

- 1 Manuscript submitted as <u>Research Report</u> to *Marine Pollution Bulletin*
- 2 Combined effects of fishing and oil spills on marine fish: role of stock
- 3 demographic structure for offspring overlap with oil
- 4 Running head: Demography and oil spill effects
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17 Abstract

18	It has been proposed that the multiple pressures of fishing and petroleum activities						
19	impact fish stocks in synergy, as fishing-induced demographic changes in a stock may						
20	lead to increased sensitivity to detrimental effects of acute oil spills. High fishing						
21	pressure may erode the demographic structure of fish stocks, lead to less diverse						
22	spawning strategies, and more concentrated distributions of offspring in space and						
23	time. Hence an oil spill may potentially hit a larger fraction of a year-class of offspring.						
24	Such a link between demographic structure and egg distribution was recently						
25	demonstrated for the Northeast Arctic stock of Atlantic cod for years 1959–1993. We						
26	here estimate that this variation translates into a two-fold variation in the maximal						
27	proportion of cod eggs potentially exposed to a large oil spill. With this information it is						
28	possible to quantitatively account for demographic structure in prospective studies of						
29	population effects of possible oil spills.						
30	Key words: oil spill; fishing; multiple stressors; fish eggs; Atlantic cod Gadus morhua;						
31	Barents Sea						
32	Highlights						
33	• We quantify maximal potential overlap between fish eggs and hypothetical oil						
34	spills						
35	• Maximal overlap is highest when the spawning stock is dominated by small fish						
36	• Fishing may thus influence sensitivity to oil spills through effect on demography						
37	Our results can be used in prospective studies to correct for this effect						

38 Introduction

Multiple stressors such as over-exploitation and pollution often impact natural systems 39 non-additively, implying a need to study such impacts in concert (Crain et al., 2008). 40 41 High fishing pressure has led to demographic changes in many fish stocks, towards 42 increased dominance of young and small spawners (Law, 2000; Longhurst, 2002; 43 Berkeley et al., 2004; Ottersen, 2008). It is feared that heavy fishing thereby increases the stocks' sensitivity to environmental influences, including effects of acute oil spills 44 (Hjermann et al., 2007). Specifically, erosion of demographic structure may reduce the 45 diversity of spawning strategies and the spatiotemporal distribution of eggs and larvae 46 (Kjesbu et al., 1992; Opdal, 2010; Opdal and Jørgensen, 2015), which are life-stages 47 48 thought to be particularly sensitive to toxic oil compounds (e.g., Carls et al., 1999; 49 Sørhus et al., 2015). Hence, the proportional overlap between these sensitive early 50 life-stages and oil in the case of an oil spill may increase. However, quantitative 51 knowledge on how erosion of spawning stock structure influences potential overlap of 52 offspring with oil is scarce.

53 The Northeast Arctic (NEA) stock of Atlantic cod Gadus morhua is currently the 54 world's largest and of high economic and ecological importance (Kjesbu et al., 2014). Spawning occurs along the west and north coasts of Norway from mid-February to 55 56 early May (Ottersen et al., 2014) and the eggs and larvae drift pelagically north- and eastwards towards the Barents Sea nursery area (Olsen et al., 2010). The drift path of 57 58 the eggs and larvae crosses areas with ongoing oil activities as well as areas that are 59 closed for such activities due to concern for fisheries and the environment – a topic of 60 political and scientific debate (Misund and Olsen, 2013; Blanchard et al., 2014).

61 Statistical analyses of egg survey data for NEA cod for 1959–1993 revealed 62 positive associations of distributional extent of cod eggs with mean weight (and 63 alternatively, age) in the spawning stock, spawning stock biomass and a liver condition 64 index (Stige et al., 2017). We here build on results of Stige et al. (2017) and use the 65 same egg survey data to quantify in more detail how changes in mean weight and 66 biomass of spawners are likely to influence the egg distribution and thereby the 67 potential overlap between eggs and oil. We first consider a case study where overlap 68 between oil and cod eggs is simulated for a large oil spill near the main spawning 69 grounds of NEA cod for one year, and assess how hypothetical changes in egg 70 distribution associated with demographic variables influences overlap calculations. 71 Subsequently, we construct an index of "worst-case" overlap rate by identifying the 72 areas with highest cod egg concentrations and calculating how large a fraction of a 73 year-class is maximally contained within an area of a given size. We then assess how 74 this fraction depends on spawning stock biomass and mean weight of spawners. We thus quantify the roles of stock size and demographic structure in influencing potential 75 76 year-class susceptibility to geographically bounded events such as oil spills. We 77 hypothesize that both low mean weight and low total biomass of spawners lead to 78 increased susceptibility to oil spills.

79 Methods

80 Data

Eggs of NEA cod were sampled during dedicated ichthyoplankton surveys by the Polar
Research Institute of Marine Fisheries and Oceanography (PINRO), Murmansk
(Mukhina et al., 2003). The survey covered main drift areas of eggs and larvae of NEA
cod between 67°30'N and 74°30'N from about 7 km (4 nautical miles) to 500 km from

the coast. From around 10 % to 25 % of the landings from the fisheries on spawning 85 fish in years 1959–1969 were from south of the survey area (Opdal, 2010), with the 86 87 long-term trends in the proportion apparently covarying with the mean age of the spawners [(Opdal and Jørgensen, 2015) but see (Sundby, 2015)]. The survey was 88 89 conducted in April–May, i.e. 0–2 months after the peak spawning of the cod (Ellertsen 90 et al., 1989), each year from 1959 to 1993, except 1967, when there was no survey. On 91 average 156 stations were sampled each year, but with considerable variability among 92 years in the extent and timing of the survey (Mukhina et al., 2003; Stige et al., 2015). 93 Cod eggs were classified into four developmental stages based on morphology. Stage-1 94 eggs could not be reliably differentiated from the eggs of haddock. Stage-1 eggs were 95 therefore classified to species according to the fraction of cod compared to haddock 96 eggs of stages 2–4 in the sample. For further details on the ichthyoplankton data we 97 refer the reader to Mukhina et al. (2003) and Stige et al. (2015).

