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The effects of seasonal processes on size spectrum dynamics

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Abstract

The recent advent of dynamic size spectrum models has allowed the analysis of life processes in marine ecosystems to be carried out, without the high complexity arising from interspecies interactions within dense food webs. In this paper we use *mizer*, a size spectrum modelling framework, to investigate the consequences of including the seasonal processes of plankton blooms and batch spawning to the model dynamics.

A multispecies size spectrum model is constructed using twelve common North Sea fish species, with growth, predation and mortality explicitly modelled, before simulating both seasonal plankton blooms and batch spawning of fish (using empirical data on the spawning patterns of each species). The effect of seasonality on

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1 the community size spectrum is investigated; it is found that with seasonal pro-
2 cesses included, the species spectra are more varied over time, while the aggre-
3 gated community spectrum remains fairly similar. Growth of seasonally spawn-
4 ing mature individuals drops significantly during peak reproduction, although
5 lifetime growth curves follow non-seasonal ones closely. On analysing properties
6 of the community spectrum under different fishing scenarios, seasonality gener-
7 ally causes more varied spectrum slopes and lower yields. Under seasonal con-
8 ditions increasing fishing effort also results in greater temporal variability of fish-
9 eries yields due to truncation of the community spectrum towards smaller sizes.
10 Further work is needed to evaluate robustness of management strategies in the
11 context of a wider range of seasonal processes and behavioural strategies as well
12 as longer term environmental variability and change.

13 **Keywords: size spectrum; marine ecosystem; seasonality; spawning; plankton bloom**

1 Introduction

2 The concept of the size spectrum, established in the pioneering work of Sheldon and
3 Parsons (1967), has initiated a whole branch of research in marine ecology. In aquatic
4 systems, neglecting taxonomy and looking only at organism weights, the abundance
5 of organisms is a negative power-law distribution of the individual mass (or equiva-
6 lently size), and plotting $\log(\text{abundance})$ against $\log(\text{mass})$ gives a roughly linear fit
7 with slope -1 (Sheldon et al. 1972; Platt and Denman 1978). This regular pattern ap-
8 pears to be robust, independent of the size scale which is investigated (within marine
9 systems, although less commonly in freshwater systems (see Sprules and Barth, this
10 issue)), and the linear relationship has been observed for phytoplankton (San Martin
11 et al. 2006; Huete-Ortega et al. 2010), zooplankton (Heath 1995; Zhou et al. 2009) and
12 fish spectra (Boudreau and Dickie 1992; Jennings and Mackinson 2003).

13 Within this broad pattern there is important seasonal variation caused by changes
14 in temperature, nutrient levels and turbulence. Such environmental factors can al-
15 ter abundances of plankton and/or larger organisms, influencing the intercepts and
16 slopes of size spectra over the year (Navarro and Thompson 1995; Mari and Burd
17 1998; Cózar and Echevarría 2005). The single biggest seasonal driver of variation in
18 size spectra is the phytoplankton bloom that occurs at some stage during the year
19 (Barnes et al. 2011), usually in the spring, although smaller blooms can also occur in
20 the autumn (see Truscott 1995; Findlay et al. 2006). The bloom is characterized by an
21 increase in the phytoplankton to 5 to 10 times its usual abundance (Gasol et al. 1992;
22 Navarro and Thompson 1995; Batten et al. 2003; Huete-Ortega et al. 2010), depending
23 upon the latitude and surrounding environment, before returning to a fairly constant
24 abundance for the rest of the year. This process can take place over several days or
25 over the course of weeks, and is followed by an increase in abundance of zooplankton
26 further along the size spectrum (Heath 1995), which in turn provides a larger food
27 source for fish larvae (Cushing and Horwood 1994; Mertz and Myers 1994).

1 Spawning patterns are another important source of temporal variation in marine sys-
2 tems. It is well established that some fish species spawn only at certain times in the
3 year, such as cod, sole and sprat (see e.g. Mertz and Myers 1994; Johnson 2000; Arm-
4 strong et al. 2001); this behaviour is to take advantage of the extra food abundance
5 from annual phytoplankton plankton blooms, if Cushing's "match / mismatch" hy-
6 pothesis is to be believed (Cushing 1975; Beaugrand et al. 2003). However, different
7 species living under the same conditions have vastly different spawning patterns (see
8 Figure 1), and it is important to include in any reproductive model the range of strate-
9 gies adopted by different species.

10 Dynamic size spectrum models are increasingly being used to understand structure
11 and dynamics of marine systems, including the effects of fishing and climate change
12 (Blanchard et al. 2012; Woodworth-Jefcoats et al. 2013) and inclusion of multispecies
13 dynamics to address questions related to fisheries (Blanchard et al. 2014; Spence et al.,
14 this issue). However, so far these models have simplified reproductive processes and
15 focused on inter-annual changes in plankton levels.

16 Modelling reproduction in a size spectrum model has been accomplished in a non-
17 seasonal setting before (e.g. Maury et al. 2007a; Hartvig et al. 2011; Law et al. 2012); in
18 short, models generally used a fraction of the assimilated body mass from predation
19 to produce eggs of a fixed weight, following the dynamic energy budget (DEB) theory
20 of Kooijman (1986, 2009). This resulted in an influx of biomass at a fixed offspring
21 weight in the spectrum, following the observation that regardless of fish species, egg
22 size is fairly constant among many pelagic fish species (Ware 1975; Cury and Pauly
23 2000). Physiologically structured population models (PSPMs) often include pulsed re-
24 production, where pre-allocated mass is transformed into a batch of new cohorts at the
25 beginning of each season (see e.g. Persson et al. 1998; De Roos and Persson 2001). More
26 recently PSPMs have allowed reproduction to occur over longer discrete time intervals
27 (Huss et al. 2012; van Leeuwen et al. 2013). More complex individual-based size struc-

1 tured models have incorporated seasonal reproduction dynamics, but either focused
2 on a specific region geographically (Marzloff et al. 2009) or on consumer-resource dy-
3 namics without the effects of predation on consumers (De Roos et al. 2009; Sun and
4 de Roos 2015).

5 A recent paper by Sainmont et al. (2014) used an ODE-model approach to investigate
6 alternative strategies for seasonal reproduction (capital versus income breeding) in
7 environments with varying feeding seasons and maturity weights. In short, income
8 breeding allocates incoming mass directly to reproduction, whereas capital breeding
9 stores reserves which may be used for spawning at a later time, independent of food
10 availability. That paper found capital breeding to be the optimum strategy in higher
11 latitude environments where food availability was more variable (with a sharp spike
12 in springtime and lower levels outside of this period), with income breeding being
13 advantageous for longer feeding seasons. For more details about capital and income
14 breeding, see Jönsson (1997); Jager et al. (2008); Ejsmond et al. (2015).

15 Clearly a comprehensive simulation of the impacts of seasonality should include capi-
16 tal breeding, since this strategy can be both optimal theoretically (Sainmont et al. 2014;
17 Ejsmond et al. 2015) and, more importantly, common empirically in seasonal environ-
18 ments (e.g. Lambert and Dutil 2000). However, the purpose of this manuscript is to
19 conduct an initial exploration of the consequences, for the consumer size spectrum,
20 of seasonality in both resource availability and consumer spawning times. Given this
21 is an introductory analysis of the qualitative effects of these seasonal patterns on the
22 dynamics and behaviour of consumer spectra, the focus will be on income breeding,
23 perhaps the simplest representation of how energy is allocated between reproduction
24 and growth in the face of seasonal variation in resource availability, as modelled pre-
25 viously (Law et al. 2012; Blanchard et al. 2014; Scott et al. 2015). We expect that the
26 results of this analysis will guide future work on including and evaluating the impact
27 of capital breeding behaviour on seasonal patterning of the consumer spectrum.

1 This study builds upon previous work on the timing of larval hatching, where a fixed
2 temporal "background" spectrum was set up, and then cohorts born at different times
3 were followed to calculate the best time of year to be born in terms of fast growth and
4 low mortality, without modelling the dynamical feedbacks to the size spectrum (Pope
5 et al. 1994). That paper found it was best to be born at or before the peak in plankton
6 abundance, to avoid increased predation mortality from the fixed wave reaching the
7 size range of predators of newborns, and to stay ahead of this wave for the rest of the
8 year.

