

Estimating the economic loss of recent North Atlantic fisheries management

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Abstract:

It is accepted that world's fisheries are not generally exploited at their biological or their economic optimum. Most fisheries assessments focus on the biological capacity of fish stocks to respond to harvesting and few have attempted to estimate the economic efficiency at which ecosystems are exploited. The latter is important as fisheries contribute considerably to the economic development of many coastal communities. Here we estimate the overall potential economic rent for the fishing industry in the North Atlantic to be B€ 12.85, compared to current estimated profits of B€ 0.63. The difference between the potential and the net profits obtained from North Atlantic fisheries is therefore B€ 12.22. In order to increase the profits of North Atlantic fisheries to a maximum, total fish biomass would have to be rebuilt to 108 Mt (2.4 times more than present) by reducing current total fishing effort by 53%. Stochastic simulations were undertaken to estimate the economic loss NA fisheries in a range of 2.5 and 32 billion of euro. We provide economic justification for maintaining or restoring fish stocks to above their MSY biomass levels. Our conclusions are consistent with similar global scale studies.

Highlights

► The economic loss of North Atlantic fisheries in 2010 was B€ 12.2. ► Fish stocks would have to recover to levels 2.4 times larger than present to achieve maximum profits. ► Macroscale assessments of governance require simplifications and assumptions. ► Stochastic simulations estimate NA fisheries loss in a range from B€2.5 to B€32. ► Securing that stocks biomass are above MSY is economically justified despite uncertainty on results.

Keywords : Economic assessment ; Fisheries, North Atlantic ; Management efficiency ; Uncertainty

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56 Marine fisheries are an important source of food and livelihood opportunities worldwide 57 (Allison et al., 2009; Garcia & Rosenberg, 2010; Rice & Garcia, 2011). The exploitation state of 58 fish stocks is hotly debated (Branch et al., 2011; Pauly et al., 2002; Worm et al., 2009), but there 59 is a general consensus that marine fisheries food production potential is not achieved (Branch et 60 al., 2011; FAO, 2012). North Atlantic fisheries are nowadays yielding less fish than in recent 61 decades and despite significant improvements (Fernandes and Cook 2013), the state of many of 62 its stocks remains poor. Traditionally, the efficiency of biomass production has been the basis of 63 fisheries management. Therefore, different regulations have aimed at maintaining fish stocks at 64 levels at which they could produce their Maximum Sustainable Yield (MSY), i.e. the exploitation 65 rate where the response of the stocks to fishing through individual growth and recruitment 66 operates at its maximum capacity. In a deterministic sense, at this level, average fish biomass 67 remains stable over time and the amount of fish that can be sustainably extracted is maximized 68 (Schaefer, 1954). Classic approaches assume that these dynamics operate at a particular stock 69 level, depending on the species' life history and thus, should fisheries management succeed in 70 maintaining each of them at their MSY, the maximum potential of food production from marine 71 ecosystems would be achieved. Using Economic Exclusive Zone (EEZ) and fish species data 72 from the Sea Around Us database, the food production potential wasted due to ineffective 73 management was estimated, i.e., the difference between catch observations and their MSY 74 estimated from historic catch series (Srinivasan et al., 2010). Srinivasan et al. (2010) estimated 75 that catch losses amounted to 7-36% of the reported annual catch, resulting in a landed value 76 loss between \$6.4 billion and \$36 billion.

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78 In reality, it is ecologically impossible to simultaneously maximize sustainable yield for all 79 species in a multiple species fishery (Link, 2009). Therefore, the productivity of marine 80 ecosystems is expected to be lower than predicted by the sum of single stocks' MSY (Link et 81 al., 2012). The overall productivity and state of exploitation of marine ecosystems have been 82 investigated previously with complex ecosystem models and indicators (Blanchard et al., 2012; 83 Blanchard et al., 2009; Coll et al., 2008; Cury et al., 2008; Merino et al., 2012; Shin et al., 84 2005), and with single species models applied to entire exploited communities (Guillen et al., 85 2013; Link et al., 2012; Mueter & Megrey, 2006; Sparholt & Cook, 2009; Worm et al., 2009). 86 For example, 'surplus production models' (SPM), have been used to produce simple 87 representations of the key ecological processes underlying fisheries (Link et al., 2012). SPM can

88 be used to estimate biological reference points (BRP's) such as the biomass level and the rate of

89 exploitation to achieve the MSY of single fish stocks or marine ecosystems.

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91 SPM have allowed the extension of fisheries assessment into other disciplines beyond ecology. 92 For example, the seminal paper by Gordon (1954) introduced the concept of Maximum 93 *Economic Yield (MEY)*, the bioeconomic reference point at which the economic profits of a 94 fishery are maximized. This concept relies on fish stocks' productivity described by SPM 95 (Schaefer, 1954), the market price of fish and the costs of fishing. A derivation of this model 96 was used to assess the economic efficiency at which the world's fisheries are exploited 97 (Arnason et al., 2009), from which global MEY was estimated based on world's catch, value 98 and costs databases. Arnason et al (2009) highlight the vast economic consequences of 99 inefficient fisheries management and the economic benefit of maintaining fish stocks at healthy 100 levels. Due to the high uncertainty in the data and the simplified model used, the numeric results 101 of Arnason et al (2009) study were presented with caution and with wide confidence intervals. 102 Nonetheless, the global cost of sub-optimal management was estimated to be in a range between 103 \$37-67 billion in 2004, with an historic accumulated loss of \$2.2 trillion between 1974 and 104 2004. Arnason et al. (2009) did not explicitly evaluate the cost of rebuilding fish stocks, i.e., the 105 cost of the necessary transition until stocks are recovered and more economic profit is obtained 106 with less fishing effort. More recent research shows that the benefit of rebuilding global 107 fisheries outweighs costs (Sumaila et al., 2012) and that investing in restoring overexploited 108 stocks is economically sound (Crilly & Esteban, 2012). However, it is important to clarify that 109 not all fish stocks are overexploited. For example, forty-three percent of assessed EU stocks 110 were considered overfished in 2012 (Fernandes & Cook, 2013; European Union 2012). In any 111 case, when fishing yields do not correspond to MSY this does not automatically mean a stock is 112 overfished (Hilborn & Stocks, 2010). Hilborn and Stokes (2010) suggest that it would be 113 reasonable to adopt a definition of being overfished as any stock size where the expected yield 114 is 80% or less than MSY, which is the level at which reductions of fishing mortality towards 115 MSY would produce measurable catch increases.

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The North Atlantic basin is a dynamic environment for physical and biological processes 117 118 (Beaugrand et al., 2002; Marshall et al., 2001; Parsons & Lear, 2001) and is home to some of 119 the largest populations of commercially exploited stocks (Trenkel et al., 2013). With this at the 120 background and due to the importance of North Atlantic global climate, BASIN (Wiebe et al., 121 2009) is a joint EU/North American research initiative with the goal of elucidating the 122 mechanisms uderlying observed changes in the North Atlantic ecosystems and their services, 123 and Euro-BASIN is a programme to implement this vision funded by the European Commission 124 7th Framework Programme (St. John et al. *introduction article of this issue*). In the context of

Euro-BASIN, this article aims to reflect the economic relevance of fisheries within the North Atlantic basin using some of the methods described above to estimate the economic cost of ineffective fisheries management, defining 'ineffective' as a deviation from maximum economic rent (Arnason et al., 2009). To do so, we tested alternative aggregations of fisheries production and economic indicators and parameterized a simple bioeconomic model. The scope and scale of this study is vast and complex and requires simplifications. The ecological complexity, regional differences and dynamics of individual fish stocks in the North Atlantic are simplified in an aggregated single stock of fish, which is exploited by an aggregated single fishery. While this approach has significant ecological difficulties, aggregated fisheries production functions are not new, and have been used to assess the economic efficiency of global fisheries as a single exploited unit (Arnason et al., 2009), at ecosystem level (Crilly & Esteban, 2012; Link et al., 2012; Sparholt & Cook, 2009) and at species-EEZ level (Srinivasan et al., 2010). The implications of this approach and justification for the use of an aggregated model will be discussed in detail throughout the manuscript. Furthermore, we explore the possible impact of parameter uncertainties and the assumptions made to obtain our numeric results. Finally, we discussed the use of multidisciplinary approaches in analyzing marine resources at the basin scale. These results provide background context to the work conducted in Euro-BASIN in the Bio-economic modeling (WP7) and Living resources (WP5) workpackages.

- **2.** Material and Methods
- *2.1. The data*

Biological parameters: Catch data from ICES FishStatPlus database (www.ices.dk), FAO Fishery Statistics (www.fao.org) and Sea Around Us catch database (www.seaaroundus.org) were used to estimate the biological parameters of the surplus production model of the North Atlantic (NA) fisheries from 1950 to 2010. The data used comprise 59 ICES stocks, 18 species and 2 habitats exploited in the North Atlantic for the ICES area (see Table 1). These data were used to explore how alternative levels of stock and taxonomic aggregation could lead to different MSY estimates and indicate the uncertainty that the aggregation process undergone for the NA bioeconomic model. The overall NA basin MSY was estimated using datasets from FAO and was used as input for the bioeconomic model. A series of all the species landed in the NA was used.

Economic parameters: Three main sources of information were used to obtain the
economic parameters of the NA fisheries. First, the Sea Around Us database was used

to obtain the value of the NA fishery as a whole. Second, a global fishing costs database
at fleet segment level (Lam et al., 2011) was used to estimate countries total profits
(Table 2). These estimates showed significant differences with the ones reported in the
Annual Economic Report of the European Fishing Fleets (JRC, 2012).

