

1 Evaluating effort regulation in mixed fisheries: a
2 Monte Carlo approach

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Abstract

This paper evaluates whether effort regulation could achieve the goal of protecting low-abundance species in mixed fisheries. We construct a two-species bio-economic model and compare the stock abundance ratio in the end of the fishing season with the ratio prior to the fishing. Fishers' profit maximization problem is governed by three key factors: (a) the overall efficiency of catching different species (catchability), (b) the price of different species, and (c) their ability to catch the favoured species separately from the less-favoured species (separability). Using a Monte Carlo sampling of feasible parameters space, we show that effort regulation has good chances (87% of the cases) of maintaining the end stock ratio near equal levels ($\frac{1}{2} < \text{stock ratio} < 2$) when the initial stock ratio is equal. If the initial stock ratio is not equal, however, there is a high risk (about 50% of the cases) that effort control increases differences in the relative species abundances, rather than diminishing them. The effects depend on whether the key factors determining fishing profitability are counteracting or reinforcing each other, and their relative strength. Our results warn against placing too much faith on the ability of effort regulation to protect species at low abundances from excessive exploitation.

KEYWORDS: mixed fisheries, effort control, bioeconomics modeling, Monte Carlo approach, fisheries management

JEL CODES: Q22, Q57

1 Introduction

Total allowable catch (TAC) regulation, a form of output control, is a single-species management approach that sets stock-specific catch quotas (Stefansson and Rosenberg, 2005). In a mixed fishery where many species are caught simultaneously, management relying on stock-specific TACs can be impractical because of data requirements and costs, and be-

38 cause of discarding of over-quota catches (Daan, 1997; Hilborn et al., 2004; Baudron et al.,
39 2010), or when discarding is not allowed, because species with restrictive quotas ('choke
40 species') prevent full utilization of species with more permissive quotas (Kuriyama et al.,
41 2016; Alzorriz et al., 2018; Mortensen et al., 2018). To address these problems, input con-
42 trol regimes, which are often species-unspecific and account for the multitude of species
43 living in an ecosystem, are sometimes favoured for the management of mixed fisheries
44 (Pope, 2002; Squires et al., 2017). Input control regimes set quotas at the operational unit
45 level; for example, quotas are given as total allowable fishing days per fleet category, fur-
46 ther split among the individual license holders (Laurec et al., 1991; Andersen et al., 2010;
47 Danielsen and Agnarsson, 2018).

48 Effort regulation has been practised in many fisheries (Squires et al., 2017), including
49 some Mediterranean fisheries (Vielmini et al., 2017; Mulazzani et al., 2018), the sole and
50 plaice fisheries in the North Sea (European Union Committee, 2008), and the demersal
51 fisheries in Faroese waters (Jákupsstovu et al., 2007; Danielsen and Agnarsson, 2018).
52 Moreover, license control, a dominant fisheries management strategy in the Global South,
53 represents a special form of effort regulation. In these fisheries, a total number of licenses
54 is specified, often in combination with other measures such as seasonal fishing bans (FAO,
55 2009; Shen and Heino, 2014; Tromeur and Doyen, 2018).

56 The fishing days system in the Faroe Islands is one of the most studied examples of
57 effort regulation (Jákupsstovu et al., 2007; Baudron et al., 2010; Danielsen and Agnarsson,
58 2018). The system was introduced in 1996 to manage the demersal fisheries with cod,
59 haddock and saithe as their main targets. While sometimes hailed as highly successful, a
60 10-years appraisal study by Jákupsstovu et al. (2007) showed that Faroese Total Allowable
61 Effort (TAE) system did not achieve some of its key objectives, namely controlling the
62 fishing mortality of the key species. This conclusion has been upheld by later assessments
63 (Danielsen and Agnarsson, 2018). Most fishers opportunistically targeted the most valu-
64 able species, cod, leading to high levels of mortality for this valuable stock even when less
65 abundant. The design of TAE system in the Faroe Islands relied on the assumption that

66 when fishers choose their target species to maximize their profit, they target the species
67 that provide the highest catch rates and ‘automatically’ protect the less abundant species
68 with lower catch rates (Jákupsstovu et al., 2007; Baudron et al., 2010; Danielsen and Ag-
69 narsson, 2018). However, this is an assumption rather than a fundamental property of
70 effort-controlled fisheries.

71 Previous studies on TAE regulation have often been conducted in a single species set-
72 ting and in comparison with TACs; for example, fishing vessel behaviour under TAE and
73 TAC regulation (Anderson, 1999; Stefansson and Rosenberg, 2005), and the effect of un-
74 certainty on the efficiency of catch or effort controls (Danielsson, 2002; Yamazaki et al.,
75 2009). Another strand of literature examined the degree of input substitution between re-
76 stricted inputs and unrestricted inputs in a TAE system. Examples include the substitution
77 intensity of restrictions on fishing days versus restrictions on vessel tonnage of the British
78 Columbia commercial salmon fishery (Dupont, 1991), and the substitution between physi-
79 cal inputs and fishing location of the UK beam trawl fishery (Pascoe and Robinson, 1998).
80 Until now, effort regulation has been assessed in a context of specific fishery systems. How-
81 ever, case-specific detail may hinder identifying the key factors that determine the success
82 (or failure) of effort regulation in mixed fisheries.

83 The key factors affecting fishers’ targeting decision in mixed-fisheries have previously
84 been studied piece by piece; for example, Katsukawa and Matsuda (2003) focused on the
85 effect of non-linear catchability, Noailly et al. (2003) on the effect of price on switching
86 harvest strategies, and Tromeur and Doyen (2018) on the effect of technical interactions.
87 As a result, the conclusions of these papers may differ depending on specific underlying as-
88 sumptions. When perfect separability is assumed, target switching may be able to protect
89 the less abundant species from being overfished (Katsukawa and Matsuda, 2003; Bisci
90 et al., 2013a,b); when joint production is unavoidable, species with lower price and growth
91 but higher intraspecific competition and catchability are more prone to overfish, as con-
92 cluded by Tromeur and Doyen (2018). We attempt to study all these key aspects together
93 to give a more holistic picture about their interplay. We are specifically interested in effort

94 regulation of mixed fisheries, but our results can be interpreted more generally as repre-
95 senting profit-oriented, effort-limited exploitation that could occur in unregulated fisheries
96 or in a single-owner case.

