adjust spatial harvest patterns, change gear types, and/or target new
64 species (Sumaila, Cheung, Lam, Pauly, & Herrick, 2011; Pinsky & Fogarty, 2012). Most
65 of our knowledge of the distributional impacts of climate change on fisheries comes from
66 examples in the developing world, with surprisingly little attention given to developed-
67 world fisheries.

68 Here we explore a case study of the distributional impacts of the 2014-2016
69 Northeast Pacific Marine Heatwave that resulted in a series of two toxic HABs (caused
70 by species in the marine diatom genus *Pseudo-nitzschia*) that severely impacted the

71 commercial Dungeness crab fishery. The Dungeness crab fishery is among the most
72 valuable on the U.S. West Coast, with annual ex-vessel revenues in the hundreds of
73 millions of dollars. Prior to the two HAB events examined here, all of the previous major
74 *Pseudo-nitzschia* HAB events on the West Coast have occurred after periods of warming
75 associated with El Nino and the Pacific Decadal Oscillation (McCabe et al., 2016;
76 McKibben et al., 2017), providing evidence for a relationship between warming and
77 extreme HABs. Therefore, continued warming due to climate change may worsen HABs
78 of *Pseudo-nitzschia* on the West Coast, potentially leading to longer, more frequent, and
79 more widespread fishery closures.

80 The two HAB events explored here occurred in 2015 and 2016, impacting the
81 2015/2016 and 2016/2017 Dungeness crab fishing seasons, respectively. The HAB events
82 were managed with localized fishery closures, that were sequentially opened once the
83 crab in their boundaries were deemed safe for human consumption. Opening individual
84 areas while others remain closed introduces heterogeneity in fishing opportunities based
85 on geography as well as characteristics of individual vessels and fishing businesses.
86 Additionally, extended closures can compress the fishing season to the advantage of
87 vessels with a greater capacity for speed in harvest.

88 We ask the question: were economic impacts of the 2015 and 2016 HAB events,
89 and the corresponding management response, equally shared across the commercial
90 Dungeness crab fishing fleet in California? We focus on the differences between impacts
91 to small and large vessels, with small vessels defined as less than 40 feet in length, or less
92 than roughly 12.2 meters. Small vessels have been shown to be less profitable (McCay,
93 2004), receive lower returns to their labor inputs (Boncoeur, Coglán, Le Gallic, &
94

95 Pascoe, 2000), and can be restricted in their ability to respond to geographic shifts in
96 fishing opportunities due to reduced mobility (Young et al., 2019), suggesting that small-
97 vessel operators and communities dominated by small fishing vessels may also be less
98 resilient to climate shocks.

99 Our analysis draws on data collected for every Dungeness crab delivery, by both
100 small and large vessels, in California from the 2010/2011 through the 2016/2017 fishing
101 seasons. First, we explore evidence of spatial shifts in vessel deliveries patterns and, if
102 shifts are detected, whether those shifts are consistent with the idea that large vessels had
103 an increased ability to respond to localized fisheries closures during the two HAB events.
104 Second, we estimate the changes in distributional outcomes, measured in terms of
105 revenue and participation, associated with the two HAB events.

106 Our results can inform the design of future fisheries policies that aim to promote
107 equity in a world of environmental change. Our findings also highlight the importance of
108 both communities-of-place, such as individual ports, and communities-of-practice, such
109 as groups of vessels with similar characteristics and constraints, in developing
110 management strategies that meet equity objectives in the face of climate shocks (Martin
111 & Hall-Arber, 2008).

112

113 **2. Background and Data**

114 Dungeness crab is the most valuable fishery on the U.S. West Coast. In California
115 specifically, the fishery generated average ex-vessel revenues of roughly \$60 million
116 USD over the last decade. California Dungeness crab are harvested in a limited-entry
117 derby fishery characterized by overcapitalization (Hackett, Krachey, Dewees, Hankin, &
118 Sortais, 2003) and a compressed season length where the majority of crab is landed

119 during the first 6 weeks of a 7-month season (Deweese et al., 2004). Despite the current
120 race to fish and the implied opportunity to improve profitability in this fishery, the
121 Dungeness crab fishery is an important source of income for participants. In a survey of
122 California Dungeness crab permit holders, over 80% of respondents stated that over 60%
123 of their gross income came from fishing Dungeness crab (Deweese et al., 2004).

124 The Dungeness crab fishery is managed at the state level with a ban on female
125 harvest, size restrictions on the harvest of male crabs, and seasonal closures to protect
126 molting crab (Porzio, 2015). The California Dungeness crab fishery is divided into
127 northern and central management districts at the Mendocino-Sonoma County border. The
128 central management district coincides with fishing district 10 and the northern district is
129 further subdivided into fishing districts 6 (Crescent City and Eureka), and 7 (Fort Bragg).
130 The central California season runs from November 15th through June 30th and the
131 northern California season runs from December 1st through July 15th. The season opening
132 dates are subject to delay if meat content is below established threshold values (i.e. <
133 25%) or if levels of the HAB toxin, domoic acid, are higher than established regulatory
134 thresholds (> 30 ppm in crab viscera).

135 Due to concern that season delays would allow some vessels to target multiple
136 openings and disadvantage vessels in delayed areas, Fair Start Provisions were
137 established in 1997 (Cal. Fish and Game Code FGC § 8279.1). The law maintains that,
138 with season delays of a California fishing district, persons landing Dungeness crab
139 outside of that district must wait 30 days after the delayed district opens before landing
140 crab in the delayed district. For example, if a district was delayed, individuals that had
141 previously fished outside of that district, either in another California fishing district or in

142 Oregon or Washington, would have to wait 30 days after the delayed district was opened
143 before participating in the delayed district, thereby affording the delayed district 30 days
144 of protections. Importantly, the regulations, as applied to the 2015/2016 and 2016/2017
145 seasons, stipulated that the open date for a delayed district is the date on which any part
146 of the delayed district is open to fishing.¹ In 2018, the law was amended to apply to any
147 delayed area regardless of whether the area is in a district that had previously opened.

148 The 2014-2016 Northeast Pacific Marine Heatwave, otherwise known as the
149 “Blob”, extended from Alaska to Mexico and produced peak near-surface (upper ~ 100
150 m) seawater temperature anomalies of 2.5°C, exceeding 3 standard deviations from
151 normal (Bond et al. 2015). The anomalously warm water caused a number of unusual
152 biological events, including a massive HAB of *Pseudo-nitzschia* that began in the spring
153 of 2015 and resulted in multiple fishery closures along the entire West Coast for many
154 species of fish and shellfish, including Dungeness crab, due to unsafe levels of domoic
155 acid (McCabe et al., 2016). Closures in the California Dungeness crab fishery were more
156 severe than in Oregon and Washington, lasting nearly 4.5 months in the 2015/2016
157 season. The 2016/2017 season was also delayed due to elevated levels of domoic acid in
158 California Dungeness crab, but the incidences were relatively localized, and closures
159 were shorter in duration and only implemented in a subset of California ports.

160 Figure 1 shows the rollout of the commercial Dungeness crab season openings for
161 the 2015/2016 and 2016/2017 fishing seasons. In the 2015/2016 season, the central
162 district opened on March 26th, roughly 4.5 months late. It was not until 36 days later
163 (May 12th) that the northern district, with the exception of Eureka, was open to fishing.

¹ On January 1st, 2017 FGC § 8279.1 was amended so that the law applied to vessels rather than persons.