Spawning stock biomass (*SSB*, tonnes) data were obtained from ICES (2009). *SSB* is computed using values for stock number at age from extended survivors analysis
(XSA) based mainly on fisheries data, weight-at-age in the stock and maturity-at-age,
calculated as weighted averages from Russian and Norwegian surveys during the
autumn and winter seasons (Marshall et al., 2006; ICES, 2009). We used logtransformed biomass, *logSSB* = log_e(*SSB*), hence assuming a log-linear relationship with
egg abundance in the statistical analysis.

105 Mean biomass-weighted weight in the spawning stock (\overline{W} , kg) was calculated 106 from abundance-at-age estimated by XSA, weight-at-age and maturity-at-age, all from 107 ICES (2009):

108 (1)
$$\overline{W}_j = \frac{\sum_{a=3}^{a=13^+} W_{aj}(N_{aj}W_{aj}M_{aj})}{\sum_{a=3}^{a=13^+} (N_{aj}W_{aj}M_{aj})}$$

Here, N_{aj} , W_{aj} and M_{aj} are, respectively, number, mean weight (kg) and proportion mature at age a in year j. The product $(N_{aj}W_{aj}M_{aj})$ is thus mature biomass-at-age and the denominator sums up to SSB_j . By weighting by biomass and not abundance of each age class, \overline{W} represents the sizes that dominate the spawning stock in terms of potential egg production. \overline{W} is highly correlated with mean age in the spawning stock (product-moment correlation, r = 0.92).

The liver condition index (*COND*, %) is liver wet weight, measured as
percentage of total wet weight for cod of lengths 41–70 cm for January–December the
year before spawning (Yaragina and Marshall, 2000).

118 Statistical analysis of how spawning stock variables influence egg distribution

119 To quantify the change in spatial distribution of cod eggs under contrasting biomass 120 and size structure in the spawning stock, we fit a spatiotemporal statistical model to 121 the cod egg data. Following results of time-series analyses by Stige et al. (2017) identifying significant predictors of cod egg distributional extent, we included sampling 122 day-of-year (Day), sampling location (Lon, $^{\circ}$ N, and Lat, $^{\circ}$ E), COND, \overline{W} and logSSB as 123 124 predictor variables. Following the same results, no abiotic environmental variables 125 were included. The spatiotemporal statistical model was used to estimate the spatial distribution of cod eggs as function of \overline{W} and *logSSB* and mean values of the other 126 127 predictor variables. Specifically, the expected stage-specific and total egg abundances at different locations in the survey area at a date representing a peak in observed egg 128 abundance halfway through their development (10th April) were calculated by 129 130 multiplying estimated probabilities from a binomial submodel with estimated 131 conditional abundances from a lognormal submodel. The spatiotemporal statistical model is described in detail in the Appendix. 132

133 Simulation of overlap between oil and fish offspring

134 To illustrate how spawning stock size and demographic structure can be accounted for 135 in oil spill simulations we used results from Vikebø et al. (2014), who modelled overlap 136 between oil compounds and eggs and larvae of NEA cod for four hypothetical oil spill scenarios, all simulated for the same year (i.e., 1997). The modelling is described in 137 138 details by Vikebø et al. (2014) and only summarised here. Specifically, 94 500 particles 139 each representing a large number of cod eggs were released at the known spawning 140 grounds and in the spawning period (1 March – 30 April) of NEA cod and transported 141 horizontally based on their vertical positioning in the water column and ocean currents. 142 Ocean currents were simulated using a regional ocean model system for the Nordic Seas with resolution 4×4 km (Lien et al., 2014). The transport and fate of oil 143 144 compounds were simulated based on the same ocean model. We here investigated 145 two oil spill scenarios representing a large oil spill at the peak of the spawning season 146 (i.e., 4500 tonnes of oil per day for 30 days, 1–30 April) but differing in oil spill location (scenario 1: N 68.67, E 13.92, scenario 2: N 68.83, E 13.45). Two other scenarios with 147 148 oil spill locations farther south investigated by Vikebø et al. (2014) were not analysed 149 here because the majority of the impacted eggs were outside of the survey area (N 150 67.5–74.5, E 8–31.5). For each particle we found the maximal concentration of total 151 polycyclic aromatic hydrocarbon (TPAH) along its drift trajectory through the egg and 152 larval stages, here using the highest concentration in the water column (Vikebø et al., 2014, also considered ambient concentrations at the depths of the particles). Overlap 153 154 was calculated as percentage of individuals having maximal TPAH concentration above 155 thresholds of 0.1 parts per billion (ppb) or 1.0 ppb, representing, respectively, order of 156 magnitude thresholds for sublethal and lethal effects. Note that work is still ongoing to refine these values for different stages and species. 157

158 The simulation results of Vikebø et al. (2014) represent a historical average situation in terms of spawning stock size and -structure. To assess the effect of 159 altering \overline{W} or *logSSB* we weighted the particles, i.e. the number of individuals each 160 particle represented, based on the results of the statistical analysis of how spawning 161 162 stock variables influence egg distribution. The weighting was based on the location of 163 the particles half-way through the egg development, i.e., around the transition from 164 the second to the third egg stage. To simulate the egg distribution expected under high 165 \overline{W} each particle was weighted by the predicted egg abundance for that location for the 90th percentile of \overline{W} divided on the prediction for the same location for mean \overline{W} . The 166 167 predictions were for total numbers of stage-2 and stage-3 eggs at April 10, which was between the peaks of abundance for these two stages. Note that effects of \overline{W} and 168 logSSB on egg distribution in the statistical model were assumed to be independent of 169 170 day-of-year and egg stage (at the linear scales of the predictors in the binomial and 171 lognormal submodels); hence the application of these weights based on locations at a single developmental stage independent of when that stage is reached is consistent 172 with the statistical model. Particles outside of the survey area (representing 20 % of 173 the individuals) were excluded from the analysis in order to avoid extrapolation. 174 175 Subsequently we calculated the fraction of the year-class exposed to sublethal or 176 lethal concentrations of oil for each oil spill scenario. Corresponding calculations were made for the 10th percentile of \overline{W} and for the 10th and 90th percentile of *logSSB*. 177

