

THE ROLE OF RESOURCE SUPPLY IN SHAPING ONTOGENETIC
GROWTH AND ALLOCATION IN FISH

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Declaration

I declare this project as my own work. The model presented here was developed in conjunction with my supervisor, Dr. Samraat Pawar, and Ph.D. students Tom Clegg and Olivia Morris. I was responsible for any simulations and data presentation.

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Abstract

With recent results showing that reproduction in fish scales hyperallometrically there is a need to update growth OGMs to reflect this fact. Current OGMs assume optimal intake, an assumption which is not always reflected in the field. In this study I develop an energy intake focused approach to explaining growth, an area which has not been covered within current literature, and shows that hyperallometric scaling of reproductive output arises when allowing for variable reproductive scaling and maximising for fitness. The model is applicable to not only fish, but any animals taxon with some simple parameter adjustments. I offer direction for improvements and areas to be developed in order to allow the model to be applicable to any temperature range.

Keywords:

allometry; functional response; metabolic theory; growth; intake; life history; metabolism; reproduction; reproductive output; supply

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1 Introduction

Body mass plays a major role in determining many biological factors. For example, larger individuals are less vulnerable to predation, have lower mass specific metabolic rates and produce more offspring in their lifetime (Barneche et al., 2018; Craig et al., 2006; Hixon et al., 2014; Magnhagen and Heibo, 2001; Marshall et al., 2006; Peters, 1983). By extension, knowing the manner in which body mass change over an organism’s lifetime is the gateway to understanding how many biological rates change throughout ontogeny. This is because so many biological rates scale with mass (Kleiber, 1932). Despite its importance, relatively little is known about the factors which determine growth trajectories (Arendt, 2011; Marshall and White, 2019).

In the case of fish, understanding growth, and the factors that play a role in determining it, is not only insightful from the perspective of understanding the world around us, but can also be used to better manage the many fisheries and marine protected areas around the world (Heino et al., 2013; Lester et al., 2009). An objective which is becoming increasingly important as the oceans’ fish stocks continue to be depleted by overfishing. This is compounded by global warming which threatens to alter the structure of marine ecosystems even if they are not fished and left in their “natural” state (Bruno et al., 2018). It is already known that metabolic rate is dependant on temperature which in turn affects fish sizes (Brown et al., 2004; Gillooly et al., 2001). This combined with increasing global temperatures, understanding in greater detail how increased metabolic rates may affect growth is useful in population management.

To date many models have been developed to predict and describe the growth of an organism throughout its lifetime. The three main approaches used are the von Bertalanffy model, the dynamic energy budget (DEB) model and the ontogenetic growth model (OGM), which will be the focus of this study (Bertalanffy, 1938; Kooijman, 1986; Pütter, 1920; West et al., 2001). All of these are energetic based models with varying assumptions, key among which is the scaling of resource supply and metabolic rate with mass.

One of the best known examples of an OGM is the model developed by West et al., 2001. This model is parameterised around the average energy content of animal tissue and asymptotic mass. Asymptotic mass being the mass at which growth has essentially stopped due to metabolic cost and energy intake equalling each other (Fig. 1a). The model hinges on the scaling with mass between energy intake ($m^{0.75}$, allometric sub-linear scaling) and maintenance cost (m^1 , isometric linear scaling). In other words, as mass increases, maintenance costs will slowly overtake the intake rate and halt growth (Fig. 1 a). The framework used by West et al. (2001) was later developed by Charnov et al. (2001) to take the cost of reproduction into account and allow the estimation of lifetime production of offspring. Hou et al. (2008) developed West et al.’s model further by expanding maintenance cost to include the cost of feeding and digestion (specific dynamic action), synthesis of new tissue and activity. In

the above OGMs intake is assumed to scale sub-linearly to the power of 0.75. This is due to the assumption that the individuals are consuming at an optimal rate at all times and therefore the only limitation is their ability to make use of that energy. In this case, intake should theoretically scale to the power of 0.75 (see West et al., 1997). However, this is not always the case in the field. It has been shown that, for non-optimal consumption, steeper scaling can occur (Pawar et al., 2012; Peters, 1983). Additionally, OGMs, like many growth and metabolic models, typically use basal or resting metabolic rate to calculate metabolic cost. This is the minimal metabolic rate of an organism and is typically thought of as the rate of the organism when relaxed and at rest. However, it has been shown, once factors such as movement are taken into account, that the scaling becomes steeper (Weibel et al., 2004).

The issue of non-optimal feeding is addressed somewhat by Hou et al. (2011). However, this growth was only investigated as, essentially, a proportion of optimal consumption and does not address a potential change in scaling of intake rate. Another limitation of the models used in previous OGMs is the dependence on asymptotic mass. The models are entirely dependent on the value of optimal intake and asymptotic mass. All other values, such as metabolic cost, are then derived in relation to these. However, organisms are not born with an inherent restriction on the size they can attain, at least not energetically. If there is surplus energy for a given mass, the organism should be able to grow. Relying on asymptotic mass to define the upper bound of attainable mass does not allow for investigation of the mechanisms that underpin asymptotic size in reality. With two of the key assumptions of current OGMs, that reproduction and metabolism scale isometrically, not holding in the field (Barneche et al., 2018; Pawar et al., 2012; Peters, 1983) there is a need to take an unexplored approach to modelling fish growth. This study focuses on developing how intake is described so as to better reflect the real world. To achieve this, a natural starting point is to model intake as a functional response (Holling, 1959) so as to better reflect real world intake rates in terms of consumed biomass over time. Non-optimal supply is a currently unexplored area within growth modelling. This is likely due to the difficulty of directly measuring intake, especially in the field. Perhaps as a result, comparatively less is known about consumption. This necessitates the use of proxy values to estimate intake, for example nutrient flux (Schiettekatte et al., 2020), or drawing broad relationships to approximate consumption, as this study will do. Changing the manner in which intake is defined also requires changing metabolic cost, since the two are dependent upon each other in current OGMs. This can be done by defining metabolic rate as a value dependent on current mass rather than asymptotic mass, as has been done in OGMs up until this point. This thought process is more mechanistic since an organism has no concept of “How large should I grow?”, but rather will acquire as much resources as possible at its current life stage and size. Taking this more bottom-up mechanistic approach also allows exploration of the factors which control fish growth, since as previously mentioned, from an energetic standpoint, an organism can grow indefinitely provided there is surplus energy available after costs have been

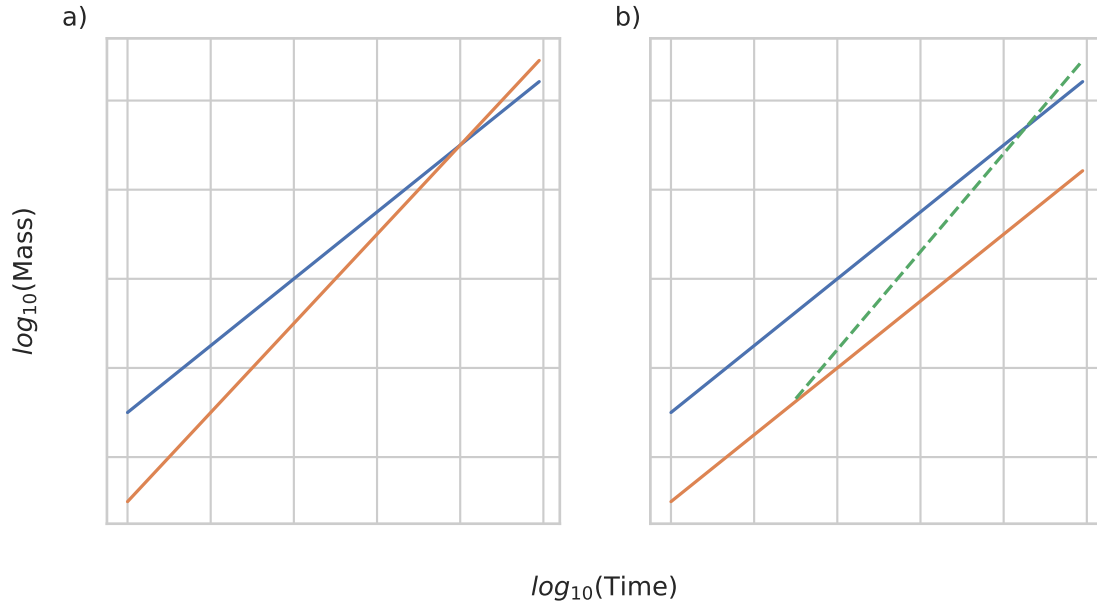


Figure 1: The effect of rate scaling exponents can be visualised within log space. The slope of the line is determined by the exponent. a) shows how maintenance cost out-scales supply in a traditional OGM. Growth only stops when maintenance (scaling exponent = 1) reaches the supply line (scaling exponent = 0.75) b) shows scaling for supply and maintenance as equal. Since scaling is equal growth will never stop until the new cost of reproduction is introduced some time during development. Blue line is supply, orange line is maintenance, green dotted line is reproduction

paid. Of course, there are also mechanical and genetic limitations upon organism size. However, once size is constrained to what is known to exist, this is not an issue.