- 167 *2.2 The models*
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169 Biological parameters estimation: We used a relatively simple method to obtain 170 plausible MSY estimates and other biological parameters from catch data, based on 171 assumptions on resilience (corresponding to the intrinsic growth rate r in the SPM) and 172 the plausible range of relative stock sizes at the beginning of the time series (Martell & 173 Froese, 2010). We used a medium resilience range as defined by Martell & Froese, i.e. 0.2<r<1, and an initial (in 1950) relative stock size range of 50-90% of carrying 174 175 capacity K or pristine biomass for all stocks (except for 'ghl-arct' ICES stock which 176 was considered of 'low' resilience, 0.05<r>0.5), and all species, habitat and the total 177 NA. The identification of pairs of r-K values compatible with the catch time series and 178 the above assumptions was performed using the R-code for batch processing made 179 publicly available in http://www.fishbase.de/rfroese/CatchMSY_2.r for 59 ICES stocks 180 ('ICESct2.csv', catch file processed and also made available by Martell and Froese), for 181 the 18 species targeted in the ICES areas and for the entire NA basin from FAO catch 182 data. The aggregation was a simple summation of catches of all the stocks of each of the 183 18 species, of all demersal and pelagic species and of all ICES stocks. Similarly, for the 184 NA estimation, all NA species catches were summed to obtain a single catch time 185 series. For each plausible r-K pair, an estimate is obtained as MSY=1/4 r K. This MSY estimation algorithm has been validated against analytical fish stock assessment 186 187 estimates of MSY (Martell & Froese, 2010). Good agreement was found between stock 188 assessment MSY estimates and the geometric mean of MSY values calculated from the 189 plausible r-K pairs (Martell & Froese, 2010).

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- Aggregated bioeconomic model (Arnason, 2007; Arnason et al., 2009). This model assumes that the stocks exploited by global fisheries can be modeled as a single fish stock with an aggregate biomass growth function and a fishing industry operating exclusively in the area. The economic performance from fisheries is estimated with the value of the global landings calculated with an aggregated harvest function (SPM by Schaefer 1954) and an aggregated fishing cost function relating current fishing effort to fisheries costs. Incorporating NA fisheries into a single fishery allows for a model with a manageable number of parameters. This model requires 4 biological parameters: (i)

199 Global MSY, (ii) total 'carrying capacity' or 'unexploited biomass level', (iii) fish 200 biomass growth in the last year and, (iv) a 'schooling' parameter; and 5 economic 201 parameters from the fishing industry: (i) Landings, (ii) value of landings and, (iii) total 202 profits from fishing in the last year, (iv) 'fixed costs ratio' and, (v) 'elasticity of 203 demand' with respect to total biomass. The MSY has already been explained and 204 'unexploited biomass', fish biomass growth, landings, value of landings and profits are 205 self-explanatory. The 'schooling' parameter describes the spatial distribution behavior 206 of fish and ranges between 0 and 1. The lower the parameter the more aggregated the fish, e.g., small pelagic stocks like anchovies, sardines, mackerel etc. When this 207 parameter is close to 1, fish are homogeneously distributed in space, e.g., demersal 208 209 species such as hake or plaice. For our analysis we fixed this parameter as 1, to assume 210 that all fish are homogeneously distributed throughout the NA. However, the impact of this parameter on the final calculations is explored in the Appendix. The 'fixed costs 211 212 ratio' describes the fraction of the total costs incurred by the fishing industry that are 213 not originated by labor, fuel, capital and other factors of production such as 214 maintenance, repair, supplies and gear costs. We considered this ratio to be 0 as in the 215 global study (Arnason et al., 2009). Assuming a zero value means that fishing effort is 216 measured as the size of fishing industry and not by its activity (if inactive, fleets would 217 still generate fixed fishing costs). The elasticity of demand to biomass expresses the 218 price of fish as dependent on the global marine commercial fish biomass. This elasticity 219 is positive: when there is overexploitation, biomass is at a low level, the proportion of 220 low value fish is higher and mean price is smaller; in the other direction, if fish stocks 221 recover from overexploitation, the average size and trophic level of caught fish 222 increased and price does so likewise. This is a manifestation of the "fishing down the 223 food web" effect (Pauly et al 1997).

224 The bioeconomic model also required additional input that was obtained as follows: 225 from the FAO catch data series we obtained catches in 2010 (Y_{2010}) and biomass growth 226 in 2010 (G(x)₂₀₁₀). We used the difference between Y_{2010} and Y_{2009} as G(x)₂₀₁₀ which 227 assumes that catch changes were only caused by abundance changes rather than 228 management or other factors. For the value of catches, the Sea Around Us database was 229 used to complement FAO data (Value₂₀₁₀). The profits of NA fisheries in 2010 was 230 obtained summing the national profits (Table 2) obtained from Lam et al (2011) applied 231 to NA value of catch.

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233 (# Insert Table 2 #)

We plot the classical equilibrium catch-biomass curve (Schaefer, 1954) and different potential profit curves ($iso_{-}\Psi$) defined by the profits in 2010 (Ψ_{2010}), price at equilibrium (p_{eq}), costs of fishing per unit of effort (c) and biomass at equilibrium (equation 1).

239 $iso_{\Psi} = \frac{\Psi_{2010}}{p_{eq} - c \cdot B_{eq}}$ (equation 1)

The points where $iso_{-}\Psi$ trajectories meet the catch and biomass equilibrium curve are a feasible sustainable profit, catch and biomass equilibrium points. The maximum feasible $iso_{-}\Psi$ is searched to identify the *MEY* of North Atlantic fisheries. Further transformations of the basic equations by Gordon-Schaefer required to plot the curves are explained in the Appendix. Economic loss is then calculated as the difference between this MEY value and realized profits in 2010.

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247 Stochastic simulations: The economic loss estimated with the deterministic model for 2010 248 was re-estimated allowing for uncertainty on the input parameters (Table 3): (i) random values 249 in a range of $\pm 30\%$ of the initial parameters ('sim 1'). (ii) lognormal distributions for MSY and 250 K as provided by the Martell and Froese (2010) estimation model and random for the others 251 ('sim 2'), (iii) lognormal distributions for MSY and K and random for Catch 2010 ('sim 3') and, 252 (iv) random within $\pm 30\%$ for all parameters but MSY, K and Catch 2010 which were kept constant ('sim 4'). For all stochastic simulations the model was run for 10⁵ iterations to 253 254 equilibrium.

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3. Results

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The total MSY for all the ICES stocks combined was estimated to be between 6.68 and 9.75 million tonnes, depending on the level of catch aggregation from which the estimates were calculated (Table 1 and Figure 1).

261 Biological parameters were estimated for each of the ICES stocks and were then aggregated into 262 species, habitat and total ICES areas. The total MSY estimate for the ICES fisheries decreases 263 exponentially with the level of aggregation, with MSY estimates 30% lower when using ICES 264 area aggregation (largest aggregation) compared to estimates from stock level aggregation 265 (lowest aggregation). Although we don't use the estimated MSY by ICES area for the basin-266 scale analysis, the differences between estimates arising from different levels of aggregation 267 were used to calculate the confidence limits of MSY in the North Atlantic, which were used as 268 inputs for the bioeconomic model.

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270 (# Insert Figure 1 #)

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We used the time-series of total aggregated NA landings applied to the algorithm by Martell and
Froese (2010) to estimate an MSY of 13.7 Mt (s.d.=0.04) (Figure 2). Historically North Atlantic
fisheries were considered to be under development up to 1970s, when total landings started to
exceed MSY considerations. From 1980 until the early 2000s, the total catch has fluctuated near
this estimated global MSY. Since then landings have decreased to levels approximately 80% of
basin MSY. This model also estimated the carrying capacity parameter (K) or unfished biomass
for North Atlantic fish resources to be 170 Mt (Table 3).

281 (# Insert Figure 2 #)

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The bioeconomic model estimated that NA fisheries could generate $B \in 12.85$ of profits compared to the current $B \in 0.63$ (Figure 3). In addition, this equilibrium model shows the biomass level (45Mt, 26% of K) if current profits were to be maintained. In summary, this figure indicates that allowing stocks to rebuild to the biomass consistent with MEY (108 Mt, 63.5% of the unexploited biomass) would allow multiplying profits 20 fold. In other words, NA fisheries are only generating 5% of their economic potential. It must be noted that the catch at MEY is estimated to be 12.66 Mt, only 25% larger than the catch level in 2010.

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291 (# Insert Table 3 #)

292 (# Insert Figures 3 and 4 #)

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Using the classic revenue-cost against fishing effort curve by Gordon (1954) (Figure 4), the effort level that would lead to the economic maximization of North Atlantic fisheries is estimated to be 47% of current effort. Thus, should fishing effort increase 10% above current levels, the NA fisheries would incur economic losses. Figure 4 also shows that assuming equilibrium conditions NA fisheries in 2010 were near the "Bioeconomic Equilibrium" (BE), the point at which the fishery rents are dissipated as fishing costs are equal to the revenues from fishing.

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302 (# Insert Figure 5 #)

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The computations above are subject to uncertainties, and thus we added a level of stochasticity to our model's input parameters, which indicated that the economic loss of North Atlantic fisheries in 2010 ranged between B \in 2.5 and B \in 32 when all parameters were randomly fluctuating with a 30% coefficient of variation (Figure 5). MSY and Y₂₀₁₀ are the most

308 important sources of variation when estimating the economic losses of fisheries (Arnason, 2007; 309 Arnason et al., 2009) and Figure A.1 (Appendix). Besides, significant uncertainty was 310 propagated into estimates of stocks' carrying capacity (sdLog=0.24). Therefore, specific 311 simulations investigating the impact of those three parameters were performed. Uncertainty in 312 estimation was moderately reduced by varying the three parameters through lognormal 313 distributions and generating random values with a uniform distribution with bounds $\pm 30\%$ for 314 the others ('sim 2') or assuming them constant ('sim 3'). The simulations 'sim 3' and 'sim 4' 315 confirmed that these parameters generated the largest uncertainty on the final estimates of 316 economic loss. For 'sim 4', fixing MSY, K and Y_{2010} the variability of the loss estimate was 317 reduced significantly, ranging between B \in 6 and 19 with 95% confidence. The most important 318 result from these simulations is that the model is more sensitive to biological parameters and 319 therefore, biological parameterization is more important than economic parameterization.

