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Diversity and complexity of angler behavior drive socially optimal input and output regulations in a bioeconomic recreational-fisheries model

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**Diversity and complexity of angler behaviour drive socially optimal input and
output regulations in a bioeconomic recreational-fisheries model**

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26 **Abstract**

27 In many areas of the world, recreational fisheries are not managed sustainably.
28 This might be related to the omission or oversimplification of angler behaviour and
29 angler heterogeneity in fisheries-management models. We present an integrated
30 bioeconomic modelling approach to examine how differing assumptions about angler
31 behaviour, angler preferences, and composition of the angler population alter
32 predictions about optimal recreational-fisheries management, where optimal
33 regulations were determined by maximizing aggregated angler utility. We report four
34 main results. First, accounting for dynamic angler behaviour changed predictions
35 about optimal angling regulations. Second, optimal input and output regulations
36 varied substantially among different angler types. Third, the composition of the angler
37 population in terms of angler types was important for determining optimal
38 regulations. Fourth, the welfare measure used to quantify aggregated utility altered the
39 predicted optimal regulations, highlighting the importance of choosing welfare
40 measures that closely reflect management objectives. A further key finding was that
41 socially optimal angling regulations resulted in biologically sustainability fish
42 populations. Managers can use the novel integrated modelling framework introduced
43 here to account, quantitatively and transparently, for the diversity and complexity of
44 angler behavior when determining regulations that maximize social welfare and
45 ensure biological sustainability.

46

47 Keywords: angler specialization; age-structured model; harvest regulations; effort
48 dynamics; utility

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50

51 **Introduction**

52 Recreational anglers are the dominant users of most freshwater and some
53 coastal fish stocks in industrialized countries (Arlinghaus and Cooke 2009).
54 Accordingly, managers are faced with the challenge of balancing the interests of
55 angling groups utilizing fisheries resources with concerns about the biological
56 sustainability of exploited fish populations (Radomski et al. 2001; Peterson and Evans
57 2003; Arlinghaus 2006b). The lack of sustainable recreational-fisheries management
58 in some areas of the world (Post et al. 2002; Lewin et al. 2006) suggests that current
59 management strategies have not always been successful in achieving this balance.
60 This may be because effectively managing a fishery requires understanding not only
61 how fish respond to exploitation, but also how anglers alter their fishing behaviour in
62 response to social and ecological changes in the fishery; consequently such
63 behavioural dynamics must be incorporated into integrated fisheries-management
64 models (Johnson and Carpenter 1994; Radomski et al. 2001; Post et al. 2008). In the
65 past, however, recreational-fisheries researchers and managers have focused on the
66 biological dimension of recreational fisheries, largely overlooking the “human
67 dimension” (Aas and Ditton 1998; Cox and Walters 2002a; Arlinghaus et al. 2008a).
68 To move forward, it is critical to quantify and integrate angler preferences and
69 resulting behavioural decisions into recreational-fisheries models designed to
70 determine optimal management policies (Radomski and Goeman 1996; Arlinghaus et
71 al. 2008a).

72 Optimum social yield (OSY) is one management objective that can
73 incorporate social and economic aspects into fisheries-management models and
74 policies (Roedel 1975). In comparison with the traditional approach of managing for
75 maximum sustainable yield (MSY) in both commercial and recreational fisheries

76 (Larkin 1977; Malvestuto and Hudgins 1996; Hilborn 2007), OSY is better suited to
77 recreational fisheries because it incorporates socio-cultural benefits a fishery provides
78 that are not measured by yield alone, such as an angler's satisfaction resulting from
79 catching a large fish (Roedel 1975; Malvestuto and Hudgins 1996; Radomski et al.
80 2001). OSY integrates such social and economic factors with biological
81 considerations, to develop a fisheries-management objective that maximizes the total
82 utility (alternatively termed benefits or social welfare; Dorow et al. 2010) that a
83 recreational fishery provides to society (Roedel 1975; Malvestuto and Hudgins 1996).
84 Hence, similar to MSY, management for OSY may provide an unambiguous
85 management objective against which to judge management developments and
86 successes (Bennett et al. 1978; Barber and Taylor 1990; Radomski et al. 2001).

87 Despite the general advantages of a socioeconomic objective such as OSY
88 over MSY for managing recreational fisheries, few recreational-fishing models based
89 on utility theory have been developed to predict the optimal social welfare generated
90 by different management schemes (e.g., Die et al. 1988; Jacobson 1996; Massey et al.
91 2006). Furthermore, angler-effort dynamics, if considered at all, are generally
92 assumed to be predominantly or exclusively driven by catch rates, or by some other
93 measure of fish abundance (Johnson and Carpenter 1994; Beard et al. 2003; Post et al.
94 2003). However, angler behaviour is likely much more complex (Carpenter and Brock
95 2004; Arlinghaus et al. 2008a). It is known from social-science research on
96 recreational fisheries that, in addition to catch rates, a diverse set of social and
97 biological attributes of a fishery – such as availability of preferred species, fish size,
98 congestion, facilities, regulations and the perceived aesthetic value of the fishery –
99 affect the participation decisions of anglers (reviewed in Hunt 2005). Therefore,
100 angler-effort dynamics driven by catch rates alone can be unrealistic (Paulrud and

101 Laitila 2004). Hence, recreational-fisheries models designed to maximize angler
102 utility should account for complexity in angler behaviour by incorporating multi-
103 attribute utility functions that describe the fishing-participation decisions of anglers.

104 Another important, yet often overlooked, aspect of recreational fisheries is
105 angler diversity (i.e., heterogeneity in angler behaviour; Anderson 1993; Jacobson
106 1996; Post et al. 2008). Various types of anglers will differ not only in their fishing
107 preferences, and therefore in the utility they derive from fishing (Fisher 1997;
108 Connelly et al. 2001; Arlinghaus et al. 2008b), but also with respect to their fishing
109 practices (Bryan 1977; McConnell and Sutinen 1979; Hahn 1991). Hence, the
110 potential impacts of fishing on fish populations likely vary with angler type (Dorow et
111 al. 2010). For example, in many fisheries a minority of anglers catches the majority of
112 fish (Baccante 1995), and this minority typically encompasses the most avid and
113 specialized angler types (Dorow et al. 2010). Human-dimension researchers have
114 repeatedly highlighted that accounting for angler diversity is important for sustainable
115 fisheries management (Fisher 1997; Aas et al. 2000; Arlinghaus and Mehner 2003).
116 While there are some examples of coupled social-ecological models that link complex
117 angler behaviour and fish population dynamics (e.g., Cole and Ward 1994; Woodward
118 and Griffin 2003; Massey et al. 2006), to our knowledge only McConnell and Sutinen
119 (1979) and Anderson (1993) considered heterogeneity either in angler preferences or
120 fishing practices in a bioeconomic modelling context. In both cases, the modelling
121 frameworks differed substantially from that presented here. In particular, these earlier
122 studies did not use random-utility models to predict angler participation under
123 different management scenarios, and the complexity of the biological and angler-
124 behaviour components were much more simplified.

125 Our goals of this study are fourfold. First, we present an integrative
126 bioeconomic modelling approach that links the ecological, socioeconomic and
127 management components driving angler-effort dynamics to a fish population model,
128 and that allowed optimal harvest regulations for various angler types to be predicted.
129 Second, we demonstrate the importance of assumptions about angler-effort dynamics
130 in fisheries management by contrasting predictions from models that make traditional
131 assumptions of static or exclusively catch-based dynamic angler behaviour with
132 models that assume more complex, multi-attribute dynamic behaviour. In this study,
133 complexity in angler behaviour is characterized by whether angler-effort dynamics
134 rely on a single fishery attribute to drive angler behaviour or on multiple fishery
135 attributes. Third, by incorporating heterogeneity in angler behaviour into a
136 bioeconomic modelling framework by accounting for the perceived utility a fishery
137 provides to an angler population,, we examine how angler diversity (i.e.,
138 heterogeneity of angler types) and the composition of the angler population (in terms
139 of these angler types) influence predictions about optimal management strategies.
140 Finally, we explore how different management objectives, represented by different
141 measures of social welfare, alter predicted optimal management regulations. Rather
142 than simulating a particular fishery, our approach is stylized in nature and is intended
143 to demonstrate the suitability of an integrated bioeconomic modelling approach for
144 investigating coupled angler-fish population dynamics.

145 **Methods**

146 We developed an integrated model in which angler-type-specific utility
147 derived from both catch- and non-catch-related attributes of the fishing experience
148 was linked to a deterministic age-structured fish population model for a single-
149 species, single-lake fishery. Our modelling framework had three components: (i) a

150 management component that described the regulations applied to the fishery system,
151 (ii) a socioeconomic component that described the effort dynamics of different angler
152 types, and (iii) a biological component that described the fish population dynamics.
153 Angler utility was used to determine changes in angling effort in the dynamic angler-
154 behaviour scenarios, and to make predictions about optimal harvest regulations. The
155 resulting impacts on the fish population under different management policies were
156 investigated to determine whether management for social optima also conserved the
157 fish population. All model equations are summarized in Table 1 and illustrated in
158 Figure 1; model parameters are listed in Tables 2 and 3.

Insert
Figure 1

159 **Management component**

160 Traditional harvest-control measures have focused on regulating the harvest
161 rates of individual anglers to achieve biological sustainability (Radomski et al. 2001).
162 However, in open-access systems, which are typical for many recreational fisheries
163 (Post et al. 2002), output-control measures that do not directly limit angler numbers
164 cannot constrain total fishing mortality (Radomski et al. 2001; Cox and Walters
165 2002a; Cox and Walters 2002b). The failure of traditional output-control measures to
166 preserve some recreationally exploited fish populations (Post et al. 2002) has led to a
167 call for input-control measures that more directly limit angling effort (Cox and
168 Walters 2002a; Cox and Walters 2002b). Therefore, we investigated two types of
169 regulatory policies over a range of values (Table 2): a traditional output-control
170 regulation, expressed in terms of a minimum-size limit, and an input-control
171 regulation, expressed in terms of the number of angling licenses issued.

172 **Socioeconomic component**

173 *Angler utility*

174 Economic utility theory assumes that human agents make choices that will
175 maximize their personal utility (alternatively termed benefits or satisfaction; Perman
176 et al. 2003). For example, from a set of potential alternatives, recreational anglers will
177 choose to fish a fishery that provides them with the greatest possible utility (Hunt
178 2005). Multiple attributes contribute to an individual angler's utility function, and the
179 relative importance of fishery attributes (such as fish size or crowding), called part-
180 worth utilities, for total angler utility vary substantially among different angler types
181 (Aas et al. 2000; Oh et al. 2005a; Oh and Ditton 2006). Choice models based on
182 random-utility theory (McFadden 1974; Manski 1977) can be calibrated with actual
183 (revealed) or hypothetical (stated) empirical site-choice data. Such models constitute
184 one approach that can be used to predict recreational-angler behavior, which can then
185 be used to predict and understand how anglers will react to changes in the attributes of
186 a fishery (Paulrud and Laitila 2004; Massey et al. 2006; Wallmo and Gentner 2008).

187 Three scenarios of angler behaviour were investigated. In the first scenario, we
188 simulated static angler behaviour, characterized by anglers that did not respond to
189 changes in a fishery's attributes (such as fish size, catch rate or congestion level), but
190 instead, participated at the maximum effort level allowed. Predictive recreational-
191 fisheries models often assume constant exploitation rates and ignore angler dynamics
192 when evaluating regulation impacts (e.g., Dunning et al. 1982). The static scenario
193 mimics this situation by keeping angling effort constant. In our two other scenarios,
194 anglers were allowed to behave dynamically, i.e., they chose to fish or not to fish
195 depending on the time-varying utility provided by the fishery. Utility functions that
196 described the preferences of a particular angler type for the fishing attributes
197 experienced were used to simulate angler-type-specific behavioural decisions. In the
198 second scenario, the utility of fishing was based on the utility gained from catch rates

199 alone (Table 1, equation 1a; and Table 3), an approach used in previous recreational-
200 fishing models (Cox et al. 2003; Post et al. 2003). In the third scenario, utility was
201 based on a more realistic multi-attribute utility function (Table 1, equation 1b; and
202 Table 3). Attributes included in this utility function were catch rates, average size of
203 fish caught, maximum size of fish caught, angler congestion, minimum-size limit
204 regulations and license costs, all of which have been shown to affect anglers' fishing
205 decisions about participating in a particular fishery (Hunt 2005). Although the multi-
206 attribute utility function was not used to determine angling effort in the static
207 scenario, for comparative purposes it was used to evaluate the quality of the fishery at
208 the end of the simulations (Table 1, equation 1b) (Figure 1).