97 The purpose of this paper is to present a generic analysis of factors determining the
98 practicality of effort regulation using a simplified two-species fishery system. We gener-
99 alize earlier models by treating separability and stock elasticity of harvest as parameters;
100 the former describes fishers' ability to target and catch a specific species and the latter the
101 degree of schooling behaviour of the fish. Our aim is to show when the effort regulation
102 approach can work in terms of achieving the biological goal of not exhausting any of the
103 species, or when it may not work, and why. The rest of the paper is organized as fol-
104 lows. The first section illustrates key factors governing fisherman's targeting decision and
105 detailed model specifications. In the subsequent sections, we first present an analytical opti-
106 mal harvest rule based on a common assumption about catchability parameter, followed by
107 an evaluation of effort regulation based on a generic set-up using Monte Carlo simulations.

108 **2 Methods**

109 **Identifying key factors**

110 We start by identifying three key factors that describe the dynamics of a mixed fishery:
111 the relative abundance and catchability of fish stocks, which determine the potential catch
112 rates; the relative price for different fish species, which relates to the revenues obtained;
113 and the ability of fishermen to catch the favoured species separately from the unfavoured
114 species. These three factors form the cornerstones of effort regulation (Fig. 1).

Catchability is a species-specific parameter that describes the efficiency of fishing op-
eration in catching a certain type of fish, therefore depending on both characteristics of the
target (e.g., fish behaviour) and the fishing operation (e.g., efficiency of the fishing gears).
Typically, catchability q of species i is defined through a linear relationship between catch
rate (C_i), effort (E), and stock abundance (N_i), $C_i = q_i E N_i$. However, this simple model is

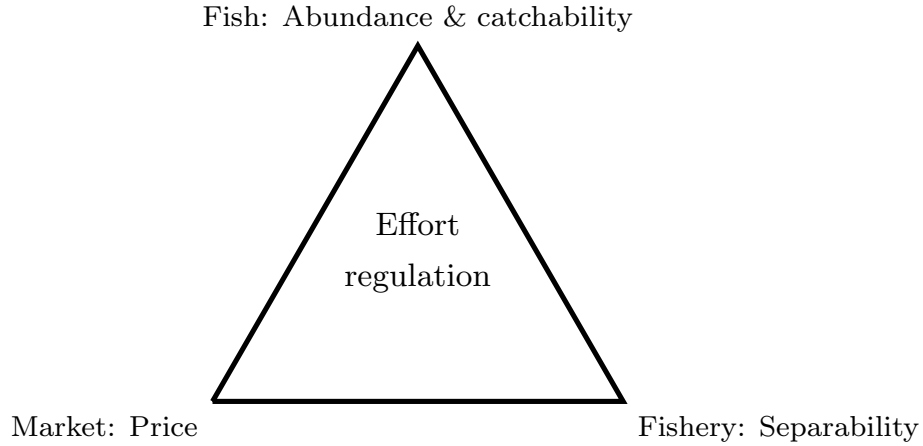


Figure 1: Key factors determining the practicality of effort regulation

often acknowledged to be inappropriate (Winters and Wheeler, 1985; Hilborn and Walters, 1992; Harley et al., 2001); for instance, a high catch per unit effort (C/E) for schooling species can still be maintained even at a low level of stock abundance. We thus follow the alternative formulation proposed by Steinshamn (2011) and Liu and Heino (2013) where catchability is measured in terms of the local stock density ρ experienced by a fisher:

$$C_i = q_i E \rho_i = \tilde{q}_i E N_i^{b_i} = a_i q_i N_i^{b_i - 1} E N_i, \text{ with } \rho_i = a_i N_i^{b_i}. \quad (1)$$

115 Thus \tilde{q}_i in Eq. 1, termed local catchability, is analogous with q_i in the classic formulation,
 116 but it has absorbed the scaling parameter a_i and is measured relative to local density ρ_i . N_i
 117 is stock abundance normalized against carrying capacity K_i . Parameter b_i is stock elasticity
 118 of harvest, with a typical value between $0 \leq b_i \leq 1$. The limit case ($b_i = 0$) refers the
 119 species that is extremely schooling and the fishers can locate the schools perfectly. $b_i = 1$
 120 represents a non-schooling species that is uniformly distributed.

The influence of the market is captured by the relative price of the two species. We consider two ways of price determination: i) exogenous price, in which price is constant,

and ii) endogenous price, in which price is a decreasing function of the catch. Specifically,

$$p_i(C_i) = \max(0, p_{i,0}(1 - \omega_i(C_i - C_{i,0}))), \quad (2)$$

121 where $p_{i,0}$ is the initial unit price and $p_{i,0}\omega_i$ is the slope of the price function. When $\omega_i = 0$,
122 $p_i = p_{i,0}$ and the price is constant (exogenous price); when $\omega > 0$, price is endogenous,
123 with higher total catches resulting in reduced unit price. The price $p_{i,0}$ is obtained when
124 $C_i = C_{i,0}$, where $C_{i,0}$ is the optimal catch on the first fishing day when $\omega_i = 0$. That is to
125 say, we express the price relative to the price obtained during the first fishing trip.

126 The final element in our model is ‘separability’ (τ). This is a fishery-level parameter
127 and refers to fishers’ ability to catch the favoured species, or in economics jargon, to the
128 degree of joint production. It depends on biological factors (e.g., intrinsic differences in
129 behaviour and micro-habitat use of the alternative species), fishers’ skill of using these
130 differences in fishing, and available fishing technologies that fishers can utilize (Branch
131 and Hilborn, 2008; Squires et al., 2017). We take $\tau = 1$ to mean perfect separability where
132 only the targeted species is caught, and $\tau = 0$ complete inability to catch species separately
133 such that the total effort is equally shared among the two harvested species. For every unit
134 of fishing effort made by a fisherman (e_i), only fraction $\frac{1+\tau}{2}$ is effectively converted into
135 catching the target species (i), whereas fraction $\frac{1-\tau}{2}$ ‘leaks’ to the other species, leading
136 to its bycatch. Thus, we differentiate the nominal effort targeted on species i , e_i , and the
137 effective effort E_i , where $E_i = \frac{1+\tau}{2}e_i + \frac{1-\tau}{2}e_{-i}$ and e_{-i} denotes effort targeting the other
138 species.

139 **Biological model and assessment criteria**

140 We model a simple fishery system consisting of two species that share the same overall
141 habitat but are otherwise independent. Discarding is assumed not to occur. Our model
142 focuses on stock dynamics in a single fishing season, which comprises a number of fishing
143 trips of equal length; the length is inconsequential for our analysis and we will assume

144 daily trips for convenience. Recruitment is assumed to occur outside the fishing season and
 145 can thus be ignored.

