2	fisheries bycatch of vulnerable marine ecosystem indicator taxa
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Estimating quantitative gear and taxa specific encounter thresholds for commercial

17 Abstract: Corals and sponges are often a component of vulnerable marine ecosystems (VMEs) 18 in the deep sea. These taxa can be impacted and removed by bottom contacting fishing gear and 19 protecting VMEs is an important component of managing ecosystems. One of the tools that is 20 routinely used to manage VME impacts from fishing gear is move-on rules triggered by bycatch 21 thresholds (encounter thresholds) of VME. Usually, these bycatch thresholds are set with little 22 information regarding the level of impact on the benthic habitat. The objective of this analysis 23 was to develop and apply methods for quantifying threshold catches of VME indicator taxa by 24 gear type and VME indicator taxa grouping. Three previously used methods based on cumulative 25 bycatch distributions and one novel method based on percentile regression of fishery bycatch and 26 density from underwater camera surveys were applied to data from the northeast Pacific Ocean 27 to determine data-based encounter thresholds that could trigger spatial fishery closures. The 28 percentile regression method suggested encounter thresholds of ~ 40 - 65 kg of Antipatharia, < 2029 kg of gorgonians and 78 - 131 kg of Porifera would equate to a density of 0.2 VME indicator 30 taxa per m^2 for bottom trawl bycatch. Threshold values were lower for longline and pot gear 31 (generally < 10 kg per set). Using the percentile regression method allowed for the definition of 32 VME encounter thresholds to be expressed in terms of density of the taxa of interest, an 33 improvement over examination of break points in the cumulative bycatch data alone. This 34 improvement allows the ecological importance (e.g. density of VME) to be defined and used to 35 estimate encounter thresholds, rather than assuming that the natural break points in cumulative 36 bycatch represent an ecological break point.

Keywords: vulnerable marine ecosystems, move-on rules, bycatch thresholds, deep-sea coral
and sponge, fisheries bycatch

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40 Introduction

41 Deep-sea corals and sponges have been identified as ecosystem components that need 42 protection from bottom-contact fishing activity, and in 2006, resolution 61/105 was adopted by 43 the UN General Assembly to prevent impacts on vulnerable ecosystems by deep-sea fisheries 44 (FAO 2009). Vulnerable marine ecosystems (VMEs) have further been defined as areas where 45 fishing activities are likely to have significant and adverse impacts on the benthic community 46 (FAO 2009). Deep-sea corals and sponges are widely recognized as important components of 47 VMEs and these taxa can increase seafloor biodiversity and contribute to fisheries production by 48 providing complex structure on the seafloor (Buhl-Mortensen et al. 2010, Watling et al. 2011, 49 Linley et al. 2017). In general, deep-sea corals and sponges are long-lived, mature after an 50 extended period (years to decades), have low capacity for recruitment and recovery and are 51 structurally fragile making them vulnerable to damage from mobile or fixed fishing gears 52 (Koslow et al. 2001, Freese 2001, Clark and Rowden 2009, Heifetz et al. 2009). Spatial 53 management of the impacts of fishing activity on VMEs is difficult due to the paucity of data on 54 the distribution of deep sea VMEs and the poor or unknown distribution and spatial resolution of 55 fishing activity.

Encounter protocols with subsequent spatial closures is a tool often used in managing fisheries and VME impacts (Hourigan 2009, Auster et al. 2011, Ardron et al. 2014, Wallace et al. 2015). Typical implementation of an encounter protocol to protect VME involves defining a threshold bycatch weight of benthic invertebrates that indicate the potential presence of a VME on the seafloor. Bycatch at or above this threshold weight then triggers a move-on rule where the fishing activity is forced to move away from the location where the encounter threshold was exceeded. In some cases, either a permanent or temporary spatial closure is adopted around the

63 encounter event. The move-on rule and spatial closure can apply to only the vessel or gear type 64 that triggered the encounter (Wallace et al. 2015, SPRFMO 2023) or it can apply to all fisheries 65 operating in the area with bottom contacting gear (NPFC 2023). Indicator taxa are typically used 66 to indicate the presence of a VME. So for example, in international waters of the Northeast 67 Pacific Ocean, by catch of > 50 kg of corals (any combination of Alcyonacea, gorgonian, 68 Antipatharian or Scleractinian corals) or > 500 kg of sponges (any combination of Hexactinellida 69 or Demospongiae) by a commercial fishing event (e.g. a bottom trawl haul) triggers a temporary 70 spatial closure within 1 nm of the trawl path and forces the vessel (and other vessels fishing the 71 same gear type) to avoid the closed area (NPFC 2023). Although there is substantial debate over 72 the effectiveness of encounter protocols with move-on rules as a mechanism to protect VME 73 (e.g. Auster et al. 2011), these protocols are a commonly used tool, especially for fisheries in 74 international waters (FAO 2016). The goal of the analyses presented here is to improve the use 75 of these tools.

76 Setting thresholds for VME indicator taxa is difficult because of the unquantified link 77 between VME indicator taxa abundance, the observed bycatch on commercial fishing gear, and 78 the degree of impact on the indicator taxa. An effective threshold is dependent on knowledge of 79 the catch efficiency of the gear (e.g. what does the weight of VME indicator taxa in a trawl haul 80 equate to in terms of abundance of the VME indicator taxa on the seafloor?). Patch size for VME 81 can be determined from underwater imagery, and can guide the placement and size of spatial 82 closures (Kenchington et al. 2014, Rowden et al. 2017, SPRFMO 2021, Piechaud and Howell 83 2022). However, bycatch thresholds have historically been put in place based on limited 84 information (Ardron et al. 2014). Thresholds that have been picked using bycatch data may not 85 have an ecological basis for the value (Geange et al. 2020). Only a handful of studies have

examined the efficiency of fishing gear in capturing VME indicator taxa (e.g. Freese 2001). Most
of these studies have examined bottom trawl gear and have found that the catch efficiency is
typically less than 5% for VME indicator taxa (Kenchington et al. 2011, SPRFMO 2022).
Assessing catch efficiency of VME indicator taxa in fishing gear would allow for more databased methods to identify when bycatch values indicate the presence of a VME.

91 The objective of this analysis was to develop and apply methods for quantifying 92 threshold catches of VME indicator taxa by gear type and VME indicator taxa grouping. Geange 93 et al. (2020) developed three methods to estimate VME indicator taxa encounter thresholds and 94 applied them to New Zealand bottom trawl fishery bycatch data from seamount systems in the 95 South Pacific Regional Fishery Management Organization (SPRFMO) Convention Area. Here 96 we apply these three methods (Geange et al. 2020) to fishery by catch data only and develop and 97 apply a new method that relates fishery bycatch data to density data independently collected 98 during stereo-camera surveys. Threshold catches are then proposed that would indicate the 99 presence of a VME based on fishery observer data and observed density data. These threshold 100 catches could then be used to trigger move-on rules and spatial closures for bottom contacting 101 fishing gears.

102 Methods

A variety of benthic invertebrates have been defined as vulnerable marine ecosystem indicator taxa by management bodies (Baco-Taylor et al. 2023). In this analysis we used VME indicator taxa groupings as defined by the Regional Fisheries Management Organization (RFMO) for international waters of the North Pacific, the North Pacific Fisheries Commission (NPFC; NPFC 2023) with two additions. The VME indicator taxa groupings used here are

108 Alcyonacea (soft corals, excepting those species defined as gorgonian corals), gorgonian corals 109 (upright, complex and branching corals from the families Primnoidae, Plexauridae, 110 Keratoisididae, Coralliidae, and Acanthogorgiidae), Antipatharia (black corals), Scleractinia 111 (stony corals), Hexactinellida (glass sponges) and Demospongiae in the phylum Porifera. The 112 two additional groups that were included in the analyses were pennatulaceans (Pennatuloidea) 113 and hydrocorals (Stylasteridae). These two taxonomic groups have been considered in the past 114 for inclusion in the NPFC indicator taxa list and are included as VME indicator taxa by some 115 other regional fisheries management organizations. Although we use these definitions of VME 116 indicator taxa here to be consistent with the current RFMO definition for the North Pacific, the 117 definitions are clearly not consistent with the currently accepted taxonomic definitions for coral 118 species (McFadden et al. 2022). The parentheses above and Table 1 contain the family level 119 names that can be used to cross-walk between the VME indicator taxa (NPFC definition) and the 120 currently accepted taxonomy (McFadden et al. 2022).