164 Eureka was finally open on May 26th, or 50 days after the central district opened. The
165 extended delays in the 2015/2016 season shortened the season length relative to previous
166 seasons by pushing the season into the period when crabs begin to molt and the season is
167 officially closed.

168 The timing of HAB closures and protections under the Fair Start Provisions in the
169 2015/2016 and 2016/2017 seasons varied over space. Specifically, in the 2015/2016
170 fishing season all fishing districts were considered delayed and protected under the Fair
171 Start Provisions. However, a section of district 6, containing Crescent City, opened on
172 May 12th while the remainder of district 6, containing Eureka, opened on May 26th. Thus,
173 according to the Fair Start Provisions, individuals who had landed Dungeness crab in
174 other California districts, or in Oregon or Washington, were eligible to fish in any part of
175 district 6 on June 11th, providing the Crescent City area with 4 weeks of protection and
176 the Eureka area with roughly 2 weeks of protection. In summary, extended delays in the
177 2015/2016 season led to greater compression of season length and, while all 3 fishing
178 districts were delayed in the 2015/2016 season, the Eureka area experienced the longest
179 delays and least protections under the Fair Start Provisions.

180 In the 2016/2017 season, the central district, with the exception of Bodega Bay,
181 opened on time (November 15th). In the northern district, Crescent City and Eureka
182 opened on time (December 1st), Bodega Bay area was delayed until December 3rd, and
183 the Fort Bragg area was opened in sections with the first opening on December 26th and
184 the final opening on January 16th. In this fishing season, only fishing district 7 (Fort
185 Bragg) was protected under the Fair Start Provisions, because even though Bodega Bay
186 was delayed, Bodega Bay is within the central district (fishing district 10) some of which

187 opened on time. Additionally, the 30-day fair start waiting period for Fort Bragg ended
188 on January 25th, offering only 9 days of protection to the last section of the Fort Bragg
189 area to open.

190 There were several mechanisms by which the two HAB events potentially
191 affected outcomes for small vessels. First, the compressed length of the 2015/2016 season
192 may have affected small vessels if the ability to compete in a compressed season is driven
193 by vessel size. Second, in the 2015/2016 season, spatial spillover of vessels from Eureka
194 (the area with the longest delays) to other ports may have may have disproportionately
195 increased competition for small vessels in other ports. Third, in the 2015/2016 season,
196 small vessels based out of Eureka may have faced increased competition from large
197 vessels fishing in multiple areas if large vessels took advantage of the relatively short
198 period of protections under the Fair Start Provisions. Fourth, in the 2016/2017 season, the
199 delayed opening of Bodega Bay may have been interpreted as a signal that northern
200 California districts would again face extended delays, for a second consecutive season.
201 Larger, mobile vessels responding to this signal could have left northern port areas and
202 increased competition for small vessels in other areas. Fifth, when Bodega Bay finally
203 opened in the 2016/2017 season, it was not protected under the Fair Start Provisions,
204 therefore larger, more mobile vessels participating in multiple openings could participate
205 in the Bodega Bay opening without restrictions, enhancing competition for small vessels.
206 Sixth, the gradual opening of the Fort Bragg area may have created opportunities for
207 larger, more mobile vessels to make their way up north and participate in multiple
208 openings, enhancing competition for small vessels as they went.

209 Indeed, fishers reported that the larger and most mobile vessels were able to
210 participate in multiple openings, providing anecdotal evidence of distributional impacts
211 from the climate shock (Callahan, 2017a, 2017b). Our analysis seeks to explore the
212 empirical support behind these anecdotal reports. Specifically, we quantify the impact of
213 the HAB closures in the 2015/2016 and 2016/2017 seasons on three response variables:
214 1) spatial landings patterns for small and large vessels; 2) the proportion of total revenue
215 going to small vessels; and 3) the proportion of small vessels participating in the fishery
216 relative to total participation. Vessel landings and revenue data come from confidential
217 fish tickets, collected by state managers, and archived in the PacFIN database housed at
218 the Pacific States Marine Fisheries Commission (PSMFC). The fish ticket data are
219 essentially sales receipts generated upon the delivery of a harvester's landings to a
220 processor. We match the fish ticket data with state data on vessel registration, which
221 includes vessel length. Large vessels are defined as vessels ≥ 40 feet in length following
222 Kasperski and Holland (2013). We first filter the data to remove any observations outside
223 of the legal fishing season. We then aggregate landings data, for small and large vessels,
224 to the week and port-group level, where port groups are defined by the PSMFC (see
225 Figure 1).² Our dataset covers all landings in the 2010/2011 through 2016/2017 fishing
226 seasons, allowing us to observe outcomes before and after the HAB closures, utilizing the
227 before-period observations in a counterfactual analysis.

² Port group definitions can be found at https://pacfin.psmfc.org/pacfin_pub/data_rpts_pub/code_lists/pc_tree.txt.