9 Here we introduce seasonality to a previously published multispecies size spectrum
10 model (Blanchard et al. 2014; Scott et al. 2015), to investigate the properties of seasonal
11 size spectra. To do this we introduce a dynamic, time-varying phytoplankton spec-
12 trum over the year, as well as seasonal spawning for twelve North Sea fish species
13 fitted to empirical spawning data. The interplay of seasonal effects with the basic life
14 processes of growth, predation, reproduction and death leads to high variability in
15 the fish species spectra compared to non-seasonal systems. As an example of the uses
16 of this dynamic model, we investigate several properties of the size spectrum under
17 seasonal and non-seasonal conditions. These include the long-term behaviour of the
18 slope of the size spectrum under different fishing regimes, as well as changes in fishing
19 yields.

20 **Methods**

21 **Setting up the size spectrum model**

22 To model seasonal reproduction of pelagic predators (typically fish) and plankton
23 blooms, a modified version of the previously published *mizer* package in R is used
24 (Scott et al. 2015), which uses the work of Blanchard et al. (2014) as its basis. The

1 model of that paper is a representation of the North Sea community, with parameters
2 calibrated to observed catch and effort data for the North Sea over the period 1985-
3 1995. Hence, this study is in part an extension of the work presented in Blanchard
4 et al. (2014), and will necessarily be conditioned by the structure and assumptions of
5 that work. Specifically:

- 6 1. model responses to both the addition of seasonal affects and to changes in fishing
7 pressure will be conditioned by how this set of North Sea species was harvested
8 over the calibration period 1985-1995;
- 9 2. the calibration procedure used in the 2014 study will have a significant effect on
10 the role that changes in egg production can have in this new study; more detail
11 on this will be provided in the description of how reproduction is represented in
12 this model.

13 The model presented here consists of two parts: a resource spectrum and a multi-
14 species consumer spectrum, both of which are dynamic and time-dependent. Here,
15 "consumer" refers to organisms within any of the fish species spectra, and the aggre-
16 gation of all species spectra is labelled the "community" spectrum. The "resource"
17 spectrum comprises both the phyto- and zooplankton communities. We summarise
18 the model used in the following subsections; for further information on the equations
19 and parameters used, see Table 1; default parameter values are used for all simula-
20 tions, as in the *mizer* vignette (Scott et al. 2015, page 67). Values for parameters related
21 to seasonal spawning and plankton blooms are given in Table 2. Supplementary ma-
22 terial S1 describes how to add seasonality to the *mizer* model, along with a link to the
23 R code used; what follows here is an in-depth look at the model structure.

1 Consumer spectrum

2 The model uses the McKendrick-von Foerster equation as its backbone for growth
3 through the spectrum (Andersen and Beyer 2006; Blanchard et al. 2009; Law et al. 2009;
4 Datta et al. 2010). Organisms feed upon smaller organisms in both the resource and
5 consumer spectra according to a Gaussian feeding preference function (4), commonly
6 used in modelling predation in marine systems (Benoît and Rochet 2004; Andersen
7 et al. 2009; Law et al. 2012). Organisms die either from being eaten by larger organ-
8 isms, or from natural causes (this encompasses starvation and natural mortality; see
9 Table 1). Starvation has often been cited as a source of high mortality for fish (although
10 results are not entirely conclusive; see Anderson (1988)), and is included as a function
11 which becomes more severe as the growth rate of consumers drops; it decreases expo-
12 nentially with body size, as larger body size makes starvation less likely (Duarte and
13 Alcaraz 1989; Leggett and Deblois 1994).

14 Since the weight of eggs spawned from marine teleost fish lie in a narrow range around
15 1mg (Ware 1975; Cury and Pauly 2000), recruitment to a single fixed egg weight (of
16 1mg) in the spectrum is used. Reproduction can either be constant or over a limited
17 period of time across the year. The reproductive rate is dependent on the predation
18 rate, in keeping with the dynamic energy budget methods (Kooijman 2009) commonly
19 used in size spectrum models to allocate incoming mass to somatic and reproductive
20 mass (e.g. Maury et al. 2007a; Blanchard et al. 2011). Thus we consider mature in-
21 dividuals to be income rather than capital breeders, the latter having spawning that
22 is relatively independent of current prey availability but strongly dependent upon
23 lagged average availability (Sainmont et al. 2014). The physiology of egg production
24 is not explicitly modelled, and simple size-based fecundity is assumed.

25 The incorporation of reproduction into the model follows the *mizer* model (Scott et al.
26 2015, equation (3.10)), although here we also include a time-dependent term to allow
27 for seasonal spawning. This time-dependent term generates temporal change for both

1 seasonal spawning and plankton pulses, and is adapted from previous work (Pope
2 et al. 1994; Datta 2011). It has the form

$$s(v_i, t_i, \psi_i, t) = \frac{e^{(1-\psi_i)v_i \cos(2\pi(t-t_i))}}{I_0((1-\psi_i)v_i)}, \quad (1)$$

3 and is a dimensionless term dependent on time t , where t_i is the time of the pulse
4 peak, and v_i describes the severity of the peak (i.e. how short and sharp spawning or
5 the bloom is over the year). For reproduction, the two parameters t_i and v_i are fitted
6 to empirical data on spawning patterns of the twelve fish species modelled (Figure
7 1), while for resource blooms the parameters are estimated from observed data (see
8 Table 2). I_0 is the modified Bessel function of order 0, and is a normalising factor
9 which fixes the total mass allocated by an individual to spawning over a year for all
10 v_i . In other words, as v_i increases, the duration of spawning becomes shorter, giving
11 a narrower Gaussian-shaped peak (the numerator of (1)). However, a higher v_i causes
12 the denominator I_0 to decrease, increasing the height of the peak to give the same area
13 under the curve (i.e. an equal number of eggs). Thus, altering spawning duration
14 for a species produces an equivalent amount of offspring (for equal feeding rates). In
15 Supplementary Material S2 $s(v_i, t_i, \psi_i, t)$ is plotted for a range of values of v_i to illustrate
16 this.

17 We introduce the following terminology to describe aspects of the reproductive pro-
18 cess in the model:

- 19 • Individual reproductive investment: the proportion of incoming mass that an in-
20 dividual uses for egg production (11). This depends upon an individual's species,
21 its weight w and time t , and is limited by incoming biomass at each time step.
- 22 • Population reproductive investment: the aggregated population production of
23 eggs for a species (13). This is summed over all mature individuals in the species
24 (the integral term of (13)), and takes into account production efficiency and the

1 sex of individuals (Scott et al. 2015, section 3.5).

- 2 • Recruitment: the total number of eggs feeding into the "newborn" weight bin of
3 a species at time t , which takes into account the maximum level of recruitment
4 for each species (14).

5 Given the parameterisation in Blanchard et al. (2014), we expected that recruitment
6 for all species would remain at or close to their maximum ($R_{max,i}$ in (14)) for all times
7 and scenarios, and this expectation was borne out in practise during initial runs of the
8 model. As a consequence, behavioural patterns observed in model output of this study
9 are interpreted in terms of the shifts in growth and mortality rates that are produced
10 by introducing seasonality.

11 The weight parameter ψ is the allocation of incoming mass to reproduction rather
12 than somatic growth, which incorporates both the maturation weight and asymptotic
13 weight of a species (Table 1, equation (9)). Without the $(1 - \psi)$ terms, equation (1)
14 is the von Mises function used by Pope et al. (1994), and for plankton blooms we set
15 $\psi_p = 0$ to give this form. The $(1 - \psi)$ terms are included for spawning to limit the
16 growth of organisms close to their asymptotic weight, by forcing them to spawn less
17 seasonally as their weight increases. This assumption is not realistic, but within this
18 modelling framework it is a necessary addition for growth behaviour of the species
19 to match the fitted growth curves of Blanchard et al. (2014). Those curves implicitly
20 assumed constant reproduction as a fraction of incoming mass, and to disentangle the
21 process would require an entirely novel parameterisation for the growth of all species,
22 and was beyond the scope of this work. The assumption is discussed more in the
23 Discussion.

24 Using empirical data on the monthly spawning rates for the twelve fish species sim-
25 ulated, von Mises distributions (Pope et al. 1994) are fitted to each species to give the
26 shape of the reproduction curves (Figure 1).

1 In Supplementary Material S3, a methodological derivation for the reproduction func-
2 tion used in *mizer* (Scott et al. 2015) is shown, using similar methods to the derivation
3 of the deterministic jump-growth model (Datta et al. 2010); in short, it is assumed that
4 the amount of weight lost in each spawning event is quite small, leading to a first-
5 order approximation which can potentially be incorporated into any McKendrick-von
6 Foerster equation to simulate both capital and income breeding processes.