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122 Fisheries data

123 Information collected by fisheries monitoring programs in Alaska, British Columbia and 124 the contiguous west coast of the United States of America (US) from 2002-2022 were the 125 primary data used in this analysis. For each fisheries monitoring program the bycatch of deep-sea 126 coral and sponge taxa are recorded from subsamples of the total catch collected from individual 127 hauls as the weight of each taxa. These weights are then expanded to the total haul catch using 128 the total weight of the bycatch. The proportion of the catch sampled for each fishing event is 129 highly variable depending on the gear type, fishery, region and species targeted. All of these data 130 are subject to privacy restrictions in Canada and the US, so the data from individual hauls were

131 provided for analyses without potentially identifying characteristics or other confidential 132 information (e.g. latitude, longitude, depth, vessel name, etc.). The data were identified by year 133 of catch and one of 5 regions; eastern Bering Sea, Aleutian Islands, Gulf of Alaska, British 134 Columbia and the US west coast which roughly follow the existing definitions of large marine 135 ecosystems for the marine waters of North America (Figure 1). The data were also identified as 136 coming from one of three fishing gear types; bottom trawl, hook and line (primarily longline 137 gear), or trap gear (pot). All of these fishing gear types have been used to target benthic fish and invertebrates in commercial fisheries in each area since the early 20th century. Typical species 138 139 targeted by bottom trawls are rockfishes, flatfishes and gadids in all 5 regions (NPFMC 2020a, 140 NPFMC 2020b, PFMC 2023, DFO 2024). These bottom trawl fisheries, although usually 141 targeting a single species or species complex (e.g. rockfishes), can have significant by catch of 142 both VME indicator taxa and other fishes and invertebrates (NOAA 2011). Hook and line gear is 143 typically used to target single species most commonly Pacific halibut (Hippoglossus stenolepis), 144 sablefish (Anoplopoma fimbria) or Pacific cod (Gadus macrocephalus) in all five regions 145 (NPFMC 2020a, NPFMC 2020b, PFMC 2023, DFO 2024). Bycatch in these longline fisheries 146 tends to be less (but can comprise significant portions of the catch) and because of the stationary 147 nature of the fishing configuration and the active capture method (baited hooks), less VME 148 indicator taxa are retained on the gear than for mobile fishing methods (NOAA 2011). Similarly, 149 pot gear is stationary and requires active capture by the target fish or invertebrates, so the amount 150 of retained VME indicator taxa is less (NOAA 2011). Pot gear is used in all five regions 151 primarily to capture Pacific cod, sablefish and multiple crab species (NPFMC 2020a, NPFMC 152 2020b, PFMC 2023, DFO 2024). Observer coverage strategies vary by region and fishery. In 153 Canada, there has been 100% observer coverage on the groundfish fleet since 1997 through

either electronic monitoring or human observers combined with dockside monitoring (Turris
2000). In Alaska coverage rates depend on the fishery and typically range from ~20-100% of
fishing events for bottom trawl fisheries and 15-100% for pot and longline fisheries. On the U.S.
west coast coverage rates are ~5-100% of fishing events for pot and longline fisheries and ~20%
from 2002 to 2010 and 100% from 2011 to 2022 for trawl fisheries. Further details on observer
protocols, coverage rates and data collection can be found in NWFSC (2023), AFSC (2023),
AMR (2005), NMFS (2023) and Somers et al. (2024).

161 The individual observations used in the analysis of fishery data were the bycatch in kg of 162 benthic taxa for each haul or set of a gear type. Zero catches were not used in this analysis since 163 they would not be relevant to setting a taxa specific threshold. Effort data from the commercial 164 fishery (e.g. hours towed by each bottom trawl) was also not used in this analysis. This was 165 primarily due to the fishery data being collected over many different gear types and target 166 fisheries, so comparisons of effort between, for example, longline gear and trawl gear were 167 impossible, without strong assumptions regarding the seafloor area contacted by the gear and the 168 impact of that contact. Even within gear types, such as bottom trawl gear, the gear types were not 169 comparable as the relative effort and catchability of benthic taxa in rockfish gear with large 170 rollers on the footrope was different than for flatfish gear with less robust footrope construction. 171 Thus standardizing the catches among gear types, fisheries and VME indicator taxa was not 172 possible. Instead, here we assumed that within a broad gear type (e.g. bottom trawl), the bycatch 173 of VME indicator taxa would reflect their relative abundance on the seafloor.

174 Observers typically identified benthic invertebrate taxa to the lowest possible taxonomic 175 resolution. In some cases this was species, but more often it was a higher taxonomic level (e.g. 176 class or higher for sponges). The taxonomic groups used for reporting also varied somewhat over

177 time and among regions, for example in Alaska sponges were lumped into a single taxonomic 178 grouping (Porifera), while in British Columbia sponges were split into Demospongiae, 179 Hexactinellida and Calcarea (Table 1). Similarly, observers in Alaska sometimes identified 180 observations of gorgonian, stony coral, black coral, and soft coral groups only as a single coral 181 group (corals and bryozoans). The majority of the occurrence records were from bottom trawls 182 (particularly in the Aleutian Islands), while longline gear and pot gear had fewer observations of 183 benthic invertebrate bycatch (Table 1). Porifera tend to be the most abundant VME indicator taxa 184 overall across all the regions studied here (Hourigan et al. 2017, Stone et al. 2011). Of these, 185 Demospongiae at shallower depths in Alaska and Hexactinellida across all depths in the more 186 southern regions are dominant, with Calcarea being relatively rare across all regions. The 187 dominant coral taxa on hard substrate in continental shelf waters of the northeast Pacific are 188 typically Primnoidae, with Keratoisididae and Antipatharia dominating in deeper slope waters 189 (Hourigan et al. 2017, Wilborn et al. 2021, Stone et al. 2023). In soft-bottom substrates sea whips 190 (pennatulaceans) tend to be the most common VME indicator taxa (Hourigan et al. 2017, 191 Wilborn et al. 2021, Stone et al. 2023).

192 Camera survey data

The other source of data used in these analyses were obtained from underwater camera transects conducted in Alaska from 2012-2019 (Table 2). These data were collected using a stereo-camera system in the Aleutian Islands (n = 216, Rooper et al. 2018), eastern Bering Sea outer shelf and slope (n = 250, Rooper et al. 2016) and Gulf of Alaska (n = 338, Sigler et al. 2023) as a part of a series of species distribution model validation studies and were only available for these three regions. The stereo-camera surveys followed roughly the same sampling protocol in each of the regions, with stations chosen using a stratified random sampling design

200 (Aleutian Islands and eastern Bering Sea) or a haphazardly stratified sampling design (Gulf of 201 Alaska). Stratification in the Aleutian Islands and the eastern Bering Sea was by depth and were 202 designed to collect data that would represent each of the regions to depths of 900 m. Transects 203 targeting 15 minutes of on-bottom time were visually surveyed at each selected location (\sim 350 – 204 420 m in length). Fish and benthic invertebrates (primarily corals and sponges) were enumerated 205 to their lowest possible taxonomic level (sub-family in most cases) and all or a subsample were 206 measured for total height using stereo-image analysis (Williams et al. 2010). The area observed 207 by the camera was calculated using the distance traveled during transect observations and the 208 median target distance for each transect, assuming 100% detection at a swath width equivalent to 209 the viewing width, typically 2 - 5 m (Rooper et al. 2016). Density was then calculated as the 210 number of each taxa observed on a transect divided by the area observed on that transect in 211 $no./m^2$. Densities of individual taxa were summed by transect into the VME indicator taxa 212 groups used for analysis (Table 2). In this analysis only transects with density greater than zero 213 were utilized (n = 196 in the Aleutian Islands, n = 183 in the eastern Bering Sea and n = 196 in 214 the Gulf of Alaska).

215 Data analysis - Fishery bycatch frequency

Bycatch data from all five regions were examined across the various VME indicator taxa
to determine the general trends and data characteristics. Mean, median, histograms and
cumulative frequency of bycatch were summarized and compared among regions (see
Supplemental Material). Naturally occurring breakpoints (Jenks breaks) and quantiles were also
computed and compared for each of the regions.

221 Geange et al. (2020) used three methods to estimate potential encounter thresholds using 222 only the shape of the cumulative bycatch curve. We applied these cumulative bycatch curve 223 threshold methods to data from the Northeast Pacific and compared among regions and across 224 taxa where the number of bycatch records within the grouping was ≥ 300 . The first of the three 225 methods used by Geange et al. (2020) fits a 3-parameter segmented regression to the cumulative 226 frequency distribution of the bycatch and was applied here to each gear type, region and VME 227 indicator taxa indicator individually. The final breakpoint of the segmented regression is used to 228 calculate the cumulative bycatch threshold. Fitting of segmented regressions for the VME 229 indicator data from fishery bycatch was completed using the segmented package in R (Muggeo 230 2008, R Core Team 2022). In the second method, the point on the cumulative frequency 231 distribution that is closest to the top-left corner (point closest to x = 0 and y = 1) was calculated 232 as

233
$$q_1 = \min_{i=1}^n \sqrt{(1 - y_i)^2 + (0 - x_i)^2}$$

referred to hereafter as the minimum distance method (Tilbury et al. 2000). The final method applied to the fishery bycatch data was to calculate the Youden Index (Youden 1950, Ruopp et al. 2008), which is the point on the cumulative distribution that is the maximum of the linear distance between the extreme points on the curve. The Youden Index is calculated as

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$$q_2 = \max_{i=1}^n (y_i + x_i - 1)$$

Variance estimates for the cumulative bycatch threshold generated using the segmented regression were taken directly from the model fit, whereas for the q_1 and q_2 variance was estimated by the bootstrap method where the bycatch data was resampled 1000 times with replacement and the variance calculated from the 1000 replicated estimates (Efron and Tibshirani1993).

244 Data analysis - Fishery-camera percentile estimation

245 An alternative and potentially improved taxa and gear specific method to estimate a 246 threshold is to compare the cumulative distribution of densities of VME indicator species from 247 camera surveys to the cumulative distribution of bycatch of the same VME indicator taxa groups. 248 The goal of this comparison was to estimate an equivalent density of VME indicator species to a 249 weight of bycatch of that taxonomic grouping. To accomplish this comparison, 20 percentiles (5th, 10th, 15th, 20th,...) from the cumulative distribution function of the observed density of VME 250 251 indicator taxa in stereo-camera surveys were used to predict the corresponding 20 percentiles (5th, 10th, 15th, 20th,...) from the cumulative distribution function of fisheries bycatch data within 252 253 each region and within each gear type in Alaska. The stereo-camera data and the bycatch data 254 were not collected at the exact same location or through the same process, so a number of 255 assumptions were required: 1) we assumed the true distribution of the density of VME indicator 256 taxa was known for each region from the stereo-camera survey, 2) we assumed that the bycatch 257 of VME indicator taxa by each gear type for each fishing event was proportional to the density of 258 VME indicator taxa at that site, 3) we assumed that the fishery events sampled from the full 259 distribution of potential densities of VME indicator taxa in a region, and 4) from this we 260 assumed that the distribution of bycatch of VME indicator taxa by a gear type in a region was 261 proportional to the distribution of density of VME indicator taxa in the region.