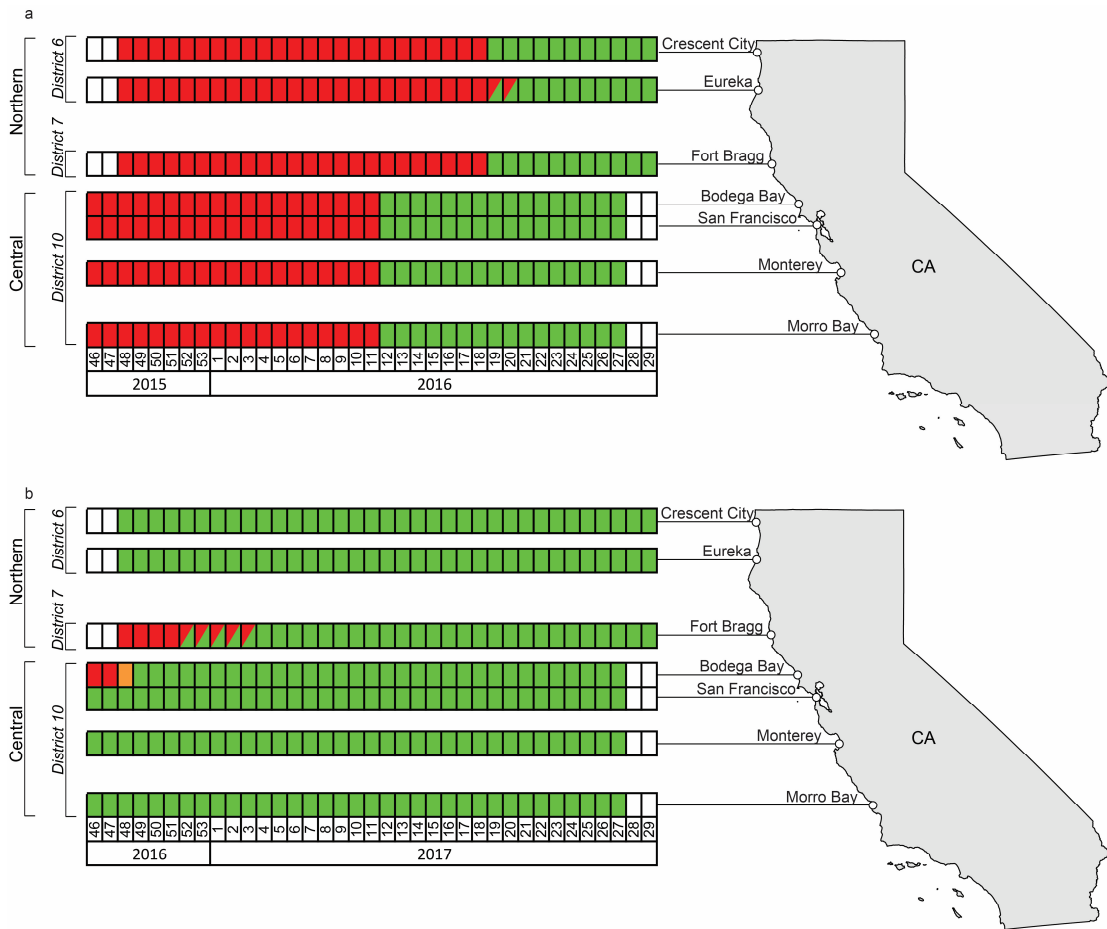


Figure 1: Map of California Dungeness crab port group closures in the 2015/2016 season (top panel) and 2016/2017 season (bottom panel). Grid cells represent weeks that the port area is opened to fishing in absence of any delays. Red represents the entire week in the port area is closed due to HABs, yellow represents a fraction of the week was delayed due to HABs, and green represents an open fishery. Cells that are half red and half green mean that a fraction of the port group was open for the entire week and a fraction of the port group was closed for the entire week.

228

229 **3. Methods**

230 **3.1 Examining Distributional Shifts in Landings**

231 As a first step, we examine spatial shifts in Dungeness crab pounds landed during the

232 2015/2016 and 2016/2017 seasons. We create a “baseline” profile for each vessel,

233 measuring the proportion of Dungeness crab pounds (based on fish tickets) landed at each

234 port group. This baseline represents vessel behavior over a 3-season period (fishing
235 seasons 2011/2012, 2012/2013, and 2013/2014).³

236 We use the baseline landings profiles to separately group large and small vessels
237 with similar delivery patterns, applying a k-means clustering analysis to the 3-season
238 baseline period. We perform two cluster analyses in total using baseline landings for all
239 vessels falling into each of the following groups: (1) small vessels active in 2015/2016
240 and/or 2016/2017; and (2) large vessels active in 2015/2016 and/or 2016/2017. Thus, in
241 order to be included in the analysis, a vessel must have been active in at least one HAB
242 season and at least one baseline season. The criteria exclude 9% and 11% of vessels
243 active in the 2015/2016 and 2016/2017 seasons respectively.

244 Clusters identified in the analysis represent groups of vessels with similar spatial
245 landings patterns, which are then used to assign a “homeport” to each vessel. In this
246 context, homeport is defined as the port group at which a vessel landed a majority of its
247 catch, of Dungeness crab, during the baseline period. The dissimilarity matrix for the
248 clustering analysis was calculated using a Bray-Curtis distance (Bray & Curtis, 1957),
249 which is well suited for data where multiple abundance or coverage measures
250 characterize a single unit. We use the Ward method (Ward Jr & Hook, 1963) for
251 clustering and determine the optimal number of clusters using a combination of
252 screeplots, the Tibshirani et al. (2001) gap statistic, and visual assessments of heatmaps
253 showing delivery frequencies per port group during the baseline period.

254 The cluster analysis allows us to examine evidence for a spatial shift in
255 Dungeness crab pounds landed the two HAB seasons, relative to the “before” period (or

³ We exclude the 2014/2015 season from our baseline period, used to define a vessel’s home port, because shifts in the spatial distribution of Dungeness crab (from north to south) led to changing deliveries patterns and vessels fishing outside of their homeports (Spencer, 2014).

256 the 2010/2011-2014/2015 seasons), for groups of vessels assigned to the same homeport.⁴
257 The spatial shift in pounds delivered for each cluster is calculated as the difference in the
258 proportion of pounds delivered to each port group during the 2015/2016 and 2016/2017
259 seasons and the proportion of pounds delivered to each port group in the before period.
260 We descriptively analyze the sign and magnitude of these spatial shifts in pounds
261 delivered for small and large vessels using heatmaps. Note that for confidentiality
262 reasons, results with fewer than three vessels in a homeport/landing port combination are
263 not shown. Therefore, for example, if two vessels from a given homeport shifted all of
264 their landed pounds into another port, these shifts would not be recorded in the heatmap.
265 Note also that the heatmaps will show the changes in total proportions over a season. It
266 does not show intra-seasonal dynamics, which may be important in this context, but are
267 beyond the scope of our analysis.

268

269 **3.2 Estimating Impacts with Robustness Checks**

270 To estimate the impacts of the two HAB events on the small-vessel proportions of
271 revenue and participation, we use a “before/after” analysis, an approach which has been
272 implemented in numerous program evaluations of fisheries including the impact of catch
273 shares on trip costs and revenues (Kroetz and Sanchirico 2010) and diversification
274 (Holland et al., 2017), and the impact of HABs on seafood prices (Jin, Thunberg, &
275 Hoagland, 2008).⁵ Our two outcome variables of interest, the small-vessel proportions of
276 revenue and participation, are naturally bounded between zero and one. Thus, to account

⁴ Note that the “before” period (2010/2011-2014/2015) is different than the baseline period (2011/2012, 2012/2013, and 2013/2014).

⁵ There are similarities in our methods and data to those in Perry (2018) except for that our before-after analysis is done in a regression framework, our data set was based on the confidential fish ticket data with fewer missing observations, and our dataset covers the entire 2016/2017 season.

277 for the bounded nature of the data, following Papke and Wooldridge (1996) and Hausman
 278 and Leonard (1997), we use fractional response models to estimate the impact of the
 279 HAB events on small vessel revenue and participation in California. We estimate both
 280 average effects, averaging across all port groups, and heterogeneous port-specific effects
 281 with the following models:

282
 283 *Average effects:*

$$284 \quad E(y_{it} | A_t, \mathbf{X}_{it}, \tau_t, \lambda_i) = F(A_t \delta, \mathbf{X}_{it}' \beta, \tau_t, \lambda_i), \quad (1)$$

285

286 *Heterogeneous effects:*

$$287 \quad E(y_{it} | A_{it}, \mathbf{X}_{it}, \tau_t, \lambda_i) = F(A_{it} \delta_i, \mathbf{X}_{it}' \beta, \tau_t, \lambda_i), \quad (2)$$

288

289 where y_{it} is the outcome variable for the port-group of landings i in season t , which is
 290 either the fraction of total revenue generated by small vessels or the fraction of the total
 291 unique vessels participating in the fishery that are classified as small vessels; A_t is a
 292 dummy variable set equal to one for all observations in the “after” or “treatment” period,
 293 i.e. during a HAB event; A_{it} are dummy variables set equal to one for observations in port
 294 groups i in the treatment period (i.e. it is the interaction between being in the treatment
 295 period A_t and a port-group fixed effect); \mathbf{X}_{it} are controls for port-specific season trends
 296 that can account for trends in outcomes over time at a specific port (e.g. trends in the
 297 relative revenues of small vessels delivering to the San Francisco port);⁶ τ_t are controls
 298 for fishing week fixed effects that can account for temporal drivers of outcomes within a

⁶ Port-specific time trends are calculated by interacting the season number (season numbers for seasons 2010/2011 – 2016/2017 are 1-7 respectively) with port-group fixed effects.