178 How does "worst-case" potential overlap rate depend on spawning stock variables?

While the oil spill simulation illustrates the role of stock size and demographic
structure for two oil spill scenarios, other scenarios with the same spatial extent but
different locations of oil could conceivably give higher overlap with cod eggs and larvae

(e.g., Carroll et al., 2018). We thus constructed an index of maximum number of cod 182 eggs contained within an area representative of a large oil spill and investigated how 183 this index varied with spawning stock variables. To construct this index we used the 184 185 spatiotemporal statistical model of how spawning stock variables influence egg distribution and calculated stage-specific as well as total egg abundance for a grid at 186 187 fixed 1° longitude and 1/3° latitude intervals over the study area. The area represented by each grid cell is given by $Area_i = 20 \cdot 1.852 \cdot 60 \cdot 1.852 \cdot \cos(\pi \cdot Lat_i/180)$ 188 km² and varied from 1111 km² in the north to 1564 km² in the south. For given values 189 190 of \overline{W} and *logSSB* we ranked the grid cells according to expected egg concentration and 191 calculated the cumulative number of expected eggs as function of cumulative area. 192 Note that these grid cells were not necessarily contiguous. We then compared the maximal proportions of the eggs contained within 10 000 km² or 40 000 km² 193 194 dependent on \overline{W} and *logSSB*. These area sizes represent the approximate range in surface coverage of oil components at lethal concentrations in a large oil spill scenario 195 196 for the region (Vikebø et al., 2014; Langangen et al., 2017).

197In the presence of strong compensatory density dependence locally (e.g.,198Ciannelli et al., 2007), distribution area may hypothetically be a better proxy for year-199class strength than abundance. We therefore also calculated maximal proportion of200the total cod egg distribution area (km²) contained within 10 000 km² or 40 000 km².201For this calculation, egg occupancy area was calculated for each grid cell by multiplying202grid cell area with expected probability of cod egg occurrence dependent on \overline{W} and203*logSSB*.

The uncertainty of the estimates was quantified by non-parametric bootstrap, whereby we generated 1000 bootstrap data sets by resampling years (with

206 replacement) and analysed the bootstrap data sets with the same procedure as the207 original data.

The analyses were performed using the programming environment R version 3.2.4 (R Core Team, 2016). The mgcv package version 1.8-12 (Wood, 2006) in R was used for generalized additive modelling.

211 **Results**

212 The size and demographic structure of the spawning stock of NEA cod varied

considerably during the studied period 1959–1993 (Fig. S1), with SSB varying by a

factor of 9 between 0.10 million tonnes (*logSSB* = 11.5, in 1965) and 0.89 million

tonnes (*logSSB* = 13.7, in 1992), and \overline{W} varying between 2.8 kg (in 1990) and 7.9 kg (in

1974). The variation in *logSSB* and \overline{W} was uncorrelated (product-moment correlation r

217 = -0.04), allowing us to study each factor independently.

The generalized additive model results showed a wider spatial distribution of cod eggs at high compared to low \overline{W} and at high compared to low *logSSB* (Fig. 1). We find that at high *logSSB*, cod egg concentrations were particularly high off the main spawning grounds at 68–70 °N, while there was no such peak at low *logSSB* (Fig. 1).

Results of the oil spill simulation showed that adjusting the egg distribution to that expected at low or high \overline{W} or low or high *logSSB* changed the calculated oil spill effects by approximately ±5 % for the scenarios considered (see **Table 1** for exact numbers and **Fig. S2** for spatial distributions of eggs with drift paths overlapping or not with above-threshold concentrations of oil).

The maximal proportion of cod eggs contained within areas of 10 000 km² or 40 000 km² was about two times higher at the low end of the \overline{W} range compared to at

the high end of the \overline{W} range (**Fig. 2A**). Specifically, the maximal proportion of cod eggs contained within an area of 40 000 km² was 0.32 (95 % confidence intervals, c.i.: 0.24, 0.40) at \overline{W} = 7.9 kg and 0.57 (c.i.: 0.42, 0.71) at \overline{W} = 2.8 kg. The corresponding proportions for 10 000 km² were 0.13 (c.i.: 0.07, 0.21) at high \overline{W} and 0.29 (c.i.: 0.16, 0.43) at low \overline{W} . Similarly, the maximal proportion of the total distribution area contained within areas of 10 000 km² or 40 000 km² were higher at low compared to high \overline{W} (**Fig. S3**).

236 Contrary to hypothesized, the maximal proportion of cod eggs contained within areas of 10 000 km² or 40 000 km² were higher at high compared to low *logSSB* (Fig. 237 238 **2B**). This result is linked to the high concentrations of cod eggs found near the main 239 spawning grounds at high logSSB (Fig. 1). According to our estimates, nearly 80 % of the cod eggs can be contained within an area of 40 000 km² in years with high *logSSB*, 240 241 compared with maximally around 40 % in years with low logSSB. On the other hand, 242 the maximal proportion of the total distribution area contained within areas of 10 000 km² or 40 000 km² were higher at low compared to high *logSSB* (**Fig. S3**), which is a 243 244 direct result of the total distribution area being smallest at low logSSB. Results for single egg stages (Fig. S4) resembled those for total egg abundance (Fig. 2). 245

246 **Discussion**

Our results show that fishing, by influencing size and demographic structure of the
spawning stock (Law, 2000; Longhurst, 2002; Berkeley et al., 2004), may affect the
potential overlap between offspring and oil in the case of an oil spill. Such an influence
of the multiple pressures of fishing and oil has been suggested before (Hjermann et al.,
2007; Rooker et al., 2013), but the quantitative value and hence the potential
importance has until now been generally unknown.