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321 4. Discussion

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We have provided an assessment of the economic losses due to the choices taken in the management of North Atlantic fisheries. We have used methods previously implemented in the assessment of the economic losses of global fisheries (Arnason et al., 2009). Such a focus on the North Atlantic, in the context of the Euro_BASIN project, is motivated by the fact that its fisheries have a long history and economic importance, with significant catch-independent and dependent data sets, which are managed at different scales and with different degrees of success and failure.

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331 The catch and value of North Atlantic fisheries have declined significantly in the last decade, 332 partially due to management restricting catches (see below). The economic opportunity lost 333 through the inefficient management of North Atlantic fisheries in 2010 was estimated to be B€ 334 12.2. This echoes the results of a bioeconomic model built imposing strong assumptions on 335 North Atlantic basin biological productivity and economic data of the fishing fleets operating in 336 the area. Arnason et al 2009 estimated the global economic loss of marine fisheries due to 337 overexploitation to be in a range of \$B 37-67. North Atlantic landings correspond to 338 approximately 12% of global catches and the economic loss of NA fisheries represents ~33% of 339 global losses. This may be caused by the relative larger price of NA fisheries in comparison to 340 other areas (Sumaila et al., 2007) and by the historical overfishing history of North Atlantic 341 fisheries (FAO, 2012). For the North East Atlantic, Crilly and Esteban (2012) estimate that 342 restoring fish stocks could deliver up to £4.43 billion per year in profits, approximately 41% of 343 our estimate for the entire NA.

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345 Fisheries assessment provides information on the state of exploitation of marine resources and is 346 generally performed at stock level, a harvested unit which dynamics are driven by recruitment, 347 growth, natural mortality and fishing. Because of the limited number of fish stocks with stock 348 assessment data at the basin-scale, the catch based approached employed in this study allow us 349 to include a wider range of fish stocks. However, the catch based approach is based on the 350 assumption that catch reflects fish abundance and productivity. This principle is controversial, 351 especially when management interventions change through the history of catch time-series 352 (Pauly et al., 2013). However, catch-based methods are widely used to assess data-poor fisheries 353 and to produce large scale overviews of the state of fisheries (Fernandes et al., 2013; FAO, 354 2012; Lleonart & Maynou, 2003; Pauly et al., 2003, Vasconcellos and Cochran, 2005). Data on 355 North Atlantic fisheries' are abundant, especially for ICES-assessed stocks. A specific problem 356 arises because data are not available at the basin scale, one of the challenges that the Euro-357 BASIN project tries to address. Also, the proportion of assessed stocks in relation to total catch 358 differs across regions of the NA. For example, more than 90% of the North Sea catch (areas 359 IVa-c) corresponds to assessed stocks but in the Celtic Sea (VIIe-k) this number is less than 360 40% (Gascuel et al., 2012). Using data that are only available from ICES statistical area may 361 thus provide a biased view of the status of fisheries in the NA basin. Besides, multi-species 362 MSY is less than the sum of single stocks', as demonstrated in this and other studies (Link et al., 363 2012; Sparholt & Cook, 2009). Multi-species MSY could have been estimated with ecosystem 364 models as well. Fish species dynamics are regulated through trophodynamic interactions and 365 energetic fluxes across trophic levels (Pauly et al., 2000; Shin & Cury, 2004), which are 366 reflected in the ecosystem's size spectra (Blanchard et al., 2009). For example, capelin, cod and 367 herring interact in the Barents Sea food web (Lindstrøm et al., 2009). However, these models 368 are relatively complex in relation to SPM (Coll et al., 2008). We favor the use of a simplified 369 aggregated surplus production model because these models can produce robust estimates of 370 multispecies environments (Sparholt & Cook, 2009) allowing for comparison across areas 371 towards the practical implementation of the ecosystem-based fisheries management (Link et al., 372 2012). Also, this model provides a consistent platform to produce a macro scale assessment of 373 North Atlantic fisheries in combination with economic information.

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We acknowledge that the use of an aggregated economic model requires significant simplifications of complex ecological processes, and masks geographical differences in ecosystems productivity and management efficiency. For example, let us look at two cod stocks in the Irish and Icelandic Seas. Recent annual landings of Irish Sea cod have been lower than 5 k t with prospects for zero catch in 2013 and a stock which is currently outside biological limits (ICES 2012a). In contrast, the Icelandic cod's TAC for 2012 was 177 k t and the stock is

considered inside safe biological limits (ICES 2012b). The overall fishing effort recommended to achieve MEY for NA fisheries would not be expected to be applied homogeneously to all stocks. The aggregative approach by-passes stock-specific responses and assumes that fishing effort reductions would have to focus those stocks catalogued as "overexploited" or "under overexploitation", and that benefits from adequate fishing management will be especially notable for the most productive areas of the North Atlantic.

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388 Based on our analysis, North Atlantic fisheries remained within the estimated MSY range from 389 1964 to 2005. Then, total landings declined significantly and the estimation model associates 390 this to overall overexploitation of NA resources. However, marine ecosystems are driven by 391 multiple drivers that change over time; therefore, a constant historical MSY may not be 392 realistic. The constant MSY estimated in the first part of this work is used for the subsequent 393 economic assessment and should be considered with caution. Catch reductions can be caused by 394 multiple factors, including overexploitation, environmental variability or implementation of 395 catch restrictions. Overexploitation is defined by Hilborn and Stokes (2010) when catches are 396 below 80% MSY, which equates to when declining yields are obtained with increased fishing 397 effort (Schaefer 1954). The same is concluded from the biomass-catch diagram shown in 398 Figure 3. When MSY is exceeded for extended periods and if fishing effort is maintained 399 beyond the level corresponding to MSY, yield will decrease as the available biomass has fallen 400 below the point at which MSY is achieved (50% of its unexploited level, in this case). That is, 401 biomass decreases with increasing catch until the point when biomass reductions will result in 402 lower catches if fishing effort is not increased. In multispecies fisheries apparent MSY levels 403 can be maintained by targeting previously undeveloped fisheries simultaneously with declining 404 stocks. Using theoretical models, it has been shown that this feature can precede a sequential 405 collapse of geographically distant fisheries (Merino et al., 2010; Merino et al., 2011).

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407 This model does not consider environmental effects on the productivity of the NA basin. In 408 reality, fish stocks, especially small pelagic fish (70-80% of total NA catch), are highly 409 vulnerable to environmental variability (Barange et al., 2009; Chavez et al., 2003; Fernandes et 410 al., 2010; Hsieh et al., 2009). However, it is also evident that the impacts of particular 411 environmental conditions differ between species. For example, Icelandic capelin catch averaged 412 1 Mt from 1979 to 2002 (13% the yields from ICES assessed stocks) when it started declining to 413 15 kt in 2008. This decline is reflected in the overall NA trend and it could be caused by 414 temperature changes (Carscadden et al., 2013 (In press)). However, other stocks such as herring 415 (yielding ~2Mt in the last decade) seem to be favored by current conditions and have recovered 416 from overexploitation faster than expected (Nash et al., 2009), which could counterbalance the 417 negative environmental impact on capelin on the basin scale trend. Another example is blue

418 whiting whose catches have displayed a dramatic "boom and bust" dynamic over the past two 419 decades (ICES, 2011). Landings during the 1980s and early 1990s were typically between 500 420 and 1000 kt, but increased to 2400 kt in 2004 as a result of a suite of good year classes. At this 421 point, blue whiting was the largest fishery in the North Atlantic, ahead of herring, and the third 422 largest marine capture fishery in the world (FAO, 2010). The subsequent decline of the fishery 423 has, however, proved to be equally dramatic (ICES, 2011). The alternation between warm and 424 cold regimes is associated to alternative species proliferation (Chavez et al., 2003), including 425 multidecadal regime shifts (Alheit et al., 2009). However, investigating each of the 426 environmental drivers affecting fish stocks in the North Atlantic in order to better estimate 427 individual MSYs would mean losing focus on the principal objective of this study and its scale.

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429 A third factor resulting in catch reduction is management restriction. Generally, closures and 430 drastic catch limitations are the consequence of overexploiting resources and subsequent fishery 431 crises (Finlayson, 1994; Lazkano et al., 2012; Nøstbakken & Bjørndal, 2003; Worm et al., 432 2009). Historically, fish stocks have collapsed due to a myriad of unfavorable environmental 433 conditions and excessive fishing pressure (Alheit et al., 2009; Chavez et al., 2003; Merino et al., 434 2013; Watson et al., 2006) and which triggered consequent catch restrictions (Worm et al., 435 2009). However, we would like to stress that, particularly the catch reduction in the last ten 436 years of the data series, should be attributed not only to historical overfishing but also to 437 management driven catch limitations. For example, under the EU framework, the Common 438 Fishery Policy and the Financial Instrument for Fisheries Guidance (FIFG) a remarkable 439 reduction of fishing boats has been accomplished (Fernandes and Cook 2013). In addition, since 440 2005 emergency and recovery plans have applied under the EU adopted MSY framework 441 aiming to reduce fishing mortality towards achieving MSY for different stocks which is already 442 improving fisheries economic indicators (Cardinale et al., 2013). Furthermore, this approach is 443 followed by the International Council of the Exploration of the Sea (ICES) and other 444 international agreements (FAO, 2012). To sum up, some of the catch reductions reflected in the 445 basin scale trend (Figure 2) are aligned to the implementation of international efforts to restore 446 fish stocks and this can potentially bias the parameter estimation procedure used in this study by 447 estimating as economic loss what in reality may be a short term economic loss "invested" in 448 stocks recovery towards more profitable fisheries.