299 fishing season (e.g. changes in the level of competition over the course of a season in a
300 derby fishery); λ_i are controls for port-group fixed effects that account for time-invariant
301 differences in outcomes across port groups; and $F(\cdot)$ is the logit cumulative distribution
302 function.⁷

303 We estimate the model parameters using Bernoulli quasi-maximum likelihood
304 (QMLE) methods. We conservatively choose cluster robust standard errors to estimate
305 average effects and robust standard errors to estimate the heterogeneous effects, where
306 clusters are defined as port groups. Specifically, we choose between robust and cluster
307 robust standard errors based on the specification that yields the largest standard errors,
308 meaning the specification that would make it less likely for us to find a statistically
309 significant impact of the HAB events. To aid interpretation, all results are converted to
310 marginal effects, or the change in the response variables y_{it} associated with the HAB
311 events.⁸

312 We estimate two specifications of equations (1) and (2), i.e. specifications with
313 and without control variables (fixed effects and time trends). Additionally, because we
314 are interested in HAB impacts during two different fishing seasons, the 2015/2016 and
315 the 2016/2017 seasons, we estimate separate models with two different definitions of A_t
316 and A_{it} . We either define A_t as being in the 2015/2016 season or being in the 2016/2017
317 season. When estimating impacts to the 2015/2016 season, we drop observations from the
318 2016/2017 season from our model and when estimating impacts to the 2016/2017 season,
319 we drop observations from the 2015/2016 season.

⁷ We test the robustness of our result to controlling for week fixed effects instead of fishing week fixed effects, where week numbers correspond to the number of the calendar week (1-52 or 1-53 in some years). We find that our qualitative results are largely unchanged. See the Supplementary Material for estimation results when controlling for week of the calendar year rather than fishing week.

⁸ Primary regression results can be found in the Supplementary Materials.

320 Our impact estimates assume that the outcome data y_{it} are not trending over time
321 after accounting for our control variables and that autocorrelation is not driving our
322 results (see Bertrand, Duflo, & Mullainathan, 2004).⁹ Following Jardine et al. (2014), we
323 examine this assumption by performing several placebo tests. Our placebo tests ask the
324 question: what if we wrongly assumed that the HAB event occurred during a season
325 which did not have a HAB event; would we find an “impact” of this placebo event? If so,
326 the placebo test would suggest that our model results may not provide reliable impact
327 estimates. Specification 2 of Equation (1) is used to estimate the placebo tests, by
328 replacing the A_t dummy variable, which is set equal to one for all observations in the
329 treatment or HAB season, with a dummy variable set equal to one for all observations in
330 a “placebo treatment” season, or a season without a HAB event. We test all placebo
331 seasons for which we have at least one season of data before and after the placebo season.

332 We note that examining changes to the distribution of revenues is not the same as
333 examining changes to the distribution of profits. Therefore, to interpret our results as
334 impacts to the distribution of profitability one must assume that the HAB event did not
335 differentially impact costs for small and large vessels or that any differential impacts to
336 costs did not outweigh the differential impacts to revenues. Dewees et al. (2004) find that
337 fuel costs comprise a major component of variable costs for Dungeness crab harvesters.
338 Given one of the mechanisms through which we describe the HAB impacting
339 distributional outcomes is through large vessels traveling to exploit distant localized
340 fishing opportunities, it makes sense to assume that fuel costs were also increased for
341 these large vessels. However, it is unlikely that the fishing location choices of large

⁹ While we do not conduct formal tests for autocorrelation, we include a rich set of fixed effects and clustered standard errors when it is conservative to do so. In addition, we use placebo tests to determine whether autocorrelation is driving our results.

342 vessels were made in such a way that negatively impacted their profitability. Therefore,
343 while we cannot measure profitability directly, it makes sense to assume that revenue and
344 profitability shifted in the same direction.¹⁰

345 Finally, we note that, as with any proportion, the proportion of revenue going to
346 small vessels or the proportion of small vessel participation can fall with a decrease in the
347 numerator, an increase in the denominator, or both. We maintain that if the proportion of
348 revenue going to small vessels, or the proportion of small vessel participation, was
349 negatively impacted by the bloom event, then small vessels were harmed relative to a
350 counterfactual where the small vessels were able to maintain their share of the
351 revenue/participation. Thus, even if, during the HAB event, small vessels experienced no
352 change in the level of revenues generated, but large vessels exclusively experienced
353 increased revenue due to the HAB, small vessels suffer from the event because they
354 could not capture their typical proportion of revenue and share in the revenue gains with
355 large vessels. Additionally, modeling proportional outcomes allows us to control for
356 factors that lead to the same proportional changes in outcomes for small and large
357 vessels.¹¹

358
359 **4. Results**

360 Our dataset contains over 1,800 weekly observations of Dungeness crab deliveries for 7
361 California ports from both the central and northern management districts. Table 1
362 summarizes revenue and participation, by port group. Mean weekly revenues vary
363 considerably across port groups, e.g. the highest average weekly revenues are generated

¹⁰ Although a reviewer notes that large vessels may have made unprofitable fishing location choices to retain their crew.

¹¹ For example, if a 20% increase in the stock of Dungeness crab leads to a 20% increase in revenue for both small and large vessels, we have controlled for the unobserved impact of the stock.

364 in the San Francisco port group and are nearly 10 times the average weekly revenue for
365 Monterey (Table 1). Mean weekly participation levels are less variable, e.g. San Francisco
366 generates nearly 10 times the mean weekly revenue of Monterey but has less than 4 times
367 the number of vessels participating (Table 1). Table 2 summarizes revenue and
368 participation across vessel size class and small-vessel proportions of revenue and
369 participation, by port group and period, as well as averages across all of the weekly
370 observations. The proportion of revenue to small vessels in the before period varies
371 considerably by port group, with small vessels accounting for the majority of revenue, on
372 average, in Eureka and Monterey, and roughly one third of revenue in Crescent City and
373 Fort Bragg. In all cases, the average proportion of revenue going to small vessels is less
374 than the proportion of small-vessel participation, which is not surprising given
375 differences in vessel capacity. While in some cases the mean weekly proportion of
376 revenue and participation fell in the both HAB seasons, relative to the before period, e.g.
377 Crescent City, Bodega Bay, and Morro Bay, in other cases the proportions remain stable
378 or even increase. We proceed to estimate the impacts of the HAB events on these
379 outcomes after controlling for other factors that impact relative outcomes for small
380 vessels.

381

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Table 1: Summary statistics of weekly data on revenue and participation in fishing seasons 2010/2011-2016/2017

	Mean	Std. Dev.	Median	No. Obs.
Crescent City				
Ex-vessel Revenue	\$544,597	\$1,269,915	\$13,716	198
Participation	29.88	24.43	22.00	198
Eureka				
Ex-vessel Revenue	\$438,936	\$874,047	\$127,250	198

Participation	35.83	24.08	31.00	198
Fort Bragg				
Ex-vessel Revenue	\$125,212	\$180,048	\$53,134	188
Participation	11.69	8.70	10.50	188
Bodega Bay				
Ex-vessel Revenue	\$326,163	\$550,193	\$110,096	212
Participation	30.20	20.93	27.00	212
San Francisco				
Ex-vessel Revenue	\$749,552	\$1,506,905	\$210,599	218
Participation	57.34	42.03	46.00	218
Monterey				
Ex-vessel Revenue	\$75,037	\$66,859	\$54,276	218
Participation	15.38	6.87	14.50	218
Morro Bay				
Ex-vessel Revenue	\$99,609	\$92,802	\$63,839	178
Participation	8.48	4.78	8.00	178

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Table 2: Sample means (standard deviations) from weekly data on revenue and participation by vessel class (small and large) and small vessel proportions, by port group and period, and for all observations