7 Fishing is incorporated as in Blanchard et al. (2014); relevant equations are shown in
8 Table 1, equations (17) and (18). In summary, fishing mortality is size- and species-
9 specific, which can remain constant throughout the simulation or can vary at specified
10 times. A standard fixed logistic selectivity function is used to describe the ability of the
11 fishery to catch each species. Parameters are selected so that baseline fishing efforts
12 give realistic distributions for the species spectra, using stock assessment data from
13 the period 1985-1995 (www.ices.dk). Some parameters are non-species-specific and
14 are assumed to be the same for all species. For a full list of parameters and values see
15 Tables S2 and S3 of Blanchard et al. (2014).

16 **Resource spectrum**

17 For simplicity, the processes driving the dynamics of phytoplankton and the acquisi-
18 tion of energy from nutrients and light are not explicitly modelled here (see Moloney
19 and Field 1991; Fuchs and Franks 2010). Instead, it is assumed that organisms in the
20 resource spectrum can be preyed upon by organisms in the consumer spectrum, but
21 replace themselves using a semi-chemostat function for replenishment to a carrying
22 capacity. Models have tested the response of the size spectrum to bottom-up pertur-
23 bations before, by increasing the height of the phytoplankton size classes uniformly
24 for a short period of time (Maury et al. 2007b; Blanchard et al. 2011). Here a similar
25 approach is taken.

26 Following the approach of Pope et al. (1994) the seasonal size-time resource spectrum

1 is characterised by the von Mises time distribution (1), as well as predation by the
 2 consumer spectrum, and the equation for the dynamics of the resource spectrum is
 3 given by

$$\frac{\partial N_R(w, t)}{\partial t} = (\omega + (1 - \omega)s(v_p, t_p, \psi_p, t)) \cdot \mathbf{N}. \quad (2)$$

4 The left hand side denotes the change over time in the resource spectrum N_R , which
 5 is a function of weight w and time t . \mathbf{N} is the shorthand introduced for the dynamics
 6 of the resource spectrum of Scott et al. (2015, equation (3.15)). The preceding term
 7 differentiates our model from the original; ω sets what proportion of the spectrum is
 8 present independent of the bloom, and $s(v_p, t_p, \psi_p, t)$ sets the shape of the bloom (there
 9 is no weight dependent behaviour for the bloom, so $\psi_p = 0$ for all size classes in the
 10 resource spectrum). Resource abundance thus remains relatively constant for most
 11 of the year but then, dependent on the value of v_p , may "bloom" for a short period;
 12 that is, the entire resource spectrum increases in abundance, whilst keeping the same
 13 gradient, before relaxing back to its original level.

14 Parameters are chosen to reflect the likely timing and abundance of blooms in real
 15 systems (Gasol et al. 1992; Huete-Ortega et al. 2010). The increased abundance affects
 16 the feeding rate, and thus growth, of consumers able to take advantage of the increase
 17 in biomass in the system. The full form of (2) is given in Table 1, equation (19).

18 The combination of bottom-up and top-down feedbacks is what the simulations focus
 19 on, and is a natural progression from the previous work by Pope et al. (1994) where
 20 the entire size spectrum was specified by a von Mises distribution; here the interplay
 21 between differently timed seasonal processes is what drives the dynamics to produce
 22 qualitatively different community spectra. Differences in the two analyses are further
 23 reviewed in the Discussion.

1 Numerical simulations

2 To numerically solve the dynamics of the model, the system of resource and fish
3 species spectra is initialised using default values from Scott et al. (2015). This is taken
4 as the initial distribution for all of the simulations. The resource and fish species spec-
5 tra through time are then calculated, both with and without seasonal plankton blooms
6 and seasonal spawning (henceforth referred to as seasonal and non-seasonal systems
7 respectively).

8 The growth trajectory of an individual is tested using the method of characteristics (see
9 Kot 2001, p.393 onwards). The body mass of a newborn through time is calculated by
10 solving

$$\frac{dw}{dt} = g_i(w(t)) \quad (3)$$

11 where $g_i(w(t))$ is taken from (10). The method of characteristics is employed to cal-
12 culate the growth trajectories (3) of offspring of the different species over the course
13 of a lifetime, as in previous work (Law et al. 2009; Rochet and Benoît 2012). In short,
14 the starting mass of the individual is set at the mass of a newborn (1mg). Then, at
15 each timestep the growth rate of the individual is set at the weight bin that the indi-
16 vidual occupied, and the new mass calculated. By tracking this mass through time,
17 we calculate how quickly organisms grow in non-seasonal and seasonal spectra.

18 The effect of seasonal reproduction on the dynamics of the consumer spectrum is
19 investigated, to simulate the spawning behaviour of twelve North Sea pelagic fish
20 species (Figure 1). A fixed resource and community spectrum, which included waves
21 of abundance from phytoplankton blooms, was previously studied, without dynam-
22 ical links between growth, mortality and reproduction dynamics (Pope et al. 1994).
23 We extend this model framework by making both spectra fully dynamic, and includ-
24 ing species-specific seasonal spawning. The growth of individuals within both non-
25 seasonal and seasonal systems are compared; this consists of growth within the spec-

1 trum, behaviour of the community spectrum slope (calculated by fitting a linear model
2 to the plot of number density against weight, between 1g and 10kg, using a natural
3 logarithm), and fishing yields over time.

4 All simulations are carried out using R (2015). For each simulation the spectrum is
5 allowed to run for 500 years with weekly time steps, so that a steady state for the non-
6 seasonal system is reached, and the seasonal system has extremely similar patterns
7 between years, before analyses are carried out.

8 **Results**

9 **The size spectrum with non-seasonal and seasonal processes**

10 The size spectra at the end of 500 years under both non-seasonal and seasonal con-
11 ditions are shown in Figure 2. The community spectrum (averaged over a year) is
12 similar for the non-seasonal and seasonal systems (Figure 2a). In either case, the com-
13 munity spectrum rapidly settles down to a ridged uneven shape above 10g, caused
14 by the range of asymptotic sizes reached by the twelve species. This shape was also
15 noted previously by Blanchard et al. (2014) when running the system to steady state.
16 There exists a discontinuity in the gradient of the slope at the boundary between the
17 resource and consumer spectra where offspring are born, consistent with similar pre-
18 vious studies (Blanchard et al. 2011). The shapes of the community spectra can be
19 better understood by examining the individual species spectra (Figure 2b), where the
20 species grow to different asymptotic weights; this leads to the irregular shape of the
21 community spectrum. Both the non-seasonal and seasonal species spectra are visually
22 similar, and hence only the seasonal system is plotted here.

23 Although the annual average of both community spectra appear similar, within a year
24 there is much variation in the species spectra in the seasonal system. This is shown in

1 Figure 3, where densities at size are plotted relative to the non-seasonal steady state
2 spectrum. At the low end of the species spectra there are high variations in abundance,
3 due to the uneven spawning patterns of the species. A wave-like pattern propagates
4 through the species spectrum, as extra biomass from the resource bloom allows for
5 faster growth of the smallest individuals in the spectrum, leading to a drop in number
6 density in the smallest size classes and increased number density further along the size
7 spectrum. These waves are damped at larger weights, as organisms undergo growth,
8 death and reproduction through the spectrum, leading to a smoothing effect due to
9 lowered biomass. At the right end of the spectrum the ratio for some species becomes
10 very large or small due to tiny abundances at high weights in both spectra, so that
11 small variations have large impacts. However, there are two groups of species with
12 different patterns. One that has lower abundance density at largest sizes (sprat, dab,
13 herring, gurnard) and the other with higher densities at large sizes (e.g. sandeel, sole,
14 plaice, haddock, cod, saithe) relative to the non-seasonal model. This is an emergent
15 result of the model due to faster growth and larger size at age reached for the latter
16 group of species (Figure 4) whereas, for the former group, growth was either very
17 similar or lower.

18 **Growth within seasonal systems**

19 In the non-seasonal system, growth curves over lifetimes were previously cross-validated
20 with empirically fitted von Bertalanffy growth curves for each species (Blanchard et al.
21 2014). Comparison of growth in a seasonal versus non-seasonal system also reveals
22 increasing temporal variation as fish become mature (Figure 4). Somatic growth is
23 slower during the spawning period than the rest of the year. Species with narrower
24 periods of spawning (see Figure 1) have less smooth growth curves; for example, com-
25 pare the trajectories of cod and sole, which have highly seasonal spawning, with those
26 of grey gurnard and Norway pout which spawn more evenly throughout. The overall

1 growth over several years is similar to the non-seasonal system, with the same asymp-
2 totic masses being reached; this is due to the setup of the reproduction function (1),
3 which makes spawning less seasonal as the asymptotic weight is reached.

