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**Title**

Resolving global patterns in zooplankton community composition with functional traits

**Authors**

Ryan F. Heneghan1,2\*, Jason E. Everett1,3,4, Patrick Sykes1, Julia L. Blanchard4, Anthony J. Richardson1,5

1 Centre for Applications in Natural Resource Mathematics (CARM), School of Mathematics and Physics, University of Queensland, Brisbane, QLD, Australia

2 Institut de Ciència i Tecnologia Ambientals (ICTA), Universitat Autònoma de Barcelona, Barcelona, Spain

3 School of Biological, Earth and Environmental Sciences, University of New South Wales, Sydney, NSW, Australia

4 Institute of Marine and Antarctic Studies, University of Tasmania, Hobart, TAS, Australia

5 CSIRO Oceans and Atmosphere, Queensland Biosciences Precinct, St Lucia, QLD, Australia

\*Corresponding Author: ryan.heneghan@gmail.com

**Keywords**

Zooplankton, functional traits, size spectrum model, marine ecosystem structure

**Highlights**

* A trait-based ecosystem model to resolve broad patterns in the global zooplankton community
* Zooplankton groups represented by body size range, size-based feeding characteristics and carbon content
* Emergent patterns of total zooplankton biomass and maximum growth rates agree well with empirical observations
* Composition of zooplankton changes from low to high productivity waters, reflecting how functional traits give rise to community structure
* Resolving different zooplankton groups has implications for the total biomass of higher trophic levels, highlighting the critical role of zooplankton as the linchpin of the marine ecosystem

**Abstract**

As the main energy pathway between phytoplankton and fish, zooplankton are a critical component of the global marine ecosystem. Despite their central role, the diversity of zooplankton are not typically resolved in marine ecosystem models. However, trait-based approaches – where zooplankton are grouped based on functional traits such as body size – are increasingly being used to explore zooplankton composition across environmental gradients, and are a promising approach to improving our understanding of zooplankton as the linchpin of the marine ecosystem. Here, we present a functional size-spectrum model that resolves nine major zooplankton functional groups (heterotrophic flagellates and ciliates, larvaceans, omnivorous and carnivorous copepods, chaetognaths, euphausiids, salps and jellyfish), with each group represented using the functional traits of body size, size-based feeding characteristics and carbon content. The model is run globally, and zooplankton community composition emerges based on the relative fitness of the different groups. The emergent pattern of zooplankton biomass and growth rates agrees with empirical data. Across the global ocean there were changes in zooplankton community composition: larvaceans, salps, carnivorous copepods, chaetognaths and jellyfish were most prevalent in the oligotrophic waters, collectively comprising around 80% of zooplankton biomass. In contrast omnivorous copepods and euphausiids dominated in eutrophic regions, together comprising over 90% of zooplankton biomass in eutrophic waters. Finally, we consider the role of the zooplankton groups in supporting higher trophic levels, by exploring the sensitivity of total fish biomass to the removal of individual zooplankton groups. Removing larvaceans caused total fish biomass to decrease by up to 20% in oligotrophic waters. In contrast, total fish biomass increased by up to 50% in oligotrophic waters when carnivorous copepods were excluded from the model. In eutrophic waters, the removal of euphausiids or omnivorous copepods caused fish biomass to decrease by over 60% in eutrophic waters. Our model demonstrates how body size and size-based feeding characteristics are major drivers of both zooplankton community composition and the potential roles of different zooplankton groups in supporting global fish biomass.

# Introduction

Zooplankton are the linchpin of the marine ecosystem, serving as the grazers of phytoplankton and bacteria, and prey of small fish (Mitra *et al.,* 2014). All marine phyla are represented in the zooplankton and this diversity is ubiquitous across the world, with most major groups present in every region of the world’s oceans (O’Brien, 2005; Bucklin *et al.,* 2010). Despite their global presence, the composition of the zooplankton community is not constant, with groups dominating under different biotic and abiotic regimes depending on their relative fitness (Barton *et al.,* 2013). Variation in the zooplankton has implications for how efficiently energy moves through the marine food web from oligotrophic (low primary production) to eutrophic (high primary production) regions, with implications for ecosystem resilience, fisheries catch, and overall productivity (Friedland *et al.,* 2012; Jennings and Collingridge, 2015; Heneghan *et al.,* 2016; Dam and Baumman, 2017). Therefore, resolving the mechanisms that drive shifts in zooplankton community composition across environmental gradients is critical to understanding and predicting how marine ecosystems function.

In current marine ecosystem models, the dynamics of the zooplankton are typically poorly resolved or ignored (Mitra *et al.,* 2014), with these models typically using highly idealized representations of zooplankton as a handful of amorphous boxes that transfer biomass from phytoplankton to fish, as inputs from independent earth system models without feedback from higher trophic levels, or ignored altogether and lumped with phytoplankton or small fish (Everett *et al.,* 2017),. The assumption implicit in current ecosystem models not resolving zooplankton is that the dynamics of this group does not affect ecosystem function. This is despite studies demonstrating that the productivity and structure of higher trophic levels is highly sensitive to the representation of lower trophic levels. For instance, Jennings and Collingridge (2015) demonstrated that the productivity and total biomass of the global fish community was highly sensitive to the trophic transfer efficiency of zooplankton. Further, Irigoien *et al.,* (2014) suggested global fish biomass could be an order of magnitude higher than previously thought and that this discrepancy could be due to a poor understanding of transfer efficiency through the plankton. Heneghan et al., (2016) used a simple model to illustrate how the size-based feeding characteristics of different zooplankton functional groups leads to changes of up to four orders of magnitude in the total fish biomass that can be sustained by a given phytoplankton community.

The choice to simplify the zooplankton is made by modellers because – among other things - the zooplankton community exhibits tremendous taxonomic diversity. However, an alternative to resolving taxonomic diversity is to model organisms based on functional traits, such as body size, body composition and feeding strategy, since these are factors that determine an organism’s relative fitness (McGill *et al.,* 2006; Litchman *et al.,* 2013; Andersen *et al.,* 2016a; McConville *et al.,* 2017). In the past 15 years, the functional trait-based approach has been applied to explain the distribution of phytoplankton groups (Follows *et al.,* 2007; Edwards *et al.,* 2013; Avevedo-Trejos et al., 2015), and there is a growing literature applying the approach to the challenge of the zooplankton (Fuchs and Franks, 2010; Brun *et al.,* 2016; Schnedler-Meyer *et al.,* 2016; Prowe *et al.,* 2018) and fish communities (Stuart-Smith *et al.,* 2013; van Denderen *et al.,* 2018; Petrik *et al.,* 2019).

Body size is a major determinant of the trophic position of zooplankton in the marine food web (Andersen *et al.,* 2016a) and the size-based feeding behaviour of different zooplankton groups structures the zooplankton community across oligotrophic and eutrophic systems (Mitra and Davis, 2010; Barton *et al.,* 2013). Zooplankton are the primary grazers of phytoplankton, which span up to nine orders of magnitude in body size – from picoplankton (0.2 - 2 m equivalent spherical diameter, ESD; 10-14.5 – 10-11.5g wet weight) to microplankton (>20 m ESD; >10-8.4 g wet weight). Moreover, the size structure of phytoplankton changes across environmental gradients (Agawin *et al.,* 2000; Brewin *et al.,* 2010; Barnes *et al.,* 2011). Phytoplankton communities in nutrient rich coastal and upwelling eutrophic systems have greater abundance and proportion of microplankton – which translates to a higher intercept and flatter abundance size spectrum slope. In low nutrient oligotrophic systems, phytoplankton is less abundant and dominated by picoplankton (<2 m ESD), with little or no microplankton (>20 m ESD). In terms of the phytoplankton abundance spectrum, this means a lower intercept and a steeper slope. Changes in the size structure of the phytoplankton have implications for the structuring of the zooplankton community, which in turn affects how primary production is transported to higher trophic levels. This is because zooplankton exhibit vast diversity in their feeding behaviour, with preferred predator-prey mass ratios (PPMRs) varying over 7 orders of magnitude across different functional groups, from for carnivorous copepods to for salps and larvaceans (Wirtz, 2012). Coupled with their range of body sizes, from single-cell flagellates, to jellyfish that reach up to 1 tonne (Levinton, 2013), the pathways of energy transfer from phytoplankton to fish could shift markedly with changes in the composition of the zooplankton community from oligotrophic to eutrophic waters.