	Revenue (10 thousand USD)			Participation (vessels)		
	Small	Large	Proportion	Small	Large	Proportion
Crescent City						
Before period	10.009	43.798	0.322	12.183	16.288	0.496
	(26.726)	(104.809)	(0.268)	(9.140)	(15.196)	(0.207)

2015/2016	10.690	41.638	0.207	16.909	22.636	0.444
	(10.862)	(42.692)	(0.101)	(7.286)	(13.344)	(0.083)
2016/2017	10.365	47.721	0.228	12.912	20.206	0.425
	(22.393)	(110.534)	(0.168)	(10.358)	(18.055)	(0.141)
Eureka						
Before period	18.758	27.209	0.591	22.948	13.182	0.700
	(33.108)	(64.613)	(0.214)	(13.734)	(12.661)	(0.145)
2015/2016	20.614	25.293	0.510	22.545	13.818	0.640
	(20.684)	(25.850)	(0.132)	(7.699)	(7.068)	(0.065)
2016/2017	16.722	16.825	0.622	22.152	12.091	0.707
	(21.934)	(25.515)	(0.218)	(10.450)	(9.655)	(0.139)
Fort Bragg						
Before period	3.158	8.330	0.324	5.262	4.872	0.522
	(6.305)	(12.541)	(0.263)	(4.435)	(3.823)	(0.220)
2015/2016	7.192	11.075	0.364	11.273	5.182	0.669
	(5.253)	(7.897)	(0.155)	(4.714)	(2.272)	(0.091)
2016/2017	6.999	8.766	0.496	10.571	7.536	0.590
	(8.052)	(12.779)	(0.219)	(6.052)	(4.574)	(0.130)
Bodega Bay						
Before period	9.406	23.038	0.410	16.387	13.226	0.591
	(14.731)	(43.626)	(0.204)	(10.358)	(11.016)	(0.145)
2015/2016	11.917	33.067	0.261	16.643	17.429	0.490
	(13.372)	(38.524)	(0.086)	(8.661)	(9.493)	(0.062)
2016/2017	9.004	18.806	0.373	18.500	13.200	0.608
	(13.757)	(27.529)	(0.182)	(12.456)	(10.456)	(0.158)
San Francisco						
Before period	19.201	54.234	0.420	31.806	24.129	0.626
	(33.770)	(122.792)	(0.177)	(20.824)	(22.776)	(0.111)
2015/2016	37.066	80.050	0.446	46.000	31.929	0.623
	(39.259)	(103.223)	(0.186)	(20.917)	(21.255)	(0.086)
2016/2017	20.878	44.318	0.474	34.676	21.206	0.652
	(29.015)	(104.494)	(0.159)	(20.009)	(17.847)	(0.097)
Monterey						
Before period	5.083	2.523	0.670	11.665	3.494	0.773
	(4.837)	(2.556)	(0.212)	(5.585)	(2.113)	(0.111)
2015/2016	10.575	3.553	0.750	17.000	4.643	0.799
	(8.421)	(3.071)	(0.111)	(4.403)	(2.468)	(0.076)
2016/2017	2.894	1.371	0.711	11.176	2.706	0.811
	(1.522)	(1.114)	(0.188)	(4.726)	(1.818)	(0.120)
Morro Bay						
Before period	3.030	6.040	0.443	4.208	3.715	0.582

	(2.405)	(7.435)	(0.293)	(2.141)	(3.240)	(0.228)
2015/2016	5.481	14.590	0.295	5.714	6.643	0.492
	(4.276)	(8.758)	(0.182)	(1.326)	(3.054)	(0.128)
2016/2017	2.518	6.684	0.341	4.500	4.529	0.517
	(1.502)	(6.009)	(0.171)	(1.600)	(2.711)	(0.171)
All observations						
	10.348	24.043	0.455	15.857	11.784	0.615
	(21.921)	(68.977)	(0.257)	(14.629)	(14.254)	(0.188)

393

394 *4.1 Shifting Deliveries Patterns*

395

396 Even during the baseline period, large vessels are generally more mobile than small
397 vessels. For example, we identify a subset of large vessels in Crescent City/San Francisco
398 and Eureka/San Francisco homeports, meaning that these vessels are consistently active
399 in two port areas over the course of a single Dungeness crab season, but identify no small
400 vessels exhibiting that spatial landings pattern (Figure 2). The existence of clusters of
401 large vessels active in multiple ports is illustrative of the relative mobility of large
402 vessels.

403 Analyzing shifts in spatial fishing patterns during the HAB events with heat maps,
404 we see that both small and large vessels shifted their pounds delivered in the two HAB
405 seasons (Figure 2). Furthermore, there are some directional similarities between shifts for
406 small and large vessels. However, the magnitude of the shifts and distance traveled are
407 greater for larger vessels (Figure 2). In the 2015/2016 season, for example, both large and
408 small vessels based out of Eureka shifted landings out of their homeport to San
409 Francisco. Large vessels show a roughly 40% reduction in pounds landed in Eureka and a
410 33% increase in pounds landed in San Francisco. In contrast, small vessels show a
411 roughly 20% reduction in pounds landed in Eureka and a 16% increase in pounds landed

412 in San Francisco. In the 2016/2017 season vessels based out of Crescent City/San
413 Francisco and Bodega Bay shifted landings out of their homeport of San Francisco and
414 into Crescent City (by roughly 22% and 26% respectively) while smaller vessels based in
415 Bodega Bay, San Francisco, and Monterey all shifted their landings to the closest port to
416 the north (similar to patterns observed for small vessels in the previous season).

417

418 ***4.2 Changes in the Proportion of Revenue and Participation for Small Vessels***

419

420 As a first step to investigating the impacts on outcomes for small vessels, we estimate
421 Equation (1) without any controls, finding no evidence that, on average across all weekly
422 observations, the HAB events impacted small-vessel revenue or participation proportions
423 (Specification 1, Figure 3). However, after controlling for port-group specific time trends
424 and port group and fishing week fixed effects, we find significant negative impacts to
425 both outcomes (Specification 2, Figure 3). Our estimated marginal effect of the HAB
426 event, on average across all ports, is a reduction in the small-vessel revenue proportion of
427 roughly 0.14 and 0.18 in the 2015/2016 and 2016/2017 seasons respectively and a
428 reduction in the small-vessel participation proportion of roughly 0.07 and 0.12 in the
429 2015/2016 and 2016/2017 seasons respectively. Therefore, across both seasons, we see
430 that relative revenue to small vessels was more greatly affected by the HAB events than
431 relative small-vessel participation. To put the results into context, the estimated reduction
432 in the proportion of revenue to small vessels during the 2015/2016 season represents a
433 30% reduction from the before-period mean proportion ($0.14/0.455 = 0.3$, see Table 2 for
434 the before-period mean).

