

1 **Role of jellyfish in the plankton ecosystem revealed using a**  
2 **global ocean biogeochemical model**

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11

12 **Abstract.** Jellyfish are increasingly recognised as important components of the marine ecosystem, yet their  
13 specific role is poorly defined compared to that of other zooplankton groups. This paper presents the first global  
14 ocean biogeochemical model that includes an explicit representation of jellyfish and uses the model to gain insight  
15 into the influence of jellyfish on the plankton community. The PlankTOM11 model groups organisms into  
16 Plankton Functional Types (PFT). The jellyfish PFT is parameterised here based on our synthesis of observations  
17 on jellyfish growth, grazing, respiration and mortality rates as functions of temperature and on jellyfish biomass.  
18 The distribution of jellyfish is unique compared to that of other PFTs in the model. The jellyfish global biomass  
19 of 0.13 PgC is within the observational range, and comparable to the biomass of other zooplankton and  
20 phytoplankton PFTs. The introduction of jellyfish in the model has a large direct influence on the crustacean  
21 macrozooplankton PFT and influences indirectly the rest of the plankton ecosystem through trophic cascades. The  
22 zooplankton community in PlankTOM11 is highly sensitive to the jellyfish mortality rate, with jellyfish  
23 increasingly dominating the zooplankton community as its mortality diminishes. Overall the results suggest that  
24 jellyfish play an important role in regulating global marine plankton ecosystems, which has been generally  
25 neglected so far.

26

## 27 1 INTRODUCTION

28

29 Gelatinous zooplankton are increasingly recognised as influential organisms in the marine environment, not just  
30 for the disruptions they can cause to coastal economies (fisheries, aquaculture, beach closures and power plants  
31 etc.; Purcell et al., 2007), but also as important consumers of plankton (Lucas and Dawson, 2014), a food source  
32 for many marine species (Lamb et al., 2017) and as key components in marine biogeochemical cycles (Crum et  
33 al., 2014; Lebrato et al., 2012). The term gelatinous zooplankton can encompass a wide range of organisms across  
34 three phyla: Tunicata (salps), Ctenophora (comb-jellies), and Cnidaria (true jellyfish). This study focuses on  
35 Cnidaria (including Hydrozoa, Cubozoa and Scyphozoa), which contribute 92% of the total global biomass of  
36 gelatinous zooplankton (Lucas et al., 2014). The other gelatinous zooplankton groups, Tunicata and Ctenophora,  
37 are excluded from this study because there are far fewer data available on their biomass and vital rates than for  
38 Cnidaria, and they only contribute a combined global biomass of 8% of total gelatinous zooplankton (Lucas et al.,  
39 2014). Cnidaria are both independent enough from other gelatinous zooplankton, and cohesive enough to be  
40 represented as a single Plankton Functional Type (PFT) for global modelling (Le Quéré et al., 2005). For the rest  
41 of this paper pelagic Cnidaria are referred to as jellyfish.

42 Jellyfish exhibit a radially symmetrical body plan and are characterised by a bell-shaped body (medusae).  
43 Swimming is achieved by muscular, “pulsing” contractions and animals have one opening for both feeding and  
44 excretion. Most scyphozoans and cubozoans, and many hydrozoans, follow a meroplanktonic life cycle. A sessile  
45 (generally) benthic polyp buds off planktonic ephyrae asexually. These, in turn, grow into medusae that reproduce  
46 sexually to generate planula larvae, which then settle and transform into polyps. Within this general life cycle,  
47 there is large reproductive and life cycle variety, including some holoplanktonic species that skip the benthic  
48 polyp stage as well as holobenthic species that skip the pelagic phase, and much plasticity (Boero et al., 2008;  
49 Lucas and Dawson, 2014).