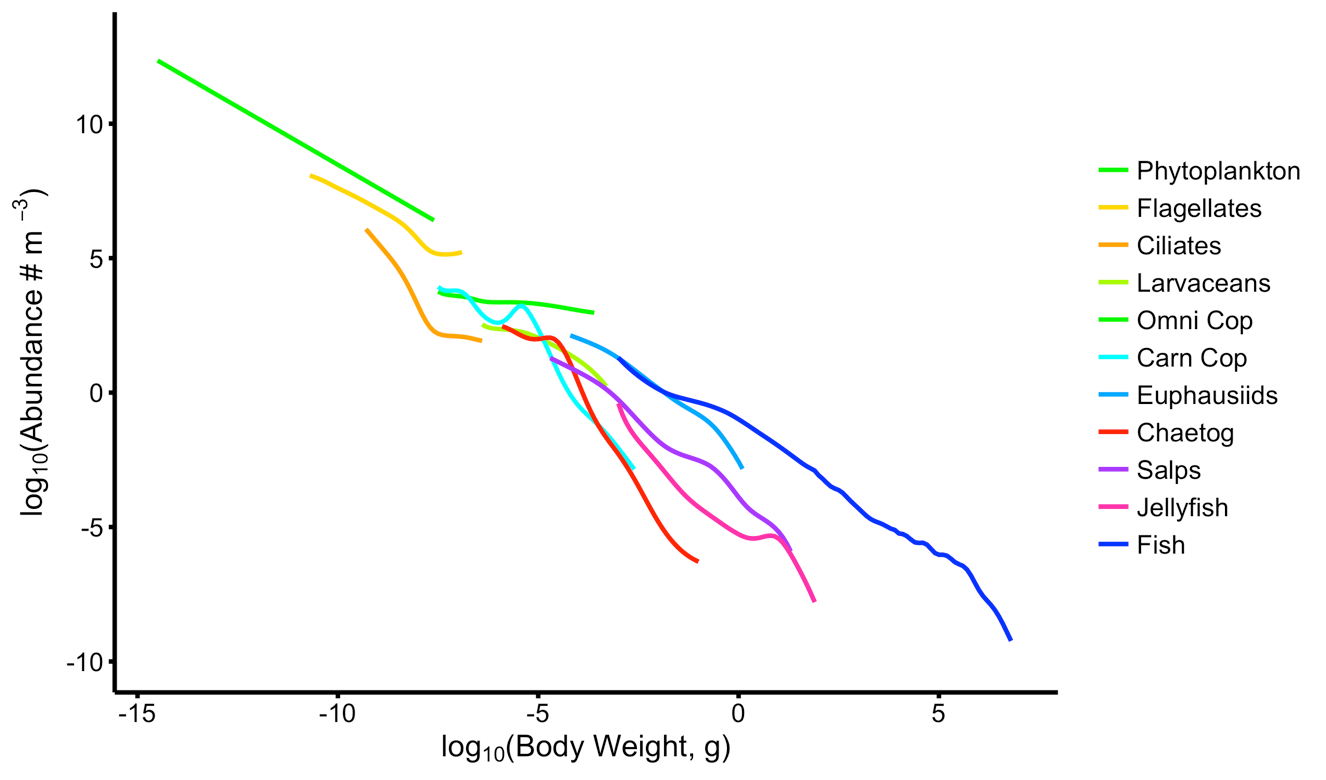
Similar to the diversity in body size and feeding strategy, zooplankton vary widely in their body composition, between gelatinous and non-gelatinous groups (Kiørboe, 2013; McConville et al., 2017). Carbon is the primary structural component of zooplankton (Kiørboe, 2013), and critical physiological and competitive processes such as metabolism, search rate, and average growth efficiency scale with carbon across zooplankton groups (Acuña *et al.,* 2011; Kiørboe, 2011; Kiørboe and Hirst, 2014; McConville *et al.,* 2017). This means that changes in carbon content affects the relative fitness of different zooplankton groups. Moreover, a zooplankton community dominated by more gelatinous groups offers less nutritional value and growth potential for higher trophic levels. This means that resolving the carbon content of the different zooplankton is necessary to resolve the implications of changing zooplankton community composition for energy transfer from phytoplankton to fish.

Using functional traits such as body size and PPMR to resolve the diversity of the zooplankton in marine ecosystem models is not an entirely new idea; 25 years ago Hansen *et al.,* (1994) hypothesised that with knowledge of the size selectivity of different zooplankton groups, it would be possible to construct a simple size-based model of the pelagic food web. Here, we use the extensive literature on zooplankton functional traits, and recent developments in functional size-spectrum modelling (Guiet *et al.,* 2016; Blanchard *et al.,* 2017) to develop a model of the marine ecosystem that resolves phytoplankton, nine zooplankton functional groups (heterotrophic flagellates and ciliates, omnivorous and carnivorous copepods, larvaceans, euphausiids, salps, chaetognaths and jellyfish) and fish. The zooplankton functional groups are resolved using their size ranges, size-based feeding characteristics and carbon content. Across the global ocean, the model is initialised with the same zooplankton composition and the zooplankton community structure emerges. Shifts in zooplankton community composition can be explained by the relative fitness of the various groups in different regions of the global ocean, which is driven by their functional traits. The model is able to reproduce global patterns of zooplankton biomass, as well as maximum growth rates from flagellates to fish. By resolving zooplankton functional diversity and resulting shifts in zooplankton community composition, the model allows us to examine the unique roles of different zooplankton groups in supporting higher trophic level biomass. To that end, we finish with a sensitivity analysis of modelled fish biomass to the absence of different zooplankton functional groups, from oligotrophic to eutrophic waters.

# Methods We used the functional size-spectrum framework (Blanchard et al., 2017) to construct a marine ecosystem model that resolves the body size ranges, size-based feeding characteristics and carbon content of nine of the most abundant zooplankton groups and three fish communities. The functional size spectrum model is run across the global ocean, split into 5x5 degree regions (Figure 1). For each region, the model is forced with annual-average satellite sea surface temperature and chlorophyll *a* from MODIS-Aqua (accessed via the GIOVANNI portal: <https://giovanni.gsfc.nasa.gov/giovanni/>).

## **The functional size spectrum model**

We represented the marine size-spectrum as three communities: phytoplankton, zooplankton and fish. The zooplankton community consists of nine of the most abundant zooplankton groups, and the fish community was made up of a small, medium and large group (Figure 1; Table 1, 2, 3).



**Figure 1:** The modelled marine size-spectrum from one of the 5x5 degree regions of the global ocean. The phytoplankton spectrum is held constant, and abundances of the zooplankton and fish communities are governed by size-dependent processes of growth and mortality.

In keeping with past size-spectrum models, the dynamics of the phytoplankton are not explicitly resolved in the model, with the phytoplankton size-spectrum serving as a static resource for zooplankton (Blanchard *et al.,* 2009; Law *et al.,* 2009; Guiet *et al.,* 2016; Heneghan et al., 2016). Instead, the size structure of the phytoplankton community is estimated directly from satellite chlorophyll *a* observations (Brewin *et al.,* 2010; Barnes *et al.,* 2011; Hirata *et al.,* 2011). The abundances of the zooplankton and fish communities are driven by the size-dependent processes of growth and mortality, with the temporal dynamics of each functional group governed by separate second-order McKendrick-von Foerster equations,

The density of individuals in group of weightat timeper is given by and their individual growth, mortality and diffusion rates are denoted by *,*  and , respectively. We used the MvF equation to govern the dynamics of both fish and zooplankton communities because it is a popular choice for modelling fish-focused size spectrum models, and it is similar to the governing equations in plankton-focused size-based models (Baird and Suthers, 2007; Fuchs and Franks, 2010; Zhou *et al.,* 2010; Ward *et al.,* 2012, Cuesta *et al.,* 2018). Looking ahead, this means that future model developments that resolve the dynamics of the phytoplankton, or more of the functional diversity of the zooplankton could be achieved with the same – or similar - governing equation as what has been employed here.