Insert Table 1

209 *Angler-effort dynamics*

210 In our second and third scenarios, anglers responded dynamically to their
211 perception of fishery quality by changing the amount of effort they devoted to the
212 fishery. In these scenarios, the utility gained from a fishing experience determined the
213 angler's probability of an angler choosing to fish over the alternative of not fishing
214 (Table 1, equation 2a). This probability was calculated as is typical in empirical
215 choice models (Oh et al. 2005b; Massey et al. 2006). The probability of fishing based
216 on angler utility, as well as the maximum time anglers would fish in a year
217 irrespective of fishing quality, were then used to determine realized annual effort of
218 anglers (i.e., the amount of time they actually fished; Table 1, equations 2b-2e; Figure
219 1). To account for the fact that anglers make decisions based on previous experiences
220 and habits, and not exclusively based on their most recent experiences (Adamowicz et
221 al. 1994), a fishing-behaviour persistence term (Table 2) was introduced to the effort
222 dynamics (Table 1, equation 2b). This term described the relative influence of last
223 year's realized fishing probability on the current year's realized fishing probability.

224 We assumed that the realized annual angling effort (Table 1, equation 2e) was limited
225 by three factors: the realized probability of fishing, the desired maximum effort that
226 an individual angler would fish irrespective of angling quality (Table 1, equation 2c),
227 and the input-control measure expressed in terms of the number of angling licenses
228 issued (Table 1, equation 2d). The instantaneous fishing effort of a given angler type
229 was assumed to be constant throughout the fishing season, and to equal zero after the
230 fishing season ended (Table 1, equation 2f).

Insert Table 2

231 *Angler heterogeneity*

232 Angler heterogeneity was introduced into our model by defining three
233 different angler types – generic, consumptive, and trophy anglers – that differed in
234 their degree of angling specialization (Bryan 1977; Ditton et al. 1992; Table 3). Our
235 parameterization of angler behaviour was based on recreational specialization theory
236 (Bryan 1977; Ditton et al. 1992). Bryan (1977) described four general angler types
237 ranging from the casually involved to the technique- and setting-specialist. As
238 specialization levels increase, skill levels improve, fish size is of greater importance,
239 and harvesting fish is of lesser importance (Bryan 1977). This can lead to differing
240 propensities to perform voluntary catch-and-release (Arlinghaus 2007), and to an
241 increased ability to catch more and larger fish (Dorow et al. 2010). Angler preferences
242 also change with specialization: for example the value of solitude relative to the social
243 aspects of the fishing experience varies with specialization (Ditton et al. 1992;
244 Connelly et al. 2001). Based on pioneering work by Bryan (1977) and subsequent
245 applications and refinements (e.g., Quinn 1992; Allen and Miranda 1996; Fisher
246 1997) we devised qualitatively realistic angler-type-specific part-worth-utility
247 functions for the various attributes of the fishing experience. Figure 2 illustrates

248 qualitative differences in preferences and tolerances for different fishery attributes
249 among angler types, while Figure 3 illustrates the resultant utility functions.

Insert
Figure 2

250 Parameters for three stylized angler types were chosen to reflect differential
251 skill, consumptive orientation and overall dedication to the recreational fishing
252 experience (Table 3). Angler types differed in both their fishing practices, and their
253 preferences for various attributes of the fishing experience (Figure 2; Table 3).
254 Generic anglers were assumed to be the least specialized, consumptive anglers were
255 intermediate, and trophy anglers were the most specialized. By definition,
256 consumptive anglers had the greatest consumptive orientation. Accordingly, generic
257 anglers were assumed to (i) be least likely to participate in angling activities, (ii) be
258 intermediate in their tolerance of restrictive minimum-size limits, (iii) be the most
259 affected by license costs, (iv) have an intermediate interest in catch rates and be least
260 interested in the challenge of catching fish, (v) be least interested in average fish size
261 and be intermediately interested in trophy-sized fish, (vi) be most tolerant of angler
262 crowding, (vii) be least skilled, and to (viii) practice some voluntary catch-and-release
263 of harvestable fish (Table 3). In contrast, consumptive anglers were assumed to (i)
264 participate at an intermediate level in angling activities, (ii) be least tolerant of
265 restrictive minimum-size limits, (iii) be intermediately affected by license costs, (iv)
266 be most interested in catch rates and intermediately interested in the challenge of
267 catching fish, (v) be intermediately interested in average fish size and least interested
268 in trophy-sized fish, (vi) be intermediately tolerant of angler crowding, (vii) have
269 intermediate skills, and (viii) practice no voluntary catch-and-release of harvestable
270 fish (Table 3). Finally, trophy anglers were assumed to (i) participate the most in
271 angling activities, (ii) be most tolerant of restrictive minimum-size limits, (iii) be least
272 affected by license costs, (iv) be least interested in catch rates but most interested in

273 the challenge of catching fish, (v) be most interested in average fish size and trophy-
274 sized fish, (vi) be least tolerant of angler crowding, (vii) have the greatest skills, and
275 (viii) practice the most voluntary catch-and-release of harvestable fish (Table 3).
276 Trophy anglers were also assumed to target larger fish relative to consumptive and
277 generic anglers (through the use of different fishing gear; Rapp et al. 2008; Table 3).
278 Parameter values and further justification for these assumptions are outlined in Table
279 3, and the resulting shapes of the angler-type-specific part-worth-utility functions are
280 illustrated in Figure 3. Although these functions might look different for particular
281 fisheries, we believe that their general features adequately reflect the angling
282 behaviour and preferences of differently specialized recreational anglers.

283 The importance of angler heterogeneity for determining optimal fishing
284 regulations was examined by first comparing model results among different
285 homogeneous angler populations, each composed of a single angler type. However,
286 because natural angler populations are likely comprised of a mixture of angler types,
287 we also considered a mixed angler population composed of all three angler types
288 mentioned above. As this aspect increases the model complexity and in an attempt to
289 simplify angler descriptions, recreational-fisheries researchers and managers may
290 wish to simplify angler descriptions by assuming some form of average angler
291 behaviour (Hahn 1991; Aas and Ditton 1998). Therefore, to examine the importance
292 of explicitly accounting for the composition of the angler population on model
293 predictions of optimal regulations, we compare model results for an average angler
294 type population with those for a corresponding mixed angler population composed of
295 three angler types. here, the average angler type was defined by a weighted average of
296 fishing preferences and fishing practices of the three angler types according to their
297 relative frequencies in the mixed angler population (Table 2). It should be noted, that

Insert
Figure 3
and
Table 3

298 this is a weighted average and therefore depends on the assumptions about the relative
299 abundance of angler types in the mixed angler population. However, this example
300 demonstrates the implications of the simplifying assumption of an average angler.

301 **Biological component**

302 Our study aimed to show how the biological and socioeconomic and
303 management components of recreational-fishery systems could be linked in an
304 integrated modelling framework. For brevity we therefore only describe the essentials
305 of the biological component in terms of growth, reproduction and survival
306 functions. Tables 1 and 2 provide further details about equations and parameters..

307 In short, an age-structured model was used to describe the fish population
308 being exploited. Individual fish within an age class were assumed to be ecologically
309 equivalent (Tables 1, equations 3a and 3b). The fish population model was
310 parameterized to be representative of a northern pike (*Esox lucius* L.) population. We
311 chose this species due to its importance for recreational fisheries in both North
312 America and Eurasia (Paukert et al. 2001; Arlinghaus and Mehner 2004a). In all
313 scenarios, the fish population reached its demographic equilibrium prior to the
314 introduction of fishing, and the results presented correspond to equilibrium conditions
315 after fishing was introduced (i.e., we investigated long-term dynamics).

316 The determination of fishing effort (Table 1, equations 2a-2f) and fish
317 reproduction (Table 1, equations 5a-5d) were assumed to occur on an annual basis at
318 the beginning of each year, and population and fishery characteristics were updated
319 annually. However, because recreational fishing is often a size-selective process
320 (Lewin et al. 2006) occurring throughout the year, we described fish mortality and the
321 growth in body size of fish by continuous functions (Table 1, equations 4a-4e). This
322 allowed our model to account for fish to grow into vulnerable size classes within each

323 year, and for the recapture and repeated exposure to hooking mortality of released
324 individuals throughout the fishing season, both of which are important aspects of
325 recreational fisheries (Coggins et al. 2007). These resultant ordinary differential
326 equations were solved numerically using the ODE45 function in Matlab (version 7.0.1
327 Mathworks, Inc.).

328 Two crucial density-dependent relationships were included to allow for
329 compensatory responses of the fish population to exploitation (Lorenzen and Enberg
330 2002): density-dependent biphasic growth in body size (Table 1, equations 4a-4d)
331 (Lester et al. 2004; Dunlop et al. 2007) and density-dependent survival from spawning
332 to post-hatch of fish of age zero. The latter was represented by a Beverton-Holt type
333 relationship, which was assumed to apply at the beginning of each year (Table 1,
334 equations 5c) (Lorenzen 2008). Fish younger than one year were assumed to
335 experience no further natural mortality (Table 2) but could experience fishing
336 mortality if they became large enough. Fish one year and older experienced a constant
337 natural mortality rate in addition to size-dependent fishing mortality (Table 2,
338 equation 7h).

339 Fishing mortality was assumed to be size-dependent in two ways that
340 quantitatively differed among angler types (see Table 3 for angler specific
341 parameters). First, catch rates were dependent on the size-dependent vulnerability of
342 fish to the specific fishing gear utilized by each angler type. Vulnerability to capture
343 therefore differed among age classes and also changed over the course of the growing
344 season (Table 1, equations 7a and 7b; see Table 3 for parameters). Catch rates were
345 also dependent on fishing effort and the skill level of the anglers (Table 1, equation
346 7b, see Table 3 for parameters). Second, harvest of fish was regulated by a minimum-
347 size limit (*MSL*; Table 1, equation 7c). While all fish above the legal *MSL* were

348 harvestable, a portion of undersized fish were also considered harvestable because of
349 non-compliance with regulations (either through ignorance or choice; Sullivan 2002).
350 Anglers chose to harvest fish based on their catch rates mediated by their propensity
351 to voluntarily release fish (Table 1, equation 7e) determined by the personal limit an
352 angler had on the number of fish they harvested in a day; (see Table 3 for angler-type-
353 specific parameters). Released fish were assumed to experience hooking mortality
354 from handling or injuries (Table 1, equation 7f; Table 3; Arlinghaus et al. 2007,
355 Arlinghaus et al. 2008c). Fish under the legal size limit, which were not part of the
356 pool of illegally harvestable fish, only experienced hooking mortality (Table 1,
357 equation 7g).

358 After fishing was introduced, the fish population was allowed to equilibrate.
359 The spawning potential ratio (*SPR*) was used to assess the biological impacts of
360 angling exploitation. *SPR*, which has previously been used in recreational-fishing
361 models (Coggins et al. 2007; Allen et al. 2009), measures reductions in the fish
362 stock's reproductive output, and can thus serve as an indicator of recruitment
363 overfishing (Goodyear 1993; Coggins et al. 2007; Allen et al. 2009). In our model, we
364 use a weighted *SPR* (Table 1, equation s 5b and 6). Depending on the life history of a
365 species, values below 0.2-0.3 are considered critically low (Goodyear 1993) and it is
366 commonly assumed that *SPR* should be maintained above 0.35-0.40 to reduce the
367 risk of recruitment failure (Goodyear 1993; Coggins et al. 2007). We used these
368 values as criterion to assess the risk of recruitment overfishing under different
369 management policies.

370 **Social-welfare measures**

371 Social welfare was used to determine optimal regulations. Social welfare is an
372 aggregation of individual utilities (Perman et al. 2003) and determines the total

373 economic value of a good or service, such as a recreational-fishing experience, as
374 perceived by anglers (Edwards 1991). A social welfare function describes how
375 individual utilities are aggregated based on their social “worth”, and it is assumed that
376 any concerns about equity are accounted for in the aggregation method (Perman et al.
377 2003). However, maximizing social welfare does not necessarily result in an equitable
378 distribution of resources among individuals, nor is there universal consensus on what
379 constitutes an appropriate social-welfare measure or function (Perman et al. 2003).
380 Managers must therefore carefully decide what social-welfare measures reflect their
381 management objectives (e.g., maximizing angler satisfaction and/or participation).

382 In most model simulations described below, a utilitarian social-welfare
383 function was used, referred to as total utility (TU), in which individual utilities were
384 weighted equally among angler types. However, in a subset of simulations, three
385 different social welfare functions, representing different management objectives, were
386 used to examine how these differences alter predictions about socially optimal
387 management regulations. The first welfare measure, TU, described the utility gained
388 by an angler type per fishing experience, multiplied by the total annual number of
389 fishing experiences (measured in terms of angling effort, and expressed in angling
390 days) by that angler type, and summed over all angler types (Table 1, equation 8a;
391 similar to McConnell and Sutinen 1979). TU reflects the realized demand for angling
392 experiences. However, TU may be influenced heavily by individuals with
393 disproportionately large utility, and a more equitable distribution of resources among
394 all anglers in the angler population may be desired (Loomis and Ditton 1993). Thus, a
395 second, more equitable utilitarian social-welfare function (EU) was examined. Here,
396 individual utility from a fishing experience was weighted by the relative abundance of
397 angler types in the angler population, to create a weighted mean utility for an

398 individual, which was then multiplied by the aggregate number of angling days (Table
399 1, equation 8b). Finally, we examined a Rawlsian approach (RU) to utility
400 maximization, where the utility of the worst-off individual was maximized,
401 emphasizing the objective of achieving the most equitable distribution of resources
402 (Perman et al. 2003). Here, the utility from the angler type with the lowest individual
403 utility was used and multiplied by the aggregate number of angling days (Table 1,
404 equation 8c). Naturally, the second and third social-welfare measures only differed
405 from the first measure in the mixed angler population composed of different angler
406 types.