146 If stock abundance in the beginning of a fishing day/trip is N_0 , it will drop to N_t in
 147 the end of that fishing trip due to natural and fishing mortality. The stock dynamics ¹ is
 148 specified in the following:

$$N_t = \begin{cases} N_0 e^{-(F_t+M)t} = N_0 e^{-(\tilde{q}E_t+M)t} & \text{if } b = 1, \\ \left\{ \left[N_0^{1-b} + \frac{\tilde{q}E_t}{M} \right] e^{-M(1-b)t} - \frac{\tilde{q}E_t}{M} \right\}^{\frac{1}{1-b}} & \text{if } 0 \leq b < 1. \end{cases} \quad (3)$$

149 where F and M are respectively fishing and natural mortality. Other notations follow Eq. 1.

150 The key question is whether profit-oriented fishing can lead to protection of the less
 151 abundant species? We use the stock size ratio as the metric to assess this question. As long
 152 as natural mortality during the fishing season is negligible, changes in this metric are driven
 153 by harvesting. If the stock ratio in the end of the fishing season $r_T = N_1^T/N_2^T$ is closer to
 154 unity than the initial stock ratio $r_0 = N_1^0/N_2^0$, then the originally less abundant species must
 155 have suffered less from fishing than the other species. In other words, it has been offered a
 156 degree of protection. This is a necessary condition for effort regulation to have the potential
 157 to be successful. We emphasize, however, that this criterion is not a sufficient condition to
 158 ensure sustainable fisheries, because it does not address simultaneous stock depletion. This
 159 could be prevented by setting the total effort quota sufficiently low; the determination of
 160 suitable total effort is beyond the scope of our current analysis.

161 We consider two basic scenarios in our analysis: (a) ‘even’ scenario in which two
 162 species in the mixed fishery have the same initial stock (i.e., $r_0 = N_1^0/N_2^0 = 1$); (b) ‘bi-
 163 ased’ scenario in which their initial stock levels are different (here $r_0 = 2$). In the former
 164 case, the success criterion is that the final stock size ratio r_T does not deviate too much
 165 from unity, and in the latter case, that the final stock size ratio r_T has moved closer to unity.

¹The expression for N_t is derived via integrating over a full fishing trip $\frac{dN}{dt} = -(F+M)N$, and $F = \tilde{q}N^{b-1}E$. For detailed derivation see Steinshamn (2011); Liu and Heino (2013)

166 **Profit maximizing fleet**

167 The goal of the fleet manager is to maximize her payoffs per fishing trip for the entire
 168 fleet. We define effort in terms of fishing days. The total allowable fishing days per fishing
 169 season allocated to a fleet is set at Q . We assume the fleet consists of n identical boats. The
 170 strategy option for the manager is to decide the number of boats to target different species
 171 $i \in (1, 2)$. The maximization will be repeated for each fishing trip until Q is exhausted or
 172 profit turns negative. We formulate trip-level profit maximization as follows:

$$\begin{aligned}
 V_t &= \max_{e_{i,t}} \sum_n \sum_i \left(\int_0^{C(E_{i,t})} p_i(C) dC - ce_{i,t} \right), i \in (1, 2) & (4) \\
 &\quad \sum_i e_{i,t} \leq n \text{ (Fleet capacity)} \\
 &\quad \sum_t \sum_i e_{i,t} \leq Q \text{ (Effort control)} \\
 &\quad s.t. \quad E_{i,t} = \frac{1+\tau}{2} e_{i,t} + \frac{1-\tau}{2} e_{-i,t} \text{ (Separability)} \\
 &\quad V_t \geq 0 \text{ (Non-negativity)}
 \end{aligned}$$

173 where price of fish p_i is determined by the price function in the Eq. 2; c is cost per unit
 174 effort, e and E denote respectively nominal effort and effective effort. Provided that the
 175 species are not fully separable, Eq. 4 gives two sources of income streams from each
 176 species. C denotes total catch per trip, given by the function by Liu and Heino (2013),
 177 based on the catchability definition in Eq. 1:

$$\begin{aligned}
 C_t &= \tilde{q}E_t \int_0^1 N_t^b dt \\
 &= \begin{cases} \frac{F_t}{F_t+M} N_0 (1 - e^{-(F_t+M)}) = \frac{\tilde{q}E_t}{\tilde{q}E_t+M} N_0 (1 - e^{-(\tilde{q}E_t+M)t}) & \text{if } b = 1, \\ \tilde{q}E_t \int_0^1 \left\{ \left[N_0^{1-b} + \frac{\tilde{q}E_t}{M} \right] e^{-M(1-b)t} - \frac{\tilde{q}E_t}{M} \right\}^{\frac{b}{1-b}} dt & \text{if } 0 \leq b < 1. \end{cases} & (5)
 \end{aligned}$$

178 The maximization problem in Eq. 4 is subject to several constraints: (1) fleet capacity
 179 constraint: total fishing boats per trip shall not exceed the fleet capacity n , assuming trip
 180 length of 1 day/trip; (2) effort constraint: total fishing days per season is no more than Q ;
 181 (3) species separability constraint: only part of the effort is converted into effective effort;

182 and (4) non-negative profit per fishing trip.

183 Because more boats targeting the same species result in faster depletion of the resource,
184 and may influence the price (if $\omega > 0$), the optimal decision of one fisherman depends on
185 what other fishermen decide to do. This can potentially lead to a game-like situation, but
186 this is avoided if we assume that the decision horizon is one fishing trip (e.g., 1 day) only,
187 and the problem reduces to a simple optimization task analogous to ‘perfect competition’
188 of identical competitors in economics, or ‘ideal free distribution’ in ecology (Fretwell and
189 Lucas, 1969). Thus, to understand the collective behaviour of individual fishermen, we can
190 simply find the distribution of effort that maximizes the collective pay-off per fishing trip.
191 The solutions of the model correspond to optimal policies in a single owner case too.

192 **Monte Carlo sampling method**

193 The model is not analytically tractable except for some special cases. To obtain an overview
194 on how key factors affect the performance of effort regulation, we turn to the Monte Carlo
195 method of repeated random sampling approach. Key factors and their assumptions are sum-
196 marized in Table 1 and are very general. We assume b and τ to follow uniform distribution
197 because they are naturally bounded between 0 and 1 (while b could exceed 1, $b = 1$ is a
198 defensible upper limit, see Steinshamn (2011)). Price and catchability ratios are assumed
199 to be log-normally distributed such that the mean ratio is one. The results are based on
200 50,000 random replicates.

201 **3 Optimal harvest decisions in a mixed fishery**

202 **A static setting**

203 For the case of uniformly distributed fish stock ($b = 1$), we can analytically show that
204 optimal harvest decision in our model is characterized by the relative marginal profit of
205 effort (MPE) of the two stocks, i.e., the difference in the marginal increase of profit when

Table 1: Key assumptions for Monte Carlo sampling. Actual levels of prices and costs are immaterial as long as fishing is profitable.