A linear model was fit to the percentiles of bycatch weights (dependent variable) and the percentiles of stereo-camera survey densities (independent covariate). Both the fisheries bycatch

weights and the stereo-camera survey densities were log-transformed prior to analyses to meet assumptions of normality. The log-transformed density and the log-transformed weight of bycatch of VME indicator taxa were ordered and the density and weight at each 5th percentile calculated (exploratory analyses were conducted using the 10th and 1st percentiles, but the effect on the results was negligible). The percentiles for the log-transformed weight of bycatch $(w_{t,r,g})$ were the density dependent variable in an analysis of covariance so that;

270
$$w_{t,r,g} = \beta * d_{t,r} + g + r + t + d_{t,r} * g + d_{t,r} * r + d_{t,r} * t + g * r + \epsilon$$

271 where *g* is gear type (bottom trawl, longline or pot), *r* is region (eastern Bering Sea, Aleutian 272 Islands or Gulf of Alaska), *t* is the VME indicator taxa found in Alaska (Alcyonacea, 273 Antipatharia, gorgonian, hydrocoral, pennatulacean, or Porifera), ϵ are normally distributed 274 errors. The second order interactions between gear and taxa and region and taxa could not be 275 included, since some taxa did not occur in all regions or gear types. The model was simplified by 276 removing insignificant variables in a backwards stepwise fashion until all remaining variables in 277 the model were significant (p < 0.05).

278 Once the best-fitting model was determined, the equation was used to generate 279 predictions of a potential encounter threshold based on the percentile regressions. Currently there 280 is no definition of a vulnerable marine ecosystem based on the density of deep-sea corals or 281 sponges. For demonstration purposes in this analysis, we defined a VME as a density of 1 282 individual coral colony or sponge per 5 m^2 . Using this definition and the best fitting model, 283 thresholds were generated using a $d_{t,r} = log(0.2)$ for each specific gear, taxon and region 284 combination. Confidence intervals were also estimated for the prediction. It is important to note 285 that the choice of example density was somewhat arbitrary, reflecting a sensible estimate of what

a relatively high density VME area might be. This example value could be easily updated if a
regional or commonly held density-based definition of a VME was determined (e.g. Rowden et
al. 2020). Percentile regression-based threshold bycatch weights were then compared among
regions, gears and VME indicator groupings.

290 Results

291 In the eastern Bering Sea there have been over 430,000 fishing events observed from 292 2002 to 2022 (58.5% of these observed events were bottom trawl hauls, 36.5% were longline sets 293 and 5% were pot sets). Sponges (unidentified Porifera, Hexactinellida, Calcarea and 294 Demospongiae combined) were the most commonly occurring taxa, but only occurred in 3% of 295 bottom trawl hauls and longline survey sets, and 2% of pot sets (Figure 2, Supplementary 296 Material). Coral taxa (Alcyonacea, Antipatharia, unidentified corals, gorgonians or Scleractinia) 297 occurred in < 1% of bottom trawl hauls and pot sets, while they occurred in $\sim 2\%$ of longline sets. 298 Pennatulaceans were more common in the eastern Bering Sea, occurring in > 5% of longline sets 299 and less than 1% of bottom trawl hauls and pot sets. Pennatulaceans also had the highest weight 300 of bycatch in longline gear compared to other gear types (361 mt, Figure 2). The weight of 301 sponge bycatch was highest in the eastern Bering Sea, with 1.3 mt observed in pot sets and 1,584 302 mt observed in bottom trawls. Coral bycatch although relatively infrequently caught had a 303 combined bycatch weight of 80 mt for bottom trawl hauls, 40 mt for longline sets and 0.1 mt for 304 pot sets. Hydrocorals were rarely observed to be caught both in terms of frequency of occurrence and total bycatch weight (< 0.03% frequency of occurrence and ~ 0.1 mt combined across all gear 305 306 types). In the eastern Bering Sea, the time series of bycatch indicates bycatch of VME indicator 307 taxa peaked from 2008-2015, but has been relatively low in the last 7 years (Figure 3). The

308 pattern in peak bycatch for the eastern Bering Sea was primarily driven by an increase in the309 observed bycatch of sponges during 2008-2015.

310 In the Aleutian Islands there were \sim 81,000 observed fishing events from 2002 – 2022, 311 mostly bottom trawling events (73%) and longline events (22%). Sponges occurred in 24% of 312 observed bottom trawl and longline events, while corals occurred in 16% of bottom trawl hauls, 313 but 27% of longline sets (Figure 2). Pot gear had low rates of occurrence for both coral 3.3% and 314 sponge 5.6% in observed sets. Hydrocorals and pennatulaceans occurred infrequently in Aleutian 315 Islands fishing, with frequency of occurrence in observed hauls ~1% for all gear types. Total 316 bycatch weight of VME indicator taxa in the Aleutian Islands was high (Figure 2), with observed 317 sponge bycatch over the 21 years of data exceeding 3,400 mt across all fisheries. Coral bycatch 318 was also relatively high in both the longline and bottom trawl fisheries with 69 mt and 579 mt 319 respectively. Pot gear retained less VME indicator taxa in the Aleutian Islands, with a total of 1.2 320 mt observed (Figure 2). In the Aleutian Islands, the total observed bycatch of VME indicator 321 taxa has remained relatively steady, except for Porifera, which has been higher since 2013, than 322 in the earlier half of the time series (Figure 3).

323 Forty-eight and 43% of the observed fishing events (n = 105,115) from 2002-2022 in the 324 Gulf of Alaska were bottom trawl hauls and longline sets respectively. The remainder (9%) were 325 pot sets. Sponges occurred in 4% of observed bottom trawl hauls, 2.5% of observed longline sets 326 and 1% of pot sets (Figure 2). Corals occurred in 1.7% of bottom trawl hauls, 2.7% of longline 327 sets and 0.5% of pot sets. Pennatulaceans were more frequently caught in longline sets (2.7%) 328 than in bottom trawl hauls (0.5%) or pot sets (0.2%). Hydrocorals occurred in < 0.2 % of all 329 observed fishing events in the Gulf of Alaska. Total observed weight of VME indicator taxa 330 bycatch in the Gulf of Alaska was lower than the other regions of Alaska, only 77 mt of sponge

was observed in bottom trawl hauls, 10 mt in longline sets and 0.1 mt in pot sets (Figure 2). For
corals, 29 mt was observed in bottom trawl hauls, 9 mt in longline sets and 34 kg in pot sets.
Almost 10 mt of pennatulaceans were observed in longline sets, but only around 1 mt in the
bottom trawl hauls and pot sets combined. Combined hydrocoral bycatch across all gears was 1.2
mt. Over time the Gulf of Alaska has seen no trend, but highly variable bycatch weight of VME
indicator taxa (Figure 3).

337 Since 1997 Canada has had 100% observer coverage, so the number of observed fishing 338 events from 2002-2022 was 625,129. Of these, 49% were bottom trawl hauls, 43% were longline 339 sets and 8% were pot sets. The frequency of occurrence of corals, pennatulaceans and 340 hydrocorals in bottom trawl hauls, longline sets and pot sets in Canada was < 1% from 2002-341 2022 (Figure 2). Only sponge by catch occurred at a higher frequency 1.5% in bottom trawl 342 hauls, while in longline sets and pot sets the frequency of sponge by catch was < 1%. The weight 343 of observed corals and sponges captured in bottom trawl hauls was 13 mt and 62 mt respectively 344 when combined over the 11 years (Figure 2). In longline sets and pot sets coral bycatch was 0.4 345 mt and 0.2 mt, while sponge catches were 0.1 mt and 0.04 mt. Hydrocorals only occurred in the 346 bottom trawl fishery and in total 15 kg of these were observed from 2002-2022 in Canada 347 (Figure 2). Combined VME indicator taxa bycatch decreased in Canada from 2002-2011 and has 348 remained at very low levels since then (Figure 3).

West Coast data was primarily from bottom trawl hauls (78%), while observed longline sets (13%) and pot sets (9%) comprise the remainder of the 193,158 observed fishing events. Similar to Canada, the frequency of occurrence of corals in observed fishing events was low (~1% for bottom trawls, longline sets and pot sets, Figure 2). Pennatulaceans were caught in 2.1% of bottom trawl hauls, 0.8% of longline sets, and 2.1% of pot sets, while sponge was

captured in 2.6% of bottom trawl hauls, 0.9% of pot sets. Hydrocoral frequency of occurrence
was < 0.03% across all gear types (Figure 2). Sponge was the most abundant VME indicator taxa
in the bycatch of bottom trawls (67.5 mt), longline sets (1.0 mt) and pot sets (0.1 mt). Coral
bycatch was 8.5 mt in bottom trawls and 94 kg in longline sets and 54 kg in pot sets.
Pennatulacean catches were 1.8 mt in bottom trawls, 50 kg in longline sets and 34 kg in pot sets
(Figure 2). Also similar to Canada, bycatch weight on the west coast peaked in 2009 and has
been at low levels since then (Figure 3).