407 **Outline of analysis**

408 Across a range of minimum-size limits and angling-license numbers, three
409 different angler-behaviour scenarios – static, catch-based dynamic and multi-attribute
410 dynamic scenarios – were considered for five different types of angler populations –
411 generic, consumptive, trophy, average, and mixed. Optimal input and output
412 regulations were identified by maximizing one of three measures of social welfare –
413 total utility TU, equitable utilitarian utility EU, and Rawlsian utility RU (Table 1,
414 equations 8a-c). With this approach, we examined the impacts of dynamic angler
415 behaviour, angler heterogeneity, and composition of the angler population on socially
416 optimal regulations and the resulting biological impacts on the fish population. In
417 most analyses presented, TU was used to determine socially optimal management
418 regulations. However, we also examined the EU and RU social-welfare measures in
419 the context of multi-attribute dynamic angler behavior and mixed angler populations,
420 to demonstrate how different management objectives alter socially optimal
421 management regulations.

422 We used sensitivity analyses to explore the importance of different attributes
423 for determining angler behaviour, optimal regulations and biological impacts, by
424 removing in turn each attribute from the multi-attribute angler-behaviour scenario.
425 However, given the hypothetical nature of the constructed angler types and their part-
426 worth-utility functions (Figure 3), we decided it would be imprudent to derive
427 generalized conclusions about the relative importance of individual attributes in
428 determining optimal regulations. Therefore, sensitivity analyses were not intensified
429 beyond the approach summarized above.

430 **Results**

431 **Impacts of dynamic angler behaviour**

432 A comparison of the three angler-behaviour scenarios showed substantial
433 differences in predictions of total utility (left to right in Figure 4). Optimal minimum-
434 size limits were predicted to be highest in scenarios with catch-based dynamic angler
435 behaviour and were generally lower (and similar) for corresponding scenarios with
436 static and multi-attribute dynamic angler behavior for angler populations composed of
437 one angler type (Table 4; Figure 4). Optimal effort regulations were lowest in the
438 static scenarios, intermediate in the multi-attribute scenarios, and highest in the catch-
439 based scenarios (Table 4). In fact, optimal license numbers in the catch-based
440 scenarios were often more than two times larger than the number predicted in the
441 other scenarios. Under predicted optimal regulations, the number of hours that anglers
442 actually fished, termed realized angling effort, were identical in the static and multi-
443 attribute scenarios when the angling population was composed of one angler type,
444 (thus following the pattern of predictions for optimal minimum-size limits). In the
445 catch-based scenario, realized effort followed a trend similar to that of optimal license
446 numbers.

Insert Figure 4 and Table 4

447 The risk of recruitment overfishing and the biological impacts of recreational
448 angling on the modelled pike population were affected by the type of angler
449 behaviour considered (Figure 5). Static angler behaviour caused the most negative
450 impacts on the fish population across the range of minimum-size limits and license
451 numbers examined, compared to the two scenarios in which anglers behaved
452 dynamically. This was because realized angling effort in the static angler-behavior
453 scenario was fixed at the maximum level allowed, whereas in the two dynamic
454 scenarios realized angling effort was less and depended on the utility anglers gained
455 from the fishery. When comparing the two dynamic scenarios, biological impacts of
456 fishing at low to moderate *MSL* levels in the catch-based scenario were generally less
457 severe than in the multi-attribute scenario, with the latter approaching recruitment
458 overfishing and fishery collapse at lower license numbers. At high *MSL* levels,
459 approaching complete catch-and-release conditions, the risk of recruitment
460 overfishing was often greater in the catch-based scenario, although the *SPR* never
461 dropped below 0.4, even when a large number of licenses were issued.

Insert
Figure 5

462 **Impacts of angler heterogeneity**

463 Not only angler dynamics, but also angler heterogeneity substantially affected
464 model-predicted optimal input and output regulations. When the three angler types
465 were compared (first three rows in Figure 4), optimal minimum-size limits were
466 generally intermediate for generic anglers, low for consumptive anglers and high for
467 trophy anglers, with the latter approaching complete catch-and-release conditions,
468 except in the catch-based scenario, in which complete catch-and-release regulations
469 were preferred by all angler types (Figure 4; Table 4). Optimal effort regulations were
470 found to be the lowest for consumptive anglers in the static and multi-attribute
471 scenarios, intermediate for trophy anglers and highest for generic anglers. However,

472 in the catch-based scenario, all angler types preferred a large number of licenses, with
473 generic anglers favouring somewhat fewer angler licenses than the other angler types.
474 Under optimal regulations, consumptive anglers were predicted to fish the least, but
475 generic and trophy anglers invested more (and similar) realized angling efforts in the
476 static and multi-attribute scenarios (Table 4). However, in the catch-based scenario,
477 consumptive anglers invested the most realized angling effort. At their optimum,
478 trophy anglers, as a homogeneous group, derived the highest utility from fishing,
479 exceeding that of the other anglers types by a factor of more than two; generic anglers
480 were intermediate, while consumptive anglers derived the least utility in the static and
481 multi-attribute scenarios (Figure 4).

482 Differences among the angler types also affected the risk of recruitment
483 overfishing. In all scenarios and across all regulation combinations, consumptive
484 anglers generally had the most negative impact and generic anglers the least, except in
485 the multi-attribute scenario at high *MSL* levels. This trend was also seen when
486 examining the biological impacts of different angler types under the different
487 regulations they perceived as optimal (Table 4). Under these optimal regulations, the
488 biological impact of consumptive anglers was greatest, occurring close to the
489 threshold levels of recruitment overfishing (0.35-0.40) and at regulation combinations
490 for which small changes in regulations could cause large changes in the risk of
491 recruitment overfishing (Figure 5). At these respective optima, generic and trophy
492 anglers impacted the fish population much less than consumptive anglers and at
493 regulation combination that imply a low risk of recruitment overfishing.

494 We found the sensitivity of results to individual attributes in the multi-attribute
495 scenario varied in their effect on optimal regulations, realized effort and *SPR*, and
496 varied greatly with angler type, without any consistent pattern becoming evident

497 (Table A1). We could tentatively conclude, however, that findings for trophy anglers
498 were strongly dependent on crowding aversion, while findings for consumptive
499 anglers were particularly sensitive to *MSL* levels and some catch attributes. It was
500 also interesting to notice that the response of mixed angler populations to the removal
501 of a particular fishery attribute sometimes exceeded that of homogeneous angler
502 populations, highlighting the importance of including heterogeneity in angler
503 preferences (Table A1).

504 **Impacts of angler-population composition**

505 Predictions of optimal input and output regulations substantially differed
506 between the average angler and the mixed angler population (bottom two rows in
507 Figure 4). Under optimal regulations, license numbers and realized angling efforts
508 were higher for the mixed angler population than for the average angler population
509 (Table 4). Optimal *MSL* levels for the mixed angler population were the same as the
510 average angler population in the static scenario, lower in the catch-based scenario and
511 higher in the multi-attribute scenario. In addition, across all scenarios, TU under
512 optimal regulations was greater in the mixed angler population than in the average
513 angler population.

514 For the average angler population was assumed, minimum-size limits and
515 realized efforts under optimal regulations were identical in the static and multi-
516 attribute scenarios. However, for the mixed angler population, minimum-size limits,
517 license numbers and realized efforts under optimal regulations were substantially
518 higher in the multi-attribute scenario than in the static scenario (Figure 4; Table 4).
519 Furthermore, in the multi-attribute scenario, predictions of optimal license sales and
520 realized efforts were generally higher than in any of the three homogeneous angler
521 populations (Table 4). The mixed angler population was also predicted to have a

522 greater biological impact than the average angler population (Figure 5). However,
523 under optimal regulations, the risk of recruitment overfishing in both cases was low
524 (Table 4).

525 Changes in the composition of the mixed angler population that fished in the
526 multi-attribute scenario were described by the changes in the proportion total realized
527 angling effort invested by each angler type (Figure 6). This shows that the
528 composition of the angling population varied depending on minimum-size limits and
529 license regulations, with trends predominantly following changes in *MSL* (Figure 6).
530 At low *MSL* levels and low license numbers, all angler types fished in approximately
531 equal proportions, whereas at low *MSL* levels and high license numbers the
532 composition of the angling population resembled that of the entire angler population
533 (i.e., 40% generic, 30% consumptive and 30% trophy). At moderate to high *MSL*
534 levels the majority of consumptive anglers in the angler population chose not to fish,
535 and thus dropped out of the angling population. Even higher *MSL* levels resulted in
536 generic anglers dropping out too, and thus in an angling population dominated by
537 trophy anglers. Under optimal regulations, the composition of the angling population
538 in the multi-attribute scenario was heavily skewed toward generic and trophy anglers,
539 with few consumptive anglers being attracted to the fishery (Table 4; Figure 6).

Insert Figure 6

540 **Impacts of social-welfare measures**

541 In the multi-attribute scenario for the mixed angler population, socially
542 optimal minimum-size limits were highest for total utility (TU), intermediate for
543 equitable utilitarian utility (EU) and lowest for Rawlsian utility (RU) (Figure 7; Table
544 4). Optimal license numbers were also highest for the TU social-welfare measure, but
545 lower (and similar) for the EU and the RU social-welfare measures, and realized
546 angling efforts under optimal conditions showed the same pattern.

547 Under optimal regulations, optimal license numbers and realized angling
548 efforts for the average angler population never exceeded those for the mixed angler
549 population, irrespective of the applied social-welfare measure (Table 4). However, the
550 optimal *MSL* was slightly higher in the average angler population than in the mixed
551 population when a RU social-welfare measure was applied (Table 4). Under optimal
552 regulations, *SPR* levels were well above 0.40, irrespective of the applied social-
553 welfare measure (Table 4); therefore, all social-welfare measures avoided recruitment
554 overfishing under optimal regulations.

Insert
Figure 7

555 **Discussion**

556 We developed a bioeconomic modelling approach that integrates angler
557 behaviour and angler heterogeneity with age-structured and density-dependent fish
558 population dynamics, to determine socially optimal input and output regulations for a
559 recreational fishery. Using this approach, we have demonstrated how angler
560 behaviour and heterogeneity affect optimal regulations, and how optimal regulations
561 varied with the social-welfare measure applied.

562 **Angler behaviour**

563 The importance of accounting for angler behaviour was demonstrated by the
564 differences observed in predicted optimal regulations (expressed in terms of
565 minimum-size limits and license numbers) among three angler-behavior scenarios that
566 describe, respectively, static, catch-based dynamic and multi-attribute angling
567 dynamics. Predicted optimal minimum-size limits and license numbers were
568 substantially higher for the catch-based scenario than for the other two scenarios.
569 However, most published recreational-fisheries models that incorporated dynamic
570 angler behaviour assumed that anglers respond to catch rates alone or some measure
571 of fish abundance (Johnson and Carpenter 1994; Beard et al. 2003; Post et al. 2003),

572 thus neglecting other attributes known to affect participation decisions of anglers
573 (Hunt 2005).

574 Our findings call into question the validity of this simplifying assumption and
575 resulting predictions of “optimal” regulations. For example, when catch rate was
576 assumed to be the only attribute determining the fishing decisions of anglers, the
577 catch-based scenario predicted optimal input and output regulations that effectively
578 imply complete catch-and-release regulatory policies at largely unlimited effort levels.
579 This prediction is clearly misleading in many situations and results from an
580 oversimplification of angler preferences. Indeed, because some angler types are
581 strongly harvest-oriented, management conflicts and dilemmas have occurred in some
582 recreational fisheries despite high catch rates, when the possibility for anglers to
583 harvest was constrained (Matlock et al. 1988; Radomski 2003; Sullivan 2003).
584 Perceived harvest constraints may result in the displacement of harvest-oriented
585 anglers to alternative fisheries (Radomski and Goeman 1996; Beard et al. 2003), an
586 important effect that cannot be captured by models that assume angler behaviour to be
587 driven by catch rates alone. In contrast, our investigations of multi-attribute dynamic
588 angler behaviour, presumably allowing a more realistic representation of angling
589 effort, showed that complete catch-and-release regulations were not always socially
590 optimal.

591 Our sensitivity analyses highlighted that, while most attributes of the fishing
592 experience (such as fish size, catch rate, crowding, aversion to regulations, etc.) were
593 important for determining angler choice and angler welfare, their relative importance
594 varied among angler types (Table A1). This underscores the importance of including
595 all relevant catch- and non-catch-related attributes affecting angler choice in

596 bioeconomic fisheries models to more accurately predict angler behaviour and fishing
597 pressure, and to derive optimal regulations that maximize angler welfare.