Parameter	Description	Value	Distribution
τ	separability	$0 \leq \tau \leq 1$	uniform
b	stock elasticity of harvest	$0 \leq b \leq 1$	uniform
$\log(\frac{p_1}{p_2})$	price ratio	$\mu = 0, sd = 0.5$	normal
$\log(\frac{q_1}{q_2})$	catchability ratio	$\mu = 0, sd = 0.5$	normal
ω_i	slope of the price function	0 or 50	n.a.
M_i	natural mortality	0.2 yr^{-1}	n.a.
Q	effort quota	1000 days	n.a.
n	fleet capacity	20	n.a.

Note: μ denotes mean and sd standard deviation.

206 effort targeting a specific species is increased. We can distinguish three cases:

- 207 • If $\text{MPE}_1 > \text{MPE}_2$, then only species 1 is harvested ($e_1 > 0$ and $e_2 = 0$).
- 208 • If $\text{MPE}_1 < \text{MPE}_2$, then only species 2 is harvested ($e_1 = 0$ and $e_2 > 0$).
- 209 • If $\text{MPE}_1 = \text{MPE}_2$, then both species are harvested ($e_1 > 0$ and $e_2 > 0$).

210 The expression for MPE in the general case is too complicated to yield insight (see the
 211 Appendix). However, if we assume that natural mortality is insignificant over short time
 212 periods and can be ignored ($M = 0$), $\text{MPE}_1 > \text{MPE}_2$ if $p_1 \tilde{q}_1 N_1 e^{-\tilde{q}_1 E_1} > p_2 \tilde{q}_2 N_2 e^{-\tilde{q}_2 E_2}$. High
 213 price p and initial stock N will therefore favour targeting one stock over the other. The
 214 role of catchability \tilde{q} is ambiguous because its effect can be either positive or negative,
 215 depending on the other parameters.

216 When fishers cannot discriminate the two species ($\tau = 0$), every unit of effort is equally
 217 divided between them, and any effort allocation satisfying the fleet capacity constraint will
 218 be optimal. This is a trivial case, and in the subsequent discussion we will focus on the
 219 case $\tau > 0$ only.

220 **Seasonal patterns of harvest decisions**

221 Because a fishing season consists of a number of fishing trips in our model, we will study
222 how MPE evolves over time. There are three generic patterns of time evolution in the
223 model:

- 224 • A single species is the sole target species during the whole fishing season. MPE of
225 one species is always greater than the other.
- 226 • A single species is the sole target species in the beginning of the fishing season. After
227 MPEs are equalized between the species, both species are targeted.
- 228 • A single species is the sole target species in the beginning of the fishing season.
229 MPEs switch the rank that leads to a target switch, i.e., the other species becomes the
230 sole target.

231 The first two cases follow naturally from the rules of optimal harvest described in the
232 previous section: the fishermen will target the more profitable stock, or if both stocks are
233 equally profitable, they will split their effort such that the equal profitability is maintained.
234 Initially, one stock is almost always more profitable. Whether targeting the more profitable
235 species leads to equal profitability during the period of interest depends on the details. First,
236 while targeting one species will often cause the MPEs to converge towards each other, this
237 is not always the case. MPEs may diverge if separability τ is low, or if one species has
238 higher natural mortality M than the other. Second, even when targeting one species is
239 causing the MPEs to converge, full equalization might not be reached during the available
240 fishing season.

241 Fig. 2a–c illustrates how targeting one species reduces its profitability, eventually re-
242 sulting in the situation in which both species have similar MPEs. In this example, species
243 1 has initially a higher MPE and is the sole target during the first part of the season. Over
244 time, MPE for the target species 1 decreases more than MPE for the bycatch species 2
245 because fishing reduces its abundance more than that of the bycatch species. When both

246 species reach the same MPE level, the effort (Fig. 2b) is divided between the two species
 247 such that the MPE equality is maintained. In this example, effort regulation first causes
 248 divergence in the relative stock abundances, followed by rebalancing and divergence to the
 249 other direction.

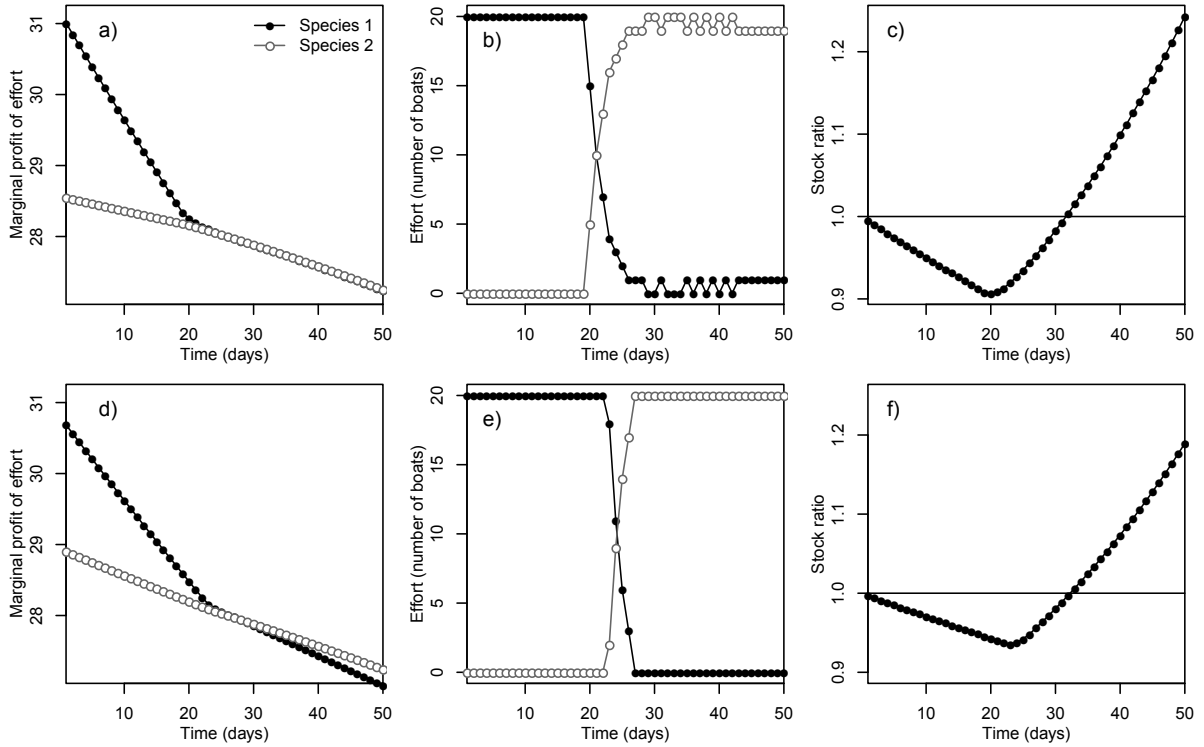


Figure 2: Two qualitatively different scenarios of marginal profit of effort and effort allocation over time, together with the corresponding relative stock size. On the top row (a–c), separability is high ($\tau = 0.8$). Two distinct phases are observed: first, only the species 1 that is more profitable is targeted, and second, once the profitability of species 1 is reduced to that of species 2, both species are targeted with an effort allocation that keeps their profitability similar. On the bottom row (d–f), separability is lower ($\tau = 0.57$). This leads to dynamics that are initially similar, but instead of profitability equalization, the initial bycatch species becomes the sole target. Parameters other than τ are equal for both scenarios: $N_0 = [0.5, 0.5]$, $b = [0.813, 0.086]$, $p = [53924, 100000]$, $\tilde{q} = [0.0003, 0.0003]$, $\omega = [0, 0]$, $c = 0.1$, $Q = 1000$.