This study takes the novel approach of using a mass-specific functional response and assimilation efficiency to describe how intake changes both throughout ontogeny and varying levels of resource availability. This study focuses on supply and growth within fish, however the same principles can be applied to other taxa.

Previous OGMs have assumed that reproduction scales isometrically with mass. This is indeed the case, within fish larger individuals produce more offspring than smaller ones. However, it has been shown that larger fish produce far more offspring than the equivalent mass composed of smaller fish. In other words, a 2kg fish will produce more offspring than two 1kg fish, i.e. reproduction scales hyperallometrically (Barneche et al., 2018). Furthermore, larger fish also use energy more efficiently than multiple smaller ones per unit mass. This is due to their lower mass specific metabolic rate (Brown et al., 2004; Kleiber, 1932; Peters, 1983). Additionally, larger mothers produce larger offspring, which are then more likely to survive to adulthood and reproduce (Hixon et al., 2014; Marshall et al., 2006). This has led to thinking rather than metabolism having steeper scaling than supply being the reason that growth stops (Fig. 1a). Instead, metabolism and supply at saturated resources should scale in

99 a similar manner, with the trigger for growth slowing and stopping being reproductive cost (Fig. 1b)
 100 (Marshall and White, 2019; Sibly and Brown, 2020).
 101 Assuming that fish have evolved to maximise reproductive output and can adapt to find an optimal
 102 strategy within the constraints of resource density, simulations can be carried out to demonstrate
 103 what conditions need to be met in order to achieve hyperallometric scaling of reproduction from an
 104 energetic perspective. This study will show that 1) hyperallometric reproduction arise is dependent
 105 upon metabolic scaling exponent, 2) possible scaling of metabolism and reproduction is dependent
 106 upon supply and by extension dimensionality.

2 Methods

2.1 Altering OGMs to Account for Resource Supply

108 In order to address the issue of supply in the context of an OGM, which can be generically described
 109 as $dm/dt = \text{gain} - \text{loss}$, some changes need to be made to the model's terms. The first is to remove
 110 the assumption of asymptotic mass and the reliance of cost upon it. Within a traditional OGM the gain
 111 term (a) and asymptotic mass are used to define the metabolic cost (b). However, since the assumption
 112 of perfect intake is going to be broken, because of the variable supply, this relationship no longer holds.
 113 As such, both intake and metabolic cost need to be redefined. Additionally, in light of recent work
 114 showing that reproduction scales allometrically and not isometrically, the reproductive cost must also
 115 be modified from the form used by Charnov et al. (2001) (Barneche et al., 2018; Marshall and White,
 116 2019)

2.1.1 Full Growth Equation

The general form of the model still follows that of an OGM, i.e. $dm/dt = \text{gain} - \text{loss}$. The gain term
 is represented by a functional response ($f(\cdot)$) modified by assimilation efficiency of biomass within
 poikilotherms (ϵ). Loss is dependent on whether the organism has reached maturity (α) or not. Prior
 to maturity, loss is resting metabolic rate (B_m) and results in growth as described by Eq. 1. Whereas
 after maturity, reproductive cost (cm^ρ) starts to be considered resulting in Eq. 2.

$$\frac{dm}{dt} = \epsilon f(\cdot) - B_m \quad t < \alpha \quad (1)$$

$$\frac{dm}{dt} = \epsilon f(\cdot) - B_m - cm_t^\rho \quad t \geq \alpha \quad (2)$$

2.1.2 Gain

119 To define supply a natural starting place is the functional response (Holling, 1959). Functional
 120 responses are used to define how much an organism consumes for a given resource density and is
 121 described by the following equation:

$$f(\cdot) = \frac{aX_r}{1 + ahX_r} \quad (3)$$

where, $f(\cdot)$ is the functional response, a is the search rate, h is handling time and X_r is resource density. For a fixed mass and increasing resource density Eq 3 produces a sigmoidal shape with intake eventually reaching an asymptote after some saturating amount of resources is reached. At lower resource densities, the intake rate is primarily defined by the search rate with higher search rates yielding higher intake rates. Conversely, at high resource densities, intake rate is approximately equal to the inverse of the handling time (h^{-1}), where lower handling times yield higher intake rates.

An organism's functional response will not remain constant throughout its life history. Search rate and handling time are affected by both the organism's mass and how it interacts with its environment (Pawar et al., 2012). Interactions can be broken into 3D and 2D, that is whether the organism consumes from a 2D "surface" e.g. a cow grazing or a 3D "volume" e.g. a pelagic consumer which consumes prey from within the water column. As such, both handling time and search rate can be defined as Eq. 5 and Eq. 4 respectively.

$$a(m) = a_0 m_t^\gamma \quad (4)$$

$$h(m) = t_{h,0} m_t^\beta \quad (5)$$

A functional response alone is not enough to fully define intake. This is because processing of consumed resources is not one hundred percent efficient which leads to inevitable loss consumed energy. As a result, to achieve the final gain term, a dimensionless efficiency term ϵ is applied. In poikilotherms assimilation efficiency is roughly 70% (Peters, 1983)

2.1.3 Loss

Metabolic cost has previously been dependant upon the gain term within traditional OGMs (see Hou et al., 2008; West et al., 2001). However, for non-maximal intake the relationship will no longer hold true. As a result, this model has taken previously measured values to be used as metabolic cost (see Eq. 6 taken from Peters, 1983 and Table 1 for further details).

$$B_m = 0.14 m_t^\mu \quad (6)$$

Next to take allometric scaling of reproduction into account, the reproductive cost term from Charnov et al. (2001) is changed from cm^1 which assumes isometric scaling to cm^ρ . c can be interpreted as the proportion of mass dedicated to reproduction, i.e. the gonadosomatic index (GSI) of the fish (Charnov et al., 2001). Just as in Charnov et al. (2001) reproductive cost is only taken into account once maturity is reached. Meaning that until a length of time (α) has passed, reproductive cost is considered to be zero.

2.2 Calculating Fitness

At any time (t) a reproducing organism devotes some amount of energy to reproduction. This is the product between the amount of mass dedicated to reproduction (cm^ρ) and a declining efficiency

term (h_t) which begins at maturity (α) and represents reproductive senescence (Benoît et al., 2018; Stearns, 2000; Vrtílek et al., 2018). In addition to amount of reproduction, the offspring are also subject to mortality (l_t). By combining the two, lifetime reproductive output can be estimated and is described by the “characteristic equation” (Eq. 7) which represents reproductive output in a non-growing population (Arendt, 2011; Roff, 1993, 2001; Roff, 1984, 1986; Stearns, 1992; Tsoukali et al., 2016)

$$R_0 = \int c m_t^\rho h_t l_t dt \quad (7)$$

Mortality is experienced differently by juvenile ($t < \alpha$) and reproducing individuals ($t \geq \alpha$). Mortality of offspring prior to maturity is described as a survival rate $l_t = e^{-Z(t)}$ which is an exponentially decreasing function bounded between zero and one. It controls how many offspring make it to maturity. After maturity, survival is again described as an exponential function which takes time to maturity into account, $l_t = e^{-Z(t-\alpha)}$. Reproductive senescence can also be estimated as an exponential function which begins after maturity and declines over time ($e^{-k(t-\alpha)}$), where k is the senescence term. When all values are inserted into the characteristic equation, it results in the equation used by Charnov et al. (2001) with the inclusion of reproductive senescence (Eq. 8).

$$R_0 = c \int_0^\alpha e^{-Z_t} dt \int_\alpha^\infty m_t^\rho e^{-(\kappa+Z)(t-\alpha)} dt \quad (8)$$

In Eq. 8, Z represents instantaneous mortality. This rate has been shown to be related to time of maturation in many taxon groups, and within it follows the relationship $\alpha \cdot Z \approx 2$. This can then be rearranged to estimate instantaneous mortality, $Z \approx 2/\alpha$

2.2.1 Maximising Reproduction

It is assumed that evolution will converge on metabolic values which maximise fitness. Fitness being defined as how much an individual is able to contribute to the gene pool (Speakman, 2008; Stearns, 2000). To this end, lifetime reproductive output is often used as a measure of fitness (Audzijonyte and Richards, 2018; Brown et al., 1993; Charnov and Berrigan, 1991; Charnov et al., 2001, 2007; Speakman, 2008; Stearns, 2000; Tsoukali et al., 2016). Therefore, by maximising for reproductive output, it should become clear what parameters will yield the highest fitness. These parameters will then show whether, within a theoretical framework, hyperallometric scaling arises.