4 The growth rate of a mature individual is shown in Figure 5, compared with a non-
5 seasonal spawner. Initially reproductive effort is low, so the majority of biomass is
6 used for somatic growth. Hence the gradient of the growth is steeper than for a non-
7 seasonal spawner, whose growth rate is relatively constant over the whole year. As
8 the reproductive period begins, overall growth decreases, and drops to zero around
9 time $t = 0.35$. Negative growth is not permitted in the model (as we are using income
10 breeding), and while spawning is near its peak growth remains at zero, with all incom-
11 ing mass being allocated to egg production. Around $t = 0.75$ growth increases again,
12 mirroring the von Mises spawning distribution closely. If a capital breeding function
13 were used (i.e. making the somatic growth rate (10) and egg production (11) indepen-
14 dent of each other) there would be the possibility for individuals to lose weight over
15 the reproductive period, which would depend upon the rates of growth and spawn-
16 ing. In laboratory studies body mass has been observed to decrease if the mass lost in
17 spawning is not compensated by the biomass assimilated through predation (Wootton
18 1977; Lambert and Dutil 2000), and the condition factor (related to weight at length)
19 of fish has been shown to drop post-spawning (Le Cren 1951; Pedersen and Jobling
20 1989).

21 **Size spectrum properties in seasonal and fished systems**

22 The effect of seasonality on the community spectrum slope is shown in Figure 6a.
23 The initial slope is approximately -1.65 (this is maintained under baseline fishing ef-
24 fort). After fishing intensity is doubled the slope for both seasonal and non-seasonal
25 systems decreases to around -1.9. The non-seasonal system settles to a steady state
26 rapidly, while the seasonal system oscillates at an average slightly above this value,

1 with a cycle of one year, owing to seasonal processes occurring at the annual timescale.
2 Halving fishing effort leads to an increase in the slope of the spectrum due to a lack of
3 fishing mortality for fish on the right hand side of the spectrum, reaching around -1.5
4 for both systems; the slope in the seasonal system oscillates around an average value
5 slightly below that observed in the non-seasonal system. These qualitative responses
6 of the slope to changes in fishing effort should be regarded as particular to our North
7 Sea model, as they are conditioned by how our model represents the species-specific
8 interplay between fishing effects and the growth pulse driven by the resource bloom.

9 At baseline fishing effort, yields for seasonal systems tend to be lower than non-
10 seasonal systems, and peak around the non-seasonal yield (Figure 6b). This is due
11 to the long-term distribution of the seasonal community generally having a higher
12 abundance of smaller individuals but fewer large individuals, resulting from the ef-
13 fect of seasonal processes on growth of individuals to the larger size classes (Figure
14 7).

15 After fishing efforts are doubled yields initially spike to around twice the original
16 quantity; however, in the following years yields quickly drop, as the increased mortal-
17 ity from fishing forces the community spectrum into a steeper shape, with lower abun-
18 dance of larger fish (Figure 6b). Yields settle at about their original quantity within ten
19 years, despite fishing pressure remaining twice as high. The opposite behaviour is
20 seen when fishing intensity is halved; an initial drop in yields is followed by an grad-
21 ual increase (due to lower mortality rates for mature fish leading to a more abundant
22 community), settling at a value of 2.95Mt per year, slightly lower than at baseline effort
23 (3.58Mt per year). Seasonal yields follow the same pattern as non-seasonal systems,
24 although they oscillate around a lower mean value; only at oscillation peaks are yields
25 roughly equivalent to the non-seasonal spectrum, and at the troughs they are approx-
26 imately 80% of the non-seasonal yield. Increasing fishing effort in seasonal systems
27 makes the variability in yields greater (comparing the difference in peaks and troughs

1 between double and half fishing effort in Figure 6b).
2 However, integrated across the entire year, whichever fishing strategy is picked, the
3 ratio of the seasonal to non-seasonal yields is close (approximately 93%). This is ul-
4 timately due to the fact that in the region of parameter space the model occupies ju-
5 venile production is fairly constant between years (due to recruitment being more or
6 less equal for non-seasonal and seasonal systems), but adult population densities are
7 more often lower for seasonal than for non-seasonal spectra, resulting in lower fishing
8 yields.

9 Discussion

10 Seasonal processes induce time-varying behaviour in the size spectra for individual
11 species, with markedly different intra-annual growth for mature individuals. Al-
12 though the aggregated community spectra remain visually similar for both seasonal
13 and non-seasonal systems, analysis of these communities reveals key differences. The
14 spectrum slope is more varied in seasonal systems, while fishing yields are up to 20%
15 lower than under non-seasonal conditions during the year. In the seasonal system, in-
16 creased fishing effort amplifies the peaks and troughs in the species size spectra, and
17 hence the yields become much more variable through time (Figure 6b). In aquatic sys-
18 tems, increases in fishing effort have led to a shift towards faster growth rates which
19 have previously been shown to cause increased temporal variability in production and
20 fisheries yields (Blanchard et al. 2012). Arguments about balanced harvesting have
21 also been made on the basis that fisheries size-selectivity affects the stability of size
22 spectra and yields (Law et al. 2012) but this has yet to be investigated in the context
23 of seasonal dynamics. Seasonal processes may thus have important implications for
24 conclusions about fisheries management.

25 Compared to the value of -1 often seen in aquatic systems (e.g. Boudreau and Dickie

1 1992; San Martin et al. 2006; Zhou et al. 2009) the slopes observed here are steeper;
2 however, heavily exploited systems have been shown to produce steeper slopes in
3 the past, due to fishing-induced truncation of the spectrum (Rice and Gislason 1996;
4 Blanchard et al. 2005). Seasonality led to a higher variability in the community spec-
5 trum slope than in its absence. Spawning strategies ranged from matching closely
6 with the timing and duration of the plankton bloom (e.g. sole and Norway pout in
7 Figure 1) to those spawning far away from the bloom (e.g. sandeel). However, the
8 balance of matches/mismatches of spawning and blooms does not account for the
9 variation in slopes. Rather, within-year variation was due to the interplay between in-
10 creased growth of smaller individuals caused by the annual plankton bloom, leading
11 to a qualitatively different community spectrum with varied intra-annual dynamics
12 but fairly similar inter-annual behaviour. We were aware that the region of parameter
13 space that the model occupied had high levels of recruitment, with R_i close to $R_{max,i}$
14 for all species and times (14), and this limited the effect that seasonal reproduction
15 had on the species spectra (as changes in recruitment were extremely minor over the
16 course of a year). In reality density-dependent recruitment is an important factor in
17 constructing size spectrum models, and requires a more in-depth approach in future
18 (Andersen et al. 2016, pages 18-19).

19 The resource bloom provided the majority of the changes in slope in this study. It
20 was observed that, regardless of what upper and lower weight limits were chosen
21 when calculating the slope, the qualitative behaviour when altering fishing effort was
22 the same (as in Figure 6a). 1g and 10kg were chosen here as representative of the
23 community. Altering these could affect the magnitude of shift in average seasonal
24 slope compared to the non-seasonal slope, but the overall trend of steeper slopes with
25 increased fishing persisted regardless.

26 Adding seasonality induced waves of biomass which moved up the species spectrum,
27 as growth rates and offspring populations rose and fell throughout the year (Figure 3).

1 There was a peak in biomass in offspring at the appropriate spawning peak from Fig-
2 ure 1; however, this was not as pronounced as the dip in biomass during the plankton
3 bloom, due to the increased growth rate of newborns caused by a more abundant food
4 supply in this period. Waves of abundance have been observed in previous models
5 (Law et al. 2009; Datta et al. 2010) when parameter values were chosen which desta-
6 bilised the steady state distribution. Adding a plankton bloom has a similar effect,
7 as organisms whose feeding range lies within the plankton spectrum are subject to
8 higher growth rates (Benoît and Rochet 2004; Maury et al. 2007a). Here it was ob-
9 served that seasonal forcing via the reproductive process also pushes the system away
10 from the steady state. With both plankton blooms and time-dependent reproduction
11 occurring simultaneously, departures from the well-established power-law distribu-
12 tion (Sheldon et al. 1972) are expected.