250 The dynamic in the third case where the MPE curves of two species cross each other
 251 and the fishermen switch their target species is more intricate. Fig. 2d–f illustrates how
 252 such target switching can result from effort ‘spill-over’. Fishermen first target the more
 253 profitable species 1. As its profitability declines, both species 1 and 2 become targeted. At
 254 this point, it is possible that even when species 2 is the sole target, species 1 is caught so

255 much as bycatch that its MPE stays lower than that of the target species. We will elaborate
256 further on this point in the Section 4 when discussing reinforcing effect.

257 **4 Assessing performance of effort regulation through Monte** 258 **Carlo simulations**

259 **The overall performance of effort regulation**

260 We assess performance of effort regulation by simulating a large number (50,000) of cases
261 generated by sampling a realistic parameter space (Table 1). When the fisheries are initiated
262 with two species at equal abundance, effort regulation can often keep the relative stock
263 levels within ‘reasonable’ bounds. Specifically, if we require that the stock ratio N_1^T/N_2^T at
264 the end of fishing season remains within the interval $[0.5, 2]$ (i.e., the less abundant species
265 has density at least 50% of that of the more abundant species), effort regulation is successful
266 in 87% of the cases (Fig. 3a). However, if the fisheries are initiated with one species
267 being twice as abundant as the other one, the chances of effort regulation meeting the
268 same success criterion are considerably lower, about 51% (Fig. 3b). In these cases, effort
269 regulation has either maintained the original biased stock ratio, or ‘corrected’ it towards
270 equality. However, there is also a sizeable proportion of cases – 40% – where the relative
271 stock abundance becomes driven towards more extreme bias in favour of the originally
272 more abundant species. In the remaining fraction of the cases, the originally less abundant
273 species becomes more abundant as a result of fishing.

274 **Single-factor effects**

275 We now investigate the characteristics of the cases where effort regulation can be success-
276 ful. Figures 4 and 5 show the density distributions of the final stock ratio against key
277 parameters that we varied. They reveal two main patterns. First, the fishers’ ability to
278 catch target species, separability τ (the only stock-unspecific parameter we varied), has

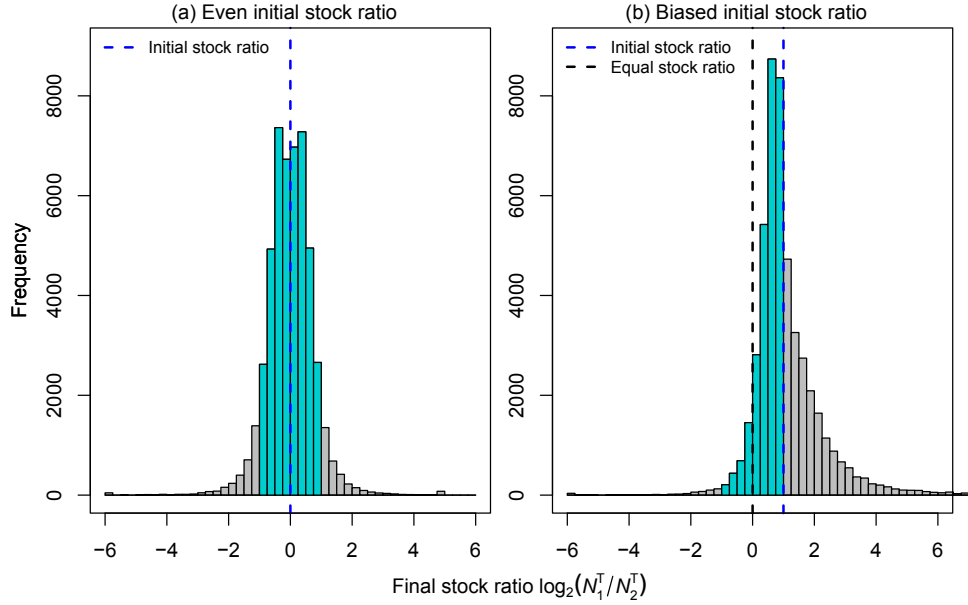


Figure 3: Histogram of stock ratio in the end of the fishing season for (a) even ($N_1^0 = N_2^0$) and (b) biased initial stock ratio ($N_1^0/N_2^0 = 2$). Cyan bands indicate the range where end stock ratio is seen as ‘reasonable’, defined as $N_1^T/N_2^T \in [0.5, 2]$. Price is constant ($\omega = [0, 0]$); other parameters as detailed in Table 1.

279 an important role. When separability is low ($\tau \rightarrow 0$), stock ratio at the end of the season
 280 usually stays near its initial value, be it equal or biased (Fig. 4a and 5a). However, when
 281 separability is high ($\tau \rightarrow 1$), the stock ratio diverges away from its initial value, in one
 282 direction or the other. If the initial stock abundances are even, then a bimodal distribution
 283 emerges (Fig. 4a). If the initial relative stock abundances are biased, then the final stock
 284 ratio shows a unimodal but skewed distribution (Fig. 5a). Across all τ , there is a tendency
 285 for the stock ratio to move closer to unity; the average final stock ratio is $N_1^T/N_2^T \approx 1.15$,
 286 although there is a long tail of cases towards more extreme bias (Fig. 5a). Nevertheless, in
 287 an average sense, effort regulation offers a degree of protection for the less abundant stock
 288 when starting from unequal initial abundances.

289 Second, the parameters that are stock-specific show similar patterns. If the relative
 290 stock abundances are initially even, the abundances at the end of season have best chances
 291 of maintaining the status quo when the species are similar in terms of their schooling pa-
 292 rameter b , catchability \tilde{q} , and price p (Fig. 4b–d). On the other hand, if the initial relative

293 stock abundances are biased, then the similarity of the species, on average, tends to hinder
 294 equalization of relative abundances. However, if the species differences are such that the
 295 initially less abundant species is a more favourable target (higher price p or catchability \tilde{q} ,
 296 or lower stock elasticity to harvest b), then equalization is more likely to happen (Fig. 5b–
 297 d). Differences in the opposite directions will consequently make equalization less likely.