598 A multi-attribute perspective on angler behavior and welfare is also likely to
599 improve predictions of the biological impacts of fishing under different regulations.
600 Historically, angler populations were expected to be self-regulating, as anglers were
601 assumed to leave a fishery when catch rates declined (Cox and Walters 2002a,
602 Radomski 2003). However, because catch rate is just one among many attributes
603 characterizing a fishing experience, such catch-based self-regulation does not
604 necessarily apply (Post et al. 2002; Paulrud and Laitila 2004; Post et al. 2008). Indeed,
605 we found that realized angling effort and the biological impacts were higher in the
606 multi-attribute scenario than in the catch-based scenario at low to intermediate *MSL*
607 levels. These finding corroborate claims that multi-attribute angler behaviour may put
608 fish populations at risk of overexploitation (Post et al. 2002), since anglers continue to
609 be attracted to particular fisheries even after catch rates have declined because other
610 attributes of the fishery (such as close proximity, social aspects of the experience)
611 provide them with utility, and thereby partly compensate for reduced catch rates. The
612 interesting features of the multi-attribute utility scenario derive from its partial
613 “decoupling” of fish and angler dynamics (Johnson and Carpenter 1994). In contrast,
614 the catch-based scenario is appropriate for describing predator-prey interactions
615 where a predator’s fitness is predominantly dependent on prey consumption. Not
616 accounting for the array of attributes that attract anglers to a fishery may therefore
617 lead to an underestimation of the biological impacts of fishing (Post et al. 2002).
618 Consequently, management decisions based on assumptions of purely catch-based
619 angler behaviour will likely be less conservative than intended with regard to limiting

620 biological impacts, and probably also less successful than intended with regard to
621 angler satisfaction and participation.

622 **Angler heterogeneity**

623 Our results have shown that accounting for the complexity of angler behaviour
624 when predicting the amount of angling effort invested in a particular fishery can
625 fundamentally improve predictions about optimal regulations. However, this
626 improvement alone might not be enough: predictions are likely even more realistic
627 when the heterogeneity of angler behaviour is considered in recreational-fisheries
628 models.

629 We found that, because of the consumptive orientation and aversion to angling
630 regulations of some angler types, minimum-size limits were particularly important in
631 determining angler utility and optimal regulations. Under less restrictive output
632 regulations, consumptive angling effort was reduced, because the fish population
633 could not support large numbers of harvest-oriented anglers while at the same time
634 maintain high catch rates. In these situations, trophy anglers fished in greater numbers
635 than consumptive anglers, because they were less concerned with harvest constraints
636 and more interested in attributes of the fishery unrelated to catch rates. Despite their
637 greater numbers, at low *MSL* levels the less consumptive nature and the reduced
638 catch rates of trophy anglers (which occurred because they used gear that targeted fish
639 of larger size) resulted in them imposing less fishing mortality on a fish stock than
640 consumptive anglers.

641 This demonstrates that both aspects of angler heterogeneity, diversity in
642 angling preferences and differences in fishing practices, are important when
643 determining optimal angling regulations. Furthermore, while managing for angler
644 diversity to enhance the recreational fishing experience of all anglers has been

645 repeatedly called for (Driver et al. 1984; Aas et al. 2000; Arlinghaus and Mehner
646 2004a), our study is the first to explicitly demonstrate the benefits of such an
647 approach when determining optimal, angler-type-specific regulations to maximize
648 social welfare.

649 Although the aim of our modelling exercise was to explore the general
650 importance of behavioural complexity and diversity in anglers, our model-based
651 results also highlight some practical implications. In particular, our model findings
652 suggest that some *MSL* regulations currently used for pike fisheries (45-75 cm in
653 North America; Paukert et al. 2001) are below the optimal levels (53-99 cm) predicted
654 by our model for the different angler types. Implementation of lower-than-optimal
655 minimum size limits could put fish populations at risk of recruitment overfishing.
656 Thus, depending on the composition of the local angler population, special regulations
657 described by Paukert et al. (2001) that are geared toward particular angler types (e.g.,
658 maximum-size limits, inverse slot length limits) may perform better than the standard
659 solution of imposing a moderately low minimum-size limit (such as 45-50 cm).

660 Despite considerable differences among angler types, we found that socially
661 optimal regulations resulted in biologically sustainable exploitation patterns. This is
662 because angler utility is partly dependent on catch-related attributes of the fishery
663 (such as catch rates or fish size), which implicitly requires a productive, biologically
664 sustainable fishery in the long term. Our results therefore indicate that socioeconomic
665 management objectives, such as maximizing social welfare, can account for the state
666 of a fish population through its influence on angler utility and thus provide
667 management advice that results in biologically sustainable exploitation. This supports
668 suggestions for a focus on optimal social yield (OSY) when managing for
669 sustainability (Roedel 1975; Malvestuto and Hudgins 1996; Carpenter and Brock

670 2004). However, the occurrence of optimal regulations in the vicinity of *SPR* levels
671 suggestive of recruitment overfishing varied with angler type. Thus, a precautionary
672 approach has to be taken in socially optimal management, to account for the
673 stochastic processes underlying any fishery.

674 **Angler–population composition**

675 The results discussed so far account for the dynamics and heterogeneity in
676 angler behaviour, they are still limited, in the sense that the angler population was
677 assumed to be composed of just one angler type. In reality, angler populations are
678 composed of different types of anglers that vary in their preferences and behaviour
679 (Hahn 1991; Fisher 1997; Connelly et al. 2001). Our study has shown that this
680 composition affects optimal regulations. Moreover, while, managers might be
681 inclined, for the sake of simplicity, to represent angler populations in terms of an
682 average angler (Hahn 1991; Aas and Ditton 1998), we found that such a simplification
683 can lead to misleading predictions of optimal regulations and biological impacts. This
684 is because different angler types dominated the realized angling effort under different
685 regulations, and because optimal regulations were consistently more restrictive for the
686 mixed angler populations than for the average populations. Shifts in the angling
687 population was also important for determining biological impacts, because of
688 differences in fishing practices and participation of the different angler types.

689 Therefore, our model results underscore the importance of considering not
690 only dynamic angler behaviour and angler heterogeneity in both angling preferences
691 and angling practices in models of recreational-fisheries management (Post et al.
692 2008) , but also how dynamics and diversity interact in angler populations containing
693 a mixture of angler types. Our findings suggest that current monitoring methods that
694 pool information about anglers need to be modified to account for the heterogeneity

695 of angler types using specific fisheries. This will allow managers to understand better
696 which types of anglers are fishing and why (Radomski et al. 2001), thus yielding
697 insights that our model results suggest could be of crucial importance for determining
698 optimal regulations and for more accurately predicting the biological impacts of the
699 angling population.

700 **Social-welfare measures**

701 A final insight from this study relates to the importance of the management
702 objectives determining optimal input and output regulations. From a welfare-
703 economics perspective, the management objective is to maximize the social welfare a
704 fishery provides to the angling community irrespective of which anglers benefit the
705 most or the least (Cole and Ward 1994; Perman et al. 2003). However, our results
706 suggest, that a strictly utilitarian economic approach may alienate some angling
707 groups from a fishery that is managed for maximum total utility. For example, we
708 found that consumptive anglers interested in fish harvest were no longer attracted to a
709 fishery that was subject to restrictive maximum-size limits. Trophy anglers, in
710 contrast, enjoyed high individual utility at high *MSL* levels, mainly because of their
711 lack of consumptive orientation and the greater importance of fishing to their lifestyle.
712 As a result, trophy anglers gained more utility, which strongly influenced the TU
713 social-welfare measure, and thus optimal regulations. Social-welfare measures that
714 reflected more equitable management objectives, such as equitable utilitarian utility
715 (EU) or Rawlsian utility (RU), rendered optimal regulations in mixed angler
716 populations more restrictive, but resulted in a more diverse composition of anglers
717 attracted to a fishery.

718 Thus, although there is no universal consensus about which social-welfare
719 functions to use to quantify welfare (Cole and Ward 1994; Perman et al. 2003), our

720 results illustrate how the optimal regulations predicted by bioeconomic models are
721 sensitive to the social-welfare measures applied. Therefore, managers need to be
722 explicit about their underlying management goals and objectives (Barber and Taylor
723 1990; Aas and Ditton 1998), and ensure that the welfare measure applied closely
724 reflects these objectives, when implementing an OSY approach to recreational-
725 fisheries management.

726 **Limitations and extensions**

727 While we hope that our study provides valuable insights about the importance
728 of angler dynamics and angler heterogeneity when managing for OSY, several
729 limitations need to be highlighted. First, our model results depend on the description
730 of angler behaviour. Application of our modelling approach to local fisheries
731 therefore requires a quantitative assessment of the local and regional angler
732 populations, e.g., using stated and revealed choice models (Hunt 2005; Massey et al.
733 2006). A second limitation is that we assumed that over time, anglers will follow the
734 same behavioural patterns and will keep occurring in the same proportions, which
735 may be in error (Baerenklau and Provencher 2005). Temporal trends in the behavior
736 of individual anglers or in the composition of the angler population could be
737 examined in future extensions of our model. Changing preferences of anglers over
738 time due to specialization or learning, could also be exciting to investigate, as angler
739 will likely adapt to changes in the fishery by altering their expectations (Arlinghaus
740 2006a). Third, to simplify an already complex, model we assumed that participation
741 decisions were made on an annual basis, whereas other time steps may be more
742 realistic (Schuhmann and Schwabe 2004; Hunt 2005). However, because we were
743 interested in long-term equilibrium conditions, our simplifying assumption seems
744 warranted. Fourth, our model described a single fishery and therefore did not account

745 for changes in utility offered by substitute sites in the vicinity of the modeled fishery.
746 Clearly, this is an unrealistic assumption, and further research is needed to broaden
747 our modelling approach to fisheries landscapes (Lester et al. 2003).

748 A final limitation of this study is that we defined social welfare in terms of
749 aggregated utility, rather than aggregated willingness-to-pay. In environmental and
750 resource economics, including recreational-fisheries economics, an aggregate of
751 individuals' willingness-to-pay for an environmental good or service is a commonly
752 used welfare measure (Edwards 1991). In empirical studies of non-marketable goods
753 and services, such as recreational fisheries, this measure of social welfare is calculated
754 using the change in utility provided by attributes of the good (such as catch rate or
755 crowding) from one condition of the fishery to another divided by the marginal utility
756 of income (such as the license cost coefficient in our model) and is expressed in
757 monetary units (Hanemann 1984). Here, we chose not to express utility in monetary
758 units, because this would necessitate making an additional assumption about the
759 baseline condition used for comparison, and because it was felt to be imprudent to put
760 a monetary value on hypothetical scenarios. However, such calculation could be
761 carried out if appropriate empirically derived parameters were available from stated-
762 or revealed-preference models for angler-type-specific part-worth-utility functions
763 (e.g., Massey et al. 2006). This would also ensure that the welfare measure has a
764 cardinal scale avoiding the potential debate of how comparable utility is among
765 individuals (Perman et al. 2003).

766 Despite these limitations, by coupling socioeconomic and biological models
767 our modelling framework is among the few that addresses the often-touted need for an
768 interdisciplinary approach to recreational-fisheries management (e.g., Anderson 1993)
769 (Johnson and Carpenter 1994; Radomski et al. 2001), and provides a basis for future

770 research. There are numerous directions in which our model can be extended,
771 including incorporating environmental stochasticity and a multi-species biology.
772 These extensions are important because deterministic models (Carpenter et al. 1994)
773 and single-species models (Worm et al. 2009) may result in erroneous conclusions
774 about appropriate management strategies. In multi-species models, incorporating
775 angling preferences for different species and indirect effects of angling on the aquatic
776 food webs (Roth et al. 2007) are promising options for complementing the predictions
777 presented here.

778 Further avenues for future research include, exploring the part-worth-utility
779 functions driving angler behaviour, examining the sensitivity of model predictions to
780 changes in fishery attributes, and investigating an even larger numbers of prototypical
781 angler types and their interactions in mixed angling populations. Because multi-lake
782 fisheries opportunities (Parkinson et al. 2004; Post et al. 2008) are more realistic than
783 the simplified single-lake perspective have adopted here, exploration of angler choice
784 within a landscape of fishing opportunities (Carpenter and Brock 2004) may be the
785 most important extension of our modelling approach.

786 **Implications**

787 Even though we have just scratched the surface, we hope that readers share
788 our optimism that the interdisciplinary approach to modeling recreational fisheries
789 introduced here constitutes a sound and extensible theoretical framework. The
790 approach builds on choice theory from welfare economics, angler-specialization
791 theory from leisure sciences and traditional ecological theory, and provides unique
792 insights into recreational-fisheries management.