361 Overall, the mean bycatch when they were observed was higher for sponges (Porifera, 362 Hexactinellida and Demospongiae) in the Aleutian Islands than the other regions (Figure 4). This 363 pattern was also apparent for most upright and branching corals (e.g. gorgonians), and some of 364 the other coral groups (stony corals and Alcyonacea as a grouped taxa). Pennatulacean bycatch 365 was higher in the eastern Bering Sea than any of the other regions across all gear types (Figure 366 4). Across all regions, the gear type with the highest bycatch overall of VME indicator taxa were 367 bottom trawls followed by longline gear and pot gear. Longline gear in the eastern Bering Sea 368 were particularly notable in having higher bycatch on average than even bottom trawls in the 369 eastern Bering Sea (mean = 42 kg, SE = 0.70 for longline gear and mean = 8 kg SE = 1.0 for 370 bottom trawls), this pattern was also true in the Gulf of Alaska and Aleutian Islands. Although 371 the range of bycatch weights generally overlapped among regions for a gear type and VME 372 indicator taxa group, there were significant differences (p < 0.05) in the mean values across all 373 regions for all gear types for each indicator taxa (Figure 4) as indicated by analysis of variance 374 and Tukey's post-hoc comparisons.

375 Fishery cumulative bycatch thresholds

376 The distributions of bycatch of VME indicator taxa for almost all gear types, regions and 377 taxa were heavily right-hand skewed. This was true for taxa with very few observations (e.g. 378 Alcyonacea in the west coast longline fishery with n = 100 observed catches, Figure 5) and large 379 numbers of observations (e.g. gorgonians in the Aleutian Islands bottom trawl fisheries, Figure 380 5). See the supplemental information for the full array of bycatch from all combinations of VME 381 indicator taxa, gear type and region. The skewness of the bycatch data resulted in distributions 382 where the median was often at least an order of magnitude lower than the mean (Figure 5). So 383 for example, the mean by catch of Demospongiae in bottom trawls in Canada was 16 kg, while 384 the median bycatch (meaning 50% of the catches were above and below) was 0.9 kg.

385 For the most part, the cumulative bycatch-based thresholds suggested by the Youden 386 Index and the minimum distance metrics were similar, if not exactly the same within taxonomic 387 group-region-gear type combinations (Figure 6 and Supplemental Figures and Tables). Where 388 the Youden Index and minimum distance metrics were slightly different, their standard error bars 389 overlapped indicating that the difference was not statistically significant (Figure 7). The 390 segmented regression tended to estimate a cumulative bycatch-based threshold (third break 391 point) that was lower (and almost always significantly lower) than the two other methods. 392 Reflecting the relative catches in each of the regions, the cumulative bycatch-based thresholds 393 were generally highest in the Aleutians and lowest on the US west coast (Figure 7). When 394 averaged across regions and break points, Porifera had the highest cumulative bycatch-based 395 threshold of any of the taxonomic groups. Pennatulaceans stood out in the longline gear, with 396 high cumulative bycatch-based thresholds (> 75 kg) in the eastern Bering Sea only (Figure 7). 397 There were not enough occurrences of any VME indicator taxa bycatch in the pot fishery to

estimate cumulative bycatch-based thresholds (see supplemental material for cumulative bycatch
curves for the pot fishery). There were also not enough occurrences of Scleractinians in any of
the fisheries to estimate cumulative bycatch-based thresholds.

401 Percentile regression thresholds based on image data

402 The regression of percentiles of log-transformed observed VME indicator taxa density against percentiles of log-transformed VME indicator bycatch in Alaskan fisheries resulted in 403 404 consistent patterns among gear types and regions for most fishing gears (Figure 8). The full 405 model included all possible interaction terms. Two could not be included due to the unbalanced 406 design (Gear-VME taxa and Region-VME taxa). The density-region term was insignificant (p = 407 0.48) and was removed from the best-fitting model. In the best fitting model all main effects 408 (gear type, VME indicator taxa and region) were significant, as well as the covariate interactions 409 between log-transformed VME density observed in the camera and VME indicator taxa and 410 region. The gear type-region interaction term was also significant (Table 3).

411 The predicted percentile regression thresholds were highest in the Bering Sea for bottom 412 trawl gear across most VME indicator taxa (Figure 9). The predicted percentile regression 413 thresholds were lower for longline gear, but also tended to also be slightly higher in the eastern 414 Bering Sea. For pot gear, predicted percentile regression thresholds were uniformly low across 415 all taxonomic groups and regions. For gorgonians, the estimated threshold ranged from 11.52 kg 416 in the Gulf of Alaska to 19.15 kg in eastern Bering Sea for bottom trawls (predicted at a camera 417 density of 0.5 colonies*m⁻², Supplemental Table S2). For Porifera, the values were larger 418 ranging from 79 kg in the Gulf of Alaska to 131 kg in the eastern Bering Sea. The threshold 419 values determined by regressing percentiles of observed density against percentiles of bycatch

420 were uniformly lower than the threshold values estimated by the minimum distance, Youden

421 Index or segmented regression for the same regions and gear types in Alaska.

422 **Discussion**

Unsurprisingly, bycatch of VME indicator taxa in bottom trawls was higher than for other
gears across multiple taxa and all of the observed regions. Sponges were the most frequently
captured VME indicator taxa and accounted for the most bycatch weight of any of the VME
indicator taxa across all regions, generally

427 due to their broader distribution, larger size and heavier body than most coral species. 428 Hydrocorals were generally the least common bycatch species in all regions by frequency of 429 occurrence and weight, but coral taxa and pennatulaceans were fairly common although with 430 regional differences (e.g. the high abundance of pennatulaceans in eastern Bering Sea catches 431 relative to other areas). Total bycatch frequency and weight was highest in the Aleutian Islands 432 and lowest in Canada and the U.S. west coast.

433 Fishery bycatch of VME indicator taxa generally agreed with the observed density where 434 underwater image data were available. Areas with high density in the images (e.g. sponges in the 435 Aleutian Islands or pennatulaceans in the eastern Bering Sea) yielded high bycatch in fisheries. 436 The shape of the distribution of both the fishery bycatch data and the camera survey density data 437 were similarly highly skewed with large right-handed tails. These general characteristics of the 438 two data sources and their agreement provides some comfort that the patterns and relationships 439 developed in the analysis are complementary. Of the two methods (bycatch data only or 440 percentile regression), the percentile regressions tended to generate lower bycatch thresholds

441 across all taxa. However, these comparisons could only be made for data in Alaska, as camera442 surveys for density were not available for the other regions.

443 Temporal trends in bycatch

444 Time trends were evident in coral and sponge bycatch over the regional time series. In 445 part this may have been due to changes in fishing regulations that were designed to protect benthic habitat in all three jurisdictions. In British Columbia an individual transferable quota 446 447 (ITQ) system was put in place to manage bycatch of corals and sponges in 2007 (Wallace et al. 448 2015). This catch accounting system succeeded in reducing the amount of bycatch overall in the 449 fishery and also caused a shift in fishing effort away from coral and sponge hotspots (Gale et al. 450 2022). In the Aleutian Islands of Alaska and on the U.S. west coast, spatial closures to protect 451 known or suspected concentrations of sponge and coral and to protect essential fish habitat were 452 implemented in 2005 and 2006 respectively. In the Aleutian Islands, these closures were focused 453 on areas outside of those historically fished, essentially freezing the fishing footprint at that 454 moment. The resulting consistency of spatial patterns in fishing effort did not, at least initially, 455 result in clear declines of coral and sponge bycatch, as was observed in British Columbia. West 456 coast closures excluded fishing in some areas that were actively fished at the time of closure, but 457 bycatch continued to increase until a peak in 2009, after which it has declined to relatively low 458 levels. This initial increase may have been caused by an increase in fishing effort along the 459 continental slope (PFMC 2023), that could have more closely overlapped the distribution of 460 VME indicator taxa (Poti et al. 2020). However the west coast bottom trawl effort also peaked in 461 2009, so it also appears likely that the decrease in bycatch of VME on the west coast was a result 462 of implementation of the groundfish ITQ program in 2011. This implementation has resulted in 463 fleetwide reductions of bottom trawl fishing effort by more than ¹/₂ since 2010 and also resulted

464 in some changes to fisher behavior, such as avoidance of prohibited species (Somers et al. 2023). 465 On the U.S. west coast footrope restrictions were also implemented in 2001 that were 466 demonstrated to have moved the trawl fishery away from hard-bottom substrates, potentially 467 making the response to spatial closures somewhat more complex (Bellman et al. 2005). All of 468 these changes in regulation resulted in behavioral changes that may have also significantly 469 reduced VME indicator taxa by catch directly or indirectly. It is also likely that in areas that have 470 been continually fished, reductions in VME indicator taxa bycatch are the result of direct 471 removals and mortality due to historic damage from fishing gear that has resulted in reduced 472 biomass of these taxa on the seafloor.