298 Most Monte Carlo replicates in Figures 4 and 5 show targeting of the same species
 299 through the fishing season. This happens because MPEs of the two stock converge so
 300 slowly that they do not meet during the available time, or because of the bycatch effect that
 301 may even make them diverge. Equalization of MPEs followed by targeting of both species
 302 for the rest of the season (similar to Fig. 2a–c) happens only in about 15% of replicates.
 303 Target swapping (Fig. 2d–f) is even rarer, occurring in about 4% of replicates for the even
 304 initial scenario and less than 1% of replicates for the biased initial scenario.

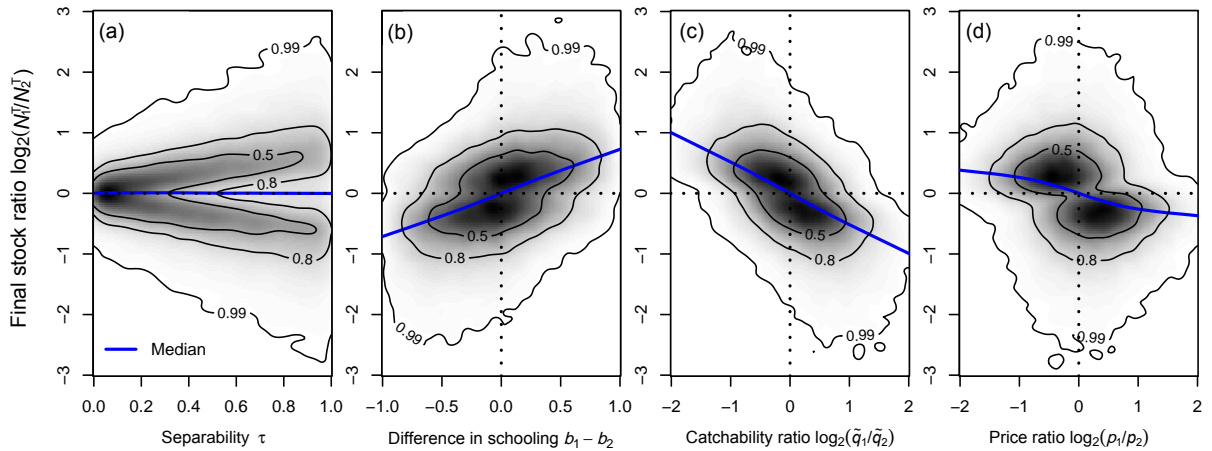


Figure 4: Density plots for the ‘even’ scenario (initial stock ratio $N_1^0/N_2^0 = 1$): black curves are contour lines, blue curves indicate the smoothed median density, and black dotted lines are reference level when the two species are symmetric. Price is declining with increasing catches ($\omega = [50, 50]$); other parameters as detailed in Table 1.

305 Interaction effects

306 When the species differ in more than one parameter, these differences could either reinforce
 307 or compensate for each other. For example, if one species has a higher unit price but

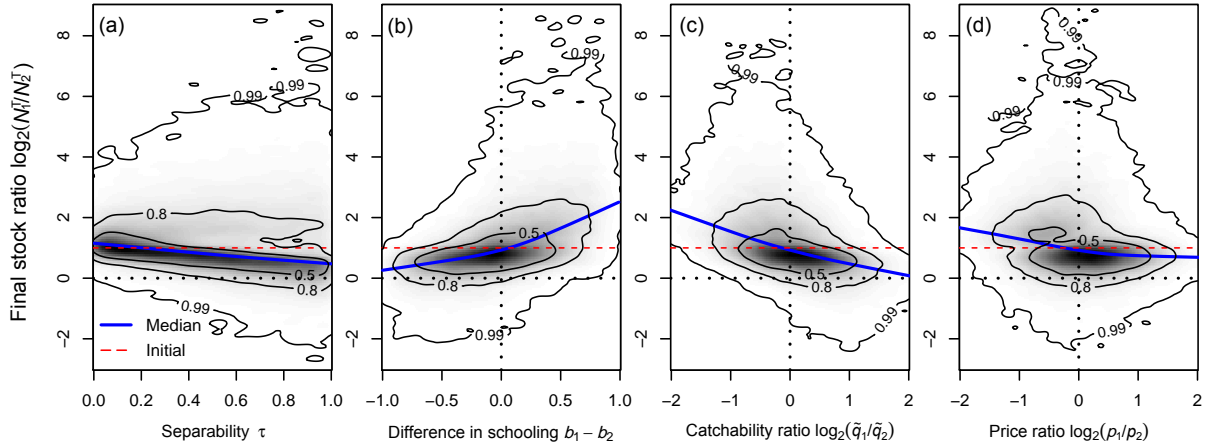


Figure 5: Density plots for the ‘biased’ scenario (initial stock ratio $N_1^0/N_2^0 = 2$ indicated by red dotted lines): black curves are contour lines, blue curves indicate the smoothed median density, and black dotted lines are reference level when the two species are symmetric. Price is constant ($\omega = [0, 0]$); other parameters as detailed in Table 1.

308 lower catchability, it can be an equally attractive target to a cheaper species with higher
 309 catchability. This ‘compensation effect’ is seen in Figure 6 as contour lines that are tilted
 310 relative to the axes. In particular, a balanced final stock ratio can be achieved not only
 311 when the two species are similar, but also when they differ such that an attractive attribute
 312 is compensated by a less attractive one.

313 Conversely, when the species differ in multiple ways that work in the same direction,
 314 we observe a ‘reinforcing effect’. This corresponds to the movement perpendicularly to
 315 the contour lines in Figure 6. In this case, even relatively small differences in the species-
 316 specific parameters may lead to large differences in the final stock ratio.