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The parameters used in the bioeconomic model can be controversial too: For example, classically, the supply-demand relation is considered as inverse: the lesser the catch, the higher the price. However, the positive elasticity parameter used here was taken from Arnason et al (2009) which aligns with a global perspective of the state of marine fisheries, as the "Fishing down the food web" concept (Pauly et al 1997) does. However, this is expected to have low

455 impact in our numeric results: The estimated catch increase when moving towards MEY would 456 be small, so the expected price changes would be small too. A different matter is the potential 457 impact of exogenous variables on North Atlantic fish demand and therefore, in the price 458 equation used in this document. We do not consider the impact of aquaculture expansion on the 459 price of wild fish nor the impact of imports that might act as less priced substitutes to North 460 Atlantic fish. Both factors could presumably reduce the price of North Atlantic fish and 461 therefore, the potential economic profit of North Atlantic fisheries would be reduced. Finally, 462 our model is based on estimates of current profits of NA fisheries, estimated with value and 463 fishing costs databases, and without considering the effects of subsidies. According to Sumaila 464 et al (2012), 31% of landed value in world fisheries is subsidized and therefore, the current 465 profits for the fishing companies are presumably larger than the $B \in 0.63$ used to parameterize 466 our bioeconomic model.

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468 The implementation of ecosystem-based fisheries management (EBFM) requires the 469 development of models to assess the economic performance of the fishing industry in 470 combination with their impact upon marine ecosystems (Gascuel et al., 2012). The bioeconomic 471 model used here was parameterized with a global estimate of NA ecosystems productivity and 472 the sum of the economic performance indicators of the countries operating in its waters. In contrast to the biological part, the aggregation of the economic parameters was additive, we 473 474 estimated the NA value of catch and net economic profits as the sum of the national estimates. 475 The values shown in table 2 were obtained collating catch and value data from the Sea Around 476 Us database and estimating the fleet specific costs of fishing using costs per tonne estimates 477 from Lam et al (2011). Fishing costs and net profit values were also available from alternative 478 reports. For example, the Annual Economic Report (AER) on the EU fishing fleet (JRC, 2012) 479 provides estimates of many fishing indicators of EU countries. However, this report aggregates 480 all EU countries fishing operations in waters beyond the NA. Using costs of fishing per tonne of 481 catch in the NA allows for assigning the fishing costs only to the operations targeting North 482 Atlantic fish. However, the cost structure provided in the AER is more detailed than in our 483 approach. The net profit of EU fleets operating in the NA estimated in the AER is B€-0.236. 484 Had this value been used as input to our bioeconomic model, our estimated loss would have 485 been even larger. Additional sources of information on the economic performance of Russian, 486 Norwegian, US and Canadian fleets (FAO, 2007; Kitts et al., 2010; NOAA, 2011) could 487 improve the economic parameter estimation process. However, as seen in Figure 5, the most 488 determinant set of parameters are those related to ecosystems productivity.

490 Our approach is based on deviations from biological and economic reference points. The 491 economic loss pivots around the concept of *Maximum Economic Yield*, an equilibrium point

492 where the net economic return from a fishery can be maximized sustainably, as assumed in 493 previous studies (Arnason et al., 2009; Crilly & Esteban, 2012; Sumaila et al., 2012). This 494 reference point is estimated with a graphical procedure (Figure 3). Large benefits will be 495 considered as unsustainable as they do not meet the parabola and; lower than the optimal will 496 cross it twice, one for high levels of biomass and the other at biomass levels below that 497 corresponding to MEY. It is important to note that the recovery of the stocks towards MEY 498 biomass would not produce major changes in the overall catch from the NA. In 2010 10.8Mt of 499 fish was landed whereas for the MEY total catch would be 12.66 Mt. Therefore, a catch increase 500 of 26% would produce a net economic gain of 2000%, but would require a 53% reduction in 501 fishing effort. According to this, the economic benefit of restoring stocks would outweigh its 502 potential food security implications (Garcia & Rosenberg, 2010; Rice & Garcia, 2011; 503 Srinivasan et al., 2010). By reducing the fishing effort, costs would reduce linearly as revenues 504 would increase potentially until the MSY peak. Then, further effort reductions would make 505 revenues reduce too until its gradient equals fishing costs lines slope. As a result, a fishing effort 506 reduction would produce a logarithmic increase in profits. Therefore, the profit increase would 507 be more substantial at the initial stages of reduction. For example, if total fishing effort was 508 reduced to 70% of current levels, total fish biomass (not each and every stock) would recover to 509 MSY and profits would increase up to B \in 10.8 (1725% more than in 2010). Therefore, accepting 510 the hard transition of reducing the size of the industry to 47% of current level, it is important to 511 note that moderate reductions would also produce large economic benefits as well as improving 512 resource conservation significantly.

513 The reduction of fishing effort will have negative short term costs in the form of reduction of 514 catch towards stocks recovery, loss of a notable number of current jobs provided by fisheries 515 and costs to dismantle a number of the fishing boats currently operating in the North Atlantic. 516 Therefore, it will require investments to reallocate fishermen in alternative activities, scrap 517 fishing vessels and other compensations to the fishing industry. Crilly and Esteban (2012) and 518 the work by Sumaila et al. (2012) demonstrate that after a short transition the benefits of 519 restoring fish stocks outweighs the costs incurred and investments required to reduce fishing 520 mortality. This conclusion holds notwithstanding the high uncertainty in estimates and the 521 assumptions made to enable large scale assessments of governance (Cash & Moser, 2000; 522 Christensen & Walters, 2004; Jennings et al., 2008; Wilbanks & Kates, 1999). In addition, 523 restoring fish stocks would avoid reducing the risk of fisheries collapses and its dramatic 524 economic consequences. For example, the collapse of cod produced an increase of 30% of 525 unemployment in some areas of Newfoundland and more than \$3 billion were spent to 526 restructuring adjustments for workers in the fishing sector, among other social implications 527 (Hamilton and Butler, 2001). However, it is also true that fishing mortality reductions haven't 528 always produced the stocks' recovery predicted by fisheries assessment models. For example, a

529 combination of environmental changes and fishing pressure are responsible of Atlantic cod530 populations failure to recover (Hilborn and Litzinger, 2009).

531

532 A single estimate of economic loss is intuitive but can be simplistic given the number of 533 parameters involved in the computation. In order to add consistency to our results and to offset 534 the uncertainty associated with our methods, four stochastic experiments were conducted with 535 the bioeconomic model. The results of these experiments provide two conclusions: First, 536 allowing as much as a 30% random variation in the input parameters, the estimated economic 537 loss of North Atlantic fisheries is measured in billions of euro. Second, the model is particularly 538 sensitive to three biological parameters: MSY, K and catch in the last year. Reducing the 539 uncertainty on these parameters reduces the standard deviation of the estimates significantly. In 540 contrast, fixing the other five parameters produces only moderate reductions of variability on 541 the economic loss of fisheries. Therefore, we emphasize the relevance of adequate commercial 542 and fishery independent data collection programs in order to improve the stock assessment 543 process. Despite uncertainties on the current scale of North Atlantic basin productivity, we 544 conclude that an overall fishing effort reduction is recommended, with not only ecological 545 benefits but significant and demonstrable economic consequences.

546

547 To conclude, our analysis supports the work conducted under the Euro-BASIN project by 548 providing a basin-scale framework for the economic analysis of the efficiency of North Atlnatic 549 fisheries management. In the future this analysis needs to take into consideration the way 550 European fisheries management, in particular, is evolving. The reform of the European 551 Common Fisheries Policy identifies MSY as a management target, consistent with our analysis. 552 It also highlights the need to implement a discard ban, which should come hand in hand with the 553 needed improvement in the monitoring and reporting of fishing activities. While our analysis is 554 conducted at the basin scale, regionalization of management is a process that would need to be 555 considered in future monitoring programs and modeling approaches. Significantly, the CFP 556 reform also identifies the need to collect environmental, social and economic data and use these 557 as criteria to allocate fishing rights. Future Euro-BASIN initiatives would have to consider the 558 above in developing their workprogramme, as well as approaches to better understand market 559 price formation (exports and competition with products from other areas) and how to influence 560 consumer demand for species that traditionally have been less preferred.

- 561
- 562

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569	
570	Appendix
571	
572	Table A1. Necessary transformations to run the bioeconomic model using the parameters shown
573	in Table 3 (Arnason, 2007; Arnason et al., 2009; Gordon, 1954; Schaefer, 1954).
574	
575	{Insert Table A1}
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576	
577	Figure A1. Sensitivity analysis of the economic loss in 2010 for different parameters.
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579	{Insert Figure A1}
580	References
581	
582	• Alheit, J., Roy, C., Kifani, S., 2009. Decada-scale variability in populations. In D.
583	Checkley, J. Alheit, Y. Oozeki, C. Roy (Eds.), Climate Change and Small Pelagic Fish
584	Stocks (p. 382): Cambridge University Press.
585	• Allison, E.H., Perry, A.L., Badjeck, M.C., Adger, W.N., Brown, K., Conway, D., Halls,
586	A.S., Pilling, G., Reynolds, J.D., Andrew, L.N., Dulvy, N., 2009. Vulnerability of
587	national economies to the impacts of climate change on fisheries. <i>Fish and Fisheries</i> ,
588	10, 173-196.
589 590	• Arnason, R., 2007. Loss of economic rents in the global fishery. <i>XVIIIth Annual EAFE Conference</i> . Reykjavik, Iceland: www.eafe-fish.eu.
591	 Arnason, R., Kelleher, K., Willman, R., 2009. The Sunken Billions: The economic
592	justification for fisheries reform. Agriculture and rural development (p. 100): The
593	World Bank and Food and Agriculture Organization.
594	• Barange, M., Bernal, M., Cercole, M.C., Cubillos, L., Cunningham, C.L., Daskalov,
595	G.M., De Oliveira, J.A.A., Dickey-Collas, M., Hill, K., Jacobson, L., Køster, F.W.,
596	Masse, J., Nishida, H., Ñiquen, M., Oozeki, Y., Palomera, I., Saccardo, S.A.,
597	Santojanni, A., Serra, R., Somarakis, S., Stratoudakis, Y., van der Lingen, C.D., Uriarte,
598	A., Yatsu, A., 2009. Current trends in the Assessment and Management of Small
599	Pelagic Fish Stocks. In D. Checkley, J. Alheit, Y. Oozeki, C. Roy (Eds.), Climate
600	Change and Small Pelagic Fish Stocks (p. 382): Cambridge University Press.
601	• Beaugrand, G., Reid, P.C., Ibañez, F., Lindley, J.A., Edwards, M., 2002. Reorganization
602	of North Atlantic marine copepod biodiversity and climate. <i>Science</i> , 296, 1692-1694.
603	• Blanchard, J., Jennings, S., Holmes, R., Harle, J., Merino, G., Allen, I., Holt, J., Dulvy,
604 605	N., Barange, M., 2012. Potential consequences of climate change for primary
605 606	production and fish production in large marine ecosystems. <i>Philosophical Transactions</i>
606	of the Royal Society B, 367, 2979-2989.