473 Catch efficiency

474 Few if any studies have measured the efficiency of different gear types in capturing 475 benthic invertebrates. The most comprehensive review of catchability of VME indicator taxa is 476 for bottom trawls and can be found in SPRFMO (2022). The authors examined published and 477 unpublished data sets from a variety of regions and substrate types and found that the 478 catchability estimates by bottom trawls were generally < 5%, but could range as high as 27% for 479 some taxa. However, SPRFMO (2022) also noted that many of these estimates were both highly 480 variable and based on very small sample sizes. Studies in Alaska have shown that a single pass 481 of a bottom trawl can remove a substantial biomass of corals and detach a high proportion 482 (~27%) of the colonies in its path (Krieger 2001). A single study that examined density of 483 sponges along experimental bottom trawl tow paths found that the densities for two types of 484 upright sponges were 16% and 31% lower in an experimentally trawled area versus background 485 densities (Freese 2001). The rate of damaged sponges remaining in the trawl path was 67% 486 (Freese 2001), and the overall density of sponges in the trawled transects had not recovered 13

487 years post-trawling (Malecha and Heifetz 2017). Moran and Stevenson (2000) estimated a 488 standard demersal trawl reduced benthic invertebrate density by ~16%, with only 4% of the 489 removed organisms retained in the net. Removals of 13.8% of sponges and 3% of gorgonians by 490 a bottom trawl in Australia was observed by Wassenberg et al. (2002), however the removals 491 varied by both organism height (with those higher than 50 cm most likely to be impacted) and 492 morphotype (with broad-based sponges more likely to be impacted). Sainsbury et al. (1997) 493 looked at the catchability of sponges >15 cm in height and found that 89% were removed by a 494 trawl. Catchability from longline, trap gear or longlined pots has not been well studied, but a 495 study by Pham et al. (2014) estimated very low removal rates for longline gear of 0.058% for 496 branched corals and 0.011% for unbranched corals. They observed impacts (ranging from minor 497 damage to non-survivable damage) on 47% of the corals located near lost fishing gear. These 498 catch efficiencies and removal rates are much lower than have been estimated for bottom trawls, 499 which explains in large part the smaller thresholds for most VME indicator taxa for longline and 500 pot gear estimated by this study (e.g. a lower catchability means that a small by catch of VME 501 indicator taxa would indicate a comparatively large density of that taxa on the seafloor).

502 Caveats for the camera density-fishery bycatch comparisons

The analysis comparing the stereo-camera data and the fishery bycatch data required strong assumptions regarding the validity of the density estimates from the underwater camera and the proportionality of the bycatch data to that density. These assumptions could not be tested during the analyses, so the results should be viewed in that context. There were no indications that the density of VME indicator taxa were biased, as the estimates were collected via random stratified sampling and should estimate the density accurately across space. However, the fishing activity in Alaska was likely spatially biased. The fisheries operating in the different regions of

510 Alaska target different species that may not fully represent available habitats. For example, the 511 majority of bottom trawls in the eastern Bering Sea were targeting Walleye Pollock or flatfish 512 assemblages. As such, they were more likely to occur in flat-soft sediment areas where structure 513 forming invertebrates, except pennatulaceans, were absent. In contrast, the majority of bottom 514 trawls in the Aleutian Islands targeted rockfish species or Atka mackerel that are more likely to 515 occur in hard-bottom areas where corals and sponges are more likely to be present. This is 516 reflected in the low overall frequency of occurrence of VME indicator taxa (except 517 pennatulaceans) for the eastern Bering Sea and the high frequency of occurrence in the Aleutian 518 Islands. The impact of spatial bias and potential bias in the habitats sampled by the fishery may 519 have been mitigated somewhat in this analysis by using only those bottom trawl hauls that 520 captured benthic invertebrates. Catchability of VME indicator taxa within a gear type likely also 521 varied (e.g. footropes with tire gear may be used for rockfish trawling on hard substrate while 522 footropes with rubber disks for flatfish trawling in soft sediments). This likely had some impact 523 on the results of this study, especially for Porifera which are known to occur in both soft and 524 hard substrates in Alaska (Rooper et al. 2016). However, for the VME indicator taxa that have 525 specific substrate associations, such as gorgonians found predominantly on hard bottom, similar 526 types of trawls and thus similar types of catchabilities would be expected. So the encounter 527 thresholds resulting from this study should be considered a generic result representing the mix of 528 broad gear types used for fishing on the west coast of North America.

Although the observer programs adhere to rigorous statistical designs for observer deployment and subsampling of the catch resulting in high quality bycatch data, the taxonomic resolution of the fishery data used in Alaska is also a source of uncertainty. Similar species identification resources are available and used across the different regions, but these have

533 changed over time (e.g. Stone 2011, Wilborn et al. 2021). In some cases, the taxonomic 534 resolution of Alcyonaceans (coral or bryozoan) recorded by observer programs in Alaska is less 535 specific than the taxonomic resolution of the camera data and includes a taxa (bryozoan) that is 536 not in fact a coral. However, given the small size and lack of hard skeletal structure in bryozoans 537 their contribution to the overall weight of bycatch may have been minimal. The broader category 538 Alcyonacean certainly included some members of the gorgonian families as well. At-sea 539 observers primarily assess and sample targeted fish and invertebrate catches in order to support 540 fisheries stock assessments. The extensive training needed to more successfully identify corals to 541 lower taxonomic levels has generally not been prioritized (Stone et al. 2015). These 542 characteristics of the bycatch data (poor taxonomic resolution, potential misidentification issues 543 and inclusion of bryozoans) made the comparisons with camera data less certain.

544 The spatial distribution of historical fishing was not considered in the random stratified 545 design for the stereo camera surveys in Alaska. This has implications for the results of this study 546 in that it is likely that some or many (depending on the region in Alaska) of the image transects 547 occurred in areas that may have been previously fished by one of the gear types. The impact of 548 using densities from previously fished areas on the results would have been to reduce the 549 densities of VME indicator taxa relative to pristine condition and thus increase the frequency of 550 lower densities in the cumulative frequency distribution. This in turn may have reduced the 551 percentile values for the observed density distribution resulting in a decrease in the 552 corresponding fisheries by catch threshold. Similarly, if the cumulative distribution of VME 553 indicator taxa in fisheries bycatch was truncated or reduced by historical fishing, the impact on 554 the analysis would be a reduction in estimated encounter threshold. As such, these data and 555 results are indicative of the current nature of VME indicator taxa abundance in Alaska, rather

than reflecting what might be found in a pristine ecosystem. Comparisons of the density
distributions of VME indicator taxa from unfished areas/times to densities after fishing has
occurred would be useful in teasing out the impacts of historical fishing on the results. However,
the spatially-explicit fishing effort data from these gears and regions date at earliest from late in
the 20th century and so cannot be used to identify unfished areas at the necessary time scales.

561 This analysis uses the best available data to determine thresholds for bycatch in the North 562 Pacific. Specifically this data-informed method could set gear and VME indicator taxa specific 563 thresholds for bycatch that would trigger implementation of a spatial closure and a move-on rule. 564 The analysis indirectly attempts to measure relative catchability of VME indicator taxa using the 565 distributions of catches. Ideally, data would be collected that could directly measure catchability 566 and damage rates of benthic organisms in the deep-sea. Selectivity for fishes in fishing gear has 567 long been studied to support stock assessment analysis (e.g. MacLennan 1992). However, these 568 data are not easily attained for non-motile VME indicator taxa. In part this is due to their 569 tendency to break apart when contacted by the gear (Freese 1999), which makes it difficult to 570 judge the original size of the organism based on the catch. Another difficulty is that the 571 individuals may not be entirely removed or even removed at all by the fishing gear, yet can still 572 experience mortality or damage (NRC 2002, Stone 2014, Malecha and Heifetz 2017). This has 573 necessitated either correlative assessments, such as the remote camera study described here, or 574 experimental studies, such as those where underwater imagery is used to look for mortality and 575 damage after known trawling events (Freese 1999, Wassenberg 2002). More of these types of 576 studies with larger sample sizes, fishing gears that include non-mobile gears, and varying 577 densities of benthic invertebrates are needed.

578 **Comparisons to other work**

579 As has been observed in other studies attempting to set data-based encounter thresholds, 580 it is difficult to estimate a threshold that is applicable across wide regions. Most previous 581 attempts to set thresholds have in part utilized some version of fisheries bycatch data (e.g. 582 Kenchington et al. 2009, Parker et al. 2009 and Geange et al. 2020 for the North Atlantic 583 Fisheries Organisation, NAFO, and SPRFMO). However, fisheries management organizations 584 have generally taken their own approach to setting thresholds, including less quantifiable 585 methods such as expert opinion (Ardron et al. 2014), an arbitrarily chosen percentile of the 586 historical bycatch (Wallace et al. 2015), or adopted encounter thresholds put forward by other 587 RFMOs (FAO 2016). The NAFO put into place encounter thresholds that were based on a kernel 588 density analysis (Kenchington et al. 2009, Kenchington et al. 2014) that examined bycatch in 589 research survey trawls scaled up to account for the differences in commercial trawling 590 techniques and gear (Kenchington et al. 2011). SPRFMO analyzed bycatch of VME indicator 591 taxa by New Zealand fishers both domestically and in international waters to develop a threshold that was based both on the 50th percentile of the cumulative weight frequency distribution of 592 593 bycatch and on the VME indicator taxa richness in the bycatch (Parker et al. 2009). Geange et al. 594 (2020) called for using area specific thresholds, but also recognized that data paucity forced them 595 to combine data across areas and taxa. In the absence of known VME indicator taxa densities 596 from underwater camera footage for British Columbia and the U.S. West Coast, the current study 597 averages across Alaska regions to provide an estimated threshold that can be applied coast-wide. 598 However, this estimate is very sensitive to the data from the Aleutian Islands. Removing the 599 Aleutian Islands from the analysis reduces the percentile-based threshold by over 50% for both 600 Porifera and the other coral taxa (gorgonians, Antipatharia, and Alcyonacea). This is due to the

601 high density of VME indicator taxa in the Aleutian Islands which has been shown to be related to 602 the high abundance of suitable hard substrates for VME indicator taxa and the relatively high 603 current speeds that deliver production to benthic invertebrates in the ecosystem (Rooper et al. 604 2016). The naturally occurring densities of some of the VME indicator taxa examined here also 605 vary substantially. For example, sea whip densities in the eastern Bering Sea have been estimated as high as 8.4 individuals * m⁻², whereas corals only ranged to 0.28 individuals * m⁻² (Rooper et 606 607 al. 2016). The differences in naturally occurring densities among VME indicator taxa also have implications for encounter thresholds, as a lower density (than the $0.2 * m^{-2}$ example used here 608 for all taxa) might be preferred for taxa that are naturally less dense. So it is clear that bycatch 609 610 and visual survey data collected from a region should be used to set a data-derived threshold for 611 that region for each VME indicator taxa. However, the lack of available image-based density 612 data that fully represents regions and substrate conditions limits development of more region 613 specific relationships for British Columbia and the U.S. West coast. There are likely both 614 published and unpublished data available from these regions that could be compiled and used in 615 future analyses.