317 5 Discussion

318 We have addressed two questions that are pertinent to any effort control system: whether
 319 species that are originally balanced (i.e., similar in stock size) can be maintained at a rel-
 320 atively balanced level as a result of profit-oriented fishing, and secondly, whether profit-
 321 oriented fishing could conserve species that are at low levels, such that relative stock levels

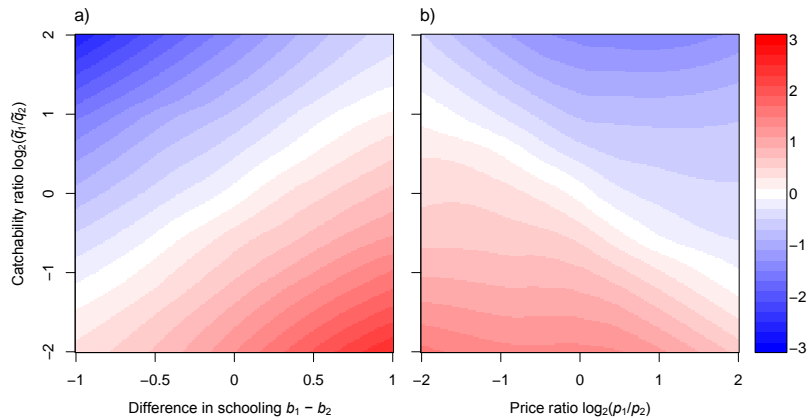


Figure 6: Parameter interactions under the even initial stock scenario for ratio of catchabilities (\tilde{q}) versus difference in the stock elasticity of (a) harvest (schooling, b) and (b) ratio of prices (p). Colour gradient shows the end stock ratio $\log_2(N_1^T/N_2^T)$. High stock differences are found when the effects of two factors are reinforcing each other. Similar stock levels appear around the mid-range area shown in white. Price is constant ($\omega = [0, 0]$); other parameters as detailed in Table 1.

322 would be more balanced after the fishing season—a necessary, but not sufficient, condition
 323 for a sustainable use of an ecosystem. We have addressed these questions using a generic
 324 two-species dynamic bio-economic model, assuming fishers that are omniscient and profit-
 325 maximizing. Our main finding is that effort regulation is prone to exaggerating the stock
 326 abundance differences, particularly when the fishers are effective in selectively catching the
 327 more profitable species. Generally speaking, it is hard to achieve balanced relative stock
 328 levels. However, effort regulation may achieve its biological conservation goal under two
 329 general conditions:

- 330 1. The species are sufficiently similar with respect to the key factors that determine the
 331 profitability of their harvest.
- 332 2. When the key parameters counteract each other such that the resulting overall prof-
 333 itability is similar.

334 Concerning the latter case, a low catchability can be compensated by a higher price, and
 335 vice versa. On the contrary, if the key parameters reinforce each other, instead of counter-
 336 acting, effort regulation can lead to increased differences in the relative stock levels, e.g.

337 fishers keep harvesting the less abundant species. The degree to which one stock gets de-
338pleted depends on the fleet's capacity to deplete the stock (the total allowable effort Q and
339catchability \tilde{q}) as well as on the degree to which high local density is maintained when
340stock is being depleted (stock elasticity of harvest b).

341 Empirical studies have established that fishers' behavioural choices can, to a large part,
342be understood based on their expectations on profits. Much of the evidence comes from
343studies on location choice (e.g., Eales and Wilen, 1986; Gillis et al., 1993; Andersen et al.,
3442010). Our model predicts that in the beginning of a fishing season, profit-maximizing
345fishers often target only a single stock. As that stock is fished down, its profitability declines
346and eventually equals that of the other stock. At this point, fishers would be expected to
347split their effort targeting both stocks. This kind of dynamic has been reported from the
348Turks and Caicos Islands, where the artisanal fishermen diversified their effort allocation
349after density of the initially favoured, more valuable target had sufficiently declined; price
350difference between the two targets was constant and did not influence targeting (Béné and
351Tewfik, 2001).

352 Because profitability reflects a range of biological and economic parameters, target
353switches can occur in response to various factors, singly or together. For example, in
354demersal fisheries of Northeast Atlantic, changes in catchability caused by technological
355change are an important factor explaining long-term changes in target species (Marchal
356et al., 2006). In a mixed coastal trawl fishery in Taiwan, the fishers responded to day-to-
357day price fluctuations by increasing catches of species with positive price signals (Liu et al.,
3582018). The failure of the Faroese fisheries to switch away from catching depleted species
359was likely caused by price compensation (Jákupsstovu et al., 2007). Studies of small-scale
360fishermen have shown simultaneous influences of seasonal fluctuations in catchability and
361changes in price that lead to target switching (Salas et al., 2004; Naranjo-Madrigal and
362Bystrom, 2019).

363 Our results show that differences in stock elasticity of harvest (b), reflecting a stock's
364spacing behaviour and the fishers' ability to find the fish, can be as important as differences

365 in the parameters traditionally emphasized when estimating revenues, namely catchability
366 and price (Fig. 4 and 5). While it is commonly acknowledged that the relationship be-
367 tween fish abundance and catch may not be linear, theoretical analyses typically assume
368 that stock elasticity of harvest is unit-elastic ($b = 1$) or perfectly inelastic ($b = 0$; see Stein-
369 shamn, 2011; Liu and Heino, 2013). Differences in b imply that the relative profitability
370 of two species might switch ranks even when they both see similar proportional reduction
371 in abundance. We are not, however, aware of any examples where changing in targeting
372 can be explained by differences in b . While empirical analyses will implicitly account for
373 this effect, it is probably difficult to detect in practice. Nevertheless, our results show that
374 effort regulation is likely to fail when a mixed fishery is composed of a schooling and a
375 non-schooling species.

376 Separability, or the ability of fishers to target and catch a species separately from oth-
377 ers, has a multifaceted role in effort regulation of mixed fisheries. The fishers' ability to
378 target the more abundant species lies at the core of the idea that effort regulation can protect
379 species that are at low abundance. Our results show that separability is indeed necessary
380 for fishery to be able to selectively harvest the more abundant species (Fig. 5a). How-
381 ever, strong separability also increases the risk of seriously depleting one of the species
382 (Figs. 4a and 5a), which can happen when one species has much higher price, catchability,
383 and/or schooling tendency than the other. When separability is poor, such extreme out-
384 comes are mostly avoided, but poor separability also prevents fishers from fishing down
385 the more abundant species (Fig. 5a). The effect of separability is approximately linear,
386 such that studies assuming perfect separability (Katsukawa and Matsuda, 2003; Bischi
387 et al., 2013a,b) and complete lack of separability (Tromeur and Doyen, 2018) capture the
388 extremes—realistic situations likely lie somewhere in between.

389 It is worth noting that effort regulation incentivizes fishers to improve separability only
390 when it helps them to increase the total value of their catches; there is no disincentive
391 per se for catching non-target species. This is in stark contrast to catch quota regulation
392 where lack of separability may prevent fisheries from fully utilizing quotas of some species

393 (e.g., Kuriyama et al., 2016; Mortensen et al., 2018), hence incentivizing investments to
394 technology that improves separability. In any case, the degree to which fishers can adapt
395 their catch profiles is a core question for mixed-fisheries management (Hoff et al., 2010).