50 Jellyfish are significant consumers of plankton, feeding mostly on zooplankton using tentacles and/or oral arms  
51 containing stinging cells called nematocysts (Lucas and Dawson, 2014). The large body size to carbon content  
52 ratio of jellyfish creates a low maintenance, large feeding structure, which, because they do not use sight to capture  
53 prey, allow them to efficiently clear plankton throughout 24 hours (Acuña et al., 2011; Lucas and Dawson, 2014).  
54 Jellyfish are connected to lower trophic levels, with the ability to influence the plankton ecosystem structure and  
55 thus the larger marine ecosystem through trophic cascades (Pitt et al., 2007, 2009; West et al., 2009). Jellyfish  
56 have the ability to rapidly form large high-density aggregations known as blooms that can temporarily dominate  
57 local ecosystems (Graham et al., 2001; Hamner and Dawson, 2009). Jellyfish contribute to the biogeochemical  
58 cycle through two main routes; from life through feeding processes, including the excretion of faecal pellets,  
59 mucus and messy-eating, and from death, through the sinking of carcasses (Chelsky et al., 2015; Lebrato et al.,  
60 2012, 2013a; Pitt et al., 2009). The high biomass achieved during jellyfish blooms, and the rapid sinking of  
61 excretions from feeding and carcasses from such blooms, make them a potentially significant vector for carbon  
62 export (Lebrato et al., 2013a, 2013b; Luo et al., 2020).

63 Anthropogenic impacts from climate change, such as increasing temperature and acidity (Rhein et al., 2013), and  
64 fishing, through the removal of predators and competitors (Doney et al., 2012), impact the plankton including

65 jellyfish (Boero et al., 2016; but see Richardson and Gibbons, 2008). Multiple co-occurring impacts make it  
66 difficult to understand the role of jellyfish in the marine ecosystem, and how the role may be changed by the co-  
67 occurring impacts. The paucity of historical jellyfish biomass data, especially outside of coastal regions and the  
68 Northern Hemisphere, has made it difficult to establish jellyfish global spatial distribution, biomass and trends  
69 from observations (Brotz et al., 2012; Condon et al., 2012; Gibbons and Richardson, 2013; Lucas et al., 2014; Pitt  
70 et al., 2018).

71 Models are useful tools to help understand the interactions of multiple complex drivers in the environment. This  
72 paper describes the addition of jellyfish to the PlankTOM10 global ocean biogeochemical model, which we call  
73 PlankTOM11. PlankTOM10 represents explicitly 10 PFTs; six phytoplankton, one bacteria and three zooplankton  
74 (Le Quéré et al., 2016). The three zooplankton groups are protozooplankton (mainly heterotrophic flagellates and  
75 ciliates), mesozooplankton (mainly copepods) and macrozooplankton (as crustaceans, mainly euphausiids; see  
76 Table 1 for definitions). Jellyfish is therefore the fourth zooplankton group and 11<sup>th</sup> PFT in the PlankTOM model  
77 series. It introduces an additional trophic level to the ecosystem. To our knowledge, this is the first and only  
78 representation of jellyfish in a global ocean biogeochemical model at the time of writing. PlankTOM11 is used to  
79 help quantify global jellyfish biomass and the role of jellyfish for the global plankton ecosystem.