616 **Conclusions and Recommendation**

617 The percentile regression thresholding method allowed the development of linear 618 relationships between density percentiles of VME indicator taxa observed on the seafloor and 619 observed bycatch in that same region. This method allows density on the seafloor to be easily 620 converted to VME indicator taxa bycatch. For example, if managers wished to protect 621 gorgonians at densities above a density of 5 individuals per 100 m^2 from bottom trawling, a 622 bycatch weight of 15.7 kg would be used to trigger an encounter based closure using the average 623 regression coefficients developed for this taxa. In contrast, encounter thresholds based on

cumulative catch from bycatch data only were able to distinguish break points, but with no
biological basis for these breakpoints being meaningful or relating to a specific VME prevalence
(Ardron et al. 2014, Geange et al. 2020). In the absence of better available data, we recommend
using the percentile regression approach for setting VME encounter thresholds even across
regions. This method is repeatable, quantifiable and easily updateable when new data become
available, an improvement over expert opinion or an arbitrarily chosen thresholds based on
historical bycatch.

631 One of the most important conclusions from this analysis is that regional-specific data 632 (bycatch and visual survey data) should be used derive encounter thresholds for that region. 633 Future studies should collect data that would allow the development of regional and gear specific 634 percentile regressions that represent the regional distribution of VME indicator taxa densities and 635 quantify the relationship between actual VME density and bycatch amounts. Ideally, this work 636 would include experiments that documented the density of VME indicator taxa along transects 637 before and after fishing and documented the amount of retained VME indicator taxa and 638 damaged VME indicator taxa after fishing. These experiments could likely be done with co-639 located (overlapping) visual surveys and fishing activity (e.g. Freese 2001, Malecha and Heifetz 640 2017) or with camera systems that are mounted on the fishing gear (e.g. headrope mounted cameras, McCarthy et al. 2023) so that the data on density and catch are collected on the same 641 642 tow path. A series of such experiments could collect data that would allow direct estimation of 643 VME indicator taxa catchability by gear type across a wide range of seafloor substrates and 644 VME densities. These improved datasets could better inform ecosystem-based fisheries 645 management by utilizing recent methods such as Rowden et al. (2020), which identified densities 646 of VME indicator taxa using visual imagery that are associated with thresholds in diversity, or

647 Baco et al. (2023), which used individual images of the benthos to define a vulnerable marine 648 ecosystem. These datasets combined could be used to support ecologically-relevant encounter 649 thresholds that would protect the important characteristics of vulnerable marine species and the 650 ecosystem services that they provide.

651 Deep sea corals and sponges are long-lived, reproduce slowly and are known to have 652 episodic recruitment. Primnoid corals and Hexactinellida are known to live for over 100 years, 653 Antipatharians have been documented to live for thousands of years, and scleractinian and 654 hexactinellid sponge reefs accumulating for tens of thousands of years (Conway et al. 1991, Levs 655 and Lauzon 1998, Andrews et al. 2002, Fallon et al. 2010, Fallon et al. 2014, Hitt et al. 2020). In 656 our current analysis we have developed a method that can be used to identify high density areas 657 of VME indicator taxa. The thresholds identified by this regression method are lower than those 658 estimated by cumulated catch curves and are generally lower than the thresholds currently in use 659 by most RFMOs (FAO 2016). This would imply that these thresholds would be more effective at 660 protecting aggregations of deep-sea corals and sponges. Given the long lifespan of these taxa and 661 the time needed for recovery it is still unclear that encounter thresholds combined with move-on 662 rules are an effective way to minimize significant and adverse impacts to VME (Auster et al. 663 2011). It has been pointed out that a more precautionary approach might be to enforce move-on 664 rules and implement closures upon any encounter of VME indicator taxa (Auster et al. 2011). 665 However, the risk in implementing move-on rules guided by any encounter threshold is that it 666 will effectively move fishing activity into new areas, spreading any significant and adverse 667 impacts across a wider area since it has been found that the majority of damage to corals and 668 sponges is caused by the first pass of fishing gear (Moran and Stevenson 2002, Wassenburg et al. 669 2002). Ultimately, alternative management tools might lead to better outcomes for VME

670 protection than simple use of move-on rules that trigger small spatial closures. For example, 671 combining comprehensive surveys of VME indicator taxa, species distribution modeling and mapping of fishing effort by gear type would allow areas of high risk to be identified and 672 673 protected. The approach used in Canada's groundfish trawl fisheries (Wallace et al. 2015) to 674 require 100% observer coverage, set an ITQ for vessels limiting the amount of VME indicator 675 taxa that can be captured, freezing the existing footprint and implementing encounter thresholds 676 with move-on rules triggering closures has been very effective at reducing the VME indicator 677 bycatch by these fisheries (Gale et al. 2022). These approaches, although likely better for 678 protection of VME, require more high quality data collections than are currently available for 679 most areas of the world's oceans, particularly those areas managed by regional fisheries 680 management organizations.

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- 906 Tables
- 907 Table 1. Taxonomic grouping of data collected from commercial fisheries 2002-2022 by gear
- 908 type in the northeast Pacific Ocean within the five study regions (Aleutian Islands, eastern
- 909 Bering Sea, Gulf of Alaska, British Columbia and the US west coast) and number of observations
- 910 (hauls) where each taxonomic grouping was recorded.

Observer	VME indicator		eastern Bering	Aleutian	Gulf of
classification	taxa	Gear type	Sea	Islands	Alaska
		Bottom			
Alcyonacea	Alcyonacea	trawl	296	852	31
		Longline	354	30	69
		Pot	2		1
		Bottom			
CoralsBryozoans	Alcyonacea	trawl	1664	7557	526
		Longline	3984	5481	1559
		Pot	120	100	39
		Bottom			
Antipatharia	Antipatharia	trawl	23	451	23

		Longline	2	103	39
		Pot	1	1	1
		Bottom			
Gorgonian	Gorgonian	trawl	253	6379	295
		Longline	349	869	244
		Pot	8	23	3
		Bottom			
Hydrocoral	Hydrocoral	trawl	12	710	58
		Longline	38	688	74
		Pot	1	23	1
_		Bottom			
Pennatulacean	Pennatulacean	trawl	1479	213	228
		Longline	15452	291	1608
		Pot	13		22
C 1		Bottom			
Calcarea	Porifera	trawl			
		Longline			
		Pot			
D	D:	Bottom			
Demospongiae	Porifera	trawi			
		Longline			
		Pot			
Havaatinallida	Domiforo	Bottom			
пехасипенна	Formera	Langling			
		Det			
		Pot			
Porifera	Porifera	trawl	11243	26984	2251
1 officia	1 officia	Longline	7296	2090 4 /830	13/10
		Dot	566	-050	100
		Bottom	500	220	100
Scleractinia	Scleractinia	trawl	1	82	4
		Longline	1	2	33
		Pot			3

VME indicator grouping	Aleutian Islands	Gulf of Alaska	eastern Bering Sea
Alcyonacea	2		
Antipatharia	54	5	
Calcarea	9	3	4
Demospongiae	177	141	107
Hexactinellida	88	95	56
Gorgonian	137	64	32
Hydrocoral	102	54	
Pennatulacean	81	99	105
Porifera		1	9

indicator taxa by region from stereo-camera surveys from Alaska in 2012-2019.

Table 2. Summary of number of transects with observations of vulnerable marine ecosystem

Table 3. Results of analysis of covariance relating the percentiles of bycatch weight in the

916 commercial fisheries to the percentiles of density for stereo-camera surveys in regions of Alaska

917 by gear type. Df are the degrees of freedom, Sum Sq are the sum of squares, Mean Sq is the

918 mean squared error and p is the probability of the outcome (significance was indicated at p < p

0.05).

Term	Df	Sum Sq	Mean Sq	F value	р
Density percentile (camera)	1	1,818.5	1,818.5	1,750.1	0
Gear type	2	864.0	432.0	415.7	0
Region	2	105.3	52.6	50.7	0
VME indicator taxa	4	377.3	94.3	90.8	0
Density percentile (camera) * Gear type	2	146.2	73.1	70.4	0
Density percentile (camera) * VME indicator taxa	4	43.3	10.8	10.4	0
Gear type * Region	4	42.8	10.7	10.3	0
Residuals	737	765.8	1.0		

923 Figures



924

- 925 Figure 1. Map of the northeast Pacific Ocean and the five study regions: Aleutian Islands,
- 926 eastern Bering Sea, Gulf of Alaska, British Columbia and the U.S west coast. Also shown are the
- 927 locations of stereo-camera transects conducted in Alaska ecosystems from 2012-2019 indicated
- 928 by red points.