253 The oil spill simulation illustrates how potential changes in egg distribution 254 caused by changes in stock size and demographic structure can be accounted for in 255 prospective studies. In these particular scenarios the effects of changes in spawning 256 stock variables on overlap between oil and fish offspring were found to be small. The 257 similarity in results for the scenarios considered is probably related to the alternative 258 oil spill locations being relatively close to one another and that only one year was considered. It should be noted that there appeared to be fewer simulated egg particles 259 260 in offshore regions than expected from the observation data (Fig. S2 cf. Fig. 1) and lack 261 of particles in these marginal areas could lead to underestimation of the effect of \overline{W} 262 on variability in egg distribution and hence overlap rate. Moreover, we found that that 263 worst-case overlap rate is more strongly dependent on stock size and demographic 264 structure than what overlap rate is for these scenarios.

265 Demographic structure has the clearest effect, and we find that potential 266 overlap rate, measured as maximal proportion of eggs contained within an area of size 267 as a large oil spill, varies by a factor of two in response to the near three-fold variation in \overline{W} observed for NEA cod in the 1959–1993 period. Potential overlap rate is highest 268 269 when \overline{W} is low, as the eggs are then concentrated in a smaller area than when \overline{W} is 270 high. Low \overline{W} signifies a low proportion of old and large fish in the spawning stock, a 271 commonly described consequence of high and often size-selective fishing pressure 272 (Law, 2000; Longhurst, 2002; Ottersen, 2008). A high proportion of old and large 273 spawners may lead to wide offspring distribution, by allowing for a high diversity in 274 spawning strategies (e.g., location and duration of spawning) and offspring traits (e.g., 275 viability and egg buoyancy) that influence offspring distribution (as discussed by Hixon et al., 2014; Stige et al., 2017). Note that our analysis mainly quantified effects of \overline{W} on 276 spatial distribution; a possible seasonal contraction of spawning at low \overline{W} (Wright and 277

Trippel, 2009) would tend to accentuate the negative effect of \overline{W} on potential overlap rate.

Spawning stock biomass has a less clear effect: An oil spill may potentially hit a larger fraction of the eggs but a smaller fraction of the egg distribution area when *logSSB* is high. This is because at high *logSSB*, there is a peak in egg abundance off the main spawning grounds, at the same time as the margins of the egg distribution area expand. The mechanisms behind this peak in egg abundance at high *logSSB* are not clear. The magnitude of *logSSB* effects on potential overlap rate is similar to that of \overline{W} .

286 The population consequences of an oil spill hitting eggs in parts of the distribution depend on the spatial pattern in natural mortality (Langangen et al., 2017). 287 288 Natural mortality in central parts of the distribution appears to be higher than in 289 marginal areas (Ciannelli et al., 2007; Langangen et al., 2014a); hence, a high fraction 290 of eggs killed if an oil spill hits these areas does not necessarily translate into high 291 cohort loss in the long-term. This is because eggs in the high-density area have very low survival anyway. For \overline{W} we get similar results if potential overlap rate is calculated 292 293 from total egg abundance or from distribution area, suggesting that results also hold if 294 local-scale compensatory density dependence in survival is strong. For *logSSB* on the 295 other hand, the direction of its effect on maximal long-term impact of an oil spill depends on how survival is regulated locally. 296

297 Prospective simulation studies of potential overlap between oil spills and early 298 life stages of fish (e.g., Vikebø et al., 2014; Carroll et al., 2018) typically, due to lack of 299 information, ignore effects of spawning stock size and demographic structure on 300 offspring distribution and exposure. We propose that future studies on NEA cod may 301 assess effects of possible changes in spawning stock size or demographic structure as

302 implemented in our oil spill simulation. Future studies on other stocks that lack long-303 term egg distribution data may use results in Fig. 2 to assess how large uncertainty is 304 introduced by ignoring these effects, e.g., quantitatively formalized in a Bayesian 305 Network model (cf. Carroll and Smit, 2011) or as a correction factor. For example, if a 306 study finds a maximal impact of 43 % (Carroll et al., 2018) of a year class assuming 307 average values for \overline{W} , we suggest that one can conservatively correct this value based 308 on the results presented here. Based on Fig. 2A, we suggest that adding 0.5 on logit 309 scale would correct for potential increase in impact caused by reduction in \overline{W} from the 310 average value to the minimum observed \overline{W} . Hence, we suggest correcting the assessed impact from 43 to 55 % (as logit(0.43) + 0.5 = logit(0.55)) if one wants to account for 311 312 the potential increased impact caused by a hypothetical future reduction in \overline{W} . In principle, hypothetical future changes in *logSSB* can be accounted for similarly, but 313 with a less solid theoretical underpinning than for \overline{W} : Our statistical results suggest 314 that the highest proportion of a year class of cod eggs can be lost in years with high 315 logSSB, but the causal basis for the distribution changes that drive this result is unclear, 316 317 as are the implications for long-term cohort loss.