From the perspective of a predator from group , the feeding rate on prey group *j* depends on the density of suitable prey (g m-3), using a simple type 1 functional response:

where (E9, Table 1; all subsequent equations are also found in Table 1) is the probability a predator of size would consumer an individual of size . The growth rate of a predator from group , of size at time is fuelled by the consumption and conversion of prey biomass to new biomass (g yr-1):

where is the predator’s search rate (E11), is the growth conversion efficiency for predators eating prey from group (E12)and is the effect of temperature on ingestion for group (E3). A type 2 functional response is more commonly used in fish-focused functional size-spectrum models (e.g., Blanchard *et al.,* 2014), however we did not find sufficient information on how maximum consumption scales with body size for the different zooplankton groups in the literature (but see Hansen *et al.,* 1997; Wirtz, 2013; Kiørboe and Hirst, 2014). The main drawback of the type 1 functional response is that it ignores the effect of satiation at high food densities. This means that the maximum growth rates of the zooplankton and fish groups could be unrealistically high in areas with high chlorophyll *a* concentration. However, the effect of high growth rates would be the same across all functional groups, and so would not affect their relative fitness.

Temperature effects are represented using a temperature coefficient:

where and are the temperatures in Kelvin of each 5x5 degree grid square, and the reference temperature where , respectively.

From the perspective of the prey, the total mortality from predation by the larger size classes () is given by:

where is the maximum size of a predator from group . Since individuals grow through time, an additional source of mortality from senescence was incorporated that increased with body size ():

where is the body size after which senescence mortality rapidly increases for an individual from group . This senescence mortality term also acts as a closure term for the largest size classes, by preventing a build-up of large individuals who are not exposed to predation (Andersen *et al.,* 2016b). For an individual of size , at time , from group , total mortality () is given by summing predation and senescence mortality ():

Finally, the second-order diffusion term for an individual from group of size at time is ():

## **Parameterizing the static phytoplankton abundance spectrum**

We split the global ocean into 5x5 degree regions. For each region, the density of phytoplankton of size is given by:

The slope *a*, intercept *b* and maximum size of the static phytoplankton spectrum were derived from temporally and spatially averaged satellite chlorophyll *a* obtained from MODIS-Aqua (accessed via the GIOVANNI portal: https://giovanni.gsfc.nasa.gov/giovanni/), using the synoptic model developed by Brewin *et al.,* (2010). The Brewin model gives an estimate of the percentage contribution of 3 phytoplankton size classes - pico (0.2-2 m ESD), nano (2-20 m ESD) and micro (>20 m ESD) – to the total chlorophyll *a* concentration (mg m-3). Pico-phytoplankton comprise up to 75% of the biomass in low chlorophyll *a* (oligotrophic) waters, declining to less than 10% in high chlorophyll *a* (eutrophic) waters as micro-phytoplankton increase from <10% in oligotrophic waters to over 75% in eutrophic waters, and nano-phytoplankton increase marginally from 20% in oligotrophic to 45% in mid-chlorophyll *a* waters, before declining to around 15% in eutrophic waters (Figure 1 a).

The contribution of micro-phytoplankton, and the phytoplankton community’s maximum size increases with chlorophyll *a* concentration (Brewin *et al.,* 2010; Hirata *et al.,* 2011; Barnes *et al.,* 2011). We incorporated this change in the size range of the micro-phytoplankton with increasing chlorophyll *a* by linearly increasing the maximum size of the micro group from 21 - 60 m ESD, depending on the percentage contribution of the micro group to total chlorophyll *a*. We used 60 m as the maximum possible ESD for the phytoplankton following Barnes et al.’s (2011) finding that 90% of phytoplankton fall below 55-65 m across polar, tropical and upwelling environments. Total chlorophyll *a* concentration for each of the 3 size classes was converted to grams wet weight (assuming 1 g chlorophyll *a* = 50 g C; Zhou *et al.,* 2010, and 1 g C = 10 g wet weight; Hansen *et al.,* 1994, Boudreau & Dickie, 1992, Woodworth-Jefcoats *et al.,* 2013) and the three size ranges were also converted from ESD to grams wet weight (assuming 1 cm3 = 1 g wet weight; Boudreau & Dickie, 1992).

Total biomass in each of the size ranges was then spread uniformly across each size range (Sheldon *et al.,* 1972; Blanchard *et al.,* 2009; Woodworth-Jefcoats *et al.,* 2013; Barange *et al.,* 2014) before being converted to numerical abundance. Finally, slope and intercept were found analytically with E8 (see Appendix 1). The phytoplankton slopes we derived ranged from -1.2 to -0.77 across the global ocean (Figure 2 b), which is similar to the range reported by previous empirical studies (Moreno-Ostos *et al.,* 2015).

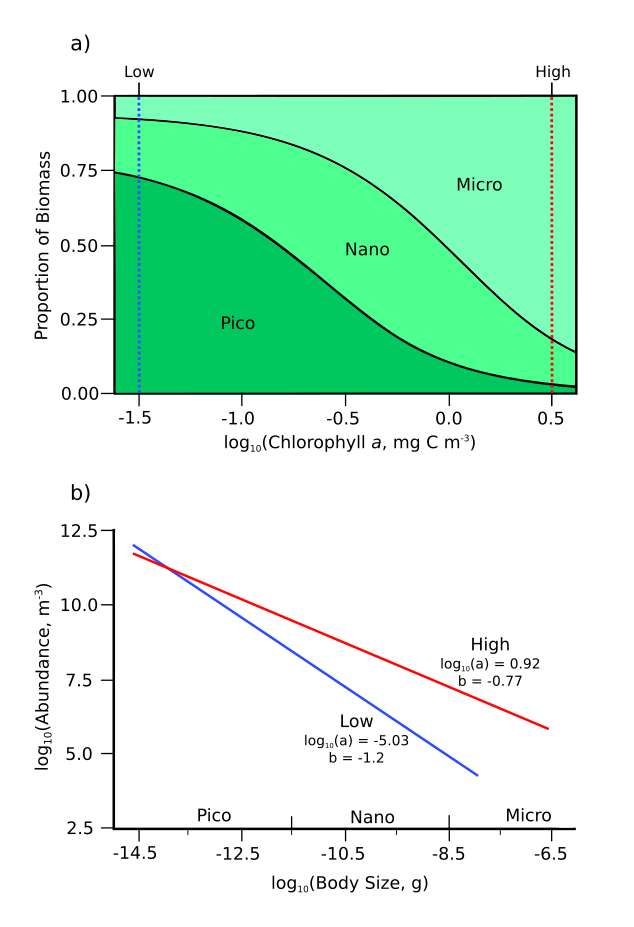


Figure 2 a) Proportion of phytoplankton that is picoplankton ( ESD), nanoplankton ( ESD) and microplankton ( ESD), against log10(Chlorophyll *a),* adapted from Brewin et al., (2015); b) Phytoplankton abundance spectrum at low (10-1.5 g m-3) and high (100.5 g m-3) chlorophyll *a*. The slope *a* and intercept *b* for the two spectra were calculated analytically using satellite chlorophyll *a* from MODIS-Aqua, and the synoptic model from Brewin et al., (2015) (see Appendix 1a for more information).