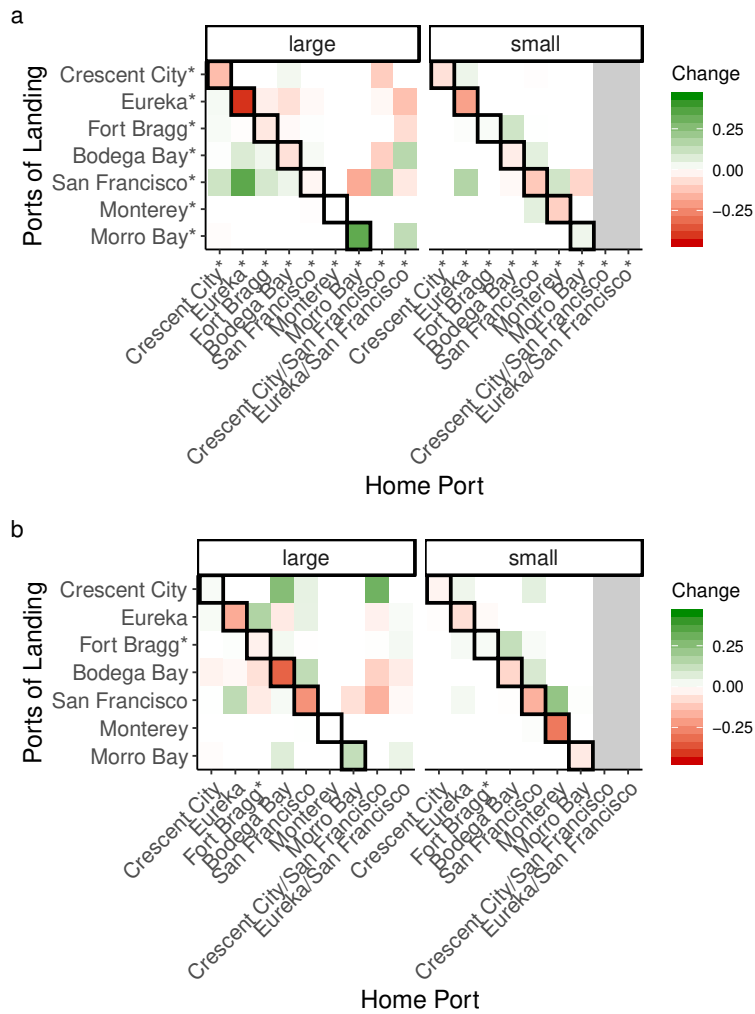


Figure 2: Difference in the proportion of Dungeness crab pounds delivered at each port group between the 2015/2016 season (panel a) or the 2016/2017 season (panel b) and the baseline period (seasons 2011/2012-2013/2014), by homeport groups. Port groups are arranged on the x and y axes in order from north to south and the asterisk (*) denotes a port group protected under the Fair Start Provisions.

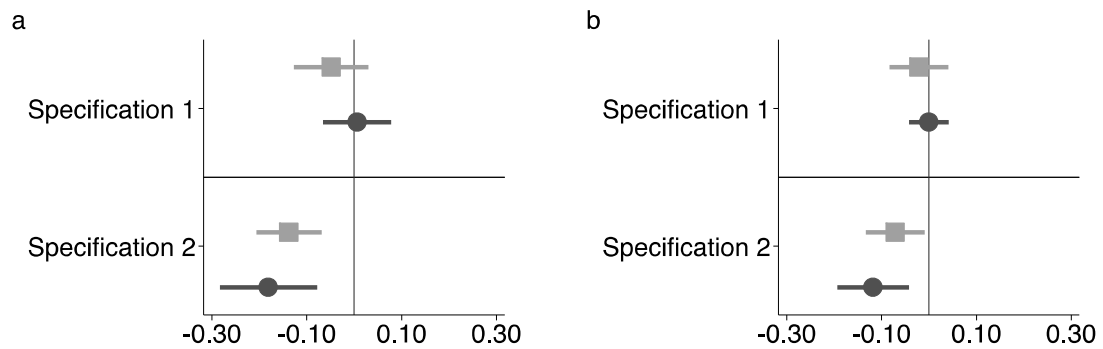


Figure 3: Average marginal impacts of HAB event. The impacts are measured as changes in proportions of: a) of revenue to small vessels and b) of small vessel participation. Square markers represent the 2015/2016 season and circle markers represent the 2016/2017 season.

438

439 Estimating heterogeneous impacts, from Equation (2), with the full set of controls,

440 we find small vessels from several port groups were negatively affected (Figure 4).

441 Specifically, impacts to the proportion of small-vessel revenue were negative and

442 statistically significant for Crescent City, Eureka, and Bodega Bay in both HAB seasons

443 and for San Francisco and Monterey in the 2016/2017 season. We find the greatest

444 impacts to the proportion of small-vessel revenue in Crescent City, with impacts over two

445 times the average impact, in both HAB seasons.

446

447 In general, as with the average impacts on participation (Figure 3), the impacts to

448 the proportion of small-vessel participation were less in magnitude than impacts to

449 revenue (Figure 4). For example, in the 2016/2017 season, the reduction in the

450 participation proportion for small vessels in the Crescent City area is equal to 63% of the

451 reduction in the proportion of revenue going to small vessels for that port.

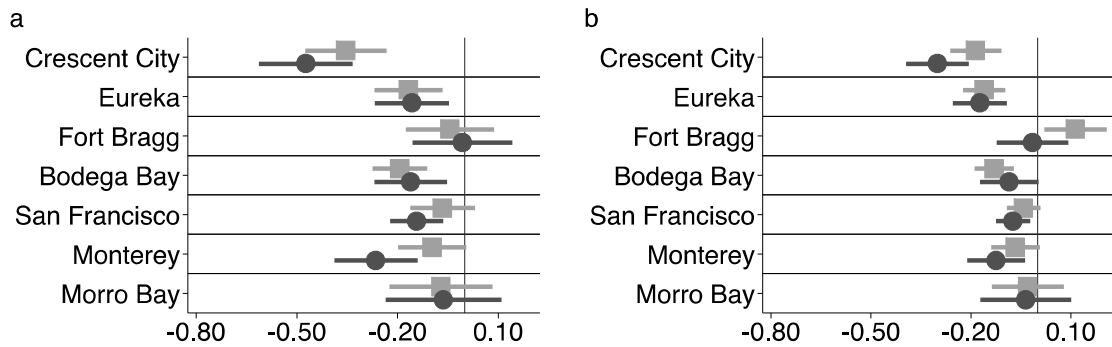


Figure 4: Heterogeneous marginal impacts of HAB event. The impacts are measured as changes in proportions of: a) of revenue to small vessels and b) of small vessel participation. Square markers represent the 2015/2016 season and circle markers represent the 2016/2017 season. Note: port groups are ordered from north to south.

452

453 Our placebo tests suggest that our outcome variables are not systematically driven

454 by variables outside of our model in a way that would confound our analysis or by

455 autocorrelated error terms. Figure 5 plots the marginal effects of the HAB event

456 calculated from the estimation of Specification 2 of Equation (1) using dummy variables

457 for placebo HAB seasons in place of dummy variables for the actual HAB seasons. All

458 95% confidence intervals around the estimated marginal effect of the placebo HAB

459 events include zero.