13 As expected, the general trend was for slopes to have a steeper gradient as fishing
14 effort increased, due to higher mortality on organisms with higher mass (Figure 6a).
15 Interestingly, doubling fishing effort initially gave higher yields but these quickly de-
16 clined to a level similar to that provided by the baseline effort; halving effort had
17 the opposite effect, with an initial decline before increasing, with an asymptote of a
18 slightly reduced yield compared to baseline effort (Figure 6b). Upon investigation,
19 the cause was not shifts in the biomass of newborns, which was not significantly per-
20 turbed. Rather, the total biomass of the larger fish (the segment of the spectrum tar-
21 geted by fishing) moved to different levels under the new fishing regimes, with dou-
22 bled effort leading to a decrease in the biomass of larger fish and halved effort leading
23 to an increase. We stress that many of the model parameters of *mizer* are derived us-
24 ing the baseline fishing effort (time-averaged estimates over 1985-1995), so we do not
25 draw heavy conclusions from this result, except that doubling these values in the long-
26 term produced yields more consistent with the flat part of a yield curve (they did not
27 increase much) but carried a greater impact on the community size spectrum.

1 In all cases (baseline, doubled or halved effort) seasonality led to yields fluctuating
2 near the non-seasonal value at their peak, but with a lower average and troughs
3 around 80% of the non-seasonal yield. It is perhaps not surprising that the non-
4 seasonal system set the upper limit for yields, as for both systems recruitment was
5 fixed by (14); with species-specific fishing pressure and asymptotic sizes, these factors
6 resulted in a similar upper limit for the number of mature individuals in the commu-
7 nity for seasonal and non-seasonal systems. Future work on the parameter sensitivity
8 and uncertainty (Spence et al., this issue) is needed to elucidate the trade-offs in yield
9 that result under different environmental conditions and management strategies. It
10 is well known that fish stocks greatly fluctuate intra-annually (Horwood et al. 2000),
11 and the simulations here show that assuming a non-seasonal system means higher
12 yield predictions than when incorporating seasonality. For more accurate fisheries
13 management, including time-dependent processes such as spawning could help im-
14 prove the predictive powers of models to judge expected catch sizes throughout the
15 year. This model could be used in conjunction with other operating models as part of
16 management strategy evaluation to assess consequence of both short-term changes in
17 seasonality as well as longer term environmental variability and change.

18 Size-based models have commonly divided incoming biomass from the predation pro-
19 cess into somatic growth and egg production (De Roos and Persson 2002; Maury et al.
20 2007a; Kooijman 2009; Blanchard et al. 2011; Hartvig et al. 2011). We followed the
21 same method here (Scott et al. 2015, section 3.4), but importantly allowing spawning
22 to take place over limited times of each year. This is a natural step forwards from the
23 early theoretical analysis of Pope et al. (1994), who used a static spectrum to inves-
24 tigate the growth and survival trajectories of cohorts born at different times of year.
25 While we have not tested the success of different spawning techniques here, we have
26 laid important groundwork in providing a more realistic framework in which the im-
27 portance of spawning strategy in maximising cohort growth and maturation can be
28 investigated. Having said that, there are several important caveats which could guide

1 further developments.

2 An assumption introduced for convenience was to make spawning less seasonal as
3 the asymptotic weight of an individual was approached. This limited the growth
4 of species to empirically reasonable values; however, in reality seasonally spawning
5 species retain seasonality, even at the largest size classes. This explains the similar-
6 ity between the steady state species size spectra in the seasonal and the non-seasonal
7 scenarios (Figure 2). Older individuals make a relatively large contribution to overall
8 population fecundity, and hence the effect of spawning seasonality (as currently de-
9 picted in the paper) on species dynamics may well be under-emphasized. Hence, an
10 important part of the follow-up work would be to incorporate more realistic growth
11 functions for seasonally spawning species, such that both seasonal reproduction and
12 a realistic upper body size for each species can be achieved.

13 Quantification of the size and timing of the plankton bloom was illustrative for the
14 model. These factors are not constant globally, and are affected, for example, by lat-
15 itude, temperature and hydrodynamics, as well as natural inter-annual variability.
16 Here a single set of parameters was selected for the model; the peak of the bloom
17 was around 6 times the abundance of outside the bloom period, and the bloom lasted
18 around 7 weeks, following temporal data on plankton abundance (Gasol et al. 1992;
19 Huete-Ortega et al. 2010). Many empirical studies sample phyto- and zooplankton
20 spectra, but far less commonly within a single year to investigate intra-annual varia-
21 tion in abundance (see studies summarised in Sprules and Barth, this issue); a wide
22 literature search revealed a broad range of values for this ratio (see e.g. Menzel and
23 Ryther 1960; Navarro and Thompson 1995; Batten et al. 2003).

24 Real systems show greater variation in growth rates when observations are recorded
25 without averaging results temporally or spatially (e.g. Heath 1995; Barnes et al. 2011).
26 In reality reproduction is sometimes subject to spatial and temporal variation, and
27 is independent of instantaneous growth (capital as opposed to income breeding, see

1 Jönsson 1997; Jager et al. 2008; Sainmont et al. 2014). Such a reproduction function
2 should be both weight-dependent (Wootton 1977; Duarte and Alcaraz 1989; Blanchard
3 2000) and time-dependent, depending on the species (Figure 1). This is not to say that
4 food supply does not affect reproductive rate in the long term; studies have shown
5 correlations between rations received by fish and egg production (e.g. Wootton 1977).
6 An in-depth modelling of the physiology of marine organisms would be required for
7 the robust modelling of egg production from food intake, and although it is beyond
8 the scope of this work, would be of future interest.

9 Differences in model assumptions can affect the coexistence and stability of multi-
10 species spectra models. For example, recent models (incorporating non-seasonal re-
11 production) have used random coupling strengths between species until stability was
12 established (Hartvig et al. 2011), senescence mortality (Maury and Poggiale 2013), or
13 density dependent stock-recruitment as in this model (although here model behaviour
14 was explored in a part of parameter space where recruitment was largely independent
15 of density; see also Andersen and Pedersen 2010). The adoption of different spawning
16 strategies could be an important process for the division of energy amongst compet-
17 ing species and adaptation to different environmental conditions (Sherman et al. 1984;
18 Longhurst 1998).

19 We used an "income" breeding method here, where biomass from predation is im-
20 mediately converted to offspring, as in many size spectrum models (e.g. Blanchard
21 et al. 2011; Hartvig et al. 2011; Law et al. 2012), hence weight loss was not possible.
22 Some models have enabled organisms to lose weight during reproduction; for exam-
23 ple, Persson et al. (1998) constructed a PSPM that incorporated a discrete reproductive
24 period once a year, where all offspring for that year were produced at the same time
25 in the spring. For this, "reversible mass" was built up over the rest of the year which
26 could be lost in reproduction (or used for metabolism in the case of low food sup-
27 ply), as opposed to "irreversible mass" which made up the bones and organs and was

1 assumed not to decrease.

2 We assume income breeding is a reasonable simplification for what has been presented
3 here: an initial exploration of the consequences of seasonality in resource availability
4 for the consumer size spectrum. It is not an ideal representation for all fish species but
5 can be justified on the grounds of simplicity in integration into the *mizer* model. For
6 capital breeding species, it is very common for females to lose 5 to 20% of their body
7 weight during the reproductive season from egg loss alone (Wootton 1977; Lambert
8 and Dutil 2000). Weight losses due to behavioural changes during breeding can be
9 similar for males, particularly for species that engage in some form of parental care.
10 In freshwater species, spawning times of capital breeders are ordered around the an-
11 nual production in ways that suggest some trade-off between optimization of feeding
12 opportunities and minimization of competition (Shuter et al. 2012).

13 Thus, a useful extension of the analysis would be to include capital breeding in the
14 model and to evaluate how results are affected, in such a way that it maintains both
15 empirical accuracy and model simplicity; as such, the growth and reproduction func-
16 tions must be made independent. Organisms must be allowed to lose weight across
17 timesteps (which is not currently feasible), and as these losses may be significant, the
18 current method of including reproduction (which is derived systematically in Supple-
19 mentary Material S3) may no longer hold (as one assumption is that the amount of
20 weight lost in a single reproductive event is always small relative to the size of the
21 organism).

22 Including capital breeding in the model would likely have the effect of exaggerating
23 the effects of seasonality. At the moment peak spawning is limited by food intake
24 (as demonstrated by the growth curve in Figure 5); with capital breeding the growth
25 curve would be able to have a negative slope as outgoing mass as eggs is greater than
26 incoming mass from food. Hence the waves of biomass moving up the spectrum could
27 become larger. Simulations of the timing of the pulsing (both resource pulse and re-

1 productive pulse) could potentially inform empirical studies, such as when the most
2 meaningful snapshot(s) should be taken within a season to observe seasonal processes
3 (see Sprules and Barth, this issue; de Kerckhove et al., this issue) for a related discus-
4 sion).