## 80 2 METHODS

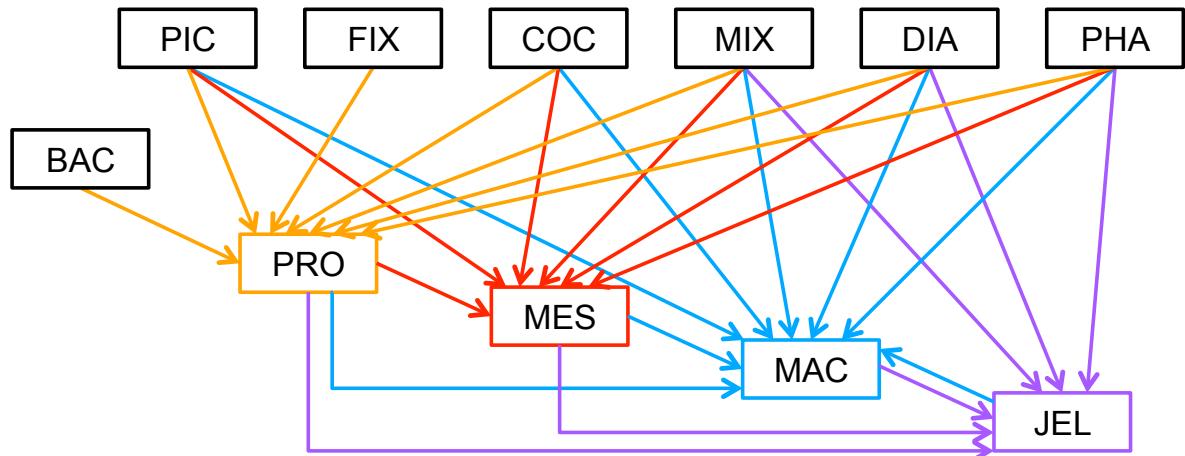
### 81 2.1 PLANKTOM11 MODEL DESCRIPTION

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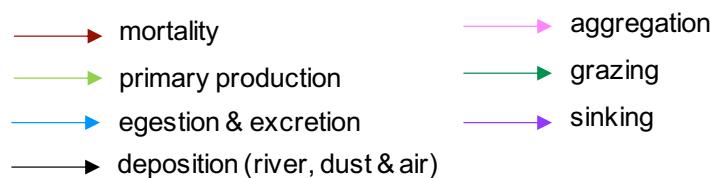
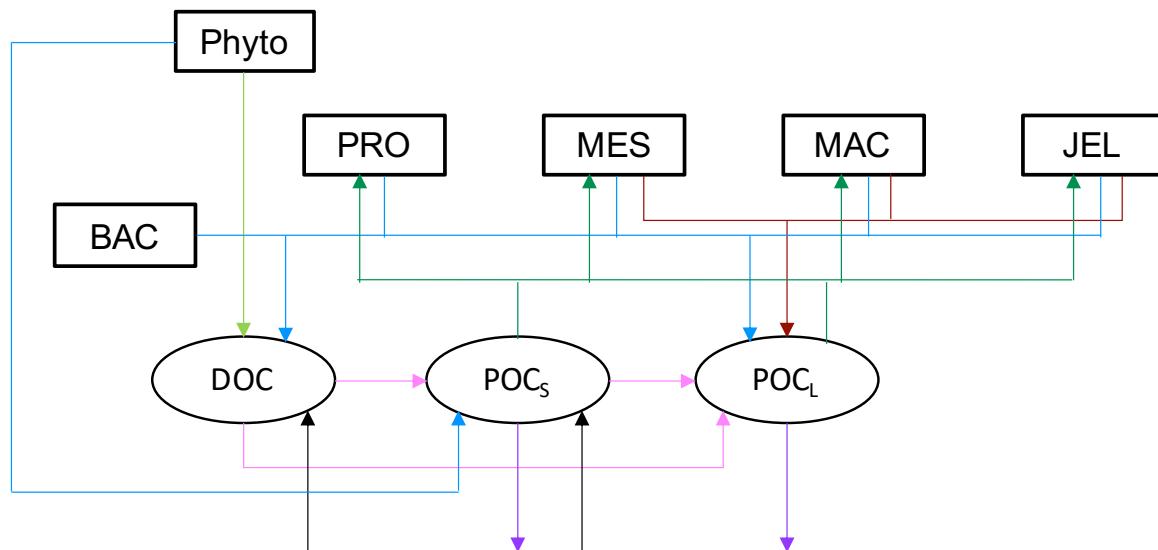
83 PlankTOM11 was developed starting from the 10 PFT version of the PlankTOM model series (Le Quéré et al.,  
84 2016), by introducing jellyfish as an additional trophic level at the top of the plankton food web (Fig. 1a). A full  
85 description of PlankTOM10 is published in Le Quéré et al. (2016), including all equations and parameters. Here  
86 we provide an overview of the model development, focussing on the parameterisation of the growth and loss rates  
87 of jellyfish and how these compare to the other macrozooplankton group. We also describe the update of the  
88 relationship used to describe the growth rate as a function of temperature and subsequent tuning. The formulation  
89 of the growth rate is the only equation that has changed since the previous version of the model (Le Quéré et al.,  
90 2016), although many parameters have been modified (Sect. 2.1.6).