793 A key finding of this study and related work (Carpenter and Brock 2004) is
794 that “one-size-fits-all” policies are likely to produce suboptimal management

795 outcomes, because they cannot account for the diversity and complexity of angler
796 behaviour that is inherent to most of the world's recreational fisheries (Cox et al.
797 2003; Arlinghaus et al. 2008a; Post et al. 2008). Furthermore, we have shown that
798 misleading predictions about optimal management can result from the omission of
799 dynamic angler behaviour and angler heterogeneity from recreational-fisheries
800 models; this can put fish populations at risk of overfishing, in line with what has been
801 suggested by other studies (Carpenter et al. 1994; Parkinson et al. 2004). In contrast,
802 although managers need to be aware that socially optimal regulations strongly depend
803 on the applied measure of social welfare and the management objectives upon which
804 it is based, managing for socially optimal regulations resulted in both social and
805 biological sustainability.

806 Managers are likely to encounter difficulties in jointly satisfying the interests
807 of the entire angling public. Decisions therefore need to be made about how to best
808 distribute access to scarce resources across angler types (Loomis and Ditton 1993;
809 Daigle et al. 1996). The benefit of an interdisciplinary bioeconomic modelling
810 approach, such as the one presented here, is that it enables managers to quantify
811 welfare changes resulting from alternative management scenarios, and to predict how
812 these regulations will affect different segments of the angling public, as well as the
813 fish population. A decision-support tool such as this one, built on clear objectives and
814 quantitative descriptions, thereby fostering transparency and defensibility in the
815 management process, can facilitate decision taking and clarify when managing for
816 diverse angling opportunities is the best strategy. Ideally, accounting for angler
817 dynamics and angler diversity in fisheries-management models will provide more
818 accurate and realistic predictions of optimal regulations that maximize angler

819 satisfaction, minimize conflicts among angling groups and result in the sustainable
820 management of recreational fisheries.

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1131 Table 1 Model equations. The modelled species was pike (*Esox lucius* L.). Variables,
 1132 parameters, parameter values and their sources are listed in Tables 2. Angler types are
 1133 specified in Table 3.

Number	Equation	Description
<i>Individual-angler utility</i>		
1a	$U_{fj} = U_{cj}$	Conditional indirect utility gained by an angler of type j from choosing to fish in the catch-based scenario only
1b	$U_{fj} = U_{0j} + U_{cj} + U_{sj} + U_{xj} + U_{aj} + U_{ij} + U_{oj}$	Conditional indirect utility gained by an angler of type j from choosing to fish in the static and multi-attribute scenarios
<i>Angler-effort dynamics</i>		
2a	$p_{fj} = \frac{\exp(\hat{U}_{fj})}{\exp(U_n) + \exp(\hat{U}_{fj})}$	Probability that an angler of type j chooses to fish, over the alternative to not fish, where (\hat{U}_{fj}) applies to the previous year
2b	$p_{Fj} = (1 - \phi)p_{fj} + \phi\hat{p}_{Fj}$	Realized probability that an angler of type j chooses to fish, where \hat{p}_{Fj} applies to the previous year
2c	$D_j = p_{Fj}D_{\max}$	Number of days an angler of type j chooses to fish during a year
2d	$A_{Lj} = A_L\rho_j$	Number of licensed anglers of type j
2e	$E_j = D_j A_{Lj} \Psi / \phi$	Total annual realized fishing effort per unit

2f
$$e_{jt} = \begin{cases} E_j / S_F & \text{if } t \leq S_F \\ 0 & \text{if } t > S_F \end{cases}$$
 area of all anglers of type j
Instantaneous fishing effort per unit area at time t of all anglers of type j

Age-structured fish population

3a
$$N_{\text{total}} = \sum_{a=0}^{a_{\text{max}}} N_a$$
 Total fish population density

3b
$$B_{\text{total}} = \sum_{a=0}^{a_{\text{max}}} N_a W_a$$
 Total fish biomass density

Growth

4a
$$h = \frac{h_{\text{max}}}{1 + B_{\text{total}} / B_{1/2}}$$
 Maximum annual growth of a fish dependent on the biomass density at the beginning of the year

4b
$$p_a = \begin{cases} 1 - \frac{G}{3+G}(1 + L_{a0}/h) & \text{if } a \geq a_m - 1 \\ 1 & \text{if } a < a_m - 1 \end{cases}$$
 Proportion of the growing season during which a fish of age a allocates energy to growth

4c
$$g_{at} = \begin{cases} h / S_G & \text{if } t \leq p_a S_G \\ 0 & \text{if } t > p_a S_G \end{cases}$$
 Instantaneous growth rate in length of a fish of age a at time t

4d
$$L_{at} = L_{a0} + g_{at}t$$
 Length of a fish of age a at time t

4e
$$W_{at} = wL_{at}^l$$
 Mass of a fish of age a at time t

Reproduction

5a
$$R_a = \begin{cases} \delta W_a GSI / W_e & \text{if } a \geq a_m \\ 0 & \text{if } a < a_m \end{cases}$$
 Annual fecundity of a female fish of age a

- 5b
$$b = \Phi \sum_{a=a_m}^{a_{\max}} R_a N_a$$
 Annual population fecundity density, pulsed at the beginning of the year
- 5c
$$s_0 = \frac{\alpha}{1 + b / b_{1/2}}$$
 Survival probability from spawning to post-hatch of fish of age zero, applied at the beginning of the year
- 5d
$$N_0 = s_0 b$$
 Density of age zero fish at the beginning of the year
- 6
$$SPR = b_F / b_U$$
 Spawning potential ratio (= relative reduction in egg production under fishing relative to the corresponding unfished condition)

Mortality

- 7a
$$v_{ajt} = [1 - \exp(-y_j L_{at})]^{z_j}$$
 Proportion of fish of age a that are vulnerable to capture by anglers of type j at time t
- 7b
$$c_{ajt} = q_j e_{jt} v_{ajt}$$
 Instantaneous per capita catch rate of fish of age a by anglers of type j at time t
- 7c
$$H_{ajt} = \begin{cases} 1 & \text{if } L_{at} \geq MSL \\ f_{nj} & \text{if } L_{at} < MSL \end{cases}$$
 Proportion of fish at age a that are harvestable by anglers of type j at time t
- 7d
$$C_{jt} = \sum_{a=0}^{a_{\max}} c_{ajt} N_a H_{ajt}$$
 Instantaneous catch rate of harvestable fish by anglers of type j at time t
- 7e
$$C_{Hjt} = \min(C_{jt}, c_{\max j} e_{jt} / \Psi)$$
 Instantaneous harvest rate by anglers of type j at time t

- 7f
$$f_{Hjt} = \frac{C_{Hjt}}{C_{jt}} + f_{hj} \frac{C_{jt} - C_{Hjt}}{C_{jt}}$$
 Proportion of vulnerable harvestable fish killed by anglers of type j at time t
- 7g
$$m_{fajt} = f_{Hjt} c_{ajt} H_{ajt} + f_{hj} c_{ajt} (1 - H_{ajt})$$
 Instantaneous per capita fishing mortality rate of fish of age a imposed by anglers of type j at time t
- 7h
$$d_{at} = m_{na} + \sum_j m_{fajt}$$
 Instantaneous per capita mortality rate of fish of age a at time t
- 7i
$$\frac{dN_a}{dt} = -d_{at} N_a$$
 Continuous rate of change in the density of fish of age a at time t

Social-welfare measures

- 8a
$$U_{TU} = \sum_j U_{fj} D_j A_{Lj}$$
 Annual total utility
- 8b
$$U_{EU} = \sum_j (U_{fj} \rho_j) \sum_j (D_j A_{Lj})$$
 Annual equitable utilitarian utility
- 8c
$$U_{RU} = \min_j (U_{fj}) \sum_j (D_j A_{Lj})$$
 Annual Rawlsian utility

1134

1135

1136 Table 2 Model variables, parameters, parameter values and their sources. The modeled
 1137 species was pike (*Esox lucius* L.). Equations are listed in Table 1. Angler types are
 1138 specified in Table 3.

Symbol	Description (unit, where applicable)	Equation	Value or range	Source
<i>Index variables</i>				
j	Angler type		Generic, consumptive, trophy, or average	
a	Age class (y)		0 - a_{\max}	
a_{\max}	Maximum age of a fish (y)		15	(1)
t	Time within the year (y)		0 - 1	
<i>Angling regulations</i>				
MSL	Minimum-size limit (cm)	7c	0 - 120	
A_L	Number of angling licenses (= number of licensed anglers)	2d	0 - 100	
<i>Angler population</i>				
ρ_j	Proportion of the angler population that is composed of anglers of type j	2d, 8b	Non-mixed: 1.0 for one j ; 0.0 for the others Mixed: (0.4, 0.3, 0.3, 0.0)	
<i>Angler-effort dynamics</i>				
U_n	Conditional indirect utility gained by an angler from choosing not to fish	2a	0	

ϕ	Persistence of fishing behaviour (= the relative influence of last year's realized fishing probability on the current year's realized fishing probability)	2b	0.5	
Ψ	Average time an angler will fish in a day (h)	2e	4	*
D_{\max}	Maximum number of days that an angler would fish per year irrespective of fishing quality	2c	40	*
ϕ	Lake area (ha)	2e	100	
S_F	Annual duration of the fishing season (y)	2f	9/12	
<i>Age-structured fish population</i>				
N_a	Density of fish of age a (ha^{-1})	3a, 3b, 5b, 5d, 7d	0 - ∞	
<i>Growth</i>				
h_{\max}	Maximum growth increment (cm)	4a	24.0	†
$B_{1/2}$	Total fish biomass density at which the growth increment is halved ($\text{kg}^{-1} \cdot \text{ha}$)	4a	100.0	†
G	Annual reproductive investment	4b	0.58	†

a_m	Age at first spawning (y)	4b, 5a	2	(4)
L_{a0}	Length of fish of age a at the beginning of a year (cm)	4b		
L_0	Length of fish at hatch (cm)	4b	0.8	(2)
S_G	Annual duration of the growing season (y)	4c	1.0	
w	Scaling constant for length-mass relationship ($\text{g}\cdot\text{cm}^{-l}$)	4e	0.0048	(6)
l	Allometric parameter for length-mass relationship	4e	3.059	(6)
<i>Reproduction</i>				
GSI	Gonadosomatic index (= gonadic mass/somatic mass)	5a	0.17	(3)
W_e	Average egg mass (g)	5a	0.0050	(3)
δ	Proportion of eggs that hatch	5a	0.75	(4)
Φ	Proportion of female fish in the spawning population	5b	0.5	(5)
α	Maximum proportion of offspring surviving from spawning to post-hatch	5c	$4.75\cdot 10^{-4}$	‡
$b_{1/2}$	Annual population fecundity density at which survival of offspring from	5c	20,325	‡

spawning to post-hatch is halved

(ha)

b_F	Annual population fecundity under fishing	6	$0 - \infty$
b_U	Annual population fecundity under unfished conditions	6	$0 - \infty$

Mortality

m_{na}	Instantaneous natural mortality rate of fish of age a (y^{-1})	7h	0.00 if $a = 0$ 0.42 if $a > 0$	(4)
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1139 Sources: (1) Craig and Kipling 1983; (2) Frost and Kipling 1967; (3) Hubenova et al.

1140 2007; (4) Kipling and Frost 1970; (5) Le Cren et al. 1977; (6) Willis 1989.

1141 * Estimated from average participation rates and average lengths of fishing trips obtained
1142 from diary data of recreational anglers in Mecklenburg-Vorpommern, Germany (Dorow
1143 and Arlinghaus, *unpublished data*) and other literature (van Poorten and Post 2005; Post
1144 et al. 2008).

1145 † Estimated from empirical length-at-age and biomass density data from various pike
1146 studies (Kipling and Frost 1970; Kipling 1983a; Tresurer et al. 1992; Pierce et al. 2003;
1147 Pierce and Tomcko 2003; Pierce and Tomcko 2005) by minimizing the sum of squares
1148 using the ‘solver’ function in Excel (Microsoft® Office Excel 2003).

1149 ‡ Estimated from modified data on female biomass and age-2 abundance in Lake
1150 Windermere (Kipling 1983b). Egg density was determined using the relative fecundity
1151 relationship reported in (Craig and Kipling 1983) and adult biomass (Kipling 1983b), and

1152 natural mortality information from Kipling and Frost (1970) was used to calculate age-1
1153 abundance from age 2 abundance.

1154 Table 3 Angler types and their angling behavior. Parameters describe four angler types (generic, consumptive, trophy, and average) in
 1155 terms of the basic utility they gain from fishing, their tolerances with regard to managerial constraints, their preferences with regard to
 1156 attributes of the fishing experience, and their fishing practices. Parameter values for the average angler type are weighted averages of
 1157 the corresponding parameter values for the three prototypical angler types, weighted by the proportion of each angler type in the
 1158 angler population (0.4 generic; 0.3 consumptive; 0.3 trophy). Parameters values for the angler-type-specific part-worth-utility (PWU)
 1159 functions (Figure 3) were chosen based on assumptions about differences among angler types reported in the angler-specialization
 1160 literature. Figure 1 illustrates qualitative differences in angler preferences, and Figure 3 illustrates the angler-type-specific utility
 1161 functions based on the parameters listed here.