To find all optimal values for reproduction would require Eq. 8 to be solved analytically. However, since no such solution is possible, I simulated the problem numerically to obtain a result. This was done by simulating across values of c and ρ , the parameters of interest between growth (Eq. 1 and 2) and reproductive output (Eq. 8). c was bound between 0 and 0.4, which encapsulates the values measured within fish (Benoît et al., 2018; Fontoura et al., 2009; Lambert and Dutil, 2000; Roff, 1983; Wootton, 1985). Though it has been shown to reach as much as 0.7 in some species (Parker et al., 2018). To search for any hyperallometry within reproduction, ρ was bound between 0 and 2. The

simulation was then run at 0.01 value intervals in both c and ρ over a lifespan of one million days. The results of each simulation were recorded and any non-viable results were discarded. A result was considered non-viable if fish had “shrunk” more than 5% in order to accommodate reproductive costs. Shrinking occurs in the model because the combined loss of energy to metabolism and reproduction is too much for the simulated values at the mass achieved by maturation so the individual experiences a deficit of energy, which is paid by loss in mass until equilibrium is achieved. Shrinking is not expected at maturity in reality. Typically, maturity will occur while the organism still has room for growth. It is the onset of reproduction which is considered to slow or stop growth (see Fig. S1). Shrinking can be thought of as starvation in a real organism. If energetic cost are not met then energy reserves in the body, such as fat and muscle, are broken down for energy. It has been shown that some fish are capable of losing up to 10% of their body mass (Berghe, 1992). However, this was during the breeding season and caused by behavioural changes due to parenting. Additionally, individuals were shown to rebound back to their “normal” body mass once the breeding season had ended.

2.3 Sensitivity Analysis

In order to determine the roles of metabolic exponent, maturation time and resource density within the model, sensitivity analyses were performed on each parameter with regard to c and ρ . This was done by simulating the parameters across multiple values and obtaining the optimal value for c and ρ as described above. The parameter values used in the analysis can be seen in table 1.

Table 1: Table describing parameters used in the model, along with values, units and sources where applicable. The units of resource density change depending on the dimension of intake. m^D represent either m^2 in 2D or m^3 in 3D

Parameter	Description	Value	Units	Range	Source
m	Mass	-	kg day^{-1}	-	
B_m	Metabolic Cost	$0.14m^\mu$	kg day^{-1}	-	Peters, 1983
μ	Metabolic Exponent	-	-	0.75 - 1.0	-
α	Age of maturity	-	day	-	-
c	Reproduction scaling constant	-	kg day^{-1}	0 - 0.5	-
ρ	Reproduction scaling exponent	-	-	0 - 1.5	-
Z	Rate of instantaneous mortality	$2/\alpha$	-	-	Charnov et al., 2001
k	Reproductive senescence	0.01	-	-	
ϵ	Assimilation Efficiency	0.70	-	-	Peters, 1983
X_r	Resource Density	-	kg/m^D	0.11 - 30	-
γ	Search rate scaling exponent	0.68 (2D) 1.05 (3D)	-	-	Pawar et al., 2012
a_0	Search rate scaling constant	$10^{-3.08}$ (2D) $10^{-1.77}$ (3D)	$\text{m}^2 \text{s}^{-1} \text{kg}^{-0.68}$ $\text{m}^3 \text{s}^{-1} \text{kg}^{-1.05}$	-	Pawar et al., 2012
β	Handling time scaling exponent	0.75	-	-	Pawar et al., 2012
$t_{h,0}$	Handling time scaling constant	$10^{3.95}$ (2D) $10^{3.04}$ (3D)	$\text{kg}^{1-\beta} \text{s}$ $\text{kg}^{1-\beta} \text{s}$	-	Pawar et al., 2012

3 Results

3.1 Growth and Maturation

In 3D, when the metabolic scaling exponent (μ) is 1, hyperallometry emerges in reproduction, i.e. $\rho > 1$, at low resources (Fig. 2 b). When resources are high, the value of ρ is lowered (Fig. 2 d). This emerges because consumption rate scaling is higher at lower resources in 3D (see Table 1) which

allows for steeper scaling within reproduction. This same pattern occurs within 3D for $\mu = 0.75$ (3). In 2D, the opposite pattern is seen for $\mu = 0.75$, with ρ at low resources lower than at saturated resources (3 a, c). This can again be explained by the difference within supply scaling at high vs. low resources in 2D. Since supply scaling is higher at high resources at low resources in 2D. However, when $\mu = 1$ in 2D, this pattern is reversed (Fig. 2 a, c). This may be caused by the very small amount of reproduction occurring at low resources, but if this were the case, the same pattern would be expected to be seen for $\mu = 0.75$, which is not the case. As resources increase, but still remain low, ρ does drop below that of the value at high resources, before climbing back up. However, the signal is not very clean (see Fig. S12 and S14)

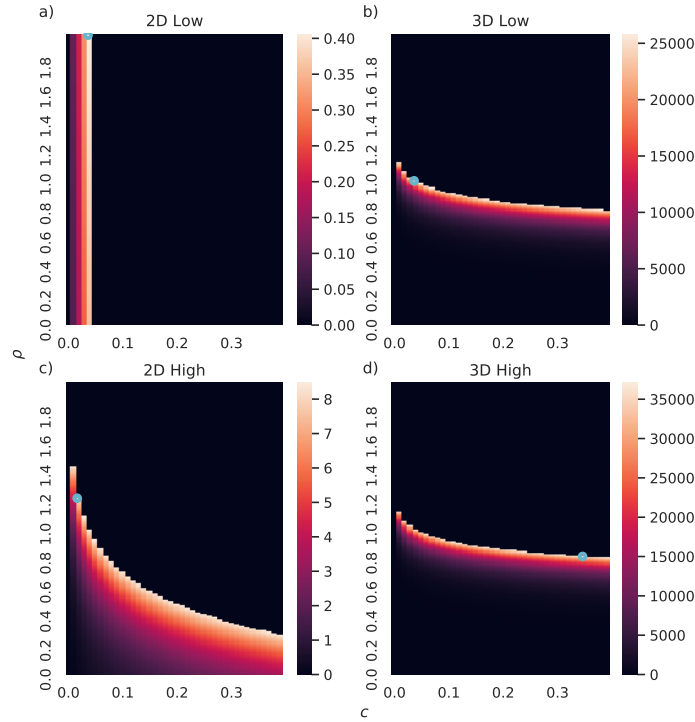


Figure 2: Metabolic exponent of 1 in 2D vs 3D at high and low resource densities. The optimal value is denoted by the blue circle. The resources used for the low resource scenario (top row) is the minimum amount of resources that allows growth with a c and ρ of 0.01. As would be expected, since 3D has steeper scaling, it allows for growth at smaller resource densities than 2D. Low resources in 2D were $\approx 0.11\text{kg/m}^2$ and 0.00035kg/m^3 in 3D. 100kg/m^D , where D is dependent on dimension, was used for high resources because, which although unrealistic, ensures that resources are not a limiting factor in the simulations. Hyperallometric scaling is observed in 2D at high (c) and low resources (a). Scaling in 3D is hyperallometric at low resources (b) and hypoallometric at high resources (d). Intensity of colour is determined by reproductive output in kg/s.

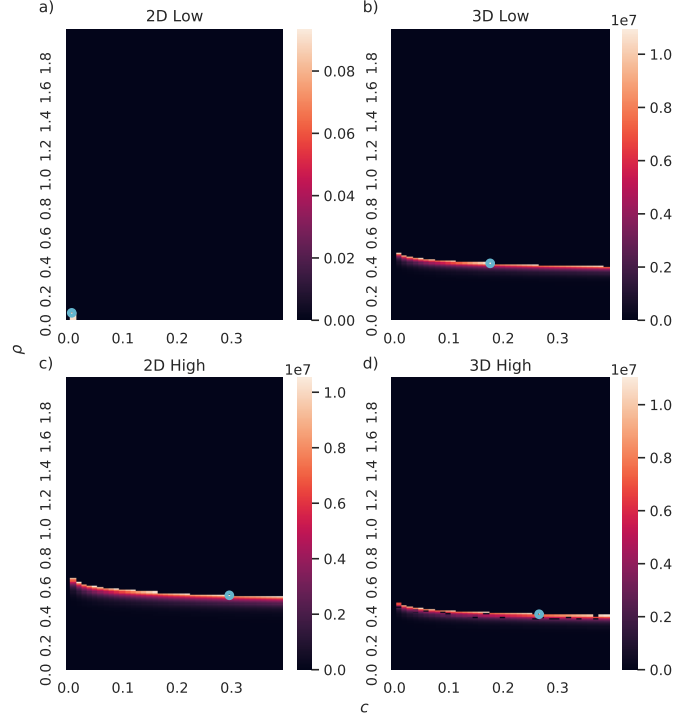


Figure 3: Metabolic exponent of 0.75 in 2D vs 3D at high and low resource densities. The optimal value is denoted by the blue circle. The resources used for the low resource scenario (top row) is the minimum amount of resources that allows growth with a c and ρ of 0.01. As would be expected, since 3D has steeper scaling, it allows for growth at smaller resource densities than 2D. Low resources in 2D were $\approx 0.11\text{kg/m}^2$ (a) and 0.00035kg/m^3 in 3D (b). 100kg/m^D , where D is dependent on dimension, was used for high resources because, which although unrealistic, ensures that resources are not a limiting factor in the simulations (c, d). Hypoallometric scaling is observed in all cases. Intensity of colour is determined by reproductive output in kg/s.

3.2 Sensitivity Analysis

3.2.1 Resource Density

The scaling relationship of ρ emerges as would be expected from the scaling of the functional response. At low resource densities the output of the functional response will scale similarly to search rate. The scaling of which is higher in 3D (see Table 1), as resources increase and the response shifts to scaling similarly to the inverse of handling time. At this point ρ starts to take values which are higher in 2D than 3D, because of the higher normalisation constant in 2D.