91 PlankTOM11 is a global ocean biogeochemistry model that simulates plankton ecosystem processes and their  
92 interactions with the environment through the representation of 11 PFTs (Fig. 1). The 11 PFTs consist of six  
93 phytoplankton (picophytoplankton, nitrogen-fixing cyanobacteria, coccolithophores, mixed phytoplankton,  
94 diatoms and *Phaeocystis*), bacteria, and four zooplankton (Table 1). Physiological parameters are fixed within  
95 each PFT, and therefore, within-PFT diversity is not included. Spatial variability within PFTs is represented  
96 through parameter-dependence on environmental conditions including temperature, nutrients, light and food  
97 availability.

(a) Plankton food web



(b) Sources and sinks for organic carbon



98

99 *Figure 1. Schematic representation of the PlankTOM11 marine ecosystem model (see Table 1 for PFT definitions). (a) The*  
100 *plankton food web, arrows represent the grazing fluxes by protozooplankton (orange), mesozooplankton (red),*  
101 *macrozooplankton (blue) and jellyfish zooplankton (purple). Only fluxes with relative preferences above 0.1 are shown (see*  
102 *Table 3). (b) Source and sinks for dissolved organic carbon (DOC) and small (POC<sub>S</sub>) and large (POC<sub>L</sub>) particulate organic.*

103

104 The model contains 39 biogeochemical tracers, with full marine cycles of key elements carbon, oxygen,  
105 phosphorus and silicon, and simplified cycles of nitrogen and iron. There are three detrital pools: dissolved organic

106 carbon (DOC), small particulate organic carbon (POCs) and large particulate organic carbon (POC<sub>L</sub>). The  
 107 elements enter through riverine fluxes and are cycled and generated through the PFTs via feeding, faecal matter,  
 108 messy-eating and carcasses (Fig. 1b; see Sect. 2.1.5. for detail; Buitenhuis et al., 2006, 2010, 2013a; Le Quéré et  
 109 al., 2016). Model parameters are based on observations where available. A global database of PFT carbon biomass  
 110 that was designed for model studies (Buitenhuis et al., 2013b) and global surface chlorophyll from satellite  
 111 observations (SeaWiFS) are used to guide the model developments.

**Table 1.** Size range and descriptions of PFT groups used in PlankTOM11. Adapted from Le Quéré et al. (2016).

Name	Abbreviation	Size Range $\mu\text{m}$	Description/Includes
<b>Autotrophs</b>			
Pico-phytoplankton	PIC	0.5 – 2	Pico-eukaryotes and non N <sub>2</sub> -fixing cyanobacteria such as <i>Synechococcus</i> and <i>Prochlorococcus</i>
N <sub>2</sub> -fixers	FIX	0.7 – 2	<i>Trichodesmium</i> and N <sub>2</sub> -fixing unicellular cyanobacteria
Coccolithophores	COC	5 – 10	
Mixed-phytoplankton	MIX	2 – 200	e.g. autotrophic dinoflagellates and chrysophytes
Diatoms	DIA	20 – 200	
<i>Phaeocystis</i>	PHA	120 – 360	Colonial <i>Phaeocystis</i>
<b>Heterotrophs</b>			
Bacteria	BAC	0.3 – 1	Here used to subsume both heterotrophic <i>Bacteria</i> and <i>Archaea</i>
Protozooplankton	PRO	5 – 200	e.g. heterotrophic flagellates and ciliates
Mesozooplankton	MES	200 – 2000	Predominantly copepods
Macrozooplankton	MAC	>2000	Euphausiids, amphipods, and others, known as crustacean macrozooplankton
Jellyfish zooplankton	JEL	200 – >20,000	Cnidaria medusae, ‘true jellyfish’