While our results provide quantitative estimates, there are some caveats and 318 limitations. First, we stress that the analysis of potential overlap rate (Fig. 2) indicates 319 320 how \overline{W} or *logSSB* affect "worst-case" effects of oil spills. These "worst-case" effects 321 can be thought of as the upper tails in the probability distribution for overlap rate if an oil spill of a given size hits a random part of the study area: With low \overline{W} this tail is 322 323 longer. The analysis therefore does not necessarily reflect how the most likely effect of a given oil-spill scenario varies with \overline{W} or *logSSB*, as indeed illustrated by the oil spill 324 325 simulation.

326 Further, our calculated potential overlap rate should be interpreted as an index and not as an accurate estimate of maximal overlap between fish offspring and oil for 327 a given oil spill size; for such estimates simulation modelling is needed for obtaining 328 329 more realistic distributions of oil and appropriately accounting for temporal aspects 330 and small-scale patchiness. To throw more light on the temporal aspect we assessed 331 the spatial distribution of simulated eggs in the oil spill model at a snap-shot in time, 332 finding that eggs having above-threshold oil concentrations at some point in their egg 333 and larval drift trajectory were concentrated in areas of similar size as used in our 334 calculation of potential overlap (Fig. S5). We therefore consider that our results are 335 also relevant for oil spills occurring near the drift paths of eggs and larvae over an 336 extended period. We find it unlikely that our simplifying assumption that overlap 337 between fish offspring and oil only depends on the horizontal, and not vertical, 338 locations of fish offspring and oil interferes with our conclusions regarding effects of \overline{W} 339 or logSSB on overlap rate. As shown by Vikebø et al. (2014, their Table 5), relaxing this 340 assumption will, in general, reduce the overlap rate by roughly a third, independent of 341 the area affected by oil. In sum, we therefore believe that our estimates are a good 342 approximation, with a clear empirical basis, of how the mean weight of the spawners and spawning stock biomass may influence maximal contact rate between fish 343 344 offspring and oil.

We conclude that loss of old and large fish from a stock due to high fishing pressure may increase its sensitivity to oil spills. Such a synergistic effect of fishing and oil is both mediated by the potential overlap between offspring and oil (this study) and by age-truncated fish stocks having reduced demographic buffering against recruitment failure (Rouyer et al., 2011; Ohlberger and Langangen, 2015). This conclusion underlines the multiple benefits of good fisheries management, which has

- 351 contributed to that the biomass of NEA cod now is high and the proportion of old and
- 352 large fish appears to be on the increase (Kjesbu et al., 2014).

353 Supplementary Material

- 354 The following figures are available as supplementary online material:
- 355 Figure S1. Spawning stock biomass (*logSSB*) and mean weight in the spawning stock
- 356 (\overline{W}) of NEA cod from 1959 to 1993.
- 357 Figure S2. Simulated distribution of NEA cod eggs at their locations halfway through
- 358 the egg development, showing which eggs experience sublethal or lethal oil
- 359 concentrations for two different oil spill scenarios.
- 360 Figure S3. Maximal proportion of the total distribution area of NEA cod eggs contained
- within areas of 10 000 km² or 40 000 km² dependent on \overline{W} or *logSSB*.
- **Figure S4.** Maximal proportion of NEA cod eggs contained within areas of 10 000 km²
- or 40 000 km² dependent on \overline{W} or *logSSB*, shown for each egg developmental stage.
- 364 Figure S5. Simulated distribution of NEA cod eggs at a snap-shot in time (April 10,
- 1997), showing which eggs experience sublethal or lethal oil concentrations at some
- 366 part of their egg and larval drift trajectories for two different oil spill scenarios.

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373 **References**

374 Berkeley, S.A., Hixon, M.A., Larson, R., Love, M.S., 2004. Fisheries sustainability via 375 protection of age structure and spatial distribution of fish populations. Fisheries 376 29, 23-32. 377 Blanchard, A., Hauge, K.H., Andersen, G., Fosså, J.H., Grøsvik, B.E., Handegard, N.O., 378 Kaiser, M., Meier, S., Olsen, E., Vikebø, F., 2014. Harmful routines? Uncertainty 379 in science and conflicting views on routine petroleum operations in Norway. 380 Mar. Policy 43, 313-320. Carls, M.G., Rice, S.D., Hose, J.E., 1999. Sensitivity of fish embryos to weathered crude 381 382 oil: Part I. Low-level exposure during incubation causes malformations, genetic 383 damage, and mortality in larval pacific herring (Clupea pallasi). Environ. Toxic. Chem. 18, 481-493. 384 385 Carroll, J., Smit, M., 2011. An Integrated Modeling Framework For Decision Support In 386 Ecosystem-Based Management: Case Study Lofoten/Barents Sea. Society of 387 Petroleum Engineers. doi: 10.2118/140431-MS Carroll, J., Vikebø, F., Howell, D., Broch, O.J., Nepstad, R., Augustine, S., Skeie, G.M., 388 Bast, R., Juselius, J., 2018. Assessing impacts of simulated oil spills on the 389 390 Northeast Arctic cod fishery. Mar. Poll. Bull. 126, 63-73. 391 Ciannelli, L., Dingsør, G.E., Bogstad, B., Ottersen, G., Chan, K.-S., Gjøsæter, H., Stiansen, 392 J.E., Stenseth, N.C., 2007. Spatial anatomy of species survival: effects of 393 predation and climate-driven environmental variability. Ecology 88, 635-646. 394 Crain, C.M., Kroeker, K., Halpern, B.S., 2008. Interactive and cumulative effects of 395 multiple human stressors in marine systems. Ecol. Lett. 11, 1304-1315. 396 Ellertsen, B., Fossum, P., Solemdal, P., Sundby, S., 1989. Relation between temperature 397 and survival of eggs and first-feeding larvae of northeast Arctic cod (Gadus 398 morhua L.). Rapp. P.-v. Réun. Cons. Int. Explor. Mer 191, 209-219. 399 Hastie, T., Tibshirani, R., 1993. Varying-coefficient models. J. R. Stat. Soc. B 55, 757-796. 400 Hixon, M.A., Johnson, D.W., Sogard, S.M., 2014. BOFFFFs: on the importance of 401 conserving old-growth age structure in fishery populations. ICES J. Mar. Sci. 71, 402 2171-2185.