Variable	Symbol and defining equation (affected equation); rationale for general shape (source)	Rationale for angler-type-specific shape (source)	Parameters values describing angler types			
			Generic	Consumptive	Trophy	Average
<i>Importance of fishing to angler lifestyle</i>						
Basic utility gained by an angler of type	U_{0j} (equation 1b); Constant function: the propensity to fish when all	As specialization increases: basic utility of fishing increases (4, 16); the assumed annual participation	Lowest $U_{0j} = -0.405$ (40%	Intermediate $U_{0j} = 0.000$ (50%	Highest $U_{0j} = 0.405$ (60%	$U_{0j} = -0.041$ (49%

j from choosing to fish	other attributes are as expected; see **†‡ for expected values.	is generally consistent with study findings (7, 10).	probability of fishing)	probability of fishing)	probability of fishing)	probability of fishing)
<i>Tolerances with regard to managerial constraints</i>						
PWU of minimum-size limit for an angler of type j	$U_{ij} = u_{1j}r + u_{2j}r^2 + u_{3j}$ (equation 1b), where r is the standardized MSL^* ; Dome-shaped quadratic function: anglers may prefer moderate minimum-size regulations, but object to too low and to too high levels (10, 16, 17).	As specialization increases: anglers become less consumptive and have a greater acceptance of stricter minimum-size regulations (6, 16), but consumptively oriented anglers are averse to harvest regulations that limit their ability to harvest fish (1, 8, 12).	Intermediate $u_{1j} = 2.321$ $u_{2j} = -3.869$ $u_{3j} = 0.271$	Lowest $u_{1j} = 3.766$ $u_{2j} = -9.414$ $u_{3j} = 0.471$	Highest $u_{1j} = 2.534$ $u_{2j} = -2.534$ $u_{3j} = -0.228$	$u_{1j} = 2.819$ $u_{2j} = -5.132$ $u_{3j} = 0.181$
PWU of annual license cost for an angler of	$U_{oj} = u_{4j}o$ (equation 1b), where o is the relative license cost**;	As specialization increases: cost aversion decreases (4, 16).	Lowest $u_{4j} = -0.015$ € ⁻¹	Intermediate $u_{4j} = -0.011$ € ⁻¹	Highest $u_{4j} = -0.008$ € ⁻¹	$u_{4j} = -0.012$ € ⁻¹

type j Linear function: license costs
usually have a negative effect on
angler utility (14, 21).

Preferences with regard to attributes of the fishing experience

PWU of daily catch rate for an angler of type j	$U_{cj} = u_{5j}c_D + u_{6j}c_D^2$ (equations 1a and 1b), where c_D is the relative daily catch rate†;	As specialization increases: focus shifts from quantity to quality and to the challenge of the catch (2, 6, 15).	Intermediate	Highest	Lowest	
			interest in catch	interest in catch	interest in catch	
	Dome-shaped quadratic function: greater utility is gained from increasing catch rates (2, 3, 15), but marginal benefits decrease at high catch rates due to the lack of challenge (1, 2, 9).		Lowest	Intermediate	Highest	
			interest in challenge	interest in challenge	interest in challenge	
			$u_{5j} = 0.968$	$u_{5j} = 1.318$	$u_{5j} = 0.825$	$u_{5j} = 1.030$
			$u_{6j} = -0.121$	$u_{6j} = -0.220$	$u_{6j} = -0.206$	$u_{6j} = -0.176$

PWU of average size of fish captured annually for an angler of type j	$U_{sj} = u_{7j}\bar{l} + u_{8j}$ (equation 1b), where \bar{l} is the relative size of fish caught†; Linear function: anglers have a general preference for catching larger fish (2, 10, 11).	As specialization increases: importance attached to the size of fish increases (2, 6, 10).	Lowest	Intermediate	Highest	
			$u_{7j} = 2.476$	$u_{7j} = 3.389$	$u_{7j} = 4.394$	$u_{7j} = 3.326$
			$u_{8j} = 0.000$	$u_{8j} = 0.000$	$u_{8j} = -0.220$	$u_{8j} = -0.066$
PWU of maximum size of fish captured annually for an angler of type j	$U_{xj} = \begin{cases} u_{9j}l_x^2 & \text{if } l_x \geq 0 \\ -u_{9j}l_x^2 & \text{if } l_x < 0 \end{cases}$ (equation 1b), where l_x is the relative maximum size (= the 95 th percentile in the size distribution of fish caught†); Piecewise quadratic function: increasing when the relative maximum size† is positive and	As specialization increases: utility gained from large-sized fish increases (2, 6, 17), but the least specialized, generic anglers gain more utility than consumptive anglers in the unlikely event that they catch a large fish (8).	Intermediate	Lowest	Highest	
			$u_{9j} = 9.414$	$u_{9j} = 6.878$	$u_{9j} = 12.207$	$u_{9j} = 9.491$

decreasing when it is negative;
 anglers gain greater utility from
 larger fish (18), and the relative
 value of large-sized fish is
 nonlinear (12).

PWU of crowding for an angler of type j	$U_{aj} = u_{10j}A + u_{11j}A^2 + u_{12j}$ (equation 1b), where A is the expected daily congestion ‡; Dome-shaped quadratic function: anglers gain utility from the social aspects of fishing, but avoid congested sites (22).	As specialization increases: desire for solitude increases (6, 7, 22); consumptive anglers recognize that areas with high catch rates will attract other anglers (13).	Highest $u_{10j} = 0.244$ $u_{11j} = -0.031$ $u_{12j} = 0.610$	Intermediate $u_{10j} = 0.149$ $u_{11j} = -0.025$ $u_{12j} = 0.396$	Lowest $u_{10j} = 0.136$ $u_{11j} = -0.034$ $u_{12j} = 0.712$	$u_{10j} = 0.183$ $u_{11j} = -0.030$ $u_{12j} = 0.577$
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Fishing practices

Skill level of an	q_j (equation 7b);	As specialization increases: skill	Lowest	Intermediate	Highest
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angler of type j	Measured in terms of catchability.	level increases (8, 10).	$q_j = 0.011$ $ha \cdot h^{-1}$	$q_j = 0.020$ $ha \cdot h^{-1}$	$q_j = 0.025$ $ha \cdot h^{-1}$	$q_j = 0.018$ $ha \cdot h^{-1}$
Size selectivity for an angler of type j	y_j and z_j (equation 7a) Measured in terms of parameters for the size-dependent vulnerability to capture (modified from 20).	As specialization increases: type of fishing gear used changes (2, 6), and gear used by more specialized anglers catches larger fish (21).	Small $y_j = 0.21$ cm^{-1} $z_j = 406$	Small $y_j = 0.21$ cm^{-1} $z_j = 406$	Large $y_j = 0.21$ cm^{-1} $z_j = 4636$	$y_j = 0.21$ cm^{-1} $z_j = 1675$
Threshold for practicing voluntarily catch-and-release fish for an angler of type j	$c_{max j}$ (equation 7e) Measured in terms of the desired average number of fish an angler will harvest daily.	As specialization increases: propensity to harvest fish decreases (6).	Highest $c_{max j} = 2$	Lowest $c_{max j} = \infty$	Intermediate $c_{max j} = 0.5$	$c_{max j} = \infty$

Hooking mortality for an angler of type j	f_{hj} (equations 7f and 7g) Measured in terms of the proportion of fish dying from hooking mortality.	As specialization increases: no differences in hooking mortality levels (5) were assumed.	$f_{hj} = 0.05$	$f_{hj} = 0.05$	$f_{hj} = 0.05$	$f_{hj} = 0.05$
Non-compliance mortality for an angler of type j	f_{nj} (equation 7c) Measured in terms of the proportion of fish under the minimum-size limit (MSL) that are harvested illegally.	As specialization increases: no differences in non-compliance were assumed; because values reported in the literature vary widely (19, 23, 24), a conservative constant value of 5% was assumed.	$f_{nj} = 0.05$	$f_{nj} = 0.05$	$f_{nj} = 0.05$	$f_{nj} = 0.05$

1162 Sources: (1) Aas and Kaltenborn 1995; (2) Aas et al. 2000; (3) Arlinghaus 2006b; (4) Arlinghaus and Mehner 2004b; (5) Arlinghaus et
1163 al. 2008c; (6) Bryan 1977; (7) Connelly et al. 2001; (8) Dorow et al. 2010; (9) Fedler and Ditton 1994; (10) Fisher 1997; (11) Gillis
1164 and Ditton 2002; (12) Jacobson 1996; (13) Martinson and Shelby 1992; (14) Massey et al. 2006; (15) Oh and Ditton 2006; (16) Oh et

1165 al. 2005a; (17) Oh et al. 2005b; (18) Paulrud and Laitila 2004; (19) Pierce and Tomcko 1998; (20) Post et al. 2003; (21) Rapp et al.
1166 2008; (22) Schuhmann and Schwabe 2004; (23) Sullivan 2002; (24) Walker et al. 2007.

1167 * $r = MSL / L_{\max}$ is the relative minimum-size limit, standardized to range between 0 and 1, where L_{\max} is the maximum size that a
1168 fish can attain at the maximum age allowed in the absence of density dependence (equations 4a-d).

1169 ** $o = (O_o - O_e)$ is the annual fishing-license cost relative to a baseline expected value, where O_o and O_e are the observed and
1170 expected values, respectively.

1171 † Attributes related to the fish population represent the proportional difference scaled relative to a baseline expected value as follows:

1172 $c_D = C_{D_o} / C_{D_e} - 1$, where C_{D_o} and C_{D_e} , respectively, are the observed and expected average daily catch rates; $\bar{l} = \bar{L}_o / \bar{L}_e - 1$, where

1173 \bar{L}_o and \bar{L}_e , respectively, are the observed and expected average sizes of caught fish in a year; $l_x = L_{x_o} / L_{x_e} - 1$, where L_{x_o} and L_{x_e} ,

1174 respectively, are the observed and expected the maximum sizes of caught fish in a year (with the latter defined as the 95th percentile of

1175 the size distribution of caught fish). Expected values are based on the literature and on unpublished data from pike fisheries. We

1176 assumed an expected daily catch rate of 0.5 fish (Kempinger and Carline 1978; Goeman et al. 1993; Arlinghaus et al. 2008c) and that

1177 anglers fished 4 h in an angling day, an expected average size of 51 cm (Kempinger and Carline 1978; Pierce et al. 1995 (harvested

1178 fish); Arlinghaus et al. 2008c), and an expected average maximum size of 69 cm (Dorow and Arlinghaus, *unpublished data*).

1179 ‡ $A = \sum_j (D_j A_{Lj}) / (365 S_F)$ is the expected average number of anglers fishing in a day (see equations 2c-d).

1180

1181 Table 4 Predicted optimal regulation and their implications. Optimal input and output
 1182 regulations maximized social welfare for various angler types and for different
 1183 assumptions about angler behaviour and social-welfare measures. Implications are
 1184 shown in terms of resulting angling efforts and biological impacts (with the latter
 1185 being measured by the spawning-potential ratio *SPR*). Three social-welfare measures
 1186 were examined for the mixed angler population: total utility (TU), an equitable
 1187 utilitarian utility (EU) and a Rawlsian utility (RU) (Table 1, equations 8 a-c). For the
 1188 non-mixed angler populations, results for the EU and R were identical to those for TU
 1189 and are therefore not repeated.

Scenario	Angler population				
	Generic	Consumptive	Trophy	Average	Mixed
<i>Optimal minimum-size limit (cm)</i>					
Static – TU	80	53	99	69	69
Catch-based – TU	104	102	101	106	98
Multi-attribute – TU	80	53	99	69	93
(EU; RU)					(69; 63)
<i>Optimal angler-license number</i>					
Static – TU	38	27	36	31	36
Catch-based – TU	92	100	99	100	100
Multi-attribute – TU	52	36	39	44	66
(EU; RU)					(48; 48)
<i>Annual realized angling effort under optimal regulations (h•ha⁻¹)</i>					
Static – TU	61	43	58	50	58
Catch-based – TU	80	112	93	94	97
Multi-attribute – TU	61	43	58	50	65

(EU; RU) (57; 57)

Composition of anglers fishing in the mixed angler population under optimal regulations

Static – TU	0.40	0.30	0.30	n.a	n.a
Catch-based – TU	0.34	0.37	0.29	n.a	n.a
Multi-attribute – TU	0.41	0.14	0.45	n.a	n.a

(EU; RU) (0.38; 0.37) (0.27; 0.29) (0.35; 0.34)

Spawning-potential ratio under optimal regulations

Static – TU	0.74	0.38	0.73	0.61	0.57
Catch-based – TU	0.78	0.54	0.61	0.67	0.63
Multi-attribute – TU	0.74	0.39	0.73	0.61	0.73

(EU; RU) (0.57; 0.48)

1190

1191

1192 **Figure captions**

1193 **Figure 1** Simplified flow diagram illustrating interactions among the three model
1194 components of our bioeconomic modelling approach: the biological component, the
1195 socioeconomic component, and the management component. The model included
1196 three angler-behavior scenarios: (a) static angler behavior, where anglers fish at the
1197 maximal rate; (b) catch-based dynamic angler behavior, where anglers responded to
1198 the fishery based on catch rates; (c) multi-attribute dynamic angler behavior, where
1199 anglers responded to the fishery based on a multi-attribute utility function. Black,
1200 solid arrows depict influences that apply across all scenarios, while gray arrows apply
1201 to the catch-based scenario only and black dashed arrows apply to either the static or
1202 multi-attribute scenarios as is also indicated by labels along the arrows. Factors in
1203 round-cornered boxes dynamically change throughout model runs, while parameters
1204 for factors in square-cornered boxes were held constant.