3.2.2 Metabolic Exponent

The expected result is for increasing values of μ for ρ to also increase. This is because the lower values of μ will result in there being a larger gap between the scaling of intake and maintenance, which allows

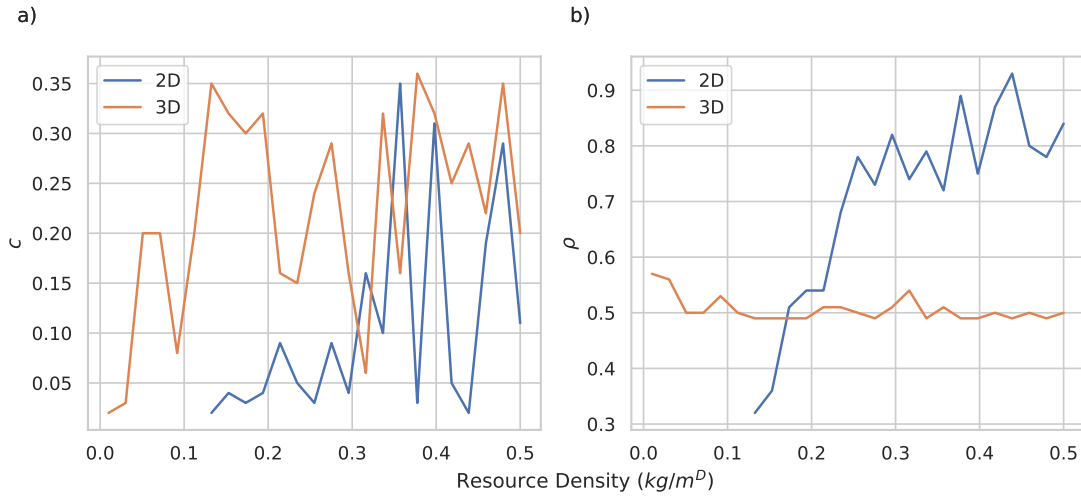


Figure 4: Effect of resource density on c and ρ where $\mu = 0.75$. Demonstrates the expected trend that under limiting resources the higher scaling of 3D search rate allows for steeper reproductive scaling (Table 1). As resources increase and supply shifts more towards being defined by the inverse of handling time, steeper scaling in 2D allows for higher ρ values. Units are kg/m^D , where D is the dimension.

for steeper scaling in reproduction (see Fig. 1 b) (Marshall and White, 2019). However, the trend suggests that increasing μ allows for higher values of ρ .

3.2.3 c Values

Estimations of c are low in some cases, especially in 2D (for example Fig. S3 and Fig. S16). While this may be low compared to the $\sim 10\%$ - 35% expected (Benoît et al., 2018; Fontoura et al., 2009; Roff, 1983; Wootton, 1985) it is not unprecedented for values of 2% to be observed in some fish (Gunderson, 1997). It may be necessary for the lower bounds of c to be adjusted based on what is expected or even viable in the organisms being simulated.

3.2.4 Shrinking

At saturated resource densities, allowing for greater proportions of shrinking enables ρ to be set to larger values in both dimensions. This is perhaps not surprising as increasing the shrinking proportion allows for a larger reproductive cost since all individuals will already be at asymptotic mass at time of maturation.

At low resources, the amount of shrinking allowed does not affect ρ or c values ??

4 Discussion

Some caveats with the results of the model. First, growth is simulated as being extremely fast. As previously stated, asymptotic size was reached by ~ 10 days. This is of course not representative of the real world, where individuals generally need several months to years to reach maturity. The rapid growth may be due to several factors. First is that metabolic cost may be underestimated. Similar to

West et al. (2001), this study used resting metabolic rate to define metabolic costs. However, this does not take other costs into account such as digestion and locomotion. This was addressed in traditional OGMs by Hou et al. (2008). However, due to the use of asymptotic mass in the parametrisation of this change, the same changes could not be used in this model. Resting metabolic rate and active metabolic rates do not scale in the same way with mass (Gillooly et al., 2001; Weibel et al., 2004). The additional cost of active metabolic rate would cause a steeper scaling within the metabolic cost term, leading to more gradual growth (Fig. 1). As such, inclusion of active metabolic rates, while challenging to measure directly and implement, is needed. Additionally, there may be behavioural or physiological factors that would lead to an altered metabolic rate. In this regard, temperature plays a critical role. It is well documented that a change in temperature will change many biological rates (see Peters (1983), Gillooly et al. (2001) etc.). It has been shown that growth is dependent on temperature within fish. For example, a 1.5°C increase in sea temperatures could result in a 15% decrease in fish lengths (Rijn et al., 2017). The functional response data used in this study is standardised around 15°C (Pawar et al., 2012). Meanwhile the metabolic cost is for an unspecified temperature (Peters, 1983). Using rates where the temperature effect is taken into account is crucial for model accuracy given the variation in rates that occurs over different temperatures. Work such as Barneche et al. (2014) has investigated this effect, however the estimate for metabolic rate is several orders of magnitude lower than what was reported by Peters (1983), which is the rate used in this study. Thus further investigation is required.

In fish metabolic rate has also been shown to drop under starvation (Cook et al., 2000). In homeotherms feeding restriction has also been shown to also lower body temperature, since metabolic rate and core temperature are closely related in homeotherms (Ballor, 1991; Blanc et al., 2003).

Another possible point of error is the estimates for supply. The parameters used from Pawar et al. (2012) are for a spectrum of animals from mammals to insects. It is possible by reanalysing the data for only marine species, or more specifically only within taxon or species, predictions of supply could be improved (Marshall and White, 2019).

A factor that is not taken into account in this model is that resources are not constant over time. This can be implemented within the model by varying resource density over time. The functional response will respond accordingly giving intake which varies through time. One concern with implementing such a response is fluctuations are likely not experienced by all organisms in the same way. For a fish with a small range a local fluctuation can be measured and described relatively simply. However, for a fish with a very large range there is the possibility leaving resource poor areas in search of richer waters.

Despite these caveats, the patterns which arise from the model are promising. The qualitative patterns seen in this study should not change even with slower growth. This is because while the exact results are not representative of reality, the relationship between the values is. For example,

285 the manner in which supply and metabolic rate interact for different values of the metabolic scaling
286 exponent. The fact that a higher metabolic scaling exponent will cause intake and metabolic cost to
287 intersect sooner does not change regardless of the absolute values.

288 In conclusion, the model presented in this study is a promising base which can be expanded upon
289 in a way that was not possible with previous OGMs allowing for much more controlled and detailed
290 explanations of the factors controlling growth. In contrast to all previous work, which assume optimal
291 supply, the concept of varying supply is addressed using functional responses. Additionally, qualitative
292 evidence is provided supporting hyperallometric scaling in fish using energy budget as the basis. The
293 model can easily be applied to any animal taxon, not just fish, with some simple changes. Additionally,
294 there are clear directions to be explored to improve the model's accuracy.

Code and Data Availability

Code is available at: https://github.com/Don-Burns/Masters_Project

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Supplementary Information

4.1 Growth Curves

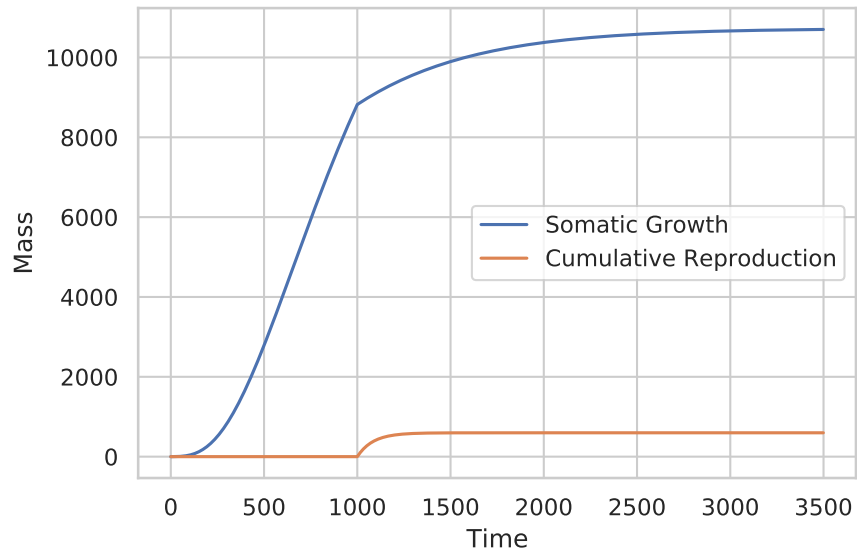


Figure S1: Example of the growth curve and cumulative reproduction expected from a traditional OGM model. Maturation occurs at 1000 days, after which growth is less steep until reaching asymptotic mass.