460

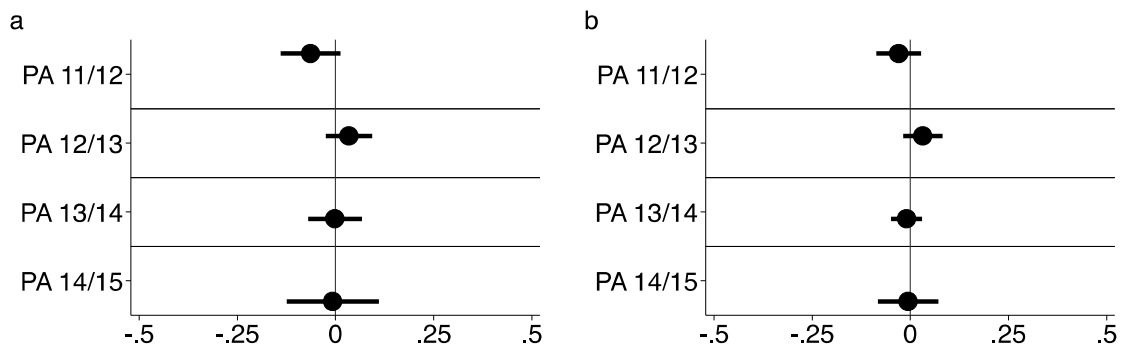


Figure 5: Average marginal impacts of placebo HAB events on the proportion of small vessel revenue (left) and the proportion of small vessel participation (right)

461

462 **4.3. Synthesis of the results**

463 In the 2015/2016 fishing season we observed substantial shifts in landings by large
464 vessels out of Crescent City and Eureka, the ports with the longest closures (Figure 1).
465 These vessels landed more Dungeness crab in Bodega Bay, San Francisco, and Morro
466 Bay than in previous years (Figure 2). At the same time, during the 2015/2016 season the
467 proportional reductions in revenue to small vessels were greatest in Crescent City,
468 Eureka, and Bodega Bay. This combination of apparently contradictory outcomes —
469 large vessels shifting landings out of some of the ports where small vessels experienced
470 the greatest impact to proportional revenues (i.e. Crescent City and Eureka)—is
471 surprising at first blush. While we cannot state definitively what caused these responses,
472 one potential explanation is that our analysis (Figure 2) misses important intra-seasonal
473 dynamics. For example, roughly half of the large vessels leaving Crescent City during
474 their closed periods and returned once the season was opened. Additionally, these two
475 port groups experienced the longest delays and, thus, greatest seasonal compression,
476 which may have impacted small-vessel outcomes even absent any changes in harvest
477 patterns.

478 In the 2016/2017 season, by contrast, we observe large shifts out of Bodega Bay,
479 which was the only port in the central district to be delayed in this season. Large vessels
480 left Bodega Bay for distant Crescent City, which then experienced the largest
481 proportional revenue impacts (Figure 4). Small vessels left Bodega Bay for the
482 neighboring port of Fort Bragg. While Fort Bragg experienced the longest delay, we do
483 not see large vessels shifting pounds into Fort Bragg. Instead it is small vessels that
484 appeared to take advantage of the extended and incremental closures in Fort Bragg. This

485 is likely due to the fact that Fort Bragg is a small port in terms of landings (Table 1) and
486 thus not attractive to large vessels.

487
488 **5. Conclusion**

489

490 We find evidence that small vessels were disproportionately impacted by the
491 2014-2016 Northeast Pacific Marine Heatwave that led to a series of HAB events
492 resulting in closures of the Dungeness crab fishery on the U.S. West Coast. Specifically,
493 the HAB events were correlated with shifting spatial fishery landings patterns in
494 California, especially for larger vessels, and a reduction in relative small-vessel revenue
495 and participation. We attribute our findings to the mechanisms of exacerbated seasonal
496 compression in the 2015/2016 Dungeness crab fishing season and localized fisheries
497 closures in both the 2015/2016 and 2016/2017 seasons, which created opportunities for
498 large vessels to participate in distant fishery openings. Our results demonstrate the
499 interconnectedness of fishing ports during a climate shock when mobile vessels can move
500 between ports.

501 We note Dungeness crab fishers also fish for other species including salmon,
502 albacore tuna, groundfish, and pink shrimp (Deweese et al., 2004; Fuller, Samhour, Stoll,
503 Levin, & Watson, 2017). Additionally, commercial fishers often earn non-fishing
504 incomes (Holland, Abbott, & Norman, 2019). Understanding the impacts of the HAB
505 events on total revenues to harvesters with diverse fishing/earnings portfolios is another
506 important area for future research that will further illuminate the capacity of fishers to
507 adapt to climate variability and change.

508 Fisheries managers will likely be faced with growing equity issues due to climate
509 change, in California and beyond. The 2014-2016 Northeast Pacific Marine Heatwave
510 also led to a rise in whale entanglements with Dungeness crab fishing gear, which in
511 2018 prompted California managers to restrict the summer harvest of Dungeness crab for
512 the first time due to ecosystem considerations, further compressing the fishing season
513 (Santora et al., 2019). With HABs in this region expected to increase in magnitude,
514 frequency, and duration under climate change (McCabe et al., 2016; Trainer et al., 2020),
515 regulatory compression of fishing seasons combined with ecological compression of
516 suitable habitat for Dungeness crab (Chan, Barth, Kroeker, Lubchenco, & Menge, 2019)
517 and protected species (Santora et al., 2019) may amplify these distributional impacts in
518 the coming decades.

519 While fisheries managers have no control over the need to restrict consumption of
520 Dungeness crabs with high levels of domoic acid or the need to reduce bycatch of
521 protected species, the disproportionate advantages for larger vessels caused by seasonal
522 compression may be mitigated through policy design. For example, since the series of
523 HAB events analyzed here, managers have amended the Fair Start Provisions to increase
524 protections for fishers participating in areas with localized closures. Our results
525 encourage a management focus on both communities-of-place, such as individual ports,
526 but also on communities-of-practice, such as groups of vessels with similar
527 characteristics and constraints, in developing management strategies that allow for
528 adaptation to climate shocks. Future work is needed to develop a deeper understanding of
529 the various mechanistic pathways through which HAB-related closures, and other

530 mitigation measures, impact distributional outcomes in order to better guide the design of
531 policies that can promote equity in fisheries impacted by climate change.

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704 **Supplementary Material**

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706 ***1. Regression results***

708 Tables SM1 and SM2 present estimates obtained from the fractional response models
709 specified in equations 1 and 2 respectively. Using the regression results reported in tables
710 SM1 and SM3 we calculate the marginal effects of the HAB events to generate figures 3
711 and 4 respectively.

Table SM1: Estimation results from Equation 1 used to create Figure 3

	Revenue proportion				Participation Proportion			
	2015/2016		2016/2017		2015/2016		2016/2017	
<i>HAB impacts:</i>								
Average	-0.195 (0.164)	-0.617** (0.158)	0.026 (0.147)	-0.803** (0.240)	-0.090 (0.134)	-0.317* (0.141)	-0.001 (0.091)	-0.525** (0.175)
<i>Time trends:</i>								
Crescent City		0.125** (0.015)		0.158** (0.034)		0.0871** (0.018)		0.133** (0.025)
Eureka		0.238** (0.014)		0.319** (0.036)		0.154** (0.015)		0.154** (0.025)
Fort Bragg		0.356** (0.015)		0.230** (0.035)		0.225** (0.014)		0.145** (0.025)
Bodega Bay		0.193** (0.017)		0.217** (0.035)		0.167** (0.016)		0.165** (0.025)
San Francisco		0.354** (0.014)		0.281** (0.033)		0.197** (0.012)		0.188** (0.023)
Monterey		-0.0160 (0.017)		0.0501 (0.038)		-0.0114 (0.017)		0.0365 (0.030)
Morro Bay		0.249** (0.015)		0.243** (0.035)		0.126** (0.015)		0.141** (0.025)
Port group fixed effects	No	Yes	No	Yes	No	Yes	No	Yes
Fishing week fixed effects	No	Yes	No	Yes	No	Yes	No	Yes
Observations	1,183	1,183	1,321	1,321	1,183	1,183	1,321	1,321
AIC	1.030	0.932	1.031	0.940	0.946	0.899	0.946	0.900

Note: Asterisk (*) and double asterisk (**) denote variables significant at 5% and 1% levels respectively. Standard errors are clustered by portgroup and reported in parentheses.