5 There is still further theoretical and empirical work needed to understand the effects
6 of seasonality on size spectrum processes, and the consequences for fishing yields, as
7 well as the feedback between increased fishing mortality and recruitment in seasonally
8 spawning species. What has been presented here is a first step towards more general
9 approaches to simulating seasonal processes in a variable environment.

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1 Tables

Table 1: Size spectrum model equations. Species-specific parameters are defined as follows for species i , in order of appearance: preferred predator prey mass ratio = β_i , width of prey size preference = σ_i , volumetric search rate = γ_i ($\text{g}^{-q} \text{yr}^{-1}$), maximum food intake rate = h ($\text{g}^{1-n} \text{yr}^{-1}$), interaction with species j θ_{ij} , maturation weight = w_i^* (g), asymptotic weight = W_i (g), standard metabolism = k_i ($\text{g}^{1-n} \text{yr}^{-1}$), population density N_i , estimated maximum recruitment parameters = $R_{max.i}$ (density yr^{-1}), fishing selectivity parameter = F_i . Constants are defined as follows, in order of appearance: critical feeding level = f_0 , exponent of search volume = q , exponent of maximum consumption = n , carrying capacity of resource spectrum = κ_r , assimilation efficiency = α , exponent of standard metabolism = p , reproductive efficiency = ϵ , egg weight = w_0 , pre-factor for background mortality = Z_0 , exponent of background mortality = z , fishing selectivity weights = S_1 and S_2 , productivity of resource spectrum = R_0 , exponent of resource spectrum = λ . For values of all parameters (except those related to seasonality) see Blanchard et al. (2014, Tables S2 and S5). For detailed information about the life processes modelled, see Scott et al. (2015, Section 3).

2

3 Encounter and consumption

4 Prey size selection

$$\phi\left(\frac{w_{prey}}{w}\right) = \exp\left[-\left(\ln\left(\frac{\beta_i w_{prey}}{w}\right)\right)^2 / (2\sigma_i^2)\right] \quad (4)$$

5 Volumetric search rate

$$V_i(w) = \gamma_i w^q; \quad \gamma_i = \frac{f_0 h_i \beta_i^{n-q}}{(1-f_0)\sqrt{2\pi\kappa_r\sigma_i}} \quad (5)$$

6 Encountered food

$$E_i(w) = V_i(w) \sum_j \theta_{ij} \int_0^\infty \phi\left(\frac{w_{prey}}{w}\right) N_j(w_{prey}) w_{prey} dw_{prey} \quad (6)$$

7 Maximum consumption rate

$$I_{max.i} = h_i w^n \quad (7)$$

8 Feeding level

$$f_i(w) = \frac{E_i(w)}{E_i(w) + I_{max.i}} \quad (8)$$

9 Growth and reproduction

1 Maturation function

$$\psi_i(w) = \left[1 + \left(\frac{w}{w_i^*} \right)^{-10} \right]^{-1} \left(\frac{w}{W_i} \right)^{1-n} \quad (9)$$

2 Somatic growth

$$g_i(w) = (\alpha f_i(w) I_{max.i} - k_i w^p) (1 - \psi_i(w) \cdot s(v_i, t_i, \psi_i(w), t)), \quad (10)$$

$$\min_{g_i(w)} = 0$$

3 Egg production

$$g_r(w) = (\alpha f_i(w) I_{max.i} - k_i w^p) \psi_i(w) \cdot s(v_i, t_i, \psi_i(w), t), \quad (11)$$

$$\max_{g_r(w)} = \alpha f_i(w) I_{max.i} - k_i w^p$$

4 Time-dependent term

$$s(v_i, t_i, \psi_i, t) = \frac{e^{(1-\psi_i(w))v_i \cos(2\pi(t-t_i))}}{I_0(1-\psi_i(w)v_i)} \quad (12)$$

5 **Recruitment**

6 Population egg production

$$R_{p.i} = \epsilon / (2w_0 N_i(w_0) g(w_0)) \int_{w_i^*}^{w_i} N_i(w) g_r(w) dw \quad (13)$$

7 Recruitment

$$R_i = R_{max.i} \frac{R_{p.i}}{R_{max.i} + R_{p.i}} \quad (14)$$

8 **Mortality**

9 Background mortality

$$\mu_0 = Z_0 W_i^z \quad (15)$$

1 Predation mortality

$$\mu_{p.i}(w_{prey}) = \sum_j \int_{w_0}^{\infty} \phi\left(\frac{w_{prey}}{w}\right) (1 - f_j(w)) V_j(w) \theta_{ij} N_j(w) dw \quad (16)$$

2 Fishing selectivity

$$S_i(w) = 1 / \left(1 + e^{(S1 - S2w)}\right) \quad (17)$$

3 Fishing mortality

$$\bar{F}_i(w) = S_i(w) F_i \quad (18)$$

4 **Resource spectrum:**

5 Population dynamics

$$\frac{\partial N_r(w, t)}{\partial t} = (\omega + (1 - \omega)s(v_p, t_p, \psi_p, t)) \cdot \left(R_0 w^{n-1} [\kappa(w) - N_r(w, t)] - \mu_{p.r}(w) N_r(w, t)\right) \quad (19)$$

6 Carrying capacity

$$\kappa(w) = \kappa_r w^{-\lambda} \quad (20)$$

7

Table 2: Parameter values used for seasonal processes in the model. The shape of plankton blooms is taken from empirical data sources (Gasol et al. 1992; Huete-Ortega et al. 2010). Parameters related to spawning are fitted to species-specific spawning data (Figure 1); numbers refer to species in order, from left to right and then top to bottom.

Term	Definition	Value (with units)
	<i>Spawning behaviour</i>	
v_i	severity of spawning peaks	[3.6047, 2.9994, 1.944, 1.141, 0.40493, 0.795, 4.973, 1.4257, 4.0495, 3.6567, 5.4732, 1.951], or all 0 (non-seasonal)
t_i	time of spawning peaks	[0.5477, 0.0643, 0.3574, 0.3825, 0.8576, 0.5021, 0.4856, 0.3716, 0.2245, 0.4181, 0.3123, 0.3333]
	<i>Plankton blooms</i>	
v_p	severity of plankton bloom	30, or 0 (non seasonal)
t_p	time of bloom peak	0.4
ψ_p	weight dependence of bloom behaviour	0
ω	proportion of plankton away from bloom	0.7

1 **Figure captions**

2 Figure 1: Fitting von Mises distributions to the twelve fish species simulated in the
3 model. The fit consists of two parameters for each species i : the time of maximum
4 spawning t_i and the severity of the spawning peak v_i . The species are, in increasing
5 asymptotic size from left to right and top to bottom: sprat, sandeel, Norway pout,
6 dab, herring, grey gurnard, sole, whiting, plaice, haddock, cod and saithe. Data ag-
7 gregated from Bowers (1954); Quéro (1984); Alheit (1988); Knijn et al. (1993); Albert
8 (1994); Brander (1994); Hunter et al. (2003).

9 Figure 2: The size spectrum after 500 years, with non-seasonal and seasonal processes
10 in place. (a) Comparing the community spectra for non-seasonal (dashed black line)
11 and seasonal (solid grey line) systems. Also included for reference is the non-seasonal
12 resource spectrum (dotted black line at left side of the spectrum). (b) The seasonal size
13 spectrum after 500 years; community spectrum given by solid black line, individual
14 species spectra shown by dashed grey lines. Asymptotic size of each species marked
15 by black squares.

16 Figure 3: The bi-monthly size spectra over a year for the twelve fish species in a
17 seasonal system, relative to their (non-seasonal) steady state abundances (horizontal
18 black lines), to elucidate the differences between the systems. Months are plotted in
19 different colours to allow comparison between times and species.

20 Figure 4: The growth curves (calculated using the method of characteristics (3)) for
21 offspring of different species hatching into spectra in both non-seasonal (dashed grey
22 lines) and seasonal (solid black lines) systems, over 15 years.

23 Figure 5: The growth trajectory for a mature sole individual in a seasonal environment
24 (solid black line), compared with a similar sized individual in a non-seasonal environ-
25 ment (dashed grey line). Shown for reference is the spawning function for sole (dotted
26 black curve), taken from Figure 1.

1 Figure 6: The effect of altering fishing intensity after a year on (a) the community
2 spectrum slope, and (b) total fishing yields. Non-seasonal spectra shown in black,
3 seasonal in grey; solid lines indicate doubling fishing intensity, dashed lines indicate
4 halving, and dot-dash lines indicate maintaining baseline fishing effort. The baseline
5 fishing effort consists of estimates used to calibrate the model across the 1985-1995
6 time period and cross-validation was shown to give realistic time-averaged species
7 size distributions and growth in non-seasonal species spectra (Blanchard et al. 2014).