403	Hjermann, D.Ø., Melsom, A., Dingsør, G.E., Durant, J.M., Eikeset, A.M., Røed, L.P.,					
404	Ottersen, G., Storvik, G., Stenseth, N.C., 2007. Fish and oil in the Lofoten-					
405	Barents Sea system: synoptic review of the effect of oil spills on fish populations.					
406	Mar. Ecol. Prog. Ser. 339, 283-299.					
407	ICES, 2009. Report of the Arctic fisheries working group (AFWG), 21 - 27 April 2009,					
408	San-Sebastian, Spain. ICES CM 2009\ACOM:01.					
409	Kjesbu, O.S., Bogstad, B., Devine, J.A., Gjøsæter, H., Howell, D., Ingvaldsen, R.B., Nash,					
410	R.D.M., Skjæraasen, J.E., 2014. Synergies between climate and management for					
411	Atlantic cod fisheries at high latitudes. Proc. Natl. Acad. Sci. USA 111, 3478-					
412	3483.					
413	Kjesbu, O.S., Kryvi, H., Sundby, S., Solemdal, P., 1992. Buoyancy variations in eggs of					
414	Atlantic cod (Gadus morhua L.) in relation to chorion thickness and egg size:					
415	theory and observations. J. Fish Biol. 41, 581-599.					
416	Langangen, Ø., Olsen, E., Stige, L.C., Ohlberger, J., Yaragina, N.A., Vikebø, F.B., Bogstad,					
417	B., Stenseth, N.C., Hjermann, D.Ø., 2017. The effects of oil spills on marine fish:					
418	Implications of spatial variation in natural mortality. Mar. Poll. Bull., doi:					
419	10.1016/j.marpolbul.2017.1003.1037.					
420	Langangen, Ø., Stige, L.C., Yaragina, N.A., Ottersen, G., Vikebø, F.B., Stenseth, N.C.,					
421	2014a. Spatial variations in mortality in pelagic early life stages of a marine fish					
422	(Gadus morhua). Progr. Oceanogr. 127, 96-107.					
423	Langangen, Ø., Stige, L.C., Yaragina, N.A., Vikebø, F., Bogstad, B., Gusdal, Y., 2014b. Egg					
424	mortality of Northeast Arctic cod (Gadus morhua) and haddock					
425	(Melanogrammus aeglefinus). ICES J. Mar. Sci. 71, 1129–1136.					
426	Law, R., 2000. Fishing, selection, and phenotypic evolution. ICES J. Mar. Sci. 57, 659-					
427	668.					
428	Lien, V.S., Gusdal, Y., Vikebø, F.B., 2014. Along-shelf hydrographic anomalies in the					
429	Nordic Seas (1960–2011): locally generated or advective signals? Ocean Dyn. 64,					
430	1047-1059.					
431	Longhurst, A., 2002. Murphy's law revisited: longevity as a factor in recruitment to fish					
432	populations. Fish. Res. 56, 125-131.					

433	Marshall, C.T., Needle, C.L., Thorsen, A., Kjesbu, O.S., Yaragina, N.A., 2006. Systematic					
434	bias in estimates of reproductive potential of an Atlantic cod (Gadus morhua)					
435	stock: implications for stock-recruit theory and management. Can. J. Fish. Aquat.					
436	Sci. 63, 980-994.					
437	Misund, O.A., Olsen, E., 2013. Lofoten-Vesterålen: for cod and cod fisheries, but not for					
438	oil? ICES J. Mar. Sci. 70, 722-725.					
439	Mukhina, N.V., Marshall, C.T., Yaragina, N.A., 2003. Tracking the signal in year-class					
440	strength of Northeast Arctic cod through multiple survey estimates of egg,					
441	larval and juvenile abundance. J. Sea Res. 50, 57-75.					
442	Ohlberger, J., Langangen, Ø., 2015. Population resilience to catastrophic mortality					
443	events during early life stages. Ecol. Applic. 25, 1348-1356.					
444	Olsen, E., Aanes, S., Mehl, S., Holst, J.C., Aglen, A., Gjøsæter, H., 2010. Cod, haddock,					
445	saithe, herring, and capelin in the Barents Sea and adjacent waters: a review of					
446	the biological value of the area. ICES J. Mar. Sci. 67, 87-101.					
447	Opdal, A.F., 2010. Fisheries change spawning ground distribution of northeast Arctic					
448	cod. Biol. Lett. 6, 261-264.					
449	Opdal, A.F., Jørgensen, C., 2015. Long-term change in a behavioural trait: truncated					
450	spawning distribution and demography in Northeast Arctic cod. Global Change					
451	Biol. 21, 1521-1530.					
452	Ottersen, G., 2008. Pronounced long-term juvenation in the spawning stock of Arcto-					
453	Norwegian cod (Gadus morhua) and possible consequences for recruitment.					
454	Can. J. Fish. Aquat. Sci. 65, 523-534.					
455	Ottersen, G., Bogstad, B., Yaragina, N.A., Stige, L.C., Vikebø, F.B., Dalpadado, P., 2014.					
456	A review of early life history dynamics of Barents Sea cod (Gadus morhua). ICES					
457	J. Mar. Sci. 71, 2064-2087.					
458	R Core Team, 2016. R: A language and environment for statistical computing. R					
459	Foundation for Statistical Computing, Vienna, Austria. URL <u>https://www.R-</u>					
460	project.org/.					
461	Rooker, J.R., Kitchens, L.L., Dance, M.A., Wells, R.J.D., Falterman, B., Cornic, M., 2013.					
462	Spatial, Temporal, and Habitat-Related Variation in Abundance of Pelagic Fishes					