Table 1 Model equations with their units and in-text reference numbers.

|  |  |  |  |
| --- | --- | --- | --- |
| Description | Equation | Units | Equation Number |
| Growth and mortality: |  |  |  |
| Density of suitable prey from group *j* for group |  |  | E1 |
| Individual growth rate for group |  |  | E2 |
| Temperature effect for group |  | *-* | E3 |
| Individual predation rate |  |  | E4 |
| Senescence mortality |  |  | E5 |
| Total mortality |  |  | E6 |
| Individual diffusion term for group |  |  | E7 |
| Phytoplankton spectrum |  |  | E8 |
| Functional traits: |  |  |  |
| Size selection for group | *=* | - | E9 |
| Feeding kernel width parameter for group |  |  | E10 |
| Search rate for group |  |  | E11 |
| Average growth efficiency for predator eating prey of group *j* |  | *-* | E12 |
| Lower boundary condition for group |  |  | E13 |

## **Incorporating functional traits**

### Body size and predator-prey mass ratio

The marine ecosystem is size-structured, with an individual organism’s position in the food chain strongly dictated by its body size. For each zooplankton group, we established body size range in wet weight using measurements and conversion equations from the literature (see Table 2).

Prey preference for predators of certain size was represented using the ratio of a predator’s body size, against its preferred prey body size (PPMR; Table 2). Across zooplankton taxa, PPMR increases with predator size, due to the non-isometric scaling of feeding-related apparatus with body size (Pearre, 1980; Wirtz, 2012). We used the mechanistic formulation from Wirtz (2012) to calculate the PPMR range for each zooplankton group. Wirtz (2012) links PPMR to a quantitative measure of the feeding mode: raptorial, active feeding is linked to a lower PPMR because predators eat prey closer to their own size. By contrast, passive, suspension feeding yields a higher PPMR. In keeping with previous studies, PPMR for the fish communities was held constant at 100 across their size ranges (Hartvig *et al.,* 2011; Andersen *et al.,* 2016b).

The body sizes and relatively high PPMRs of salps, larvaceans and omnivorous copepods means that these groups feed exclusively on phytoplankton, heterotrophic flagellate and ciliate communities. Euphausiids also have high a PPMR range, but not as large as salps and larvaceans, which means that their largest size classes also access smaller copepods, larvaceans and euphausiids, consistent with their diet in the oceans (Schmidt and Atkinson, 2016). The low PPMR ranges of carnivorous copepods, chaetognaths and jellyfish, coupled with their larger body size means that these groups are almost totally carnivorous, we further restricted their diets so that they do not feed on phytoplankton at all, which is consistent with most current understandings of their diets (Terazaki, 2000; Purcell and Arai, 2001).

### Prey size selectivity

The range of available prey sizes for an individual predator of body size from group is defined by a log-normal feeding kernel, centred on the predator’s preferred predator-prey mass ratio (PPMR; ) and a standard deviation () given by the kernel width parameter for that predator’s group (Table 2):

A wider feeding kernel means a predator can feed from a larger size range of prey. For zooplankton, feeding kernel width is positively correlated with PPMR (Hansen *et al.,* 1994; Fuchs and Franks, 2010; Kiørboe, 2016); filter feeders such as larvaceans or salps with a large average PPMR feed over a wider size range than carnivorous copepods or heterotopic flagellates. We used the empirical model developed by Fuchs and Franks (2010) to link the feeding kernel width of each zooplankton group (), to that group’s average PPMR ():

The feeding kernel width for the fish communities were held constant at 1.3, in keeping with previous studies (Andersen *et al.,* 2016b).

### Growth conversion efficiency and carbon content

Straile (1997) found that average growth conversion efficiency – as a measure of prey carbon converted to predator carbon – was fixed at around 0.25 across a large range of zooplankton taxa. This agrees with the hypothesis that prey groups with a comparatively higher carbon content and energy density contribute more to predator wet-weight growth in comparison to lower carbon groups such as jellyfish (Spitz *et al.,* 2010; Kiørboe, 2013; Mitra *et al.,* 2014). In terms of wet weight, the growth conversion efficiency () of a predator feeding on prey from group *j* prey is

where is the carbon-wet weight ratio of group (Table 2).

### 

Figure 3 Overview of the size ranges of the model’s functional groups (solid boxes), and their prey size ranges (dashed boxes).

Table 2 Parameter values for the nine zooplankton and three fish groups.

|  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| Group | Min. Size, | | | Max Size, | | | log10PPMR range, | Feeding Kernel Width, | Carbon -  Wet Weight Ratio, |
| Length | ESD | log10(g)\* | Length | ESD | log10(g)\* |
| Hetero. Flagellates | - | 3x10-4 cm*a* | -10.7*a* | - | 7x10-3 cm*a* | -6.8*a* | 0.2 – 0.72 | 0.363^ | 0.154 |
| Hetero. Ciliates | - | 1x10-3 cm*b* | -9.3*b* | - | 1x10-2 cm*b* | -6.3*a* | 2.5 – 2.92 | 0.473^ | 0.154 |
| Larvaceans | 8x10-3 cm*c* | 1x10-2 cm*c* | -6.3*c* | 3x10-1 cm*c* | 1x10-1 cm*c* | -2.3*c* | 6.8 – 7.87 | 0.73^ | 0.019 |
| Omni. Cop. | - | 6x10-3 cm*d* | -7.5*d* | 2.8x10-1 cm*e* | 9x10-2 cm*e* | -3.5*e* | 3.6 – 4.62 | 0.573^ | 0.129 |
| Carn. Cop. | - | 6x10-3 cm*d* | -7.5*d* | 6x10-1 cm*e* | 1.8x10-1 cm*e* | -2.5*e* | 0.8 – 1.92 | 0.43^ | 0.129 |
| Euphausiids | - | 6x10-2 cm*f* | -4.2*f* | 6 cm*g* | 1.5 cm*g* | 0.3*g* | 6.6 – 7.83,15 | 0.703^ | 0.129 |
| Chaetognaths | 1x10-1 cm*h* | 1.5x10-2 cm*h* | -5.9*h* | 4 cm*h* | 6x10-1 cm*h* | -0.9*h* | 1.9 - 3.416 | 0.463^ | 0.049 |
| Salps | 5x10-2 cm*i* | 5x10-2 cm*i* | -4.3*i* | 1.9 cm*i* | 1.6 cm*i* | 0.6*i* | 6.8 – 8.52 | 0.73^ | 0.019 |
| Jellyfish | - | 1x10-1 cm*j* | -3*j* | - | 15 cm*j* | 2.6*j* | 2.7 – 4.71 | 0.523^ | 0.0059 |
| Small Fish | - | 1x10-1 cm*k* | -3*k* | - | 6 cm | 2 | 221 | 1.321 | 0.1022 |
| Medium Fish | - | 1x10-1 cm*k* | -3*k* | - | 27 cm | 4 | 221 | 1.321 | 0.1022 |
| Large Fish | - | 1x10-1 cm*k* | -3*k* | - | 125 cm | 6 | 221 | 1.321 | 0.1022 |

\* g wet weight calculated from ESD, assuming 1 gram = 1 cm3.

^ Feeding kernel widths were calculated with the empirical equation derived in (3), using mean log10(PPMR) for this group.

Size range source notations: *a*: From Table 3 in (1), *b*: From figure 1 in (5), *c*: Minimum and maximum larvacean trunk lengths taken from (6) and (8) respectively, and converted to ESD and wet weight using equation derived in (7), *d*: Carbon mass obtained from supplementary material in (10), converted to wet weight and ESD using carbon: wet weight ratio from (9) *e*: Maximum omnivorous and carnivorous copepod lengths taken from (11) and converted to ESD and then wet weight using equation derived in (12), *f*: Euphausiid embryo ESD from figure 2 in (13), *g*: Maximum length taken from supplementary material in (3) and converted to ESD and wet weight using equation from (14), *h*: ESD from supplementary material in (3), derived using head width: body length ratio from (16) *i*: Minimum and maximum salp length taken from (17) and converted to ESD and wet weight using equation derived in (18), *j*: Taken from supplementary material in (19), *k*: Taken from (20).