8 Figure 7: The abundance of individuals for non-seasonal and seasonal systems over
9 ten years with baseline fishing effort, once systems have reached regular inter-annual
10 patterns. Shown are total number densities for organisms (a) less than 40g, and (b)
11 greater than or equal to 40g. Black solid lines indicate non-seasonal systems, and grey
12 dashed lines indicate seasonal systems.

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1 Figures

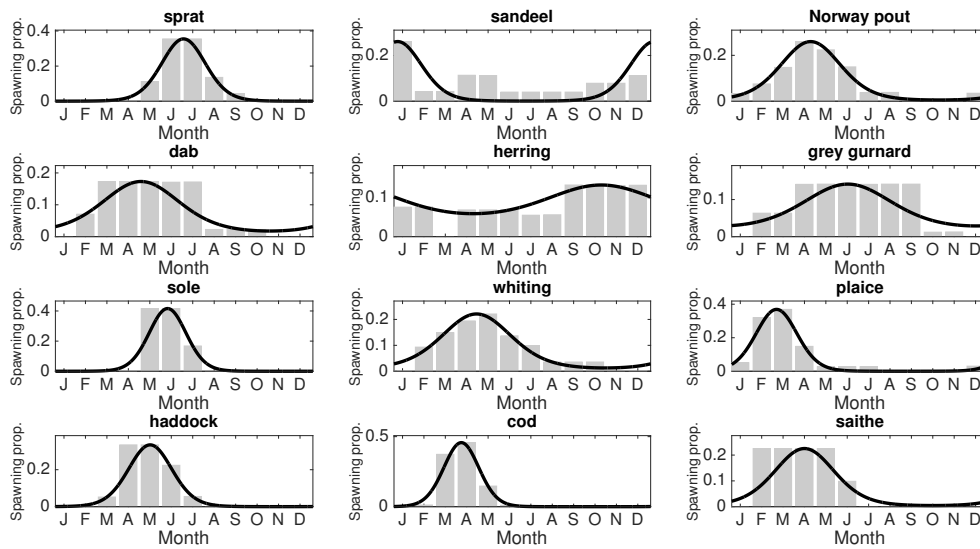


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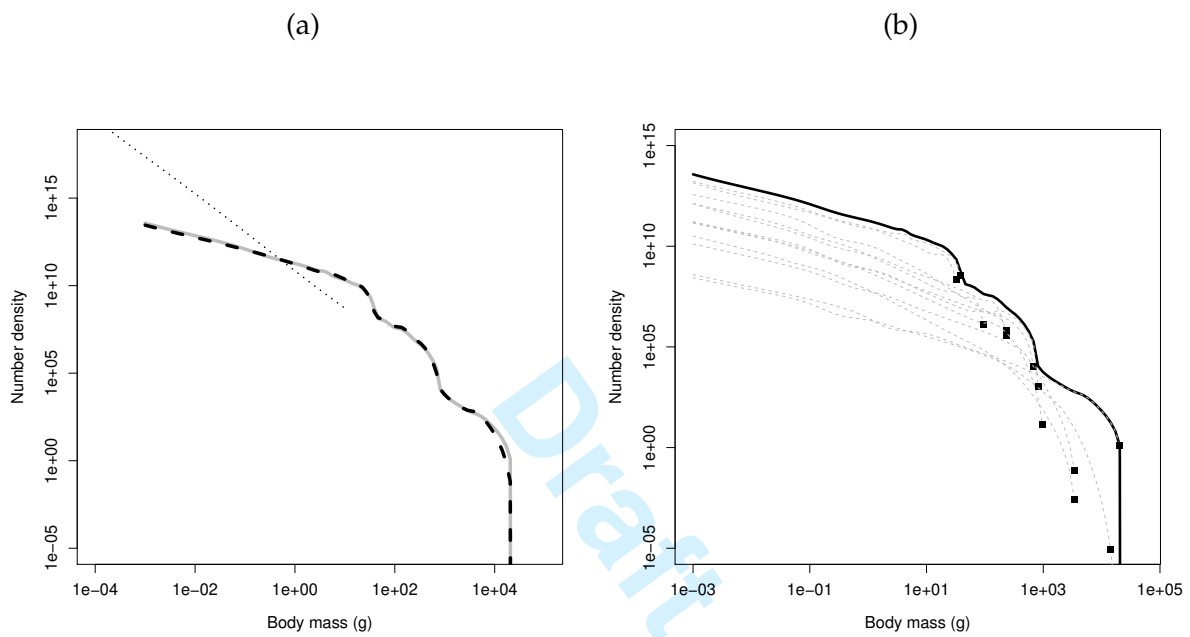


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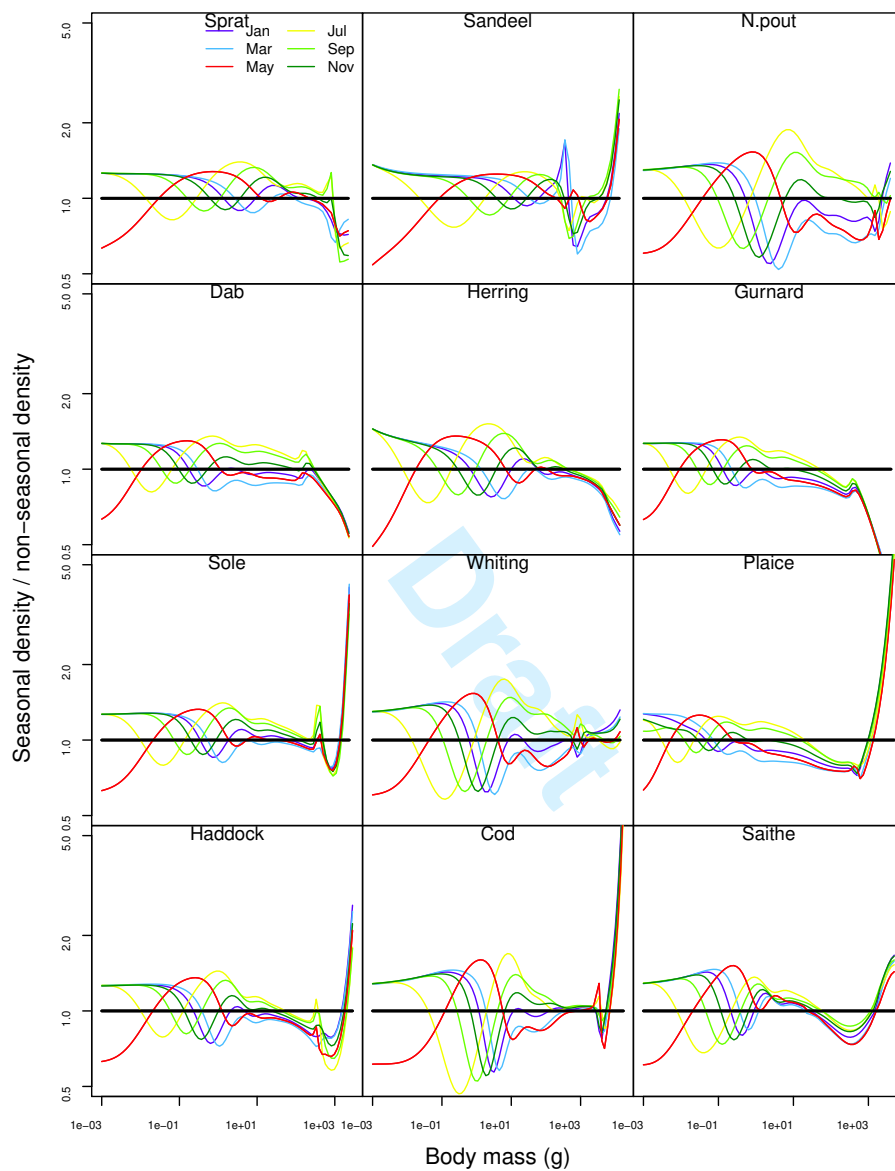