- 463 in the Gulf of Mexico: Potential Implications of the Deepwater Horizon Oil Spill.
 464 PLOS ONE 8, e76080.
- Rouyer, T.A., Ottersen, G., Durant, J.M., Hidalgo, M., Hjermann, D.Ø., Persson, J., Stige,
 L.C., 2011. Shifting dynamic forces in fish stock fluctuations triggered by age
- 467 truncation? Global Change Biol. 17, 3046-3057.
- 468 Stefánsson, G., 1996. Analysis of groundfish survey abundance data: combining the
 469 GLM and delta approaches. ICES J. Mar. Sci. 53, 577-588.
- 470 Stige, L.C., Langangen, Ø., Yaragina, N.A., Vikebø, F.B., Bogstad, B., Ottersen, G.,
- 471 Stenseth, N.C., Hjermann, D.Ø., 2015. Combined statistical and mechanistic
 472 modelling suggests food and temperature effects on survival of early life stages
- 473 of Northeast Arctic cod (*Gadus morhua*). Progr. Oceanogr. 134, 138-151.
- Stige, L.C., Yaragina, N.A., Langangen, Ø., Bogstad, B., Stenseth, N.C., Ottersen, G.,
 2017. Effect of a fish stock's demographic structure on offspring survival and
- 476 sensitivity to climate. Proc. Natl. Acad. Sci. USA 114, 1347-1352.
- 477 Sundby, S., 2015. Comment to 'Opdal AF, Jørgensen C (2015) Long-term change in a
- 478 behavioural trait: truncated spawning distribution and demography in

479 Northeast Arctic cod. Global Change Biology, 21:4, 1521–1530, doi:

480 10.1111/gcb.12773'. Global Change Biol. 21, 2465-2466.

- 481 Sørhus, E., Edvardsen, R.B., Karlsen, Ø., Nordtug, T., van der Meeren, T., Thorsen, A.,
- Harman, C., Jentoft, S., Meier, S., 2015. Unexpected Interaction with Dispersed
 Crude Oil Droplets Drives Severe Toxicity in Atlantic Haddock Embryos. PLOS
 ONE 10, e0124376.
- Vikebø, F.B., Rønningen, P., Lien, V.S., Meier, S., Reed, M., Ådlandsvik, B., Kristiansen,
 T., 2014. Spatio-temporal overlap of oil spills and early life stages of fish. ICES J.
 Mar. Sci. 71, 970-981.
- Wood, S.N., 2006. Generalized additive models: An introduction with R. Chapman and
 Hall/CRC, Boca Raton, FL.
- Wright, P.J., Trippel, E., 2009. Fishery-induced demographic changes in the timing of
 spawning: consequences for reproductive success. Fish Fisheries 10, 283-304.

Yaragina, N.A., Marshall, C.T., 2000. Trophic influences on interannual and seasonal
variation in the liver condition index of Northeast Arctic cod (*Gadus morhua*).
ICES J. Mar. Sci. 57, 42-55.

496 Appendix

497 Spatiotemporal statistical cod egg distribution model

498 To quantify the change in spatial distribution of cod eggs under contrasting biomass 499 and size structure in the spawning stock, we fit a variable-coefficient generalized 500 additive model (Hastie and Tibshirani, 1993; Wood, 2006) to the spatiotemporal cod egg data. Following results of time-series analysis by Stige et al. (2017), we included 501 502 sampling day-of-year (*Day*), sampling location (*Lon*, $^{\circ}N$, and *Lat*, $^{\circ}E$), *COND*, \overline{W} and 503 logSSB as predictor variables. As the survey data contained many stations with no eggs, 504 the data were considered to originate from two different processes: one process 505 determining the probability of a positive tow (i.e., non-zero abundance of eggs of a 506 given stage at a station) and another determining the abundance conditional on a 507 positive tow (see Langangen et al., 2014b). To account for the two processes we used a 508 hurdle model approach (Stefánsson, 1996), whereby a binomial model quantified the 509 probability of a positive tow and a lognormal model quantified abundance in positive 510 tows.

The binomial submodel quantified the probability *p* of catching at least one egg of a given stage at a station. Each data point represents presence (coded as 1) or absence (coded as 0) of one out of four egg developmental stages at one station in one year. Each station is thus represented by four data points in the analysis, one for each egg stage. This submodel can be written as:

516 (2)
$$logit(p_{s\,ij}) = \alpha_s + f_s(Day_i) + g_s(Lon_i, Lat_i) + \beta COND_j + h(Lon_i, Lat_i) \overline{W}_j + l(Lon_i, Lat_i) logSSB_j$$

518 where subscripts s, i and j represent stage, station and year, respectively. α_s is a stage-519 specific intercept. f_s and g_s are stage-specific smooth functions correcting for sampling date and location (g_s being a two-dimensional anisotropic smooth modelled as a 520 521 tensor-product of two smooth basis functions with maximally 5 knots each). Stagespecific smooths were modelled by using the flag "by=Stage" when specifying the 522 smooth. β is the coefficient for the effect of *COND*. The coefficients for the effects of 523 \overline{W} and *logSSB* were allowed to vary smoothly as functions of location. The smooth 524 function h(Lon, Lat) thus gives a location-dependent coefficient that \overline{W} is multiplied 525 526 with and l(Lon, Lat) the corresponding function for *logSSB*. The number of samples 527 taken at the station was included as offset. This binomial submodel is similar to a 528 corresponding model with a spatially-variable \overline{W} term used by Stige *et al.* (2017, Fig. 3) to visualize how cod egg distribution varies with \overline{W} . We here allowed the effects of 529 both \overline{W} and *logSSB* to vary spatially in order to focus on two factors directly influenced 530 by human activities, while the effect of COND was spatially invariable in order to limit 531 the number of parameters in the model. 532