112  
 113 The PlankTOM11 marine biogeochemistry component is coupled online to the global ocean general circulation  
 114 model Nucleus for European Modeling of the Ocean version 3.5 (NEMO v3.5). We used the global configuration  
 115 with a horizontal resolution of 2° longitude by a mean resolution of 1.1° latitude using a tripolar orthogonal grid.  
 116 The vertical resolution is 10m for the top 100m, decreasing to a resolution of 500m at 5km depth, and a total of

117 30 vertical z-levels (Madec, 2013). The ocean is described as a fluid using the Navier-Stokes equations and a  
118 nonlinear equation of state (Madec, 2013). NEMO v3.5 explicitly calculates vertical mixing at all depths using a  
119 turbulent kinetic energy model and sub-grid eddy induced mixing. The model is interactively coupled to a  
120 thermodynamic sea-ice model (LIM version 2; Timmermann et al., 2005).

121 The temporal ( $t$ ) evolution of zooplankton concentration ( $Z_j$ ), including the jellyfish PFT, is described through  
122 the formulation of growth and loss rates as follows:

$$123 \frac{\partial Z_j}{\partial t} = \sum_k g_{F_k}^{Z_j} \times F_k \times MGE \times Z_j - \sum_{k=1}^4 g_{Z_j}^{Z_k} \times Z_k \times Z_j - R_{0^\circ}^{Z_j} \times d_{Z_j}^T \times Z_j \quad (1)$$

124 *growth through grazing – loss through grazing – basal respiration*

$$125 - m_{0^\circ}^{Z_j} \times c_{Z_j}^T \times \frac{Z_j}{K_{1/2}^{Z_j} + Z_j} \times \sum_i P_i$$

126 *– mortality*

127 For growth through grazing,  $g_{F_k}^{Z_j}$  is the grazing rate by zooplankton  $Z_j$  on food source  $F_k$ . This is a temperature-  
128 dependent Michaelis-Menten term that includes grazing preference (see Sect. 2.1.2.).  $MGE$  is the modelled growth  
129 efficiency (Buitenhuis et al., 2010). For loss through grazing,  $g_{Z_j}^{Z_k}$  is the grazing of other zooplankton on  $Z_j$ . For  
130 basal respiration,  $R_{0^\circ}^{Z_j}$  is the respiration rate at 0°C,  $T$  is temperature,  $d_{Z_j}$  is the temperature dependence of  
131 respiration ( $d^{10} = Q_{10}$ ). Mortality is the closure term of the model and is mostly due to predation by higher trophic  
132 levels than are represented by the model.  $m_{0^\circ}^{Z_j}$  is the mortality rate at 0°C,  $c_{Z_j}$  is the temperature dependence of  
133 the mortality ( $c^{10} = Q_{10}$ ) and  $K_{1/2}^{Z_j}$  is the half saturation constant for mortality.  $\sum_i P_i$  is the sum of all PFTs,  
134 excluding bacteria, and is used as a proxy for the biomass of predators not explicitly included in the model. More  
135 details on each term are provided below and parameter values are given in Tables 2 through 5.