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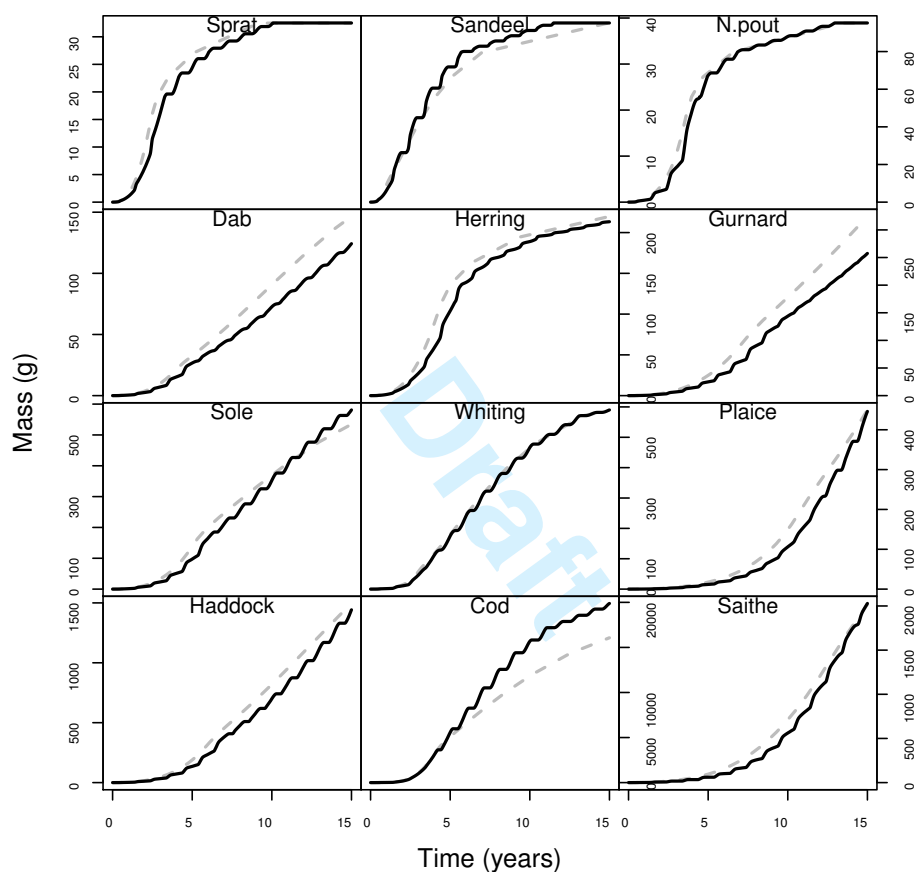


Figure 4: The growth curves (calculated using the method of characteristics (3)) for offspring of different species hatching into spectra in both non-seasonal (grey dashed lines) and seasonal (black solid lines) systems, over 15 years.

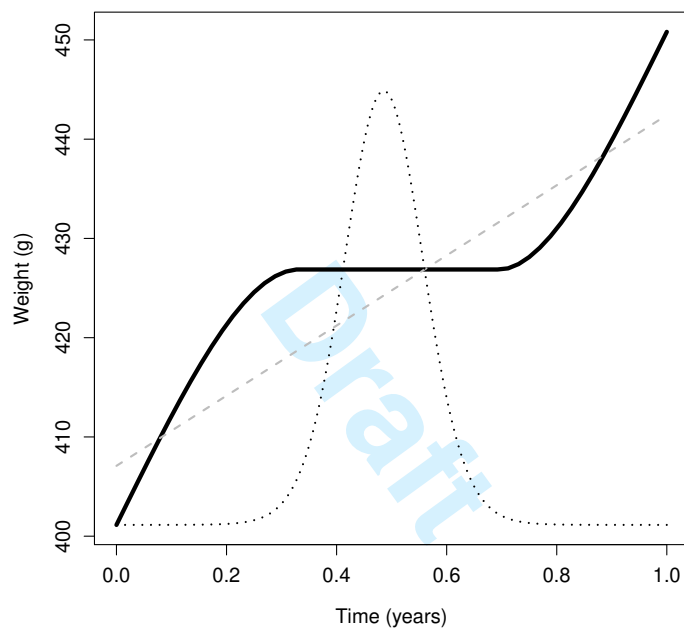


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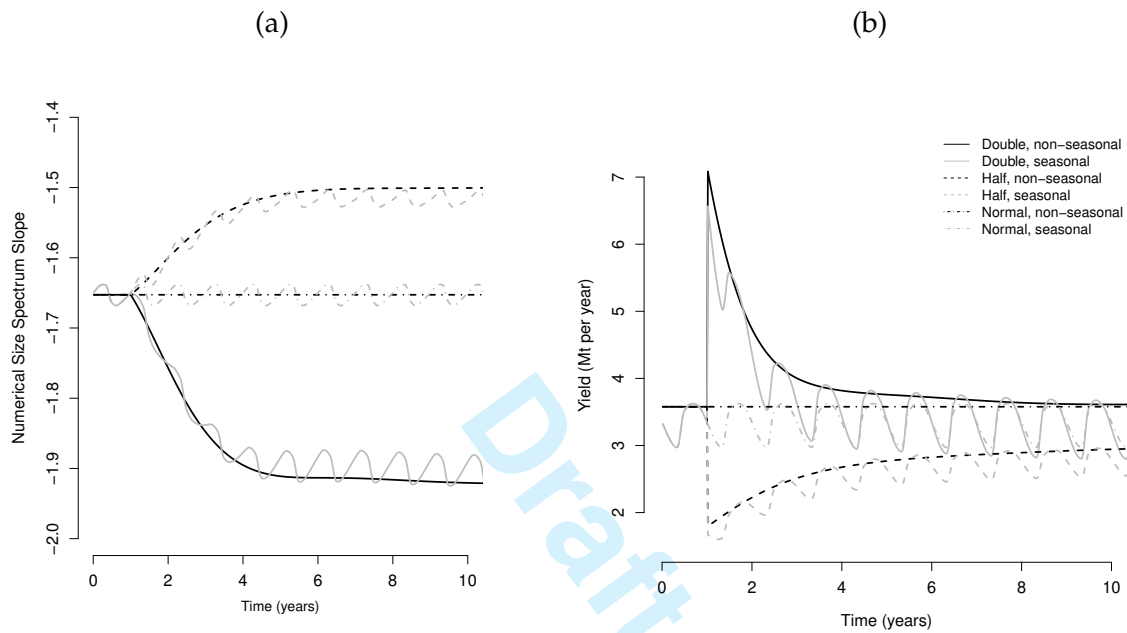


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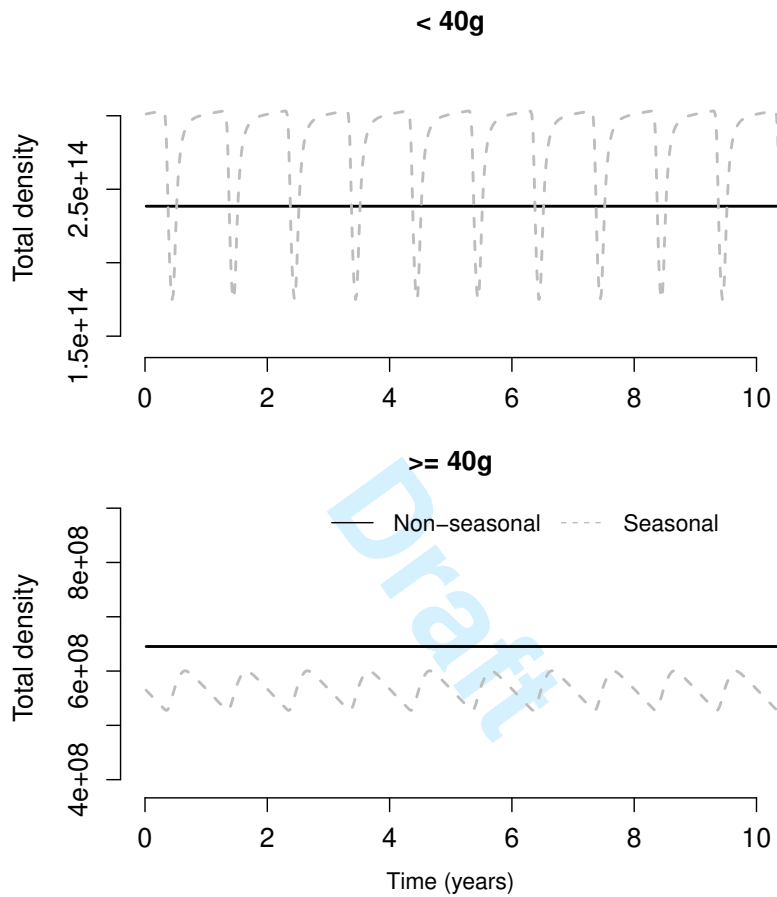


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