607 •	Blanchard, J., Jennings, S., Law, R., Castle, M.D., McCloghrie, D., Rochet, M.J.,
608	Benoît, E., 2009. How does abundance scale with body size in coupled size-structured
609	food webs? Journal of Animal Ecology, 78, 270-280.
610 •	Branch, T.A., Jensen, O.P., Ricard, D., Ye, Y., Hilborn, R., 2011. Contrasting global
611	trends in marine fishery status obtained from catches and stock assessments.
612	Conservation Biology, 25, 777-786.
613	Cardinale, M., Dorner, H., Abella, O., Andersen, J.L., Casey, J., Doring, R., Kirkegaard,
614	E., Motova, A., Anderson, J., Simmonds, E.J., Stransky, C., 2013. Rebuilding EU fish
615	stocks and fisheries, a process under way? Marine Policy, 39, 43-52.
616 •	Carscadden, J.E., Gjøsæter, H., Vilhjálmsson, H., 2013 (In press). A comparison of
617	recent changes in distribution of capelin (Mallossus villotus) in the Barents Sea, around
618	Iceland and in the Northwest Atlantic. <i>Progress In Oceanography</i> .
619 •	Call, 2000, 2000, 2000, 2000, 2000, 2000, and 1000 search assigning a frame
620	assessment and management processes. Global Environmental Change, 10, 109-120.
621 •	Coll, M., Libralato, S., Tudela, S., Palomera, I., Pranovi, F., 2008. Ecosystem
622	Overfishing in the Ocean. <i>PLoS ONE</i> , 3, e3881.
623 •	Crilly, R., Esteban, A., 2012. No catch investment: Investing to restore European fish
624	stocks. In N.E. Foundation (Ed.). London, UK.
625 •	Cury, P.M., Shin, YJ., Planque, B., Durant, J.M., Fromentin, J.M., Kramer-Schadt, S.,
626	Stenseth, N.C., Travers, M., Grimm, V., 2008. Ecosystem oceanography for global
627	change fisheries. Trends in Ecology and Evolution, 23, 338-346.
628 •	Chavez, F.P., Ryan, J., Lluch-Cota, S.E., Niquen C, M., 2003. From anchovies to
629	sardines and back: multidecadal change in the Pacific ocean. Science, 299, 217-221.
630 •	Christensen, V., Walters, C.J., 2004. Trade-offs in ecosystem scale optimization of
631	fisheries managament policies. Bulletin of Marine Science, 74, 549-562.
632 •	European Union. 2012. Communication from the Commission to the Council
633	concerning a consultation on Fishing Opportunities for 2013. COM(2012) 278 final, 17
634	pp.
635 •	FAO, FISHSTAT Statistical collections. FAO.
636 •	FAO, 2007. National fishery sector overview: The Russian Federation. (p. 17).
637 •	FAO, 2010. Statistics and Information Service of the Fisheries and Aquaculture
638	Department. FAO yearbook. Fishery and Aquaculture Statistics 2008. Rome, Italia.
639	72 p .
640 •	FAO, 2012. The State of World Fisheries and Aquaculture. In FAO (Ed.). Rome.
641	Fernandes J.A., Cheung W.W.L., Jennings S., Butenschön M., de Mora L., Frölicher
642	T.L., Barange M., Grant A., 2013. Modelling the effects of climate change on the
643	distribution and production of marine fishes: accounting for trophic interactions in a
644	dynamic bioclimate envelope model. <i>Global Change Biology</i> 19(8), 2596-2607.
645	Fernandes, P.G., Cook, R.M., 2013. Reversal of fish stock decline in the Northeast
646	Atlantic. Current Biology, 23(15), 1432-1437.
647 •	
648	2010. Fish recruitment prediction, using robust supervised classification methods.
649	Ecological Modelling, 221(2), 338-352.
650 •	
651	assessments from 1977-1990. St. John's, Newfoundland, Canada: Memorial University
652	of Newfoundland.

653	٠	Garcia, S.M., 2009. Rising to depletion? Towards a dialogue on the state of national
654		marine fisheries. Preliminary report. In T.W. Bank (Ed.), Global Program on Fisheries
655		(<i>PROFISH</i>) (p. 65).
656	•	Garcia, S.M., Rosenberg, A.A., 2010. Food security and marine capture fisheries:
657		characteristics, trends, drivers and future perspectives. Philosophical Transactions of
658		the Royal Society B, 365, 2869-2880.
659	٠	Gascuel, D., Merino, G., Döring, R., Druon, J.N., Goti, L., Guénette, S., Macher, C.,
660		Soma, K., Travers-Trolet, M., Mackinson, S., 2012. Towards the implementation of an
661		integrated ecosystem fleet-based management of European fisheries. Marine Policy, 36,
662		1022-1032.
663	•	Gordon, H.S., 1954. The economic theory of a common-property resource: the fishery.
664		Journal of Political Economy, 62, 124-142.
665	•	Guillen, J., Macher, C., Merzéréaud, M., Bertignac, M., Fifas, S., Guyader, O., 2013.
666		Estimating MSY and MEY in multi-species and multi-fleet fisheries, consequences and
667		limits: an application to the Bay of Biscay mixed fishery. Marine Policy, 40, 64-74.
668	٠	Hamilton, L.C., Bulter, M.J., (2001) Outport adaptations: social indicators through
669		Newfoundland's cod crisis. Research in Human Ecology, 8(2), 1-11.
670	٠	Hilborn, R., Stokes, K., 2010. Defining overfished stocks: Have we lost the plot?
671		Fisheries, 35, 113-120.
672	٠	Hsieh, CH., Kim, H.J., Watson, W., Di Lorenzo, E., Sugihara, G., 2009. Climate
673		driven changes in abundance and distribution of larvae of oceanic fishes in the southern
674		California region. Global Change Biology, 15, 2137-2152.
675	٠	ICES, 2011. Report of the Working Group on Widely Distributed Stocks (WGWIDE).
676		ICES CM 2011/ACOM:15. 624 pp.
677	•	ICES. 2012b. Report of the ICES Advisory Committee 2012. ICES Advice, 2012. Book
678		5. 459 pp.
679	٠	ICES. 2012b. Report of the ICES Advisory Committee 2012. ICES Advice, 2012. Book
680		2. 114 pp.
681	٠	Jennings, S., Mélin, F., Blanchard, J.L., Forster, R.M., Dulvy, N.K., Wilson, R.W.,
682		2008. Global-scale predictions of community and ecosystem properties from simple
683		ecological theory. Proceedings of the Royal Society B: Biological Sciences, 275, 1375-
684		1383.
685	٠	JRC, 2012. The 2011 Annual Economic Report on the EU fishing fleet. In J. Anderson,
686		J. Guillen, J. Virtanen (Eds.), Scientific, technical and economic committee for fisheries.
687	•	Kitts, A., Bing-Sawyer, E., Walden, J., Demarest, C., McPherson, M., Christman, P.,
688		Steinback, S., Olson, J., Clay, P., 2010. Performance of the Northeast multispecies
689		(groundfish) fishery. In N.N.M.F. Service (Ed.). Woods Hole, MA.
690	•	Lam, V.W.Y., Sumaila, R.U., Dyck, A., Pauly, D., Watson, R., 2011. Construction and
691		first applications of a global cost of fishing database. ICES Journal of Marine Science,
692		68, 1996-2004.
693	•	Lazkano, I., Nøstbakken, L., Prellezo, R., 2012. Past and future management of a
694		collapsed fishery: The bay of Biscay anchovy. Natural Resource Modeling.
695	٠	Lindstrøm, U., Smoutb, S., Howellc, D., Bogstadc, B., 2009. Modelling multi-species
696		interactions in the Barents Sea ecosystem with special emphasis on minke whales and
697		their interactions with cod, herring and capelin. Deep-Sea Research Part II: Topical
698		Studies in Oceanography, 56, 2068-2079.