1205

1206 **Figure 2** Qualitative differences in angler preferences for fishery attributes among
1207 the three different prototypical angler types (generic, consumptive, and trophy
1208 anglers). Gray circles indicate the relative preference levels or tolerance levels (low,
1209 intermediate, or high) of angler types for a particular fishery attribute.

1210

1211 **Figure 3** Part-worth-utility functions describing the preferences of generic,
1212 consumptive, trophy and average anglers for various attributes of the fishery.

1213

1214 **Figure 4** Total utility (TU) over a range of input (license number) and output
1215 (minimum-size limit) regulations. Columns illustrate results for three angler-
1216 behaviour scenarios (left column: static angler behaviour, where anglers fished at the

1217 maximal rate; middle column: catch-based dynamic angler behaviour, where anglers
1218 responded to the fishery based on catch rates; right column: multi-attribute dynamic
1219 angler behaviour, where anglers responded to the fishery based on a multi-attribute
1220 utility function). Rows illustrate results for five different angler populations (first row:
1221 generic anglers; second row: consumptive anglers; third row: trophy anglers; fourth
1222 row: average anglers; and fifth row: mixed angler population composed of 40%
1223 generic, 30% consumptive, and 30% trophy anglers). Blue diamonds indicate the
1224 optimum regulations at which total utility was maximized.

1225

1226 **Figure 5** Spawning-potential ratio (*SPR*) of fished populations over a range of input
1227 (license number) and output (minimum-size limit) regulations. *SPR* values below
1228 0.35-0.4 indicate a potential for recruitment overfishing. Columns show results for
1229 three angler-behavior scenarios (left column: static angler behaviour, where anglers
1230 fished at the maximal rate; middle column: catch-based dynamic behaviour, where
1231 anglers responded to the fishery based on catch rates; right column: multi-attribute
1232 dynamic behaviour, where anglers responded to the fishery based on a multi-attribute
1233 utility function). Rows show results for five different angler populations (first row:
1234 generic anglers; second row: consumptive anglers; third row: trophy anglers; fourth
1235 row: average anglers; fifth row: mixed angler population composed of 40% generic,
1236 30% consumptive, and 30% trophy type anglers). Blue diamonds indicate the
1237 optimum regulations at which total utility was maximized.

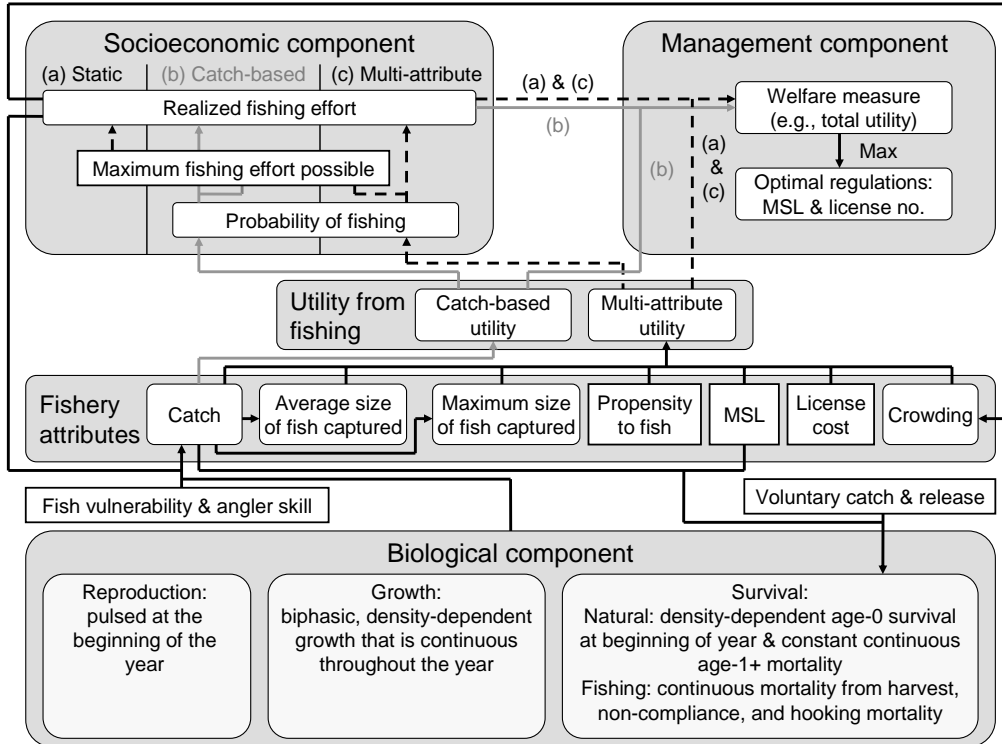
1238

1239 **Figure 6** Proportion of the total realized angling effort contributed by each angler
1240 type in a mixed angler population over a range of input (license number) and output
1241 (minimum-size limit) regulations. The mixed angler population was composed of

1242 40% generic, 30% consumptive, and 30% trophy type anglers. Anglers responded to
1243 the fishery based on a multi-attribute utility function; see (o) panels in Figures 4 and
1244 5. Blue diamonds indicate the optimum regulations at which total utility was
1245 maximized.

1246

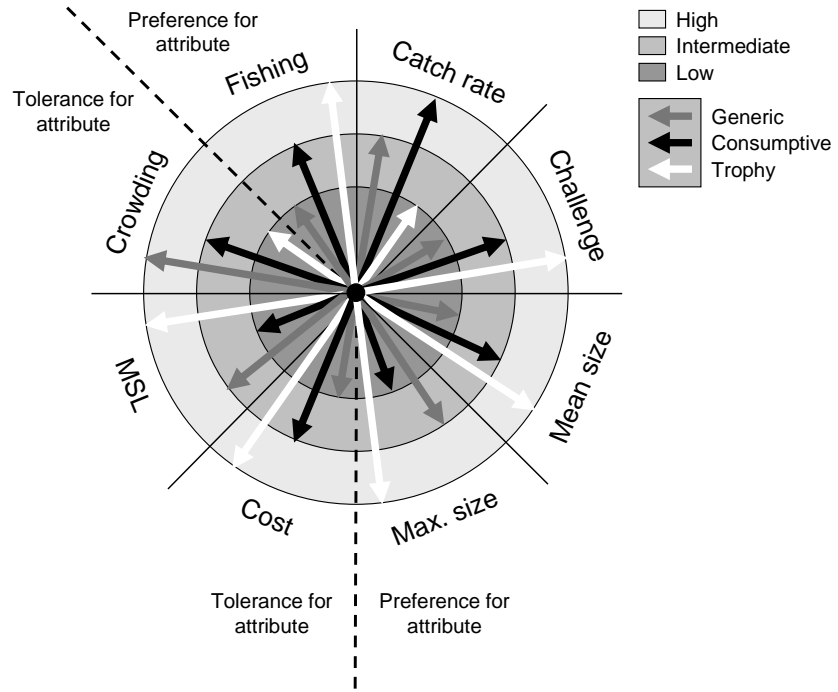
1247 **Figure 7** Social-welfare measures in a mixed angler population with multi-attribute
1248 dynamic angler behavior over a range of input (license number) and output
1249 (minimum-size limit) regulations. The mixed angler population was composed of
1250 40% generic, 30% consumptive, and 30% trophy anglers. Results are shown for three
1251 social-welfare measures (total utility, TU;, egalitarian utilitarian utility, EU; Rawlsian
1252 utility, RU; see Table 1, equations 8a-c). Blue diamonds indicate the optimum
1253 regulations at which the social-welfare measures were maximized.