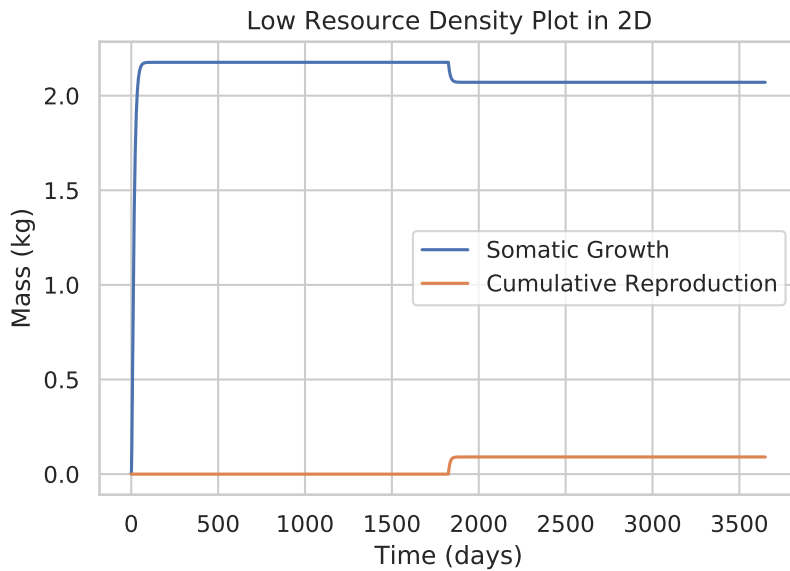


Figure S2: The growth over a fish which consumes in 2D. Maturation occurs at 5 years (1825 days). The fish was allowed to shrink by 5% at the onset of reproduction.

4.1.1 Maturation Time

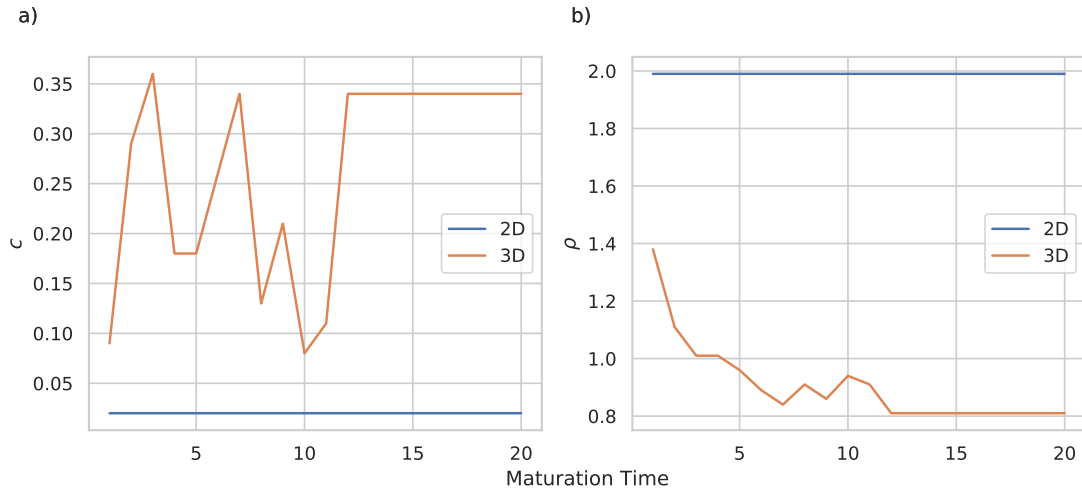


Figure S3: Effect of maturation time on c and ρ where $\mu = 1$ and resource density is low (0.11 kg/m^D, where D is the dimension).

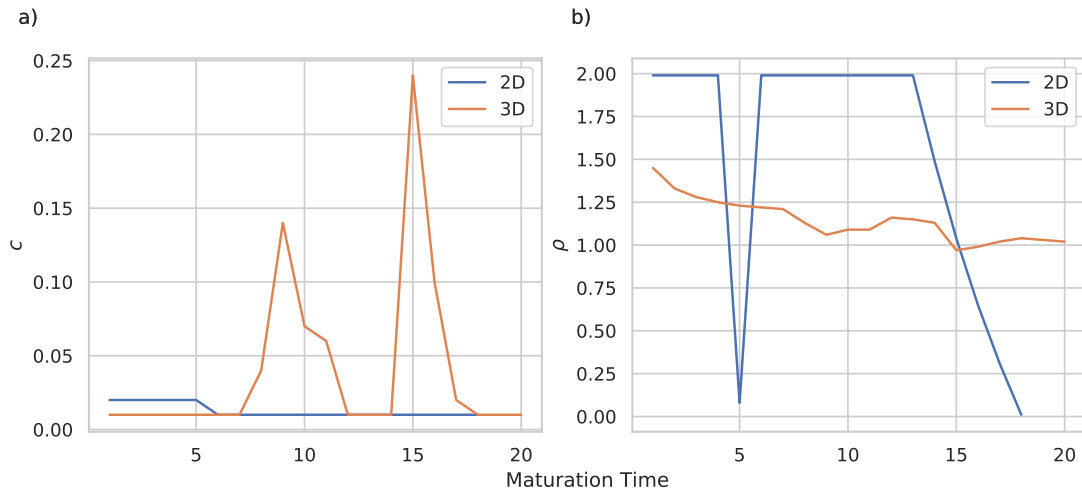


Figure S4: Effect of maturation time on c and ρ where $\mu = 0.75$ and resource density is low (0.11 kg/m^D, where D is the dimension).

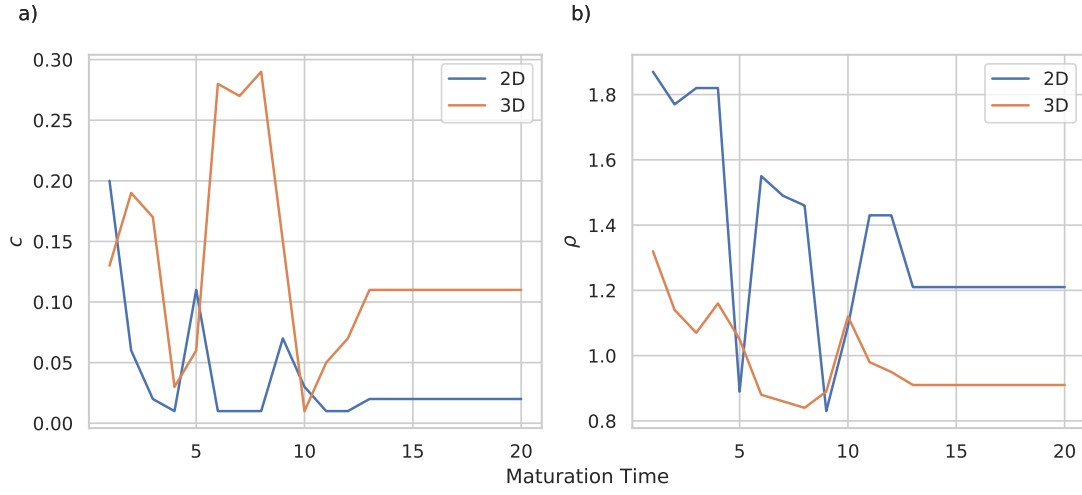


Figure S5: Effect of maturation time on c and ρ where $\mu = 1$ and resource density is high (100 kg/m^D, where D is the dimension).

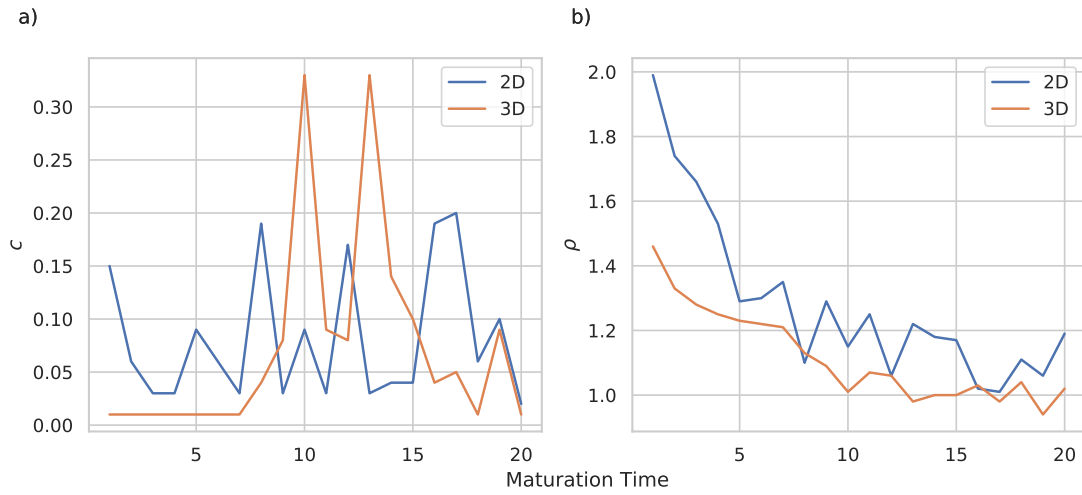


Figure S6: Effect of maturation time on c and ρ where $\mu = 0.75$ and resource density is high (100 kg/m^D, where D is the dimension).