699 •	Link, J.S., 2009. Ecosystem-Based Fisheries Management: Confronting Tradeoffs. New
700	York: Cambridge University Press.
701 •	Link, J.S., Gaichas, S., Miller, T.J., Essington, T., Bundy, A., Boldt, J., Drinkwater, K.,
702	Moksness, E., 2012. Synthesizing lessons learned from comparing fisheries production
703	in 13 northern hemisphere ecosystems: emergent fundamental features. Marine Ecology
704	Progress Series, 459, 293-302.
705 •	Lleonart, J., Maynou, F., 2003. Fish stock assessments in the Mediterranean: state of the
706	art. Scientia Marina, 67 (Suppl. 1), 37-49.
707 •	Marshall, J., Kushnir, Y., Battisti, D., Chang, P., Czaja, A., Dickson, R., Hurrell, J.,
708	McCartney, M., Saravanan, R., Visbeck, M., 2001. North Atlantic climate variability:
709	phenomena, impacts and mechanisms. International Journal of Climatology, 21, 1863-
710	1898.
711 •	Martell, S., Froese, R., 2010. A simple method for estimating MSY from catch and
712	resilience. Fish and Fisheries.
713 •	Merino, G., Barange, M., Mullon, C., Rodwell, L., 2010. Impacts of global
714	environmental change and aquaculture expansion on marine ecosystems. Global
715	Environmental Change, 20, 586-596.
716 •	Merino, G., Barange, M., Rodwell, L., Mullon, C., 2011. Modelling the sequential
717	geographical exploitation and potential collapse collapse of marine fisheries through
718	economic globalization, climate change and management alternatives. Scientia Marina,
719	75, 779-790.
720 •	Merino, G., Barange, M., Blanchard, J., Harle, J., Holmes, R., Allen, I., Allison, E.H.,
721	Badjeck, M.C., Dulvy, N., Holt, J., Jennings, S., Mullon, C., Rodwell, L., 2012. Can
722	marine fisheries and aquaculture meet fish demand from a growing human population in
723	a changing climate? Global Environmental Change,
724	http://dx.doi.org/10.1016/j.gloenvcha.2012.03.003.
725 •	Merino, G., Barange, M., Mullon, C., 2013. Role of reduction fisheries in the world
726	fishmeal production. In K. Ganias (Ed.), Biology and ecology of anchovies and
727	sardines. Enfield, New Hampshire, USA: Science Publishers.
728 •	Mueter, F.J., Megrey, B.A., 2006. Using multi-species surplus production models to
729	estimate ecosystem-level maximum sustainable yields. Fisheries Research, 81, 189-
730	201.
731 •	Nash, J.F., Dickey-Collas, M., Kell, L.T., 2009. Stock and recruitment in North Sea
732	herring (Clupea harengus); compensation and depensation in the population dynamics.
733	Fisheries Research, 95, 88-97.
734 •	NOAA, 2011. Fisheries Economics of the United States 2009. In U.D.o. Commerce
735	(Ed.), Vol. NMFS-F/SPO-118: National Oceanic and Atmospheric Administration.
736 •	Nøstbakken, L., Bjørndal, T., 2003. Supply functions for North Sea herring. Marine
737	Resource Economics, 18, 345-361.
738 •	Parsons, L.S., Lear, W.H., 2001. Climate variability and marine ecosystem impacts: a
739	North Atlantic perspective. Progress In Oceanography, 49, 167-188.
740 •	Pauly, D., Alder, J., Bennett, E., Christensen, V., Tyedmers, P., Watson, R., 2003. The
741	future for fisheries. Science, 302, 1359-1361.
742 •	Pauly, D., Christensen, V., Dalsgaard, J., Froese, R., Torres, Jr. F. 1997. Fishing down
743	marine food webs. Science, 278, 860-863.

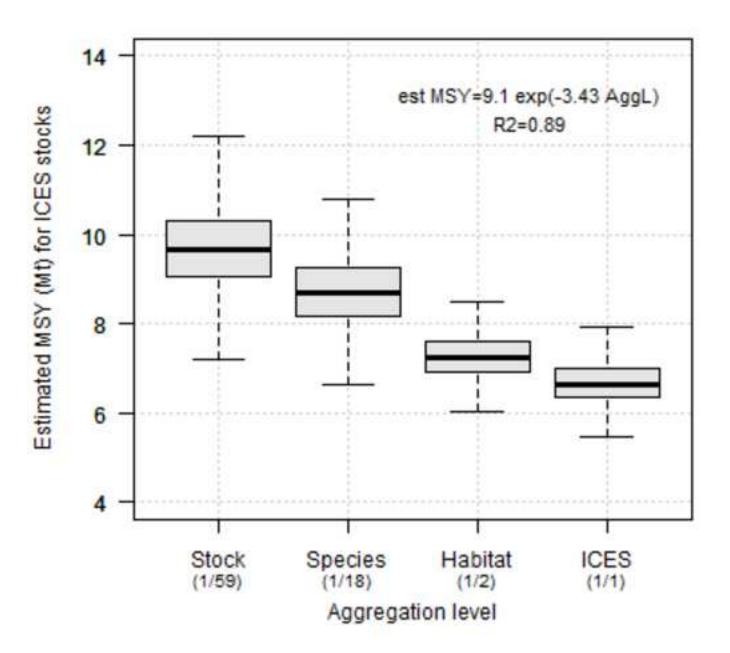
744 745 746	•	Pauly, D., Christensen, V., Guenette, S., Pitcher, T.J., Sumaila, U.R., Walters, C.J., Watson, R., Zeller, D., 2002. Towards sustainability in world fisheries. <i>Nature</i> , 418, 689-695.
747	•	Pauly, D., Christensen, V., Walters, C.J., 2000. Ecopath, Ecosim, and Ecospace as tools
747	•	for evaluating ecosystem impact of fisheries. <i>ICES Journal of Marine Science</i> , 57, 697-
748 749		706.
749 750	•	
750 751	•	Pauly, D., Hilborn, R., Branch, T.A., 2013. Fisheries: Does catch reflect abundance? <i>Nature</i> , 694, 303-306.
752	•	Rice, J.C., Garcia, S.M., 2011. Fisheries, food security, climate change and biodiversity:
753		characteristics of the sector and perspectives of emerging issues. ICES Journal of
754		Marine Science, 68(6), 1343-1353.
755	•	Schaefer, M., 1954. Some aspects of the dynamics of populations important to the
756		management of the commercial Marine fisheries. Bulletin of the Inter-American
757		Tropical Tuna Commission, 1, 27-56.
758 759	•	Seijo, J.C., Defeo, O., Salas, S., 1998. Fisheries bioeconomics. Theory, modelling and management. Rome.
760	•	5
760 761	•	Shin, YJ., Bundy, A., Shannon, L.J., Blanchard, J.L., Chuenpagdee, R., Coll, M., Kright, B., Lymon, C., Diet, C.L., Bishandoon, A.J., Chuenpagdee, R., Coll, M.,
761		Knight, B., Lynam, C., Piet, G.J., Richardson, A.J., Group, a.t.I.W., 2012. Global in
762		scope and regionally rich: an Indiseas workshop helps shape the future of marine
		ecosystem indicators. <i>Reviews in Fish Biology and Fisheries</i> , 22, 835-845.
764	•	Shin, YJ., Cury, P.M., 2004. Using an individual-based model of fish assemblages to
765 766		study the response of size spectra to changes in fishing. <i>Canadian Journal of Fisheries</i>
766		and Aquatic Sciences, 61, 414-431.
767	•	Shin, YJ., Rochet, MJ., Jennings, S., Field, J.G., Gislason, H., 2005. Using size-
768		based indicators to evaluate the ecosystem effects of fishing. <i>ICES Journal of Marine</i>
769		<i>Science</i> , 62, 384-396.
770	•	Sparholt, H., Cook, R., 2009. Sustainable exploitation of temperate fish stocks. <i>Biology</i>
771		Letters.
772	•	Srinivasan, U.T., Cheung, W.W.L., Watson, R., Sumaila, R.U., 2010. Food security
773		implications of global marine catch losses due to overfishing. <i>Journal of Bioeconomics</i> ,
774		12, 183-200.
775	•	Sumaila, R.U., Cheung, W.W.L., Dyck, A., Gueye, K., Huang, L., Lam, V., Pauly, D.,
776		Srinivasan, T., Swartz, W., Watson, R., Zeller, D., 2012. Benefits of rebuilding global
777		marine fishereis outweigh costs. PLoS ONE, 7, e40542.
778	•	Sumaila, R.U., Marsden, A.D, Watson, R., Pauly, D., 2007. A Global ex-vessel fish
779		price database: Construction and Applications. Journal of Bioeconomics, 9, 29-51.
780	•	Trenkel, V.M., Huse, G., MacKenzie, B.R., Alvarez, P., Arrizabalaga, H., Castonguay,
781		M., Goñi, N., Grégoire, F., Hátun, H., Jansen, T., Jacobsen, J.A., Lehodey, P.,
782		Lutcavage, M., Mariani, P., Melvin, G.G., Neilson, J.D., Nøttestad, L., Óskarsson, G.J.,
783		Payne, M.R., Richardson, D.E., Senina, I., Speirs, D.C., 2013 (In press in this issue).
784		Comparative ecology of widely distributed pelagic fish species in the North Atlantic:
785		implications for modelling climate and fisheries impacts. Progress in Oceanography.
786	•	Vasconcellos, M., Cochrane, K., 2005. Overview of world status data-limited fisheries:
787		inferences from landings statistics, In: Fisheries assessment and management in data-
788		limited situations. Proceedings; Lowell Wakefield Symposium, 21, Anchorage, AK
789		(USA), 22-25 Oct 2003. Kruse, G.H. (ed.) Gallucci, V.F. (ed.) Hay, D.E. (ed.) Perry,
790		R.I. (ed.) Peterman, R.M. (ed.) Shirley, T.C. (ed.) Spencer, P.D. (ed.) Wilson, B. (ed.)