1. Hansen et al. (1997), 2. Wirtz (2012), 3. Fuchs and Franks (2010), 4. Menden-Deuer and Lessard (2000), 5. Taylor (1978), 6. López-Urrutia (2004), 7. Deibel (1998), 8. Hopcroft et al. (1998), 9. Kiørboe (2013), 10. Kiørboe & Hirst (2014), 11. Benedetti et al. (2016), 12. Azevedo and Dias (2012), 13. Kawaguchi et al. (2011), 14. Meyer and Teschke (2016), 15. Schmidt and Atkinson (2016), 16. Pearre (1982), 17. Henschke et al. (2016), 18. Heron (1988), 19. Acuña et al. (2011), 20. Heneghan et al. (2016), 21. Andersen *et al.,* (2016b), 22. Pauly and Christensen (1995).

Table 3 Model parameter values.

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| Symbol | Definition | Value | Unit | Source |
|  | Coefficient of search rate |  |  | 1 |
|  | Exponent of search rate |  | - | 1,2 |
|  | Body size at which senescence mortality begins for group |  | g | 5,6 |
|  | Coefficient of senescence mortality | 1 |  | 5,6 |
|  | Exponent of senescence mortality | 0.3 | - | 5,6 |
|  | Relative abundance of smallest size class | Flagellates = 1 Ciliates = 0.1 Larvaceans = 0.1 Omni. Copepods = 0.1 Carn. Copepods = 0.1 Euphausiids = 0.1  Chaetognaths = 0.1 Salps = 0.01 Jellyfish = 0.01 | - | See 2.1.5.1 |
|  | Temperature scaling coefficient | 2 | - | - |
|  | Reference temperature | 303.15 | K | - |

Z, zooplankton; F, fish.

1. Hansen et al., 1997, 2. Peters 1983, 3. Jennings et al., 2008, 4. Blanchard et al., 2012, 5. Hall et al., 2006, 6. Heneghan et al., 2016.

## **Numerical implementation**

### Boundary conditions for the size-spectrum model

For each zooplankton group, the abundance of the smallest size class at time , , was fixed with respect to the total abundance of the other groups in that size class:

where is the relative abundance of group in size class , with respect to the total abundance of the other groups. For the smallest group, heterotrophic flagellates, the abundance of their smallest size class was fixed to be equal to the abundance of the phytoplankton community in the same size class, in keeping with past size spectrums studies (Heneghan et al., 2016). Similarly, the smallest size class abundance for the total fish community was fixed at the total zooplankton abundance in that size class, divided equally among the three fish groups. For all other zooplankton groups, except for salps and jellyfish, was fixed at 0.1. Salps and jellyfish are the two largest zooplankton groups, and was set at 0.01 to prevent these groups from dominating the biomass of the zooplankton community.

### Running the size-spectrum model

Dynamics of the zooplankton and fish groups were modelled with separate second order McKendrick-von Foerster equations, which we solved numerically using a second order semi-implicit upwind finite difference scheme (Press *et al.,* 2007). For numerical implementation we discretised the zooplankton and fish community size ranges into equal 0.1 log10 size intervals. The model is initialised with the same zooplankton community, then integrated forward through time for 500 years, with a half weekly time step. We chose these values to discretise the weight and time intervals after experimenting with smaller and larger interval widths, to ensure convergence in our numerical implementation, whilst minimising the time required to run the simulation.

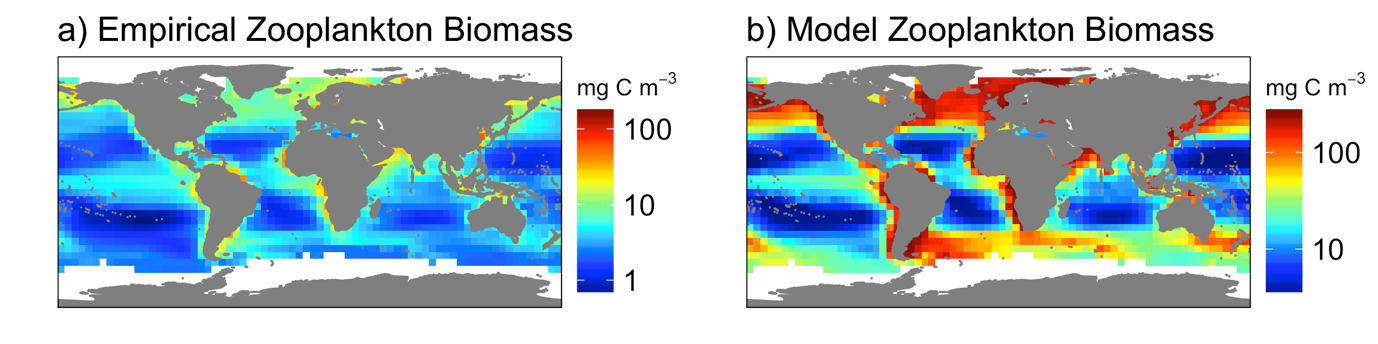
### Size-spectrum model assessment

Global scale datasets of zooplankton biomass separated into different functional groups, that could serve to evaluate our model, are quite scarce. However, Strömberg et al., (2009) created a map of the distribution of total zooplankton carbon biomass, using observations from the COPEPOD database and the Continuous Plankton Recorder (CPR) dataset, which we can use to assess the performance of the model in capturing the global pattern of total zooplankton biomass. We will focus on comparing our model’s emergent patterns of biomass with Stromberg et al.’s distributions, over the absolute biomass from either approach. This is because of the difference between the sampled biomass data underlying Strömberg et al.’s (2009) global distribution, which only captures a fraction of the total biomass of the entire zooplankton community. In contrast, the total biomass from the size-spectrum model will be the sum of all the zooplankton in the model, so we would expect our model’s biomass to be higher than that of Strömberg et al. (2009).

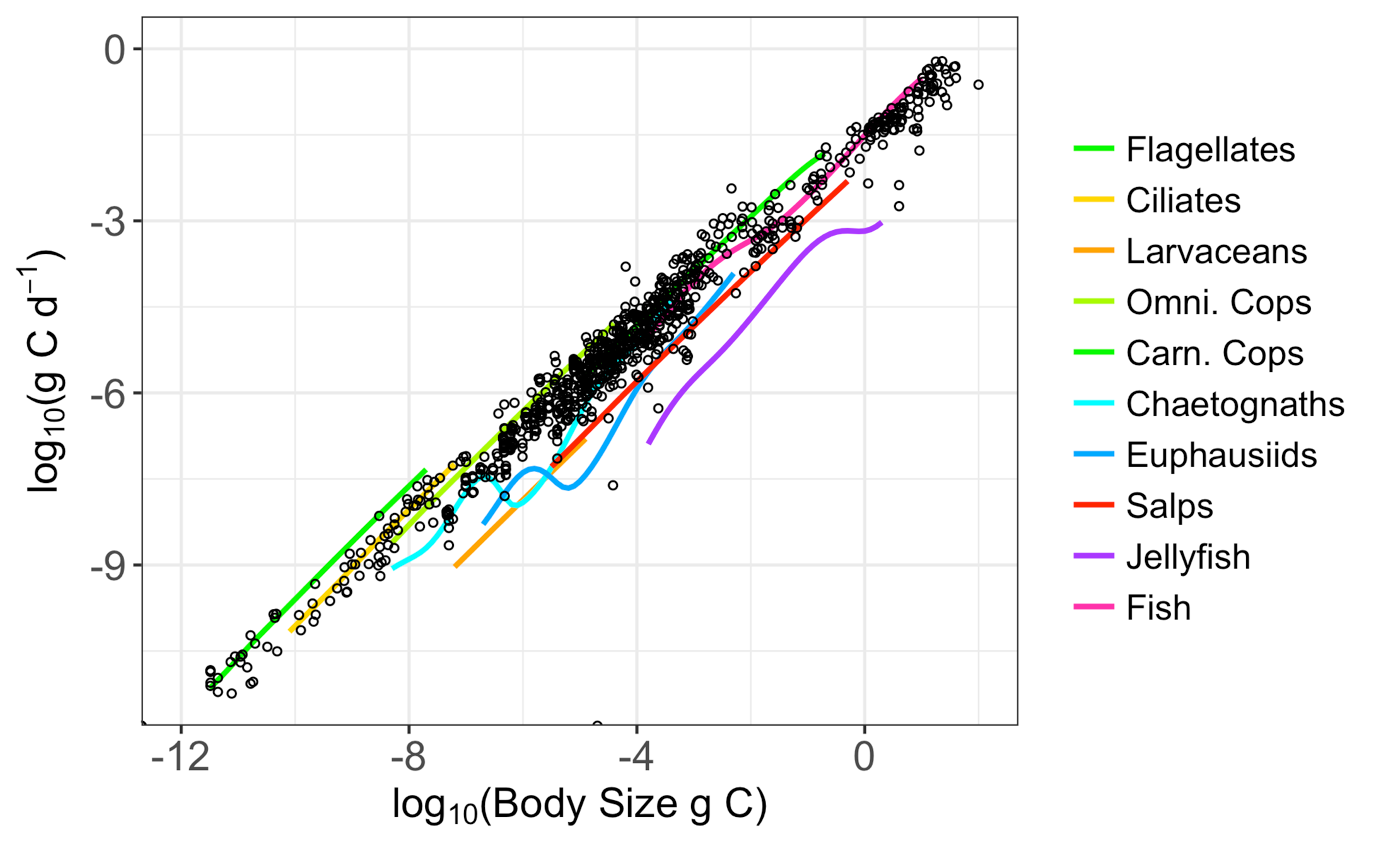
Energy transfer through the zooplankton from phytoplankton to fish is driven by the growth rates of the different zooplankton functional groups. To test whether the model produced realistic rates of energy transfer through the zooplankton functional groups, we compared modelled maximum growth rates with data from Kiørboe and Hirst, 2014 (deposited in Pangaea, http:// doi.pangaea.de/10.1594/PANGAEA.819855). The dataset from Kiørboe and Hirst, 2014 contains maximum growth rates with body size for many zooplankton functional groups, from flagellates to planktivorous fish. Comparing the model’s maximum growth rates with data also allows us to assess whether or not a simple type 1 functional response is a sufficient parameterization of zooplankton feeding rates at this scale.