Figure 2. Frequency of occurrence calculated as the number of observed occurrences divided by
the number of observed hauls (top panels) and total weight of bycatch aggregated across all

observed hauls (bottom panels) of each VME indicator taxa captured by gear type and region.

935 The data used in this figure can be found in Table S1 in the supplementary material.



Figure 3. Time series of total annual bycatch of vulnerable marine ecosystem indicator taxa in
each of the five regions of the NE Pacific Ocean from 2002-2022. The sum of the weight of the
bycatch (in kg) for each year has been standardized by the number of fishing events (bottom
trawl hauls, longline sets and pot sets) observed in each year combined across gear types.



944 Figure 4. Mean bycatch weight of VME indicator taxa by gear type and region in instances
945 where there was positive bycatch in commercial fishing gear. Data are combined across all

- 946 years (2002-2022) and values are plotted on a log-scale. In this figure Porifera includes
- 947 unidentified Porifera, Hexactinellida, Demospongiae and Calcarea combined.



949 Figure 5. Histograms of bycatch data for example VME indicator taxa by gear type in four

- 950 example regions of the NE Pacific Ocean from 2002-2022. Dashed lines indicate the 90%
- 951 quantile (red), the mean bycatch (orange) and the median bycatch (blue). Additional
- 952 combinations of taxa and gear type by region can be found in the supplemental material (Figure

953 *S1*).



955 Figure 6. Cumulative frequency distributions of bycatch in commercial fisheries from 2002-2022

- 956 for four example vulnerable marine ecosystem indicator taxa by gear type in four example
- 957 regions of the NE Pacific Ocean. The full set of combinations of taxa and gear type by region
- 958 can be found in the supplemental material (Figure S2).



Figure 7. Vulnerable marine ecosystem indicator taxa bycatch thresholds estimated from the

cumulative frequency of bycatch data in each region and gear type in 2002-2022. Where the

- 963 number of data points were < 300 a threshold value was not calculated. Threshold values used
- 964 for this plot are shown in Table S2.



Figure 8. Linear regressions of percentile-percentile plots of log commercial fishery bycatch and

- 968 log density from stereo-camera surveys of vulnerable marine ecosystem indicator taxa by gear
- 969 type in each of the three regions of Alaska. Gray shaded areas indicate 95% confidence
- *intervals*.



972

973 Figure 9. Predicted thresholds by gear type, region in Alaska and taxonomic grouping of VME

974 using the percentile regression method. Error bars indicate 95% confidence intervals. Values

⁹⁷⁵ used for this plot are shown in Table S2.



979 Figure S1. Histograms of bycatch data from the eastern Bering Sea from 2002-2022. Dashed
980 lines indicate the 90% quantile (red), the mean bycatch (orange) and the median bycatch (blue).



Figure S1 (cont.). Histograms of bycatch data from the Aleutian Islands from 2002-2022.
Dashed lines indicate the 90% quantile (red), the mean bycatch (orange) and the median
bycatch (blue).



Figure S1 (cont.). Histograms of bycatch data from the Gulf of Alaska from 2002-2022. Dashed
lines indicate the 90% quantile (red), the mean bycatch (orange) and the median bycatch (blue).



Figure S1 (cont.). Histograms of bycatch data from Canada from 2002-2022. Dashed lines
indicate the 90% quantile (red), the mean bycatch (orange) and the median bycatch (blue).



Figure S1 (cont.). Histograms of bycatch data from the U.S. West Coast from 2002-2022.
Dashed lines indicate the 90% quantile (red), the mean bycatch (orange) and the median
bycatch (blue).



Figure S2. Cumulative frequency distributions of bycatch of VME indicator taxa by gear type in
the eastern Bering Sea from 2002-2022. Points indicate the fit threshold values (where n > 300)
for the minimum distance (MinDist), segmented regression (Segmented) and Youden Index

999 (YoudenIndex) methods.



1001 Figure S2 (cont.). Cumulative frequency distributions of bycatch of vulnerable marine ecosystem

1002 indicator taxa by gear type in the Aleutian Islands from 2002-2022. Points indicate the fit

- 1003 threshold values (where n > 300) for the minimum distance (MinDist), segmented regression
- 1004 (Segmented) and Youden Index (YoudenIndex) methods.



1006 Figure S2 (cont.). Cumulative frequency distributions of bycatch of vulnerable marine ecosystem

- 1007 indicator taxa by gear type in the Gulf of Alaska from 2002-2022. Points indicate the fit
- 1008 threshold values (where n > 300) for the minimum distance (MinDist), segmented regression
- 1009 (Segmented) and Youden Index (YoudenIndex) methods.



1011 Figure S2 (cont.). Cumulative frequency distributions of bycatch of vulnerable marine ecosystem

- 1012 indicator taxa by gear type in Canada from . Points indicate the fit threshold values (where n >
- 1013 300) for the minimum distance (MinDist), segmented regression (Segmented) and Youden Index

1014 (YoudenIndex) methods.



1015

1016 Figure S2 (cont.). Cumulative frequency distributions of bycatch of vulnerable marine ecosystem

- 1017 indicator taxa by gear type on the U.S. West Coast from 2002 2022. Points indicate the fit
- 1018 threshold values (where n > 300) for the minimum distance (MinDist), segmented regression
- 1019 (Segmented) and Youden Index (YoudenIndex) methods.



Figure S3. Cumulative frequency distributions of density of VME indicator taxa in camera
surveys of Alaska regions from 2012 - 2019.

- *Table S1. Frequency of occurrence calculated as the number of observed occurrences divided by*
- 1024 the number of observed hauls and total weight (kg * 10⁻³) of catch aggregated across all
- *observed hauls of each VME indicator taxa captured by gear type and region.*

			Weight (kg	Frequency of
Gear type	Region	VME indicator taxa	* 10 ⁻³)	occurrence
Bottom trawl	Bering Sea	Alcyonacea	1.9770	0.001150
		Antipatharia	0.1007	0.000091
		CoralsBryozoans	75.1722	0.006270
		Gorgonian	3.0996	0.000984
		Hydrocoral	0.0962	0.000047
		Pennatulacean	11.7234	0.005535
		Porifera	1583.7927	0.029177
		Scleractinia	0.0037	0.000004
	Aleutian Islands	Alcyonacea	25.2340	0.013399
		Antipatharia	1.7419	0.007453
		CoralsBryozoans	460.1901	0.095516
		Gorgonian	91.0732	0.078132
		Hydrocoral	68.2431	0.011322
		Pennatulacean	1.5069	0.003484
		Porifera	3251.2513	0.243757
		Scleractinia	0.8151	0.001373
	Gulf of Alaska	Alcyonacea	0.1669	0.000617
		Antipatharia	0.0822	0.000458
		CoralsBryozoans	22.9991	0.010188
		Gorgonian	5.9575	0.005731
		Hydrocoral	1.0328	0.001154
		Pennatulacean	0.9107	0.004517
		Porifera	77.7599	0.041331
		Scleractinia	0.0222	0.000080
	Canada	Alcyonacea	3.5789	0.000737
		Antipatharia	0.0135	0.000082
		Calcarea	1.0895	0.000333
		Demospongiae	1.1641	0.001833
		Hexactinellida	7.6375	0.003096
		Gorgonian	9.1735	0.003318
		Hydrocoral	0.0153	0.000088
		Pennatulacean	4.7493	0.006613
		Porifera	51.9082	0.010335
		Scleractinia	0.6733	0.000525
	West Coast	Alcyonacea	6.7832	0.003231
		Antipatharia	0.9594	0.003191

		Gorgonian	0.1385	0.002405
		Hydrocoral	0.0186	0.000053
		Pennatulacean	1.7874	0.020504
		Porifera	67.4655	0.026327
		Scleractinia	0.5967	0.000540
Longline	Bering Sea	Alcyonacea	0.8002	0.002102
-	-	Antipatharia	0.0116	0.000013
		CoralsBryozoans	36.3506	0.021688
		Gorgonian	2.5242	0.002083
		Hydrocoral	0.0249	0.000234
		Pennatulacean	360.5305	0.054751
		Porifera	58.4796	0.033740
		Scleractinia	0.0047	0.000006
	Aleutian Islands	Alcyonacea	0.0235	0.001683
		Antipatharia	1.9492	0.005778
		CoralsBryozoans	55.7987	0.251977
		Gorgonian	11.0263	0.044539
		Hydrocoral	0.6854	0.034723
		Pennatulacean	2.6882	0.016267
		Porifera	85.6134	0.235486
		Scleractinia	0.0001	0.000112
	Gulf of Alaska	Alcyonacea	0.0982	0.001381
		Antipatharia	0.0327	0.000780
		CoralsBryozoans	8.1752	0.024687
		Gorgonian	0.6547	0.004144
		Hydrocoral	0.1555	0.001515
		Pennatulacean	9.8567	0.027272
		Porifera	10.2809	0.025133
		Scleractinia	0.0024	0.000668
	Canada	Alcyonacea	0.0872	0.000206
		Calcarea	0.0024	0.000011
		Demospongiae	0.0068	0.000026
		Hexactinellida	0.0106	0.000018
		Gorgonian	0.1008	0.000299
		Pennatulacean	0.0317	0.000125
		Porifera	0.1077	0.000472
		Scleractinia	0.2088	0.000494
	West Coast	Alcyonacea	0.0321	0.003982
		Antipatharia	0.0008	0.000319
		Gorgonian	0.0588	0.005734
		Hydrocoral	0.0016	0.000199
		Pennatulacean	0.0503	0.008362
		Porifera	0.9585	0.019591
		Scleractinia	0.0022	0.000358
Pot	Bering Sea	Alcyonacea	0.0015	0.000092