1254

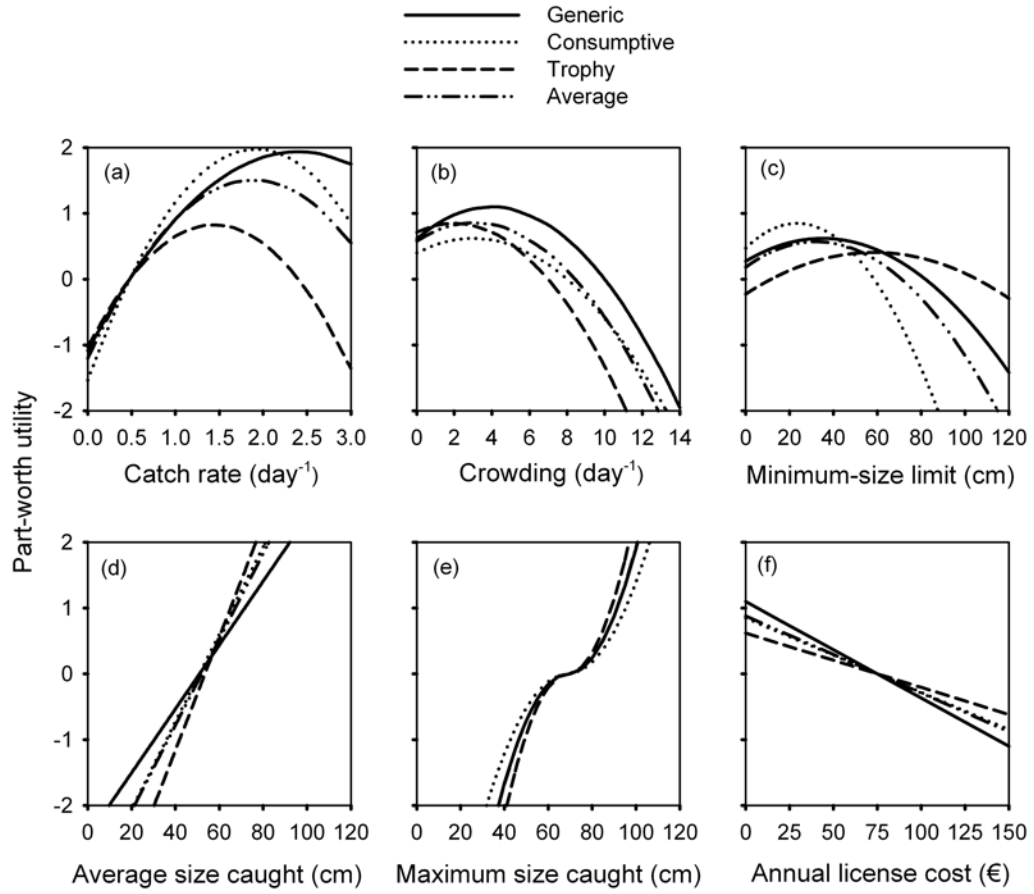
1255 **Figure 1**

1256



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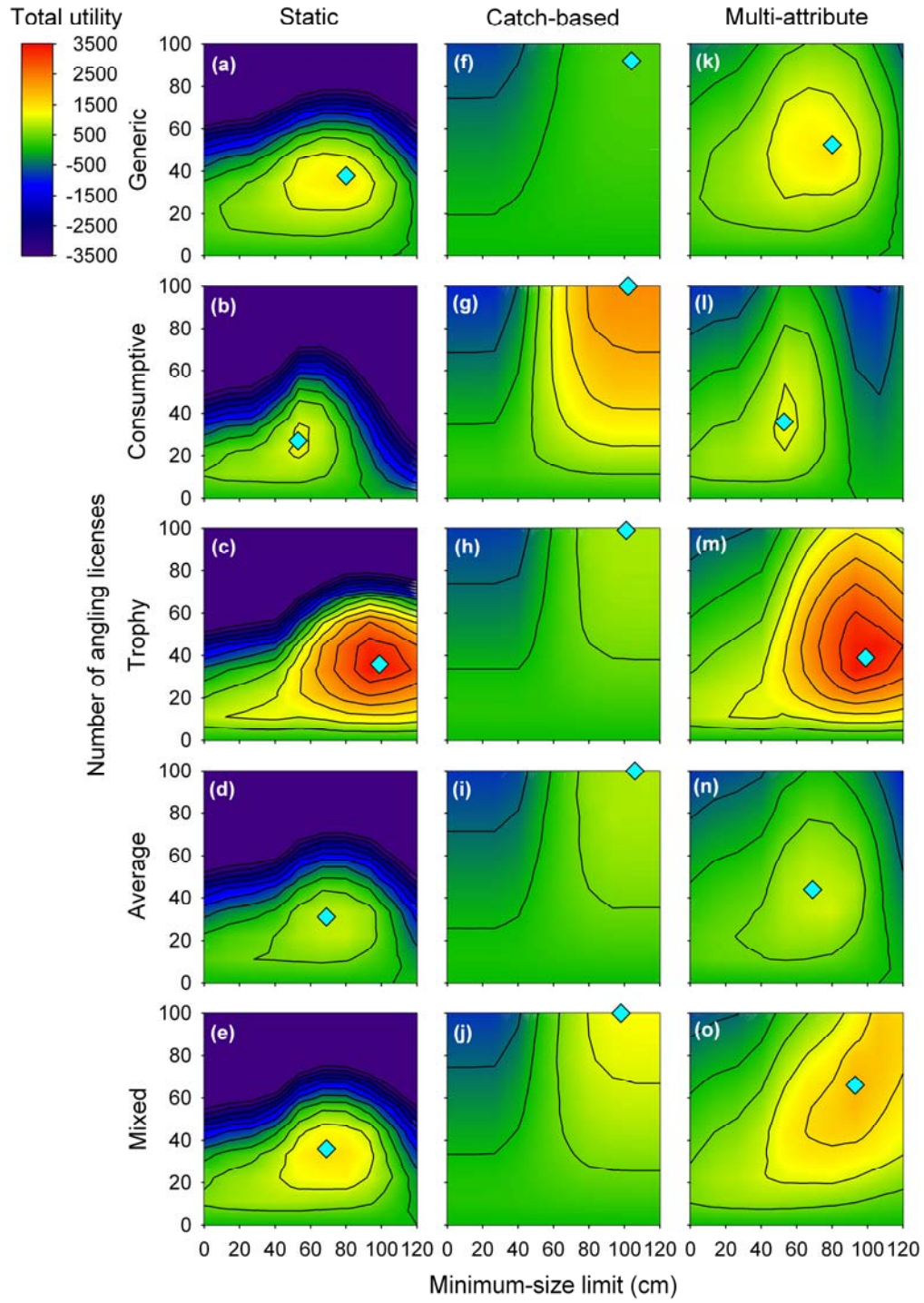
1258 **Figure 2**



1259

1260 **Figure 3**

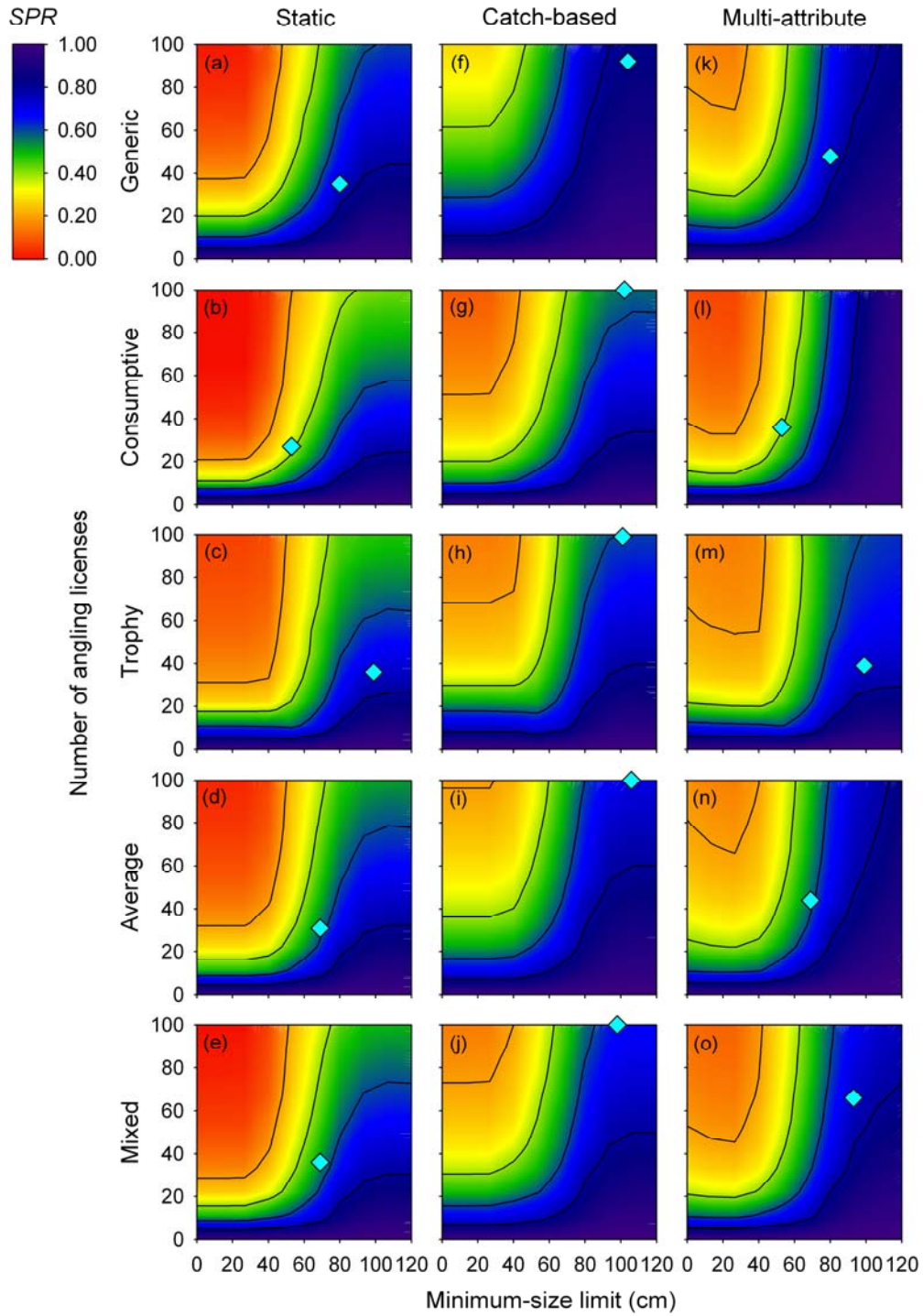
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1263 **Figure 4**

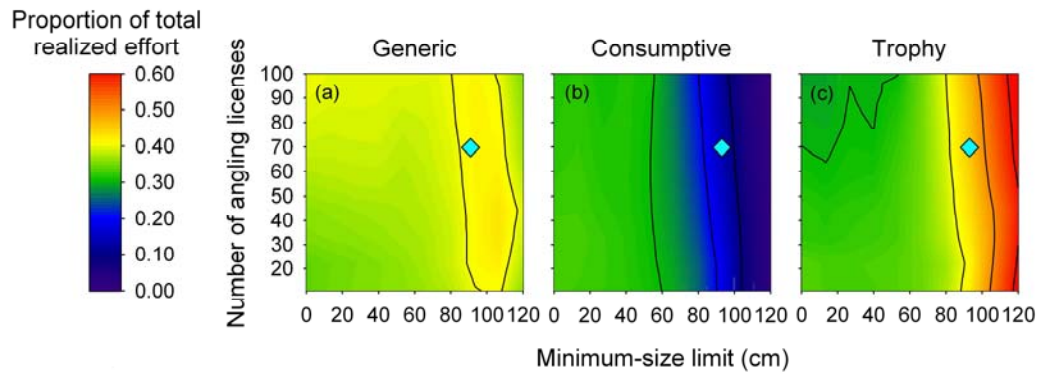
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1265

1266 **Figure 5**

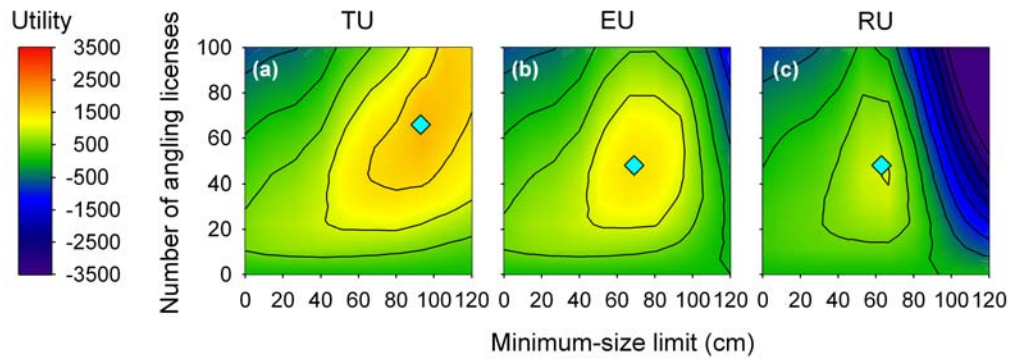
1267



1268

1269 **Figure 6**

1270



1271

1272 **Figure 7**

1273

1274 **Appendix A** Sensitivity of predicted optimal regulations to fishery attributes
 1275
 1276 Table A1 Sensitivity of predicted optimal regulations, and of the conditions that occur
 1277 under these regulations, to the removal of single fishery attributes from the multi-
 1278 attribute utility function (Table 1, equation 1b). Results shown are for the multi-
 1279 attribute scenario, assuming total utility (TU) as the maximized social-welfare
 1280 measure. Parentheses show changes relative to results for the multi-attribute scenario
 1281 with all fishery attributes being included (Table 4).

Removed attribute	Angler population				
	Generic	Consumptive	Trophy	Average	Mixed (TU)
<i>Optimal minimum-size limit (cm)</i>					
Minimum-size limit	104 (+30.0%)	103 (+94.3%)	104 (+5.1%)	105 (+52.2%)	99 (+6.5%)
Crowding	60 (-25.0%)	51 (-3.8%)	96 (-3.0%)	50 (-27.5%)	99 (+6.5%)
Catch	51 (-36.3%)	23 (-56.6%)	100 (+1.0%)	52 (-24.6%)	93 (0.0%)
Average size	55 (-31.3%)	53 (0.0%)	101 (+2.0%)	61 (-11.6%)	61 (-34.3%)
Maximum size	62 (-22.5%)	52 (-1.9%)	86 (+13.1%)	69 (0.0%)	69 (-25.8%)
<i>Optimal angler-license number</i>					
Minimum-size limit	49 (-5.8%)	50 (+38.9%)	41 (+5.1%)	45 (+2.3%)	53 (-19.7%)
Crowding	20 (-61.5%)	31 (-13.9%)	88 (+125.6%)	12 (-72.7%)	100 (+51.5%)
Catch	56 (+7.7%)	40 (+11.1%)	42 (+7.7%)	47 (+6.8%)	75 (+13.6%)
Average size	55 (+5.8%)	44 (+22.2%)	42 (+7.7%)	48 (+9.1%)	46 (-30.3%)
Maximum size	51 (-1.9%)	39 (+8.3%)	44 (+12.8%)	44 (0.0%)	50 (-24.2%)
<i>Annual realized angling effort under optimal regulations (h•ha⁻¹)</i>					
Minimum-size limit	61 (0.0%)	67 (+55.8%)	60 (+3.4%)	61 (+22.0%)	68 (+4.6%)
Crowding	19 (-68.9%)	33 (-23.3%)	114 (+96.6%)	13 (-74.0%)	70 (+7.7%)

Catch	63 (+3.3%)	44 (+2.3%)	59 (+1.7%)	49 (-2.0%)	64 (-1.5%)
Average size	64 (+4.9%)	55 (+27.9%)	59 (+1.7%)	53 (+6.0%)	57 (-12.3%)
Maximum size	58 (-4.9%)	46 (+7.0%)	61 (+5.2%)	49 (-2.0%)	59 (-9.2%)

Composition of anglers fishing in the mixed angling population under optimal regulations

Minimum-size limit	0.35 (-14.6%)	0.31 (+121.1%)	0.34 (-24.9%)	n.a.	n.a.
Crowding	0.31 (-23.8%)	0.09 (-38.7%)	0.60 (+34.1%)	n.a.	n.a.
Catch	0.45 (+8.6%)	0.06 (-55.6%)	0.49 (+9.7%)	n.a.	n.a.
Average size	0.38 (-7.2%)	0.30 (+111.2%)	0.32 (-28.6%)	n.a.	n.a.
Maximum size	0.38 (-7.9%)	0.27 (+91.6%)	0.35 (-21.8%)	n.a.	n.a.

Spawning-potential ratio under optimal regulations

Minimum-size limit	0.83 (+11.7%)	0.68 (+77.0%)	0.73 (-0.6%)	0.76 (+25.7%)	0.72 (-1.2%)
Crowding	0.76 (+2.2%)	0.42 (+10.0%)	0.56 (-23.1%)	0.66 (+9.3%)	0.71 (-2.3%)
Catch	0.42 (-43.8%)	0.13 (-65.6%)	0.72 (-0.7%)	0.38 (-37.3%)	0.74 (+0.8%)
Average size	0.43 (-41.8%)	0.34 (-12.5%)	0.72 (-0.9%)	0.49 (-18.5%)	0.48 (-34.7%)
Maximum size	0.56 (-24.5%)	0.37 (-3.9%)	0.68 (-7.2%)	0.61 (+0.2%)	0.57 (-22.3%)