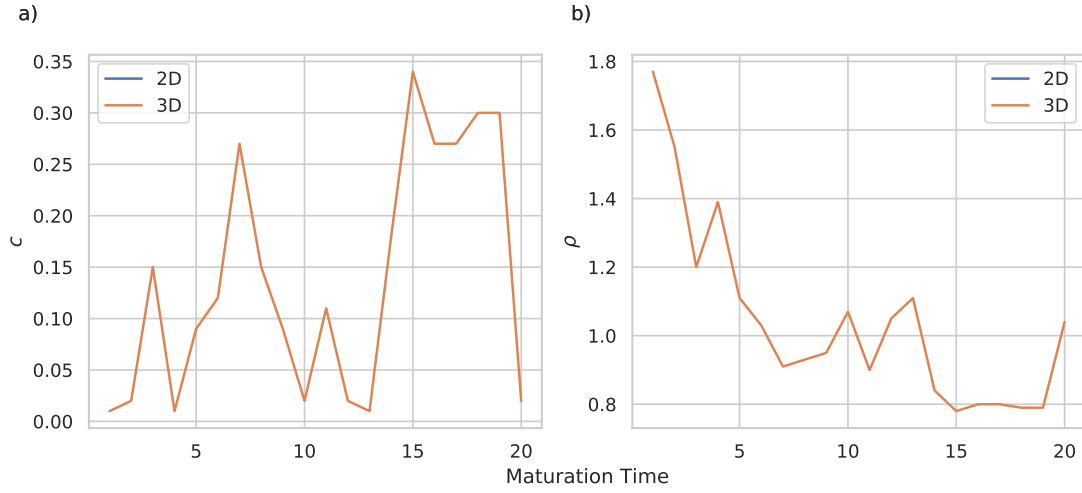


Figure S7: Effect of maturation time on c and ρ where $\mu = 1$ and resource density is very low (0.01 kg/m^D, where D is the dimension). At this resource density reproduction can only occur in 3D.

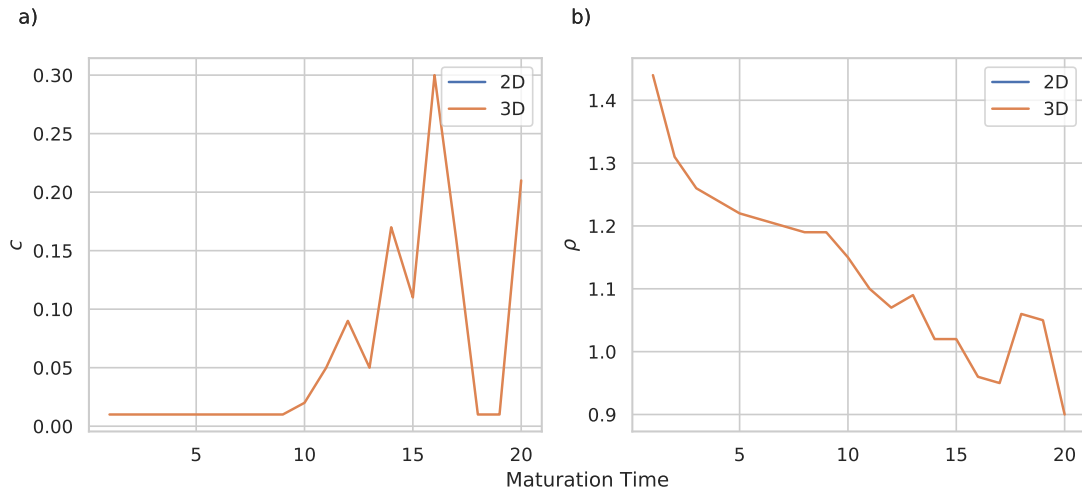


Figure S8: Effect of maturation time on c and ρ where $\mu = 0.75$ and resource density is very low (0.01 kg/m^D, where D is the dimension). At this resource density reproduction can only occur in 3D.

4.1.2 Metabolic Exponent (μ)

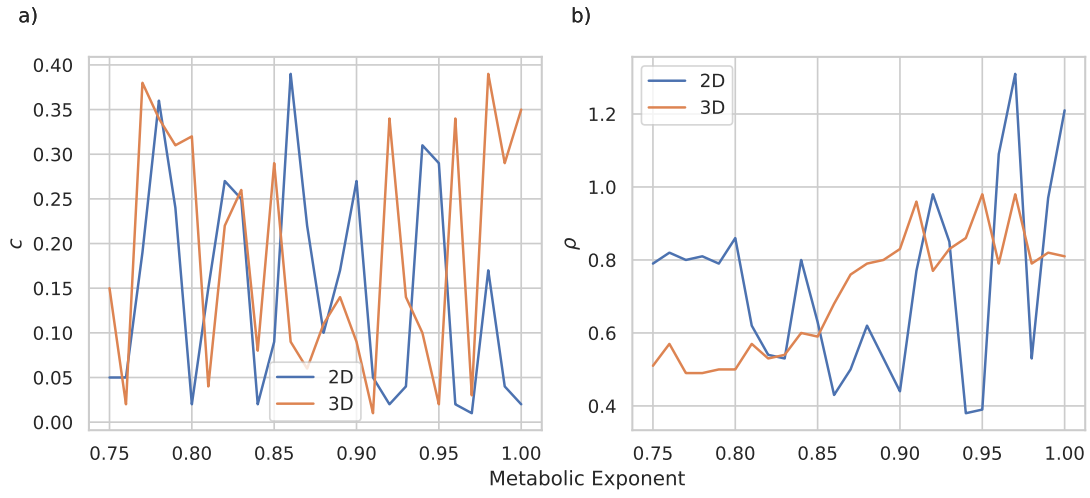


Figure S9: Effect of metabolic on c and ρ where resource density is high (100 kg/m^D, where D is the dimension)

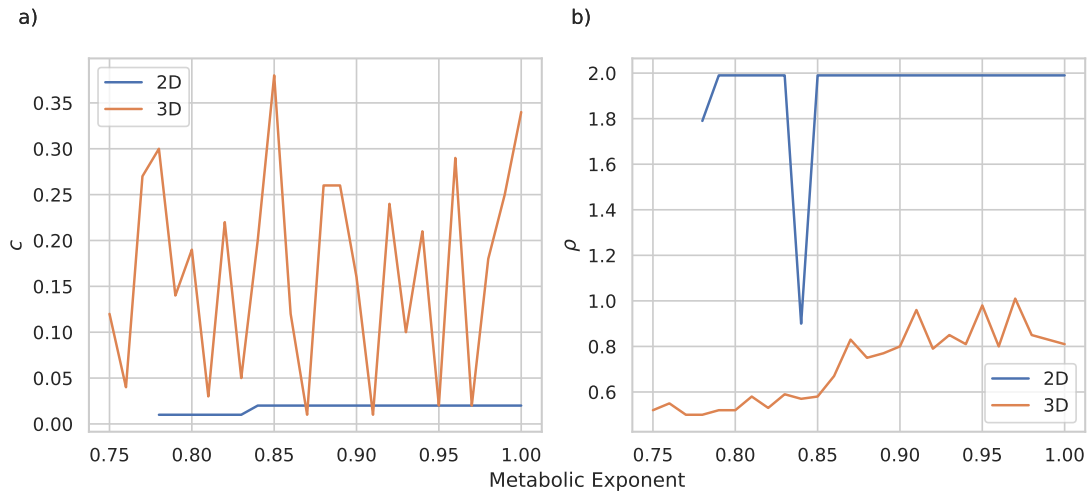


Figure S10: Effect of metabolic on c and ρ where resource density is low (0.11 kg/m^D, where D is the dimension)

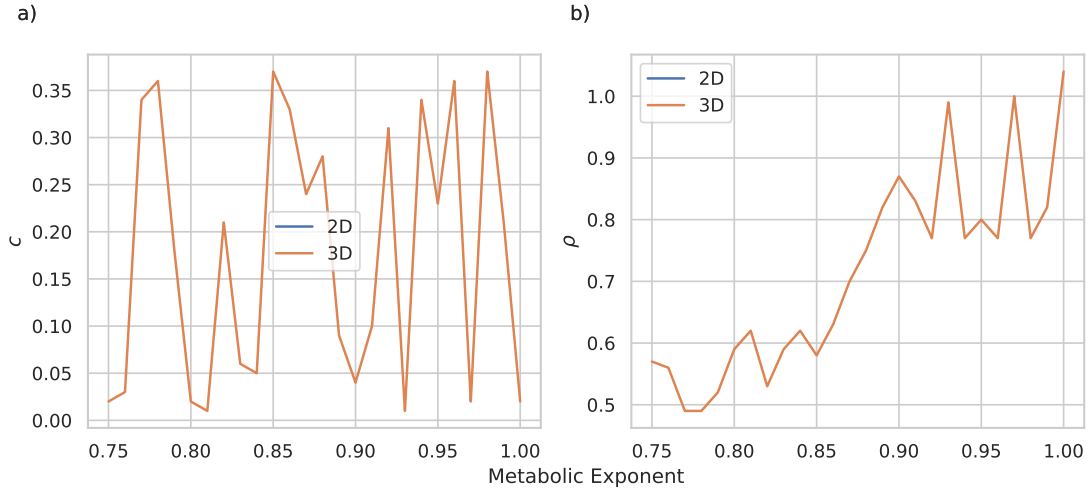


Figure S11: Effect of metabolic on c and ρ where resource density is very low (0.01 kg/m^D , where D is the dimension). At this resource density reproduction can only occur in 3D.