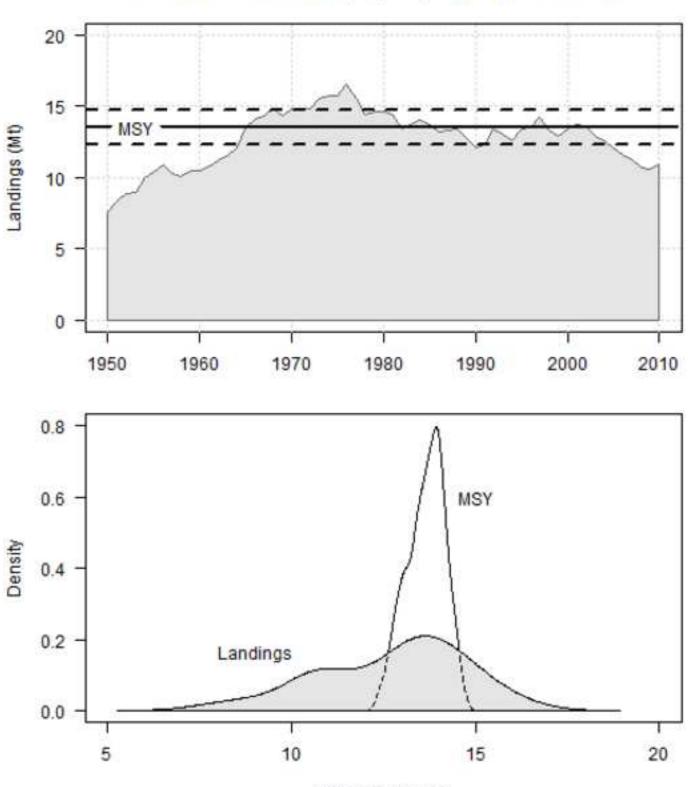
791 792	Woodby, D. (ed.) / Alaska Univ., Fairbanks (USA). Alaska Sea Grant College Program, 2005, p. 1-20.
793 •	Watson, R., Alder, J., Pauly, D., 2006. Fisheries for forage fish, 1950 to the present. In
794	J. Alder, D. Pauly (Eds.), On the multiple uses of forage fish: from ecosystems to
795	markets (pp. 1-20).
796 •	Wilbanks, T.J., Kates, R., 1999. Global change in local places: how scale matters.
797	<i>Climatic Change</i> , 43, 601-628.
798 •	Wiebe, P.H., Harris, R.P., St. John, M.A., Werner, F.E., de Young, B., 2009. P.P.E.
799	(Eds.), 2009. BASIN: Basin-scale Analysis, Synthesis and Integration. Science Plan and
800	Implementation Strategy.
801 •	Worm, B., Hilborn, R., Baum, J.K., Branch, T.A., Collie, J.S., Costello, C., Fogarty,
802	M.J., Fulton, E.A., Hutchings, J.A., Jennings, S., Jensen, O.P., Lotze, H.K., Mace, P.M.,
803	McClanahan, T.R., Minto, C., Palumbi, S.R., Parma, A.M., Ricard, D., Rosenberg,
804	A.A., Watson, R., Zeller, D., 2009. Rebuilding Global Fisheries. Science, 325, 578-585.
805	
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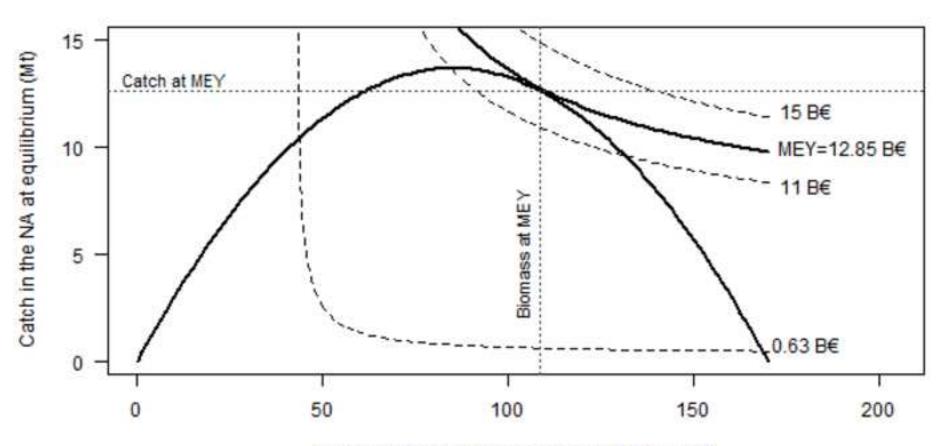
807 808 Figure 1. Maximum Sustainable Yield estimates for the stocks assessed by ICES at different 809 aggregation levels. The aggregation level is the number of units considered: 59 stocks, 18 810 species, 2 habitats and 1 for all the ICES stocks as a single fishery. Boxes show the geometric 811 mean of the estimate, 0.25 and 0.75 quantiles. The intervals limit the estimate to a 99.5% 812 confidence (see text). 813 814 Figure 2. Top: Historical landings of North Atlantic fisheries according to FAO and estimated 815 corresponding MSY. Below: Density distribution of annual total landings and distribution of 816 plausible total MSY values using the approach by Martell and Froese (2013). 817 818 Figure 3. Graphical estimation of North Atlantic fisheries Maximum Economic Yield. The 819 crossing point between different potential profit trajectories (*iso* $_{\Psi}$) and the biomass-catch 820 equilibrium curve determines the catch and biomass level that will lead to MEY. The MEY for 821 North Atlantic fisheries is B \in 12.85. The *iso* $\Psi = 0.63B \in$ corresponds to profits in 2010. 822 823 Figure 4. Gordon-Schaefer's model equilibrium for North Atlantic fisheries. Current (2010) and 824 economically optimum fishing efforts indicated with dotted lines. Net profits are calculated as 825 the difference between value of catch and costs of fishing. MEY is the Maximum Economic 826 Yield, i.e. the maximum difference between value of catch and costs of fishing. 827 828 Figure 5. Results of stochastic estimates of the economic loss of North Atlantic fisheries in 829 2010: 'sim 1' random fluctuation ($\pm 30\%$) of the parameters of the bioeconomic model (see table 830 3); 'sim 2' log-normally distributed MSY, K and Y₂₀₁₀ and random fluctuation for the others 831 $(\pm 30\%)$; 'sim 3' same as previous but with other parameters kept constant at values shown in 832 table 3; 'sim 4' MSY, K and Y_{2010} constant and random fluctuation for the others (±30%). 833 834 Figure A1. Sensitivity analysis of the economic loss in 2010 for different parameters. 835 836 837





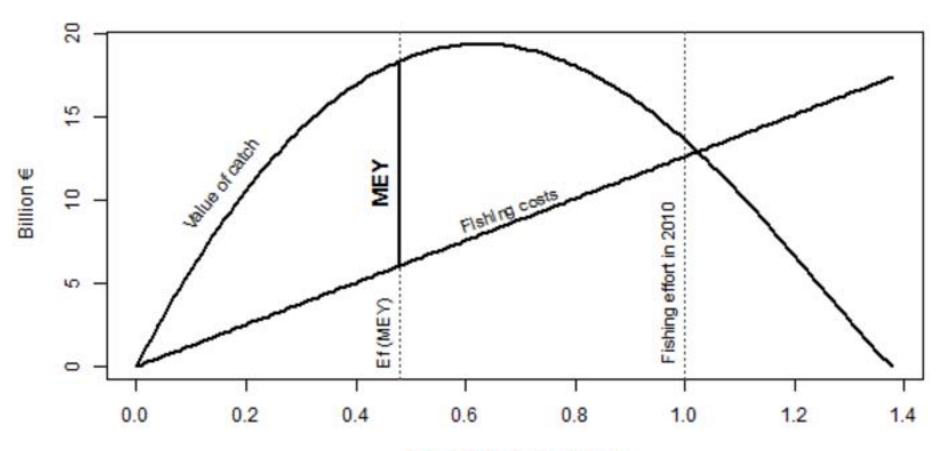
North Atlantic landings (FAO) and MSY estimates

Millions of tonnes



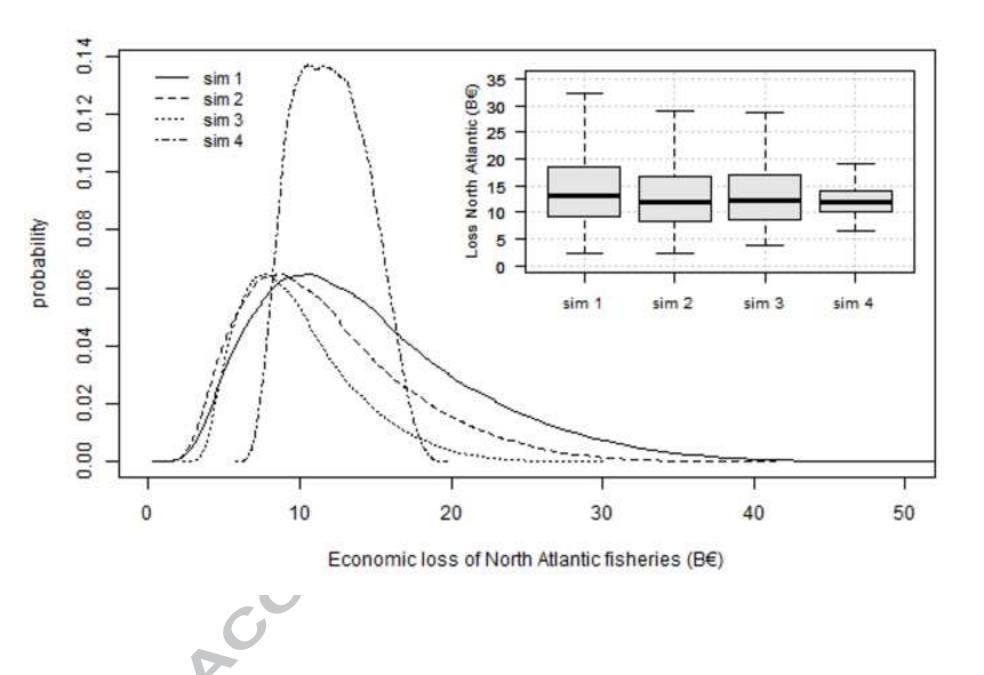
North Atlantic fish biomass at equilibrium (Mt)

G

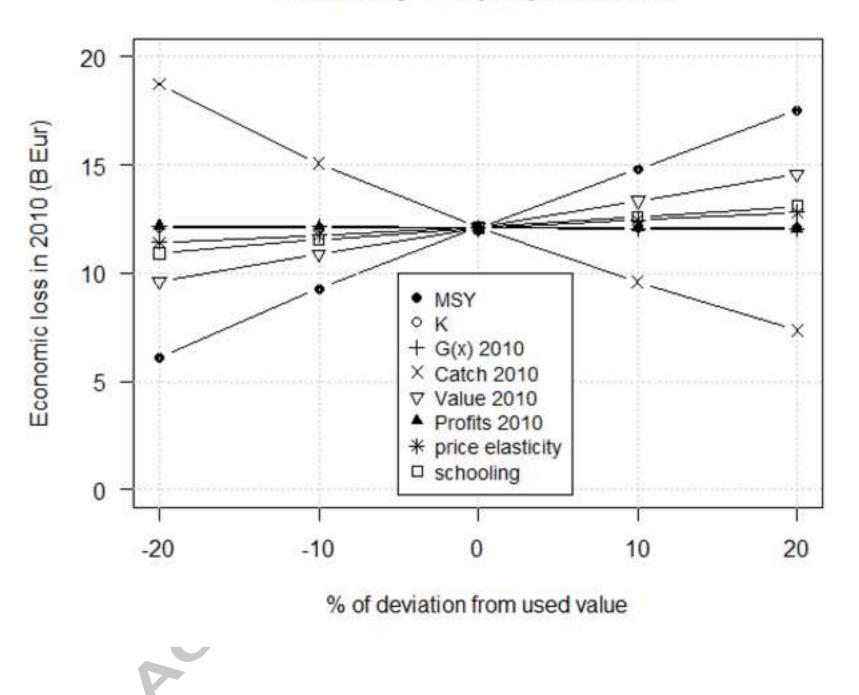


Normalized fishing effort





Sensitivity to input paramaters



Tables

838 839 840 841 842 843 Table 1. Estimated Maximum Sustainable Yield for different levels of aggregation of ICESstocks.