# Results

**Global zooplankton biomass**

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Figure 4** Comparison of model map of global zooplankton carbon biomass (mg C m-3), with a map derived from observations. a) Global annual average (1998-2005) zooplankton carbon biomass map (mg C m-3), calculated with zooplankton biomass observations from the COPEPOD database and the Continuous Plankton Recorder. Reproduced with permission from Strömberg et al., (2009). b) Functional size-spectrum model total zooplankton carbon biomass map (mg C m-3).

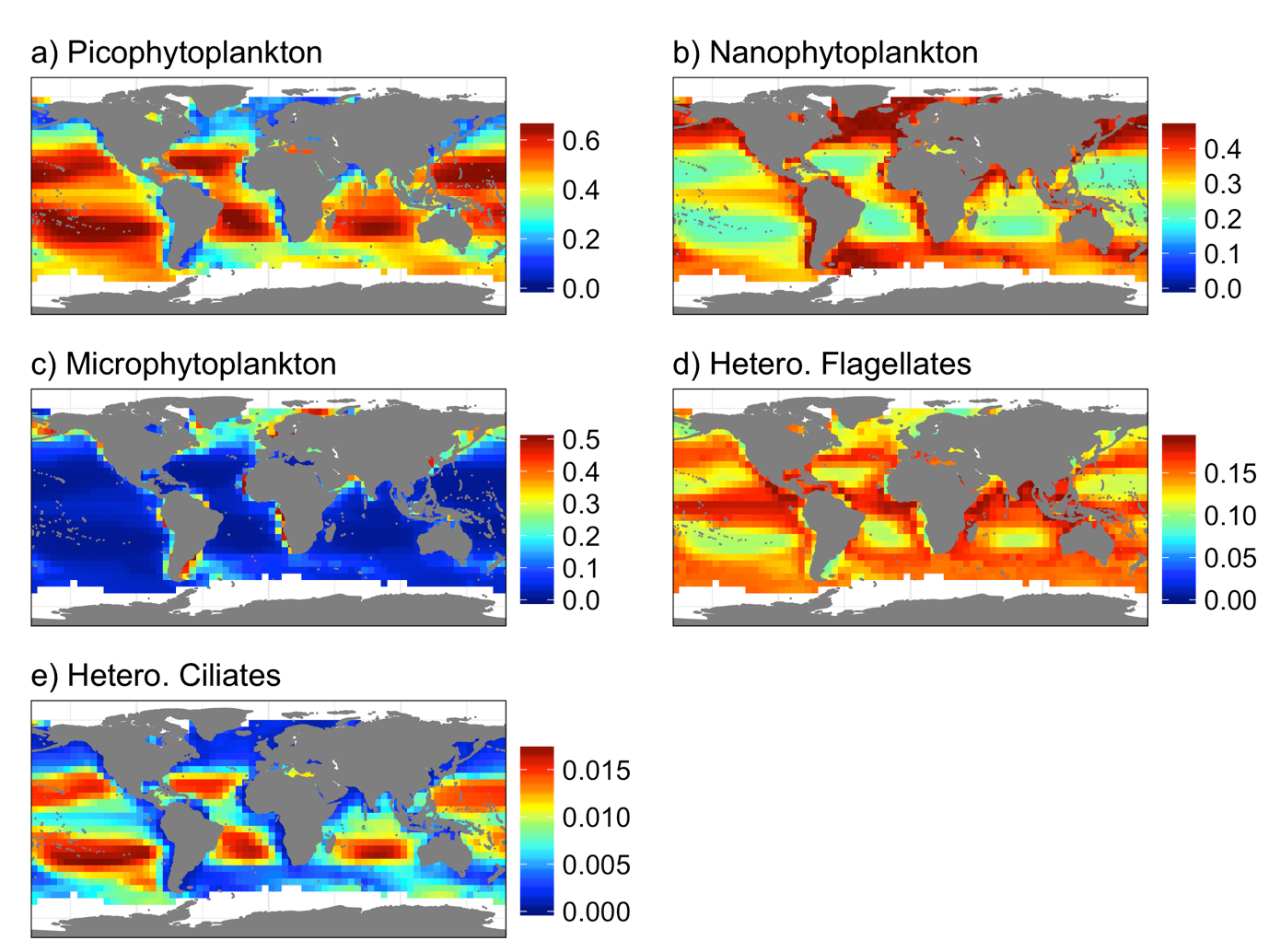
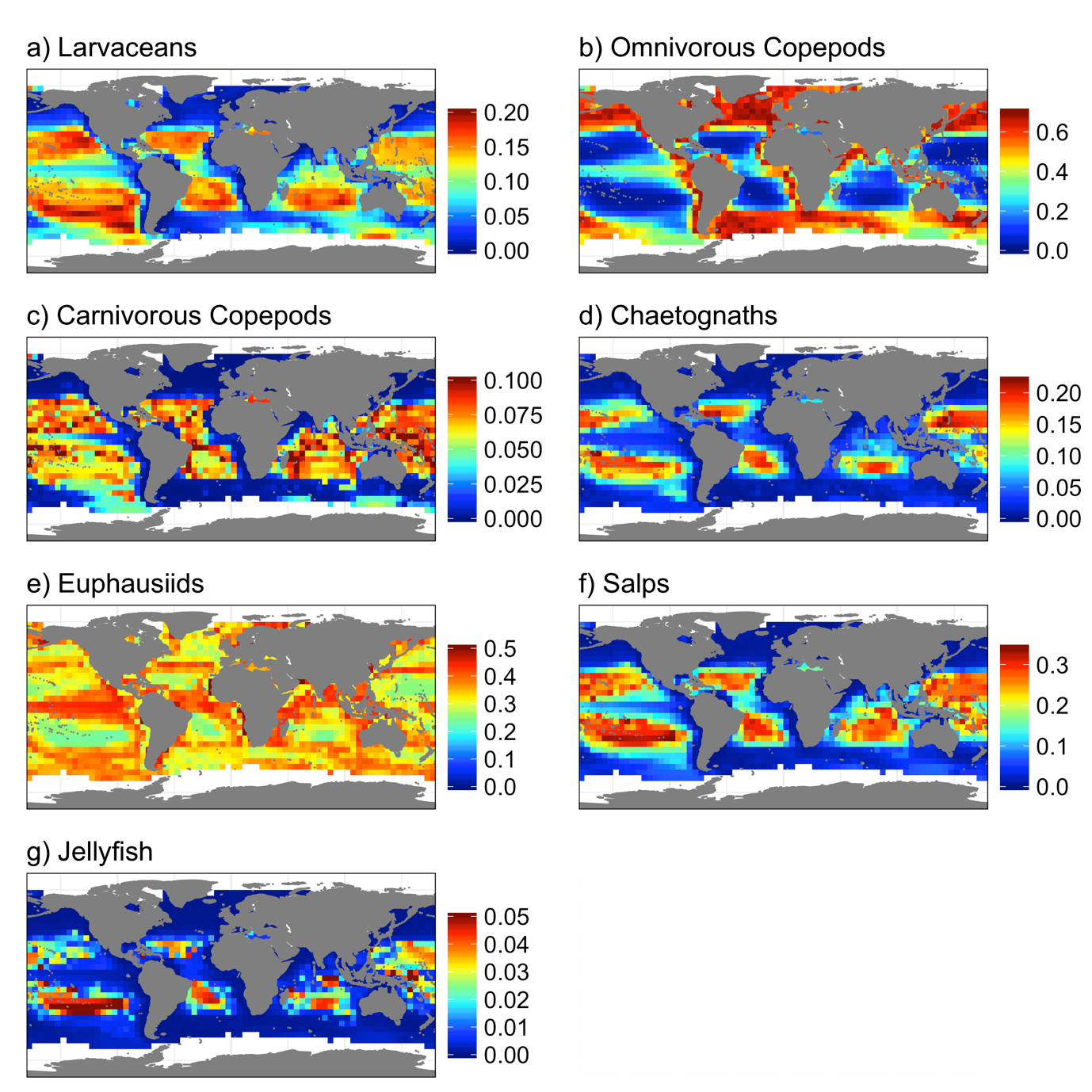
Global distribution of zooplankton carbon biomass from the size-spectrum model is shown in comparison to an empirically derived estimate in Figure 4. There is strong agreement in the spatial pattern between the model’s global zooplankton biomass and the empirically derived map, with a correlation coefficient of 0.84 between the two. The model qualitatively captures global-scale patterns of empirically derived total zooplankton biomass (Figure 4 a,b), showing the lowest levels of biomass in the oligotrophic ocean gyres, and highest in upwelling regions and coastal shelves. Both data-derived and model distributions of global biomass is similar to that of satellite-derived chlorophyll *a* at this scale. However, our model has a larger number of grid-squares with maximum or near maximum zooplankton biomass in high-productivity regions, compared to the empirically derived estimate, this could indicate that our model is overestimating the biomass of zooplankton in eutrophic waters. On the other hand, it could also mean that zooplankton biomass is underestimated in the observations underpinning the empirically derived estimate.