4.1.3 Resource Density

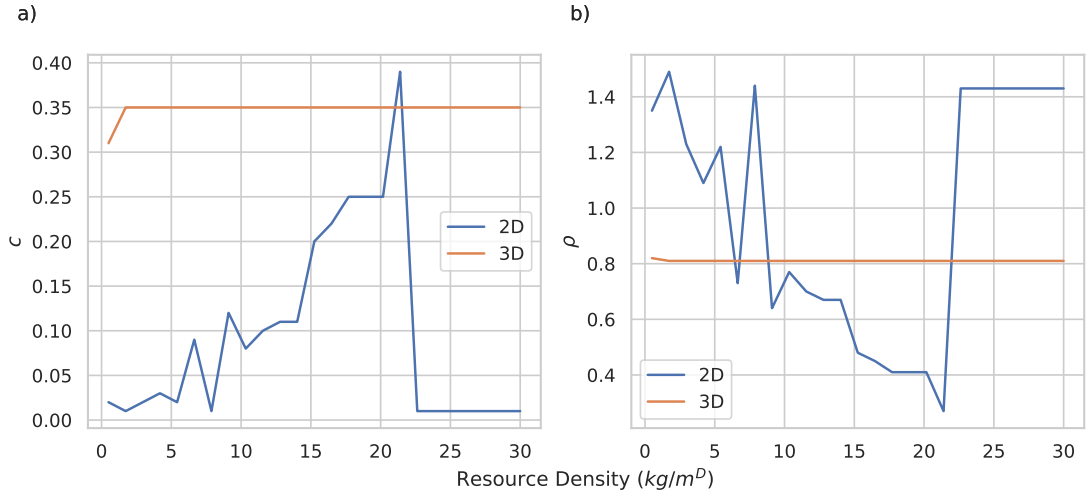


Figure S12: Effect of resource density on c and ρ where $\mu = 1$. Over larger values for resource density. 3D quickly saturates at this density, thus is a nearly straight horizontal line. See Fig. 4 for detail at lower resource density. Units are kg/m^D , where D is the dimension.

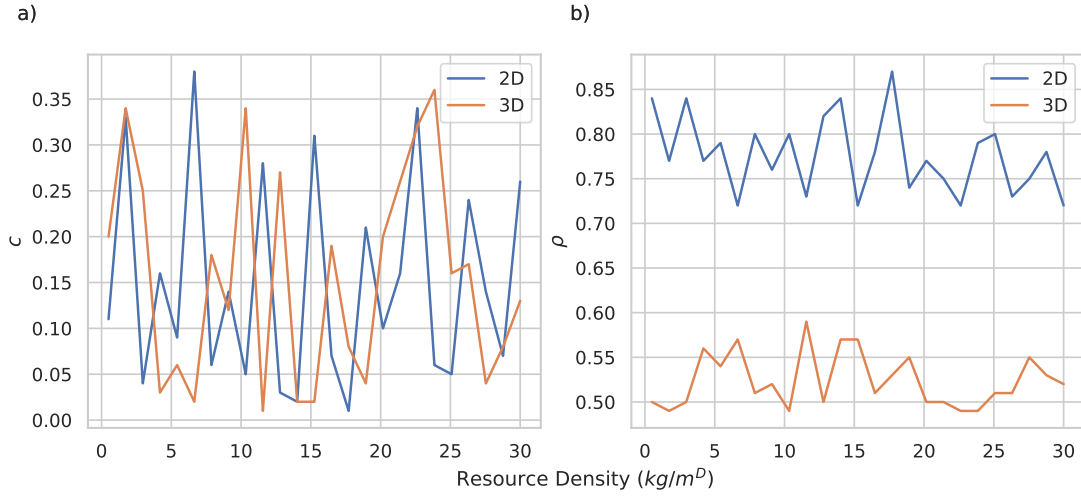


Figure S13: Effect of resource density on c and ρ where $\mu = 0.75$. Over larger values for resource density. There is a lot of numeric instability across resource densities, but the trend appears to be somewhat stable around ~ 0.8 in 2D and ~ 0.53 in 3D See Fig. 4 for detail at lower resource density. Units are kg/m^D , where D is the dimension.

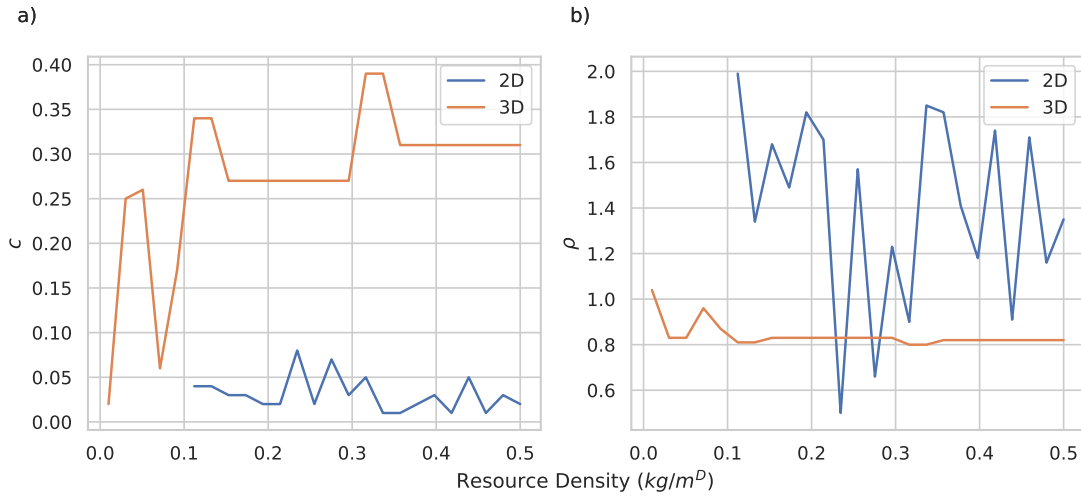


Figure S14: Effect of resource density on c and ρ where $\mu = 1$. Demonstrates the expected trend that under limiting resources the higher scaling of 3D search rate allows for steeper reproductive scaling (Table 1). As resources increase and supply shifts more towards being defined by the inverse of handling time, steeper scaling in 2D allows for higher ρ values. Units are kg/m^D , where D is the dimension.

4.1.4 Proportion of Shrinking Allowed

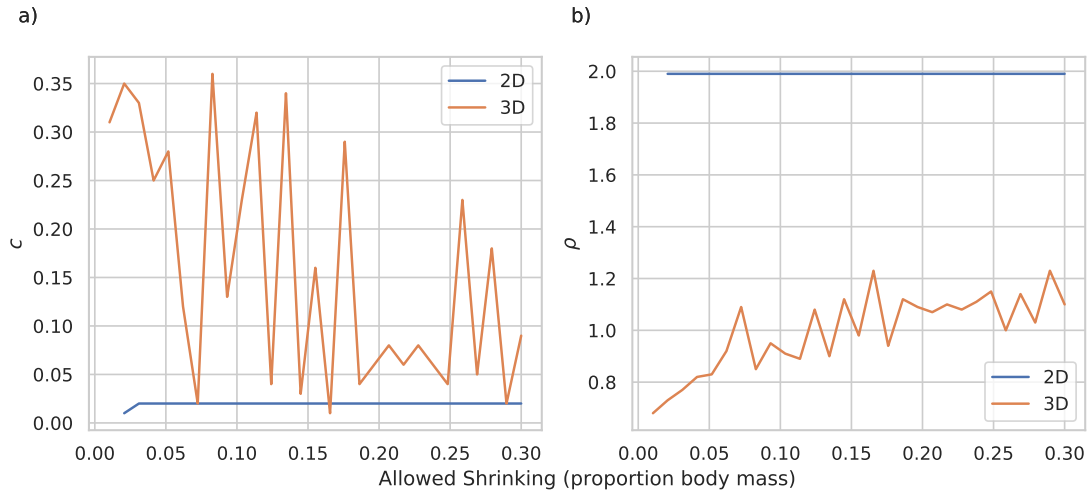


Figure S15: Effect of proportion of shrinking allowed on c and ρ where $\mu = 1$ and resource density is low (0.11 kg/m^D, where D is the dimension)