SIOCKS.			
Total MSY (Mean Mt, sdLog)	Habitat MSY (Mean Mt, sdLog)	Species MSY (Mean kt, sdLog)	ICES-Stock MSY (Mean kt, sdLog)
		()	cod-2224 (38.91, 0.09)
			cod-2532 (181.77, 0.06)
			cod-347d (256.87, 0.05)
			cod-arct (737.22, 0.05)
		Cod (1364.61, 0.03)	cod-coas (59.71, 0.1)
		(1504.01, 0.05)	cod-farp (24.48, 0.05)
			cod-iceg (374.87, 0.06)
			cod-iris (8.72, 0.07)
			cod-scow (16.41, 0.12)
		Greenland halibut (20.38, 0.20)	ghl-arct (20.16, 0.2)
			had-34 (281.57, 0.07)
			had-7b-k (12.61, 0.18)
			had-arct (164.18, 0.13)
		Haddock (419.40, 0.04)	had-faro (18.11, 0.06)
			had-iceg (81.20, 0.18)
			had-rock (4.67, 0.12)
	Demersal2.26, 0.03)	•	had-scow (23.36, 0.07)
		Megrim (1.23, 0.11)	mgb-8c9a (1.23, 0.11)
ICES (6.68, 0.07)		Anglerfish (0.46, 0.14)	mgw-8c9a (0.46, 0.14)
		Nephrops	nep-8ab (5.68, 0.11)
		(5.15, 0.09)	nop-34 (0.18, 0.13)
			ple-celt (1.18, 0.07)
			ple-eche (4.32, 0.07)
C C C		Plaice (122.48, 0.04)	ple-echw (1.68, 0.08)
		(122.10, 0.01)	ple-iris (3.38, 0.05)
			ple-nsea (114.97, 0.04)
			sol-bisc (5.27, 0.08)
			sol-celt (1.12, 0.06)
			sol-eche (4.73, 0.11)
·		Sole (31.93, 0.06)	sol-echw (0.97, 0.05)
		(01.70, 0.00)	sol-iris (1.40, 0.08)
			sol-kask (0.77, 0.09)
			sol-nsea (21.4, 0.04)
		Capelin (867.50, 0.07)	cap-icel (866.05, 0.07)
	Pelagic	Herring	her-2532-gor (245.28, 0.09)
	(4.99 Mt, 0.08)	(2028.94, 0.13)	her-30 (61.63, 0.15)
			her-3a22 (121.54, 0.15)

			L 47.12 (520.740.07)
			her-47d3 (538.76, 0.07)
			her-irls (20.81, 0.05)
			her-noss (1202.19, 0.2)
			her-riga (33.45, 0.12)
			her-vasu (92.75, 0.09)
			her-vian (79.26, 0.06)
		Horse Mackerel (260.41, 0.09)	hom-west (259.44, 0.09)
		Mackerel (697.43, 0.07)	mac-nea (696.15, 0.07)
			sai-3a46 (160.77, 0.06)
		Saithe	sai-arct (193.75, 0.08)
		(366.03, 0.04)	sai-faro (53.57, 0.15)
			sai-icel (68.40, 0.09)
			san-ns1 (357.36, 0.09)
		Sandeel (646.50, 0.08)	san-ns2 (72.04, 0.11)
		(040.30, 0.08)	san-ns3(259.59, 0.09)
		Sardine (149.70, 0.10)	sar-soth (150.34, 0.1)
		Sprat (378.03)	spr-2232 (375.41, 0.13)
		Blue whiting	whb-comb (1241.12, 0.16)
		(1254.76, 0.16)	whg-47d (37.18, 0.14)
		Whiting (39.74, 0.09)	whg-7e-k (13.09, 0.08)
844 845			
P			

Table 2. Data used to parameterize the bioeconomic model. Catch and value were built combining data from international databases (FAO and Sea Around Us). Cost parameters are from Lam et al (2011). Additionally, the net profits of EU member states as reported in the Annual Economic Report on the European Fishing Fleet (JRC, 2012).

850		· · · · · · · · ·	I	8	-, - ,		
Country	Catch (t)	Value of catch	Operational	Total Costs	Estimated Op.	Estimated	Net Profits
	(2010)	(MEur) (2010)	costs (Eur/t)	(Eur/t)	profits (MEur)	Net Profits	(AER, 2010)
						(MEur)	
Belgium	21,907	78.40	1,007.9	2,139.1	56.32	31.54	-8.2
Denmark	827,936	309.88	1,001.4	1,147.4	-519.23	-640.12	-34.1
Estonia	89,752	32.61	765.2	936.1	-600.90	-742.46	8.1
Finland	121,169	24.54	980.1	1,176.7	-63.42	-81.07	1.2
France	312,162	731.22	1,053.5	1,259.9	603.58	578.57	-5.5
Germany	193,536	211.31	1,108.4	1,324.9	-134.68	-202.26	2.0
Iceland	1,057,988	697.00	1,119.6	1,352.6	480.32	435.23	-
Ireland	285,527	63.65	981.6	1,169.0	-974.82	-1,173.11	-33.4
Latvia	77,085	14.72	644.0	803.1	-169.16	-214.59	4.5
Lithuania	21,371	19.92	673.0	828.5	-31.96	-43.95	8.2
Netherlands	283,377	49.83	1,004.8	1,206.4	28.36	24.05	4.6
Norway	2,555,186	5,027.08	610.7	733.8	4,854.01	4,819.15	-
Poland	113,579	223.46	874.1	1,067.1	-2,009.95	-2,503.27	30.9
Portugal	201,730	102.33	801.4	949.8	11.31	-5.55	-38
Russian Fed	997,827	1,305.52	1,276.0	1,533.7	1,048.10	996.12	-
Spain	399,448	245.59	1,211.3	1,427.4	-963.10	-1,178.74	-250.2
Sweden	210,552	137.83	567.3	757.6	-88.77	-164.80	-1.5
UK	578,677	293.55	973.5	1,167.5	88.57	47.72	74.6
Canada	1,046,985	1,416.26	1,652.3	1,825.5	460.11	359.91	-
US	743,143	1,444.67	1,435.4	1,604.5	377.93	252.28	-

Table 3. Parameters for the bioeconomic model (Arnason, 2007; Arnason et al., 2009),

primarily derived from data for 2010 (see text).

		X 7 1	Unit. 8 58
Parameter	Explanation	Value	Unitsoso
MSY		13.7	Mt 859
K	Carrying capacity	170	Mt 860
G(x) ₂₀₁₀	Biomass growth	0.283	t/yr 861
Y ₂₀₁₀	Fisheries catches	10.14	Mt
Value ₂₀₁₀	Value of catches	12.43	M€
Price ₂₀₁₀	Base price of catch	1.22	€/t
Profits ₂₀₁₀	Fishery profits	0.63	B€
C ₂₀₁₀	Cost of effort	11.8	M€
Price elasticity	Price elasticity	0.2	€/Mt
Schooling (b)	Schooling parameter	1	-
Ef ₂₀₁₀	Fishing effort	1	Normalized
	D	2	

- 862 Table A1. Necessary transformations to run the bioeconomic model using the parameters shown
- 863 in Table 3 (Arnason, 2007; Arnason et al., 2009; Gordon, 1954; Schaefer, 1954).

Biological parameters	
	$\alpha = 4 \cdot \frac{MSY}{K}; \ \beta = \frac{\alpha}{K}$
	K = K
Biomass in 2010	
	$B_{2010} = (\alpha/2\beta) \cdot \left(1 - \left(1 - \frac{4\beta(Y_{2010} + G(x)_{2010})}{\alpha^2}\right)^{0.5}\right)$
	$B_{2010} = (\alpha/2\beta) \cdot (1 - (1 - \frac{\alpha^2}{\alpha^2}))$
Price in 2010	
	$p_{2010} = Value_{2010}/Y_{2010}$
	P2010 · · · · · · · · · · · · · · · · · ·
Cost of unit of effort	
Cost of unit of effort	
	$c = \frac{Value_{2010} - \Psi_{2010}}{Ef_{2010}}$
<u>Ostabal ilitaria 2010</u>	
Catchability in 2010	
	$q_{2010} = Y_{2010} / (B_{2010} \cdot Ef_{2010})$
Biomass at equilibrium	
	$B_{eq} = (\alpha - q_{2010} E f_{eq}) \cdot K/\alpha$
Catch at equilibrium	
	$Y_{eq} = q_{eq} E f_{eq} B^b_{eq}$
Price at equilibrium	
	Beg .
	$p_{eq} = p_{2010} (\frac{B_{eq}}{B_{2010}})^b$
7	
Profits at equilibrium	$\Psi_{eq} = Y_{eq} p_{eq} - cE f_{eq}$
··· ··· · 1·· ·····	