**Maximum growth rates  
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**Figure 5** Modelled maximum growth rates (g C d-1) as a function of body size (g C) and functional group. The black circles represent the food saturated maximum growth rates (g C d-1) with body size (g C) of juveniles from various functional groups (nanoflagellates, ciliates, non-calanoid and calanoid copepods, other crustaceans, tunicates, cnidarians and ctenophores, chaetognaths and planktivorous fish), taken from Kiørboe and Hirst 2013 (available at Pangaea, ﻿http:// doi.pangaea.de/10.1594/PANGAEA.819855).

In Figure 5, modelled maximum growth rates across body size and functional groups were plotted with observed maximum growth rates from Kiørboe and Hirst (2013). The observed maximum growth rates come from various functional groups (nanoflagellates, ciliates, non-calanoid and calanoid copepods, other crustaceans, tunicates, cnidarians and ctenophores, chaetognaths and planktivorous fish), allowing us to compare across the functional groups in our model. Modelled maximum growth rates fell within the range of the observed maximum growth rates, with the model’s maximum growth rates not noticeably exceeding the empirical data for any body size or functional group. Although this is not a comprehensive assessment of the growth rates from the model, it does indicate that our model is not overestimating the growth rates and resulting energy flow through the zooplankton to higher trophic levels.

**Zooplankton community composition**

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Figure 6** Emergent relative biomass (wet weight composition) of the microzooplankton (heterotrophic flagellates and ciliates) and phytoplankton community. Biomass’ of the phytoplankton groups are not emergent from the model, but come from splitting total phytoplankton biomass using Brewin et al.,’s (2015) synoptic model (see section 2.2). Proportion of total wet weight for these five groups from a) Picophytoplankton, b) Nanophytoplankton, c) Microphytoplankton, d) Heterotrophic Flagellates, e) Heterotrophic Ciliates.   


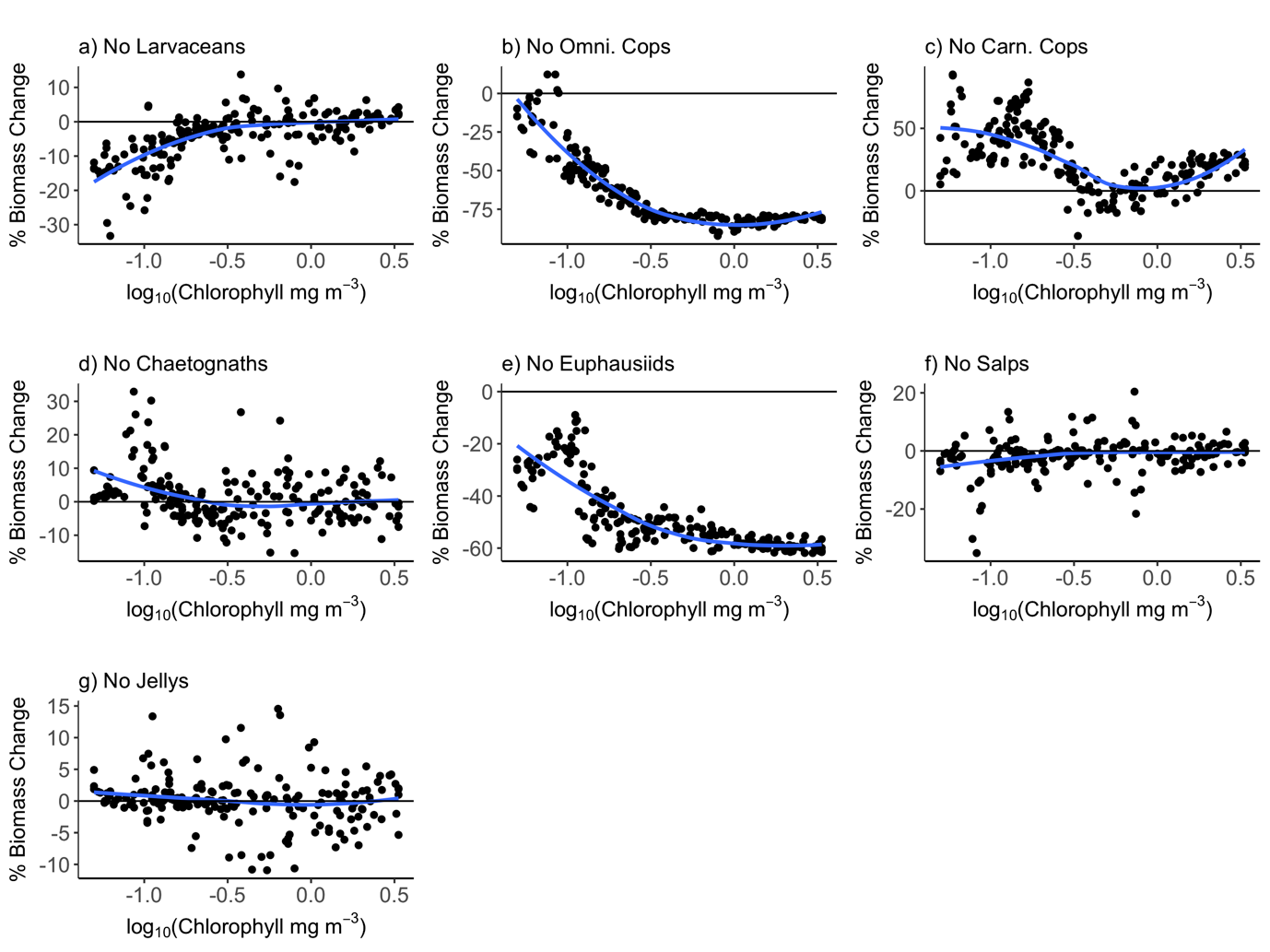
**Figure 7** Emergent relative biomass (wet weight composition) of the macrozooplankton community (excluding flagellates and ciliates) from the size-spectrum model. Proportion of total macrozooplankton wet weight (excluding flagellates and ciliates) from a) Larvaceans, b) Omnivorous Copepods, c) Carnivorous Copepods, d) Chaetognaths, e) Euphausiids, f) Salps and g) Jellyfish.