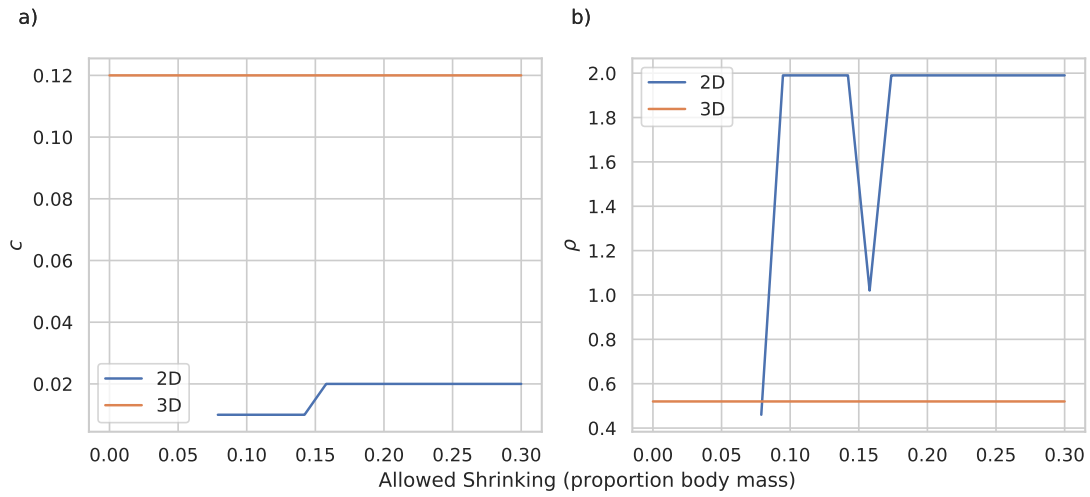


Figure S16: Effect of proportion of shrinking allowed on c and ρ where $\mu = 0.75$ and resource density is low (0.11 kg/m^D, where D is the dimension)

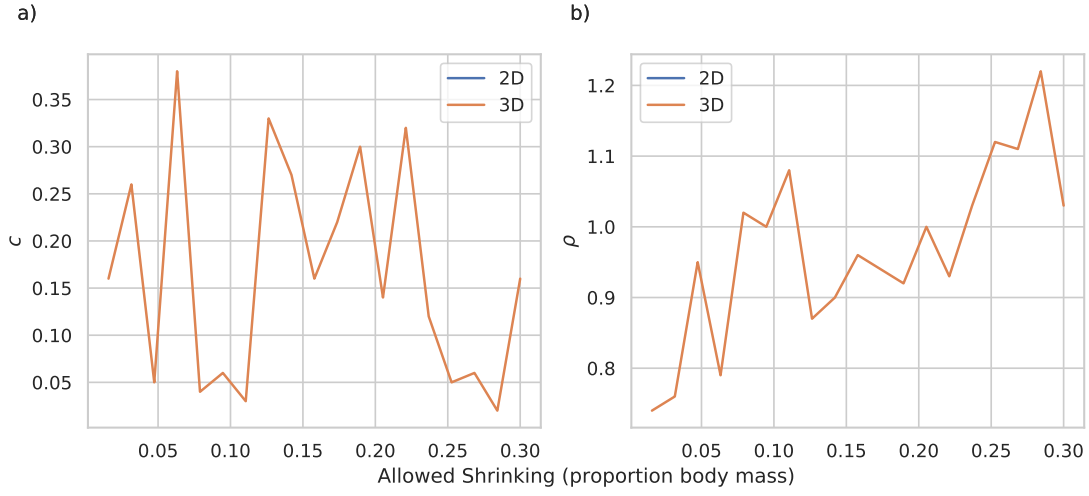


Figure S17: Effect of proportion of shrinking allowed on c and ρ where $\mu = 1$ and resource density is very low (0.01 kg/m^D , where D is the dimension). The resource density only allows for reproduction to occur on 3D.

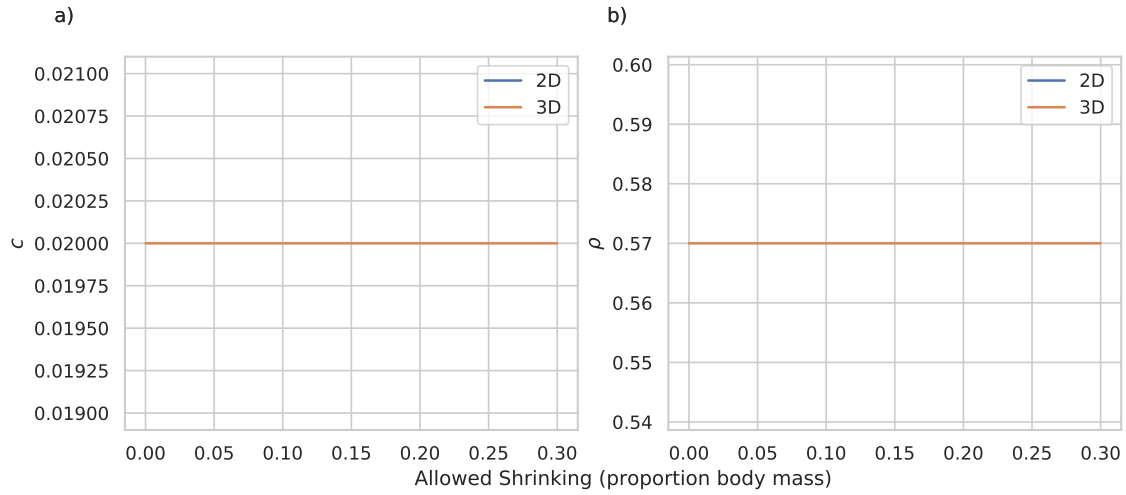


Figure S18: Effect of proportion of shrinking allowed on c and ρ where $\mu = 0.75$ and resource density is very low (0.01 kg/m^D , where D is the dimension). The resource density only allows for reproduction to occur on 3D.

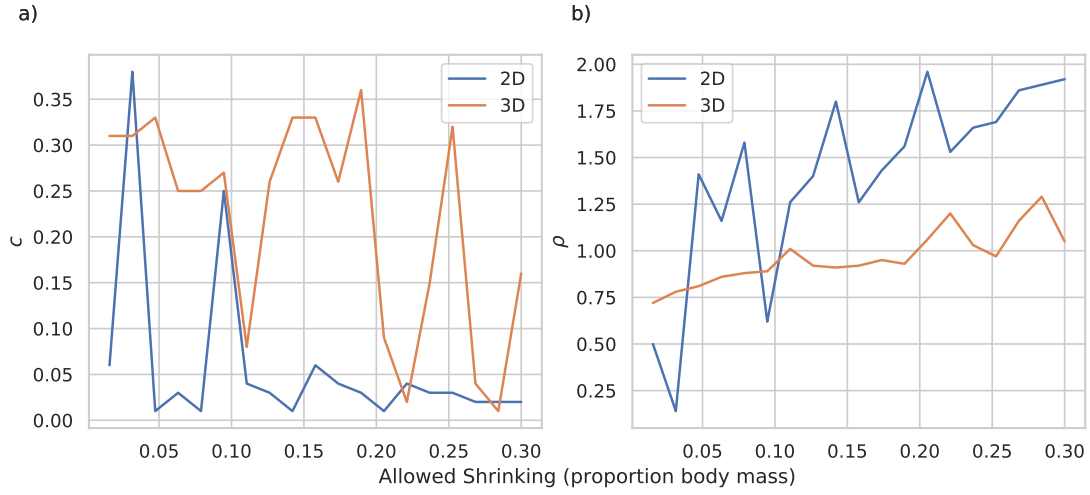


Figure S19: Effect of proportion of shrinking allowed on c and ρ where $\mu = 1$ and resource density is high (100 kg/m^D , where D is the dimension).

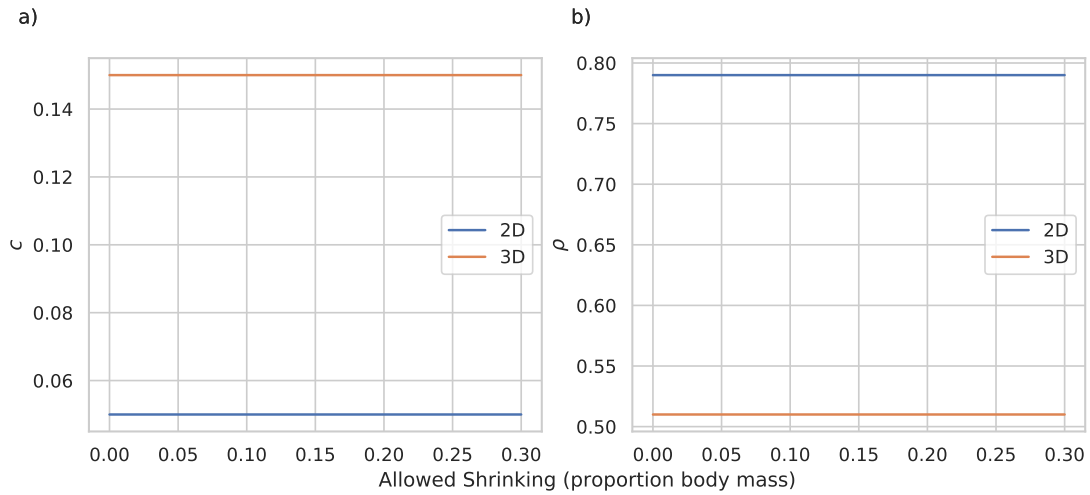


Figure S20: Effect of proportion of shrinking allowed on c and ρ where $\mu = 0.75$ and resource density is high (100 kg/m^D , where D is the dimension).