Furthermore, in contrast to Stige et al. (2017), we also quantified how cod egg abundance within the distribution area varies with \overline{W} and *logSSB*. To do so, we modelled the natural logarithm of cod egg abundance in positive tows, $log_e(N)$, but using only non-zero counts and assuming a normal error distribution (ε). This submodel can be summarized as:

538 (3)
$$\log_e(N_{sij}) = \gamma_s + m_s(Day_i) + n_s(Lon_i, Lat_i) + \delta COND_j +$$

539
$$p(Lon_i, Lat_i) \overline{W}_j + q(Lon_i, Lat_i) \log SSB_j + \varepsilon_{sij}$$
 for $N_{sij} > 0$

540 The notation is analogous to Eq. 2. For this analysis, the natural logarithm of 541 the number of samples taken at the station was offset.

- This hurdle model was used to map the expected total egg abundance as function of \overline{W} and *logSSB*. Specifically, we calculated total egg abundance (sum of all stages) by multiplying predicted probabilities from Eq. 2 with predicted conditional abundances from Eq. 3 (taking into account that expected *N* from Eq. 3 is the exponent
- 546 of the expected $\log_e(N) + \frac{1}{2}$ variance of $\log_e(N)$).

547Table 1. Overlap of NEA cod eggs and larvae with oil as function of mean weight in548the spawning stock (\overline{W}) and spawning stock biomass (*logSSB*) for two oil spill549scenarios in year 1997 and alternative threshold concentrations of oil compounds.

550		Percentage overlap [Overlap relative to baseline]					
551	Scenario / threshold	Baseline	Low \overline{W}	High \overline{W}	Low logSSB	High <i>logSSB</i>	
552	Oil spill scenario 1 / 0.1 ppb	20.3 [1]	21.3 [1.05]	19.6 [0.97]	21.0 [1.03]	19.6 [0.97]	
553	Oil spill scenario 1 / 1.0 ppb	5.7 [1]	5.9 [1.05]	5.4 [0.96]	5.8 [1.06]	5.5 [0.96]	
554	Oil spill scenario 2 / 0.1 ppb	29.8 [1]	31.3 [1.05]	28.6 [0.96]	31.5 [1.02]	28.4 [0.96]	
555	Oil spill scenario 2 / 1.0 ppb	11.1 [1]	11.8 [1.06]	10.6 [0.96]	12.0 [1.07]	10.5 [0.95]	

556 Baseline: \overline{W} = 5.5 kg and *logSSB* = 12.4, low \overline{W} : 3.8 kg, high \overline{W} : 6.9 kg, low *logSSB*: 11.7, high *logSSB*: 13.0



558 Figure 1. Expected distribution of Northeast Arctic (NEA) cod eggs at different combinations of mean weight in the spawning stock, \overline{W} , and spawning stock 559 biomass, *logSSB*. Panels A and B represent 10th and 90th percentile of \overline{W} and 560 mean *logSSB*. Panels C and D represent 10th and 90th percentile of *logSSB* and 561 562 mean \overline{W} . Sizes of circles scale with total expected egg abundance per net haul, with grey circles representing 95 % bootstrap confidence intervals. Red and 563 blue circles show the grid cells representing, respectively, the 10 000 km² and 564 40 000 km² with highest egg abundance. 565



567Figure 2. Maximal proportion of NEA cod eggs contained within areas of 10 000568 km^2 or 40 000 km^2 dependent on (A) mean weight in the spawning stock, \overline{W} , or569(B) spawning stock biomass, *logSSB*. Hatched lines represent 95 % confidence570intervals (horizontal blue lines for 40 000 km^2 , oblique red lines for 10 000 km^2).

571 Supplementary Information

- 572 This online supplementary information file accompanies the paper:
- 573 Stige, L.C., Ottersen, G., Yaragina, N.A., Frode B. Vikebø, Stenseth, N.C., and Langangen,
- 574 Ø., 2018. Combined effects of fishing and oil spills on marine fish: role of stock
- 575 demographic structure for offspring overlap with oil. Marine Pollution Bulletin.

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579Figure S1. Spawning stock biomass (*logSSB*) and mean weight in the spawning580stock (\overline{W}) of Northeast Arctic cod from 1959 to 1993. Crosses are at 10th

581 percentile, mean and 90th percentile of variables.







594Figure S3. Maximal proportion of the total distribution area of Northeast Arctic595cod eggs contained within areas of 10 000 km² or 40 000 km² dependent on (A)596 \overline{W} or (B) *logSSB*. Hatched lines represent 95 % confidence intervals.



599Figure S4. Maximal proportion of Northeast Arctic cod eggs contained within600areas of 10 000 km² or 40 000 km² dependent on (A, C, E, G) mean weight in601the spawning stock, \overline{W} , or (B, D, F, H) spawning stock biomass, *logSSB*. Each602row shows results for one egg developmental stage. Hatched lines represent60395 % confidence intervals (horizontal blue lines for 40 000 km², oblique red604lines for 10 000 km²).



606 Figure S5. Simulated distribution of Northeast Arctic cod eggs at their locations at 10th April 1997. Red points in each panel represent eggs that experience oil 607 concentrations above a threshold concentration of 0.1 ppb (left column) or 1.0 608 609 ppb (right column) at some point in their egg and larval life. Blue points (partly 610 overlaid by the red points) represent remaining eggs. Upper and lower rows 611 represent two different oil spill scenarios with filled yellow circles showing the 612 release location. Grey lines represent the grid used to predict egg distribution 613 from the statistical model (Fig. 1). Grid cells with >50 % of particles experiencing above-threshold oil concentrations are shown in black. 614