396 Our model includes only two species, while most fisheries are considerably more di-
397 verse. Our analysis indicates that profit-oriented exploitation can help to maintain a bal-
398 ance between two exploited species, but also that this requires a fortuitous balance between
399 a number of biological and economic parameters. It is worth emphasizing that the higher
400 is the number of exploitable species in a system, the more likely is that at least one of them
401 does not meet this fortuitous balance. Therefore, challenges in using effort to regulation to
402 manage mixed fisheries will increase with increasing species diversity.

403 In our analysis, we have solely focused on stock dynamics during a single fishing sea-
404 son. Dynamics between fishing seasons could either exacerbate or alleviate the differences
405 in stock levels, depending on the processes related to biomass gain and loss (i.e., gain from
406 recruitment of new individuals and body growth of existing individuals, and loss through
407 mortality) that are stock-specific. Because of density-dependent effects, we can offer some
408 general insights. If a stock is below the stock level that corresponds to maximum biomass
409 production—and maximum sustainable yield (MSY)—then its biomass production will
410 usually increase with increasing stock size. This implies that the relatively more depleted
411 stock will also, on average, have lower biomass production, exacerbating the already ex-
412 isting difference in the stock levels. On the other hand, if two stocks are larger than their
413 respective MSY levels, then the larger stock will see less growth, and the difference in the
414 stock levels is expected to decline. This suggests that effort regulation is more likely to
415 afford a degree of protection to less abundant stocks when most stocks are at healthy levels
416 (near MSY or higher), but that trusting effort regulation to conserve already depleted stocks
417 will be particularly pernicious.

418 Based on our numerical results and the arguments presented above, we suggest that ef-
419 fort regulation is most likely to succeed when (1) the fishery is catching only a few species,
420 (2) these species are biologically similar and have similar market niches (such that the

421 parameters determining revenues are similar), and (3) the species are at ‘healthy’ abun-
422 dance levels. The first two conditions are probably most likely to be fulfilled in cold-water
423 ecosystems where species diversity is low and taxonomically closely related species such
424 as gadids dominate catches. Conversely, effort regulation is likely to fail to protect at least
425 some species in species-rich temperate and tropical ecosystems—even more so when the
426 fisheries already are overexploited, such as in the Mediterranean Sea, where effort reg-
427 ulation is a part of the management toolbox and the overall status of fisheries resources
428 remains poor (FAO, 2018). However, the fact that the effort control system for the Faroese
429 demersal fisheries came close to fulfilling the first two desiderata and yet failed suggests
430 that effort control lacks robustness even under relatively favourable conditions.

431 The fragility of effort regulation is further increased if markets are paying a price pre-
432 mium for rare species, as is the case for, e.g., Pacific bluefin tuna in Japan (Tokunaga,
433 2017). This effect could be countered by consumer awareness campaigns, which have been
434 successful in the past in promoting more sustainable fishing practices (Jacquet et al., 2010).

435 Effort regulation is sometimes advocated as the solution to problems in managing mixed
436 fisheries that is simpler, more flexible, and easier to implement than species-specific TAC
437 regulation (Pope, 2002; European Commission, 2012). While some of these advantages are
438 undeniable, our results warn against placing too much confidence on the ability of effort
439 regulation to provide automatic protection for species that are depleted. Protection is only
440 expected to occur when the targeted species initially offer similar profitability. This re-
441 quires a fortuitous balance between a range of biological and economic parameters, which
442 can be easily broken by exogenous or endogenous changes in prices. In this regards, sup-
443plementary regulations such as area management and gear restriction are needed, and a
444 combination of TAC and TAE is sometimes more favourable (Squires et al., 2017).

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576 Appendix: Derivation of the optimal harvest rules

577 We rely on the Lagrangian conditions to characterize the optimal harvest rules. The scope
 578 of analysis in this section is limited to the case where the Baranov catch equation (also
 579 known as Beverton–Holt model) applies (i.e., $b = 1$ in Eq. 5) and when the price is exoge-
 580 nous.

We formulate the Kuhn-Tucker (KT) Lagrangian function of Eq. 4, and derive following
 KT conditions:

$$\frac{\partial \bar{L}}{\partial e_i} = \overbrace{A_i \frac{1+\tau}{2} + A_{-i} \frac{1-\tau}{2}}^{\text{Marginal profit of effort}} - c - \lambda \leq 0 \quad (6a)$$

$$e_i \frac{\partial \bar{L}}{\partial e_i} = 0 \text{ and } \lambda \left(\sum_i e_i - n \right) = 0 \quad (6b)$$

$$\lambda \geq 0, e_i \geq 0 \text{ and } \sum_i e_i \leq \min(n, Q - \sum_{t=1}^{t-1} e_{i,t}) \quad (6c)$$

581 where $A_i = p_i q_i N_i \left(\frac{M}{(E_i q_i + M)^2} (1 - e^{-(q_i E_i + M)}) + \frac{E_i q_i}{E_i q_i + M} e^{-(q_i E_i + M)} \right)$.

582 Note that $A_i \frac{1+\tau}{2} + A_{-i} \frac{1-\tau}{2} - c$ in Eq. 6a can be interpreted as marginal profit of effort
 583 (MPE). Analysis of Kuhn-Tucker conditions reveals a general optimal harvest rule ($\tau \neq 0$):

- 584 • If $e_1 = 0$ and $e_2 > 0$, then $\frac{\partial \bar{L}}{\partial e_1} < 0$ and $\frac{\partial \bar{L}}{\partial e_2} = 0$, such that $\text{MPE}_1 < \text{MPE}_2$.
- 585 • If $e_1 > 0$ and $e_2 = 0$, then $\frac{\partial \bar{L}}{\partial e_2} < 0$ and $\frac{\partial \bar{L}}{\partial e_1} = 0$, such that $\text{MPE}_1 > \text{MPE}_2$.
- 586 • If $e_1 > 0$ and $e_2 > 0$, then $\frac{\partial \bar{L}}{\partial e_1} = \frac{\partial \bar{L}}{\partial e_2} = 0$, such that $\text{MPE}_1 = \text{MPE}_2$.

587 The above rule states that fishermen will target the species that gives the higher MPE, or
 588 both species will be targeted if their MPEs are same. If $\lambda > 0$, the value of additional boat
 589 is positive and the fleet capacity constraint becomes binding; i.e., $e_i + e_{-i} = n$.

590 The optimal rule for $\tau = 0$, a special case, is slightly different from the above general
 591 rule. In this case only $\frac{\partial \bar{L}}{\partial e_1} = \frac{\partial \bar{L}}{\partial e_2} = 0$ can be satisfied, hence $e_i > 0, i = 1, 2$. Any effort
 592 allocation satisfying the fleet capacity constraint will be optimal, because every unit of

593 effort is equally divided between two species. We consider $\tau = 0$ as a trivial case, hence
594 the subsequent discussion will focus on the cases when $\tau > 0$.