	Antipatharia	0.0001	0.000046
	CoralsBryozoans	0.1304	0.005386
	Gorgonian	0.0045	0.000368
	Hydrocoral	0.0002	0.000046
	Pennatulacean	0.0019	0.000598
	Porifera	1.2600	0.021543
Aleutian Islands	Antipatharia	0.0009	0.000268
	CoralsBryozoans	0.1831	0.026831
	Gorgonian	0.0186	0.006171
	Hydrocoral	0.0114	0.006171
	Porifera	1.0314	0.055541
Gulf of Alaska	Alcyonacea	0.0000	0.000100
	Antipatharia	0.0002	0.000100
	CoralsBryozoans	0.0345	0.003808
	Gorgonian	0.0001	0.000301
	Hydrocoral	0.0000	0.000100
	Pennatulacean	0.0188	0.002004
	Porifera	0.1342	0.009820
	Scleractinia	0.0001	0.000301
Canada	Alcyonacea	0.0085	0.000148
	Antipatharia	0.0009	0.000042
	Calcarea	0.0005	0.000021
	Hexactinellida	0.0014	0.000063
	Gorgonian	0.1709	0.002051
	Pennatulacean	0.0093	0.000254
	Porifera	0.0430	0.001015
	Scleractinia	0.0344	0.000677
West Coast	Alcyonacea	0.0137	0.003012
	Antipatharia	0.0084	0.002231
	Gorgonian	0.0292	0.005075
	Hydrocoral	0.0013	0.000279
	Pennatulacean	0.0336	0.020579
	Porifera	0.1355	0.008700
	Scleractinia	0.0025	0.000167

1027 Table S3. Threshold results from using fishery data only (no stereo-camera) to develop the

1028 Youden Index, minimum distance and segmented regression points. The bycatch weight threshold

1029 values in this table correspond to the points in Figure S2.

Method	Bycatch weight threshold	VME_taxa	Region	Gear
MinDist	72	Gorgonian	Aleutian_Islands	Bottom trawl
YoudenIndex	72	Gorgonian	Aleutian_Islands	Bottom trawl
Segmented	36	Gorgonian	Aleutian_Islands	Bottom trawl
MinDist	30	Gorgonian	Aleutian_Islands	Longline
YoudenIndex	26	Gorgonian	Aleutian_Islands	Longline
Segmented	15	Gorgonian	Aleutian_Islands	Longline
MinDist	38	Gorgonian	BC	Bottom trawl
YoudenIndex	38	Gorgonian	BC	Bottom trawl
Segmented	27	Gorgonian	BC	Bottom trawl
MinDist	2	Gorgonian	West Coast	Bottom trawl
YoudenIndex	2	Gorgonian	West Coast	Bottom trawl
Segmented		Gorgonian	West Coast	Bottom trawl
MinDist	151	Alcyonacea	Aleutian_Islands	Bottom trawl
YoudenIndex	234	Alcyonacea	Aleutian_Islands	Bottom trawl

Method	Bycatch weight threshold	VME_taxa	Region	Gear
Segmented	30	Alcyonacea	Aleutian_Islands	Bottom trawl
MinDist	29	Alcyonacea	West Coast	Bottom trawl
YoudenIndex	29	Alcyonacea	West Coast	Bottom trawl
Segmented	19	Alcyonacea	West Coast	Bottom trawl
MinDist	12	Antipatharia	Aleutian_Islands	Bottom trawl
YoudenIndex	12	Antipatharia	Aleutian_Islands	Bottom trawl
Segmented	8	Antipatharia	Aleutian_Islands	Bottom trawl
MinDist	8	Antipatharia	West Coast	Bottom trawl
YoudenIndex	8	Antipatharia	West Coast	Bottom trawl
Segmented	1	Antipatharia	West Coast	Bottom trawl
MinDist	5	Demospongiae	BC	Bottom trawl
YoudenIndex	5	Demospongiae	BC	Bottom trawl
Segmented		Demospongiae	BC	Bottom trawl
MinDist	23	Hexactinellida	BC	Bottom trawl
YoudenIndex	23	Hexactinellida	BC	Bottom trawl
Segmented	6	Hexactinellida	BC	Bottom trawl

Method	Bycatch weight threshold	VME_taxa	Region	Gear
MinDist	35	Pennatulacean	Bering_Sea	Bottom trawl
YoudenIndex	35	Pennatulacean	Bering_Sea	Bottom trawl
Segmented	15	Pennatulacean	Bering_Sea	Bottom trawl
MinDist	137	Pennatulacean	Bering_Sea	Longline
YoudenIndex	150	Pennatulacean	Bering_Sea	Longline
Segmented	96	Pennatulacean	Bering_Sea	Longline
MinDist	10	Pennatulacean	BC	Bottom trawl
YoudenIndex	10	Pennatulacean	BC	Bottom trawl
Segmented		Pennatulacean	BC	Bottom trawl
MinDist	3	Pennatulacean	West Coast	Bottom trawl
YoudenIndex	4	Pennatulacean	West Coast	Bottom trawl
Segmented	1	Pennatulacean	West Coast	Bottom trawl
MinDist	1,015	Porifera	Aleutian_Islands	Bottom trawl
YoudenIndex	1,136	Porifera	Aleutian_Islands	Bottom trawl
Segmented	423	Porifera	Aleutian_Islands	Bottom trawl
MinDist	105	Porifera	Aleutian_Islands	Longline
Method	Bycatch weight threshold	VME_taxa	Region	Gear
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YoudenIndex	126	Porifera	Aleutian_Islands	Longline
Segmented	46	Porifera	Aleutian_Islands	Longline
MinDist	1,002	Porifera	Bering_Sea	Bottom trawl
YoudenIndex	927	Porifera	Bering_Sea	Bottom trawl
Segmented	224	Porifera	Bering_Sea	Bottom trawl
MinDist	57	Porifera	Bering_Sea	Longline
YoudenIndex	57	Porifera	Bering_Sea	Longline
Segmented	25	Porifera	Bering_Sea	Longline
MinDist	8	Porifera	Bering_Sea	Pot
YoudenIndex	8	Porifera	Bering_Sea	Pot
Segmented	7	Porifera	Bering_Sea	Pot
MinDist	156	Porifera	Gulf_Of_Alaska	Bottom trawl
YoudenIndex	176	Porifera	Gulf_Of_Alaska	Bottom trawl
Segmented	64	Porifera	Gulf_Of_Alaska	Bottom trawl
MinDist	32	Porifera	Gulf_Of_Alaska	Longline
YoudenIndex	32	Porifera	Gulf_Of_Alaska	Longline

Method	Bycatch weight threshold	VME_taxa	Region	Gear
Segmented	17	Porifera	Gulf_Of_Alaska	Longline
MinDist	74	Porifera	BC	Bottom trawl
YoudenIndex	46	Porifera	BC	Bottom trawl
Segmented	13	Porifera	BC	Bottom trawl
MinDist	70	Porifera	West Coast	Bottom trawl
YoudenIndex	80	Porifera	West Coast	Bottom trawl
Segmented	39	Porifera	West Coast	Bottom trawl
MinDist	4	Porifera	West Coast	Longline
YoudenIndex	4	Porifera	West Coast	Longline
Segmented	4	Porifera	West Coast	Longline

1033 Table S2. Predicted thresholds encounter weights (and confidence intervals) for VME indicator taxa in the regions of Alaska by gear

1034 type using percentile regression method. Threshold weights use the average regression parameters from relationships in Figure 8 and

1035 a density of $log(0.5 individuals * m^{-2})$ (x axis) to predict the associated by catch weight (y axis), which is then exponentiated to

1036 *calculate a bycatch weight for the threshold.*

Region	Gear	Antipatharia	Gorgonian	Hydrocoral	Pennatulacean	Porifera
Aleutian_Islands	Bottom trawl	64.75 (40.86 - 102.62)	17.83 (13.5 - 23.55)	11.29 (8.38 - 15.21)	44.57 (32.15 - 61.78)	121.58 (92.15 - 160.42)
Aleutian_Islands	Longline	23.62 (14.88 - 37.51)	6.5 (4.91 - 8.61)	4.12 (3.05 - 5.57)	16.26 (11.71 - 22.58)	44.36 (33.54 - 58.66)
Aleutian_Islands	Pot	2.18 (1.37 - 3.47)	0.6 (0.45 - 0.8)	0.38 (0.28 - 0.53)		4.09 (3.06 - 5.47)
Bering_Sea	Bottom trawl		19.15 (13.75 - 26.67)		47.86 (34.07 - 67.24)	130.57 (94.13 - 181.11)
Bering_Sea	Longline		13.94 (9.97 - 19.48)		34.84 (24.72 - 49.1)	95.04 (68.33 - 132.19)
Bering_Sea	Pot		0.36 (0.26 - 0.51)		0.91 (0.64 - 1.29)	2.48 (1.78 - 3.46)
Gulf_Of_Alaska	Bottom trawl	41.82 (25.95 - 67.4)	11.52 (8.66 - 15.31)	7.29 (5.46 - 9.74)	28.79 (20.89 - 39.66)	78.53 (59.06 - 104.41)
Gulf_Of_Alaska	Longline	10.36 (6.42 - 16.71)	2.85 (2.14 - 3.8)	1.81 (1.35 - 2.41)	7.13 (5.16 - 9.85)	19.44 (14.58 - 25.92)
Gulf_Of_Alaska	Pot	0.7 (0.43 - 1.16)	0.19 (0.14 - 0.26)		0.48 (0.34 - 0.68)	1.32 (0.97 - 1.81)

1037