Across the global ocean, the emergent composition of the plankton was not uniform. To see shifts in community composition more clearly, we split the total phytoplankton and zooplankton communities into micro (phytoplankton, heterotrophic flagellates and ciliates; Figure 6) and macroplankton groups (all zooplankton groups except heterotrophic flagellates and ciliates. Size classes in the phytoplankton community were set and are not emergent from the model, but help drive changes in the zooplankton community. Picophytoplankton comprise over 60% of the microplankton biomass in oligotrophic waters, declining to around 10% in eutrophic waters as microphytoplankton increase from <5% in oligotrophic waters to around 50% in eutrophic waters, and nanophytoplankton increase from 25% of macroplankton biomass in oligotrophic to around 50% in eutrophic waters (Figure 6 a, b, c).

Heterotrophic flagellates comprise between 10-15% of the microplankton biomass in oligotrophic waters, increasing to around 20% in eutrophic waters (Figure 6 d). In contrast, heterotrophic ciliates were most prevalent in oligotrophic regions, comprising around 1.5% of microplankton biomass, but decreasing to less than 0.1% in eutrophic regions (Figure 6 e). Taken together, heterotrophic flagellates and ciliates comprised at least 15% of the total microplankton biomass across the global ocean.

In the seven macrozooplankton groups (Figure 7), omnivorous copepods and euphausiids were a significant component of the zooplankton in all regions, comprising at least 10% and 25% respectively of the macrozooplankton biomass in oligotrophic waters. They were most prevalent in eutrophic regions, where nanophytoplankton and microphytoplankton - the two size classes of phytoplankton that fall in these groups’ prey size ranges (Figure 3) - dominate the microplankton community (Figure 6 b,c), with their respective biomass’ increasing to around 60% and 40% in these regions (Figure 7 b, e). In contrast, salps and larvaceans, have similar body size ranges, but larger predator-prey mass ratios than omnivorous copepods and euphausiids (Table 2; Figure 3), which means that they can access the picophytoplankton for food (Figure 3). As a result salps and larvaceans are most prevalent in the oligotrophic open ocean gyres, where picophytoplankton comprise over 60% of microplankton biomass (Figure 6 a), respectively making up about 20% and 30% of macrozooplankton biomass in these waters, declining to less than 1% in eutrophic regions. Finally, the three carnivorous groups – carnivorous copepods, chaetognaths and jellyfish – were most prevalent in oligotrophic waters, altogether making up over 30% of microzooplankton biomass in these regions (Figure 7 c, d, g). These three groups only consume other zooplankton, and their dominance in oligotrophic waters suggests increasing top-down grazing control in the macrozooplankton community in these regions. In contrast the dominance of omnivorous copepods and euphausiids, which can take advantage of the increasing concentration of larger phytoplankton, suggests more bottom-up control in eutrophic shelf and upwelling regions.

**Fish community biomass sensitivity**



**Figure 8** Change in total fish biomass against chlorophyll *a* concentration when a) Larvaceans, b) Omnivorous copepods, c) Carnivorous Copepods, d) Chaetognaths, e) Euphausiids, f) Salps and g) Jellyfish are omitted from the model, compared to total fish biomass when all zooplankton groups are present. Each dot represents an individual 5x5 degree grid square from the global ocean. To clarify the trend with chlorophyll *a* in each figure, a blue line was fit with a spline smoother. The solid black line at 0% shows where there is no change in total fish biomass.

To assess the sensitivity of fish community biomass to the composition of the zooplankton from oligotrophic to eutrophic waters, we calculated the change in total fish biomass when each of the zooplankton functional groups (excluding flagellates and ciliates) were removed individually, and plotted this against chlorophyll *a* (Figure 8).

Removing larvaceans, salps, carnivorous copepods, chaetognaths or jellyfish - groups that were most prevalent in oligotrophic regions of the global ocean (Figure 7 a, c, d, f, g) - caused the greatest change in fish biomass in low chlorophyll *a* grid squares. The removal of larvaceans caused declines in fish biomass of up to 30% in grid squares with low chlorophyll *a*, however as chlorophyll *a* increased the effect of larvaceans being removed did not cause a consistent increase or decrease in fish biomass (Figure 8a). The removal of salps caused a similar pattern of fish biomass change as larvaceans, however the declines in oligotrophic waters were less pronounced (Figure 8 f). In contrast, removing carnivorous copepods led to an increase of up to 90% in fish biomass in low chlorophyll *a* areas, and in regions of high productivity removing this group caused an increase of around 10% in fish biomass (Figure 8 c). This is despite carnivorous copepods comprising less than 1% of the biomass of macrozooplankton in eutrophic regions (Figure 7 c). Removing chaetognaths also caused up to a 30% increase in fish biomass in low chlorophyll *a* regions, however the trend with chlorophyll *a* was weak, with no consistent pattern of biomass change for most regions when this group was removed (Figure 8 d). Finally, when jellyfish were removed there was a 1-2% average increase in fish biomass in low chlorophyll *a* regions, however this increase was small compared to the 10% changes in different grid squares, from low to high chlorophyll *a* regions(Figure 8 g).

Omnivorous copepods and euphausiids comprised over 90% of macrozooplankton biomass in eutrophic upwelling and shelf regions (Figure 7 b, e), and removing these groups caused the greatest declines in fish biomass in high chlorophyll *a* regions. When omnivorous copepods were removed, there was no change in fish biomass in low chlorophyll *a* regions, however as chlorophyll *a* increased total fish biomass was reduced by up to 80%, compared to when this group was included in the zooplankton community (Figure 8 b). Removing euphausiids caused declines of up to 60% in high chlorophyll *a* waters, however unlike omnivorous copepods low chlorophyll *a* waters saw a reduction of 20% in total fish biomass(Figure 8 e), reflecting the fact that euphausiids still comprised about 25% of microzooplankton biomass in oligotrophic waters (Figure 7 e).

# ﻿Discussion

Implications of the work

The challenge of resolving the zooplankton in ecosystem models lies in their diversity of species, life histories and ecological strategies (Litchman *et al.,* 2013). A significant advantage of the trait-based approach is that community structure can emerge based on the functional traits of the community, and environmental conditions. Our results demonstrate the importance of functional traits in governing global patterns in zooplankton community composition.   
  
This work corroborates existing hypotheses about zooplankton community structure, it also raises new hypotheses.

Gaps  
Reproductive processes

Phytoplankton dynamics, motivate this gap by discussion of growth rates of salps and larvaceans

Next steps

# Testing emergent patterns of zooplankton composition with data

Exploring energy pathways from phytoplankton to fish across oligotrophic and eutrophic waters.

How changes in the zooplankton community affects the global carbon cycle.

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# Scrap writing

Feeding strategies – they matter at this scale (Brun et al., 2017; Prowe et al., 2018). Only been looked at for copepods, whereas PPMRs across different groups… not sure if this is an important point.

Temperature.

# Schnedler-Meyer et al., 2016 – jellyfish increase with productivity, turbidity and fishing.

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