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Table SM2: Estimation results from Equation 2 used to create Figure 4

	Revenue proportion				Participation Proportion			
	2015/2016		2016/2017		2015/2016		2016/2017	
<i>HAB impacts:</i>								
Crescent City	-1.174** (0.180)	-1.592** (0.279)	-1.051** (0.165)	-2.113** (0.322)	-0.697** (0.100)	-0.829** (0.176)	-0.776** (0.100)	-1.343** (0.216)
Eureka	0.211 (0.155)	-0.753** (0.235)	0.670** (0.162)	-0.703** (0.252)	0.100 (0.085)	-0.715** (0.146)	0.405** (0.117)	-0.774** (0.186)
Fort Bragg	-0.386* (0.195)	-0.198 (0.302)	0.157 (0.166)	-0.0299 (0.339)	0.228 (0.121)	0.508* (0.213)	-0.112 (0.103)	-0.0672 (0.246)
Bodega Bay	-0.867** (0.119)	-0.869** (0.187)	-0.348* (0.144)	-0.716** (0.247)	-0.514** (0.069)	-0.581** (0.136)	-0.0365 (0.122)	-0.378 (0.199)
San Francisco	-0.0446 (0.196)	-0.297 (0.222)	0.0676 (0.113)	-0.638** (0.181)	0.0272 (0.098)	-0.186 (0.115)	0.155* (0.077)	-0.327** (0.117)
Monterey	1.271** (0.156)	-0.436 (0.234)	1.073** (0.158)	-1.181** (0.283)	0.907** (0.125)	-0.295 (0.166)	0.984** (0.134)	-0.554** (0.198)
Morro Bay	-0.698** (0.228)	-0.317 (0.352)	-0.488** (0.132)	-0.282 (0.393)	-0.505** (0.135)	-0.131 (0.246)	-0.405** (0.118)	-0.157 (0.311)
<i>Time trends:</i>								
Crescent City		0.145** (0.046)		0.146** (0.045)		0.111** (0.034)		0.113** (0.0339)
Eureka		0.206** (0.070)		0.205** (0.0687)		0.0938 (0.0504)		0.0917 (0.0494)
Fort Bragg		0.427** (0.064)		0.425** (0.0653)		0.264** (0.0427)		0.265** (0.0438)
Bodega Bay		0.204** (0.046)		0.203** (0.0469)		0.199** (0.0322)		0.200** (0.0333)
San Francisco		0.338** (0.052)		0.334** (0.0516)		0.195** (0.0340)		0.192** (0.0335)
Monterey		-0.0498 (0.108)		-0.0456 (0.106)		-0.0357 (0.0885)		-0.0335 (0.0869)
Morro Bay		0.221** (0.033)		0.219** (0.0336)		0.114** (0.0203)		0.113** (0.0205)
Port group fixed effects	No	Yes	No	Yes	No	Yes	No	Yes
Week fixed effects	No	Yes	No	Yes	No	Yes	No	Yes
Observations	1,183	1,183	1,317	1,317	1,183	1,183	1,317	1,317

AIC 1.031 1.009 1.022 1.006 0.952 0.976 0.945 0.969

Note: Asterisk (*) and double asterisk (**) denote variables significant at 5% and 1% levels respectively. Robust standard errors in parentheses.

730 **2. Fishing week fixed effects versus week fixed effects**

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Key model results in figures 3 and 4 depend on estimating equations 1 and 2 using fishing week fixed effects. The variable fishing week is equal to 1 in the first week that a port group is open to fishing, 2 in the second week, and so on. Fishing week fixed effects control for the effects of competition in a derby fishery, where early in the season abundance and competition are high and later on in the season abundance and competition decrease. Here we explore the robustness of our results to replacing fishing week fixed effects in equations 1 and 2 with fixed effects for week of the calendar week (1-52 or 1-53 in some years). Week fixed effects control for seasonal weather patterns that are similar across seasons, the opening of other fisheries that Dungeness crab vessels also participate in, and other factors that may impact the relative success of small vessels that vary by calendar week. Overall, we find our qualitative results are unchanged, but that there are some differences in the magnitude and significance of our impact estimates. Figures SM1 and SM2 are the counterparts to Figure 3 and Figure 4 in the article respectively.

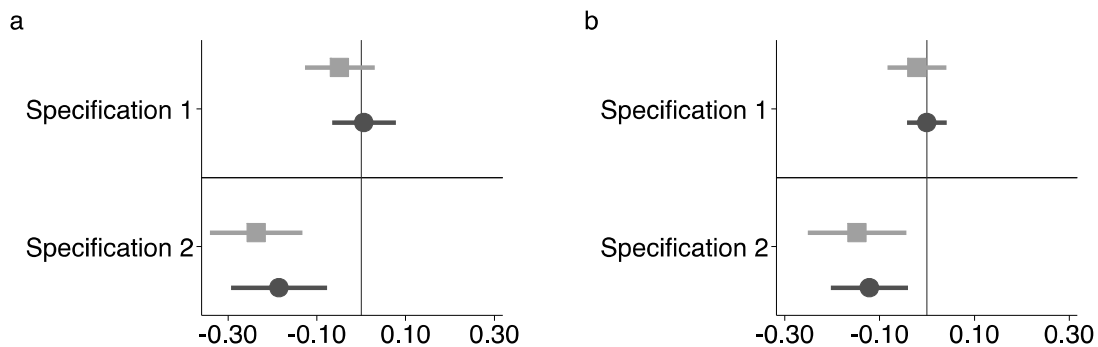


Figure SM1: Average marginal impact of HAB event estimated using week fixed effects. The impacts are measured as changes in proportions of: a) of revenue to small vessels and b) of small vessel participation. Square markers represent the 2015/2016 season and circle markers represent the 2016/2017 season. Note: port groups are ordered from north to south.

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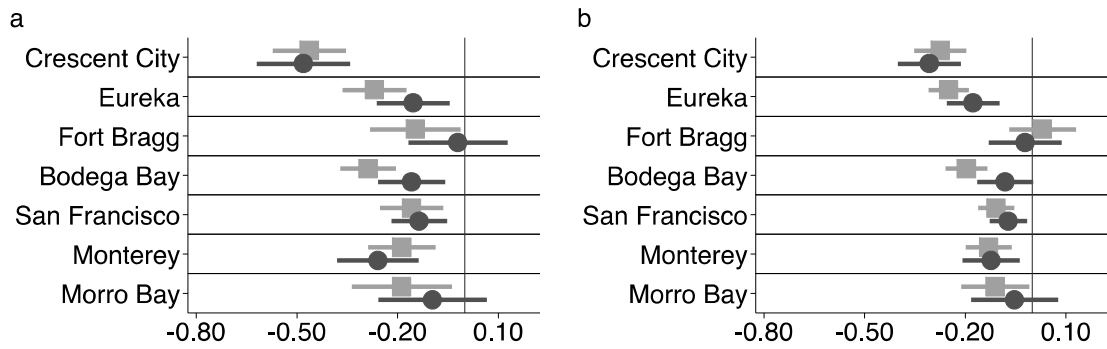


Figure SM2: Heterogeneous marginal impacts of HAB event estimated using fishing week fixed effects. The impacts are measured as changes in proportions of: a) of revenue to small vessels and b) of small vessel participation. Square markers represent the 2015/2016 season and circle markers represent the 2016/2017 season. Note: port groups are ordered from north to south.

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