136

### 137 2.1.1 PFT Growth

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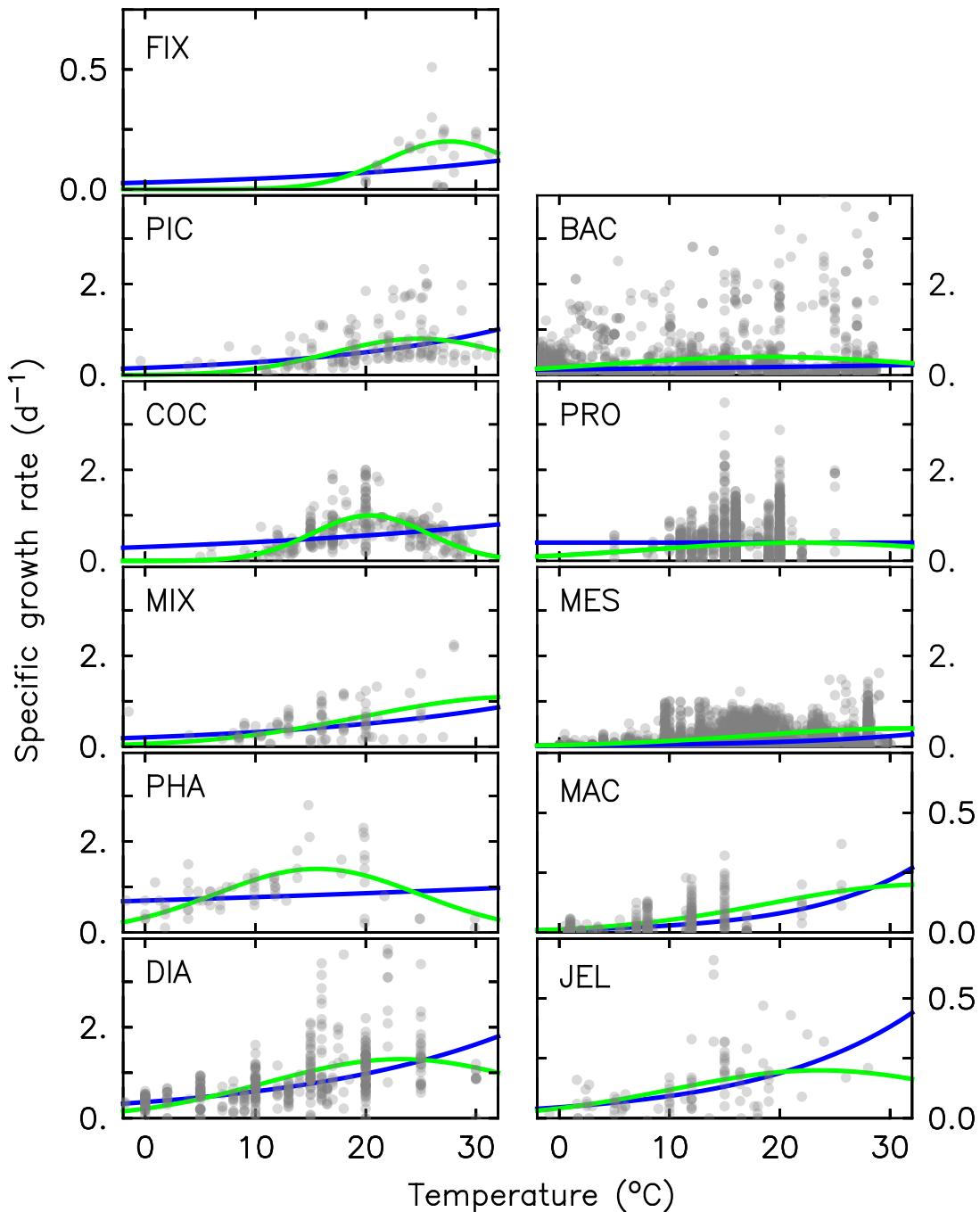
139 Growth rate is the trait that most distinguishes PFTs in models (Buitenhuis et al., 2006, 2013a). Jellyfish growth  
140 rates were compiled as a function of temperature from the literature (see Appendix Table A1). In previous  
141 published versions of the PlankTOM model, growth as a function of temperature ( $\mu^T$ ) was fitted with two  
142 parameters:

$$143 \mu^T = \mu_0 \times Q_{10}^{\frac{T}{10}} \quad (2)$$

144 where  $\mu_0$  is the growth at 0°C,  $Q_{10}$  is the temperature dependence of growth derived from observations, and  $T$  is  
145 the temperature (Le Quéré et al., 2016). Jellyfish growth rate is poorly captured by an exponential fit to  
146 temperature. To better capture the observations, the growth calculation has now been updated with a three-  
147 parameter growth rate, which produces a bell-shaped curve centred around an optimal growth rate at a given  
148 temperature (Fig. 2 and Table 2). The three-parameter fit is suitable for the global modelling of plankton because

149 it can represent an exponential increase if the data support this (Schoemann et al., 2005). The growth rate as a  
 150 function of temperature ( $\mu^T$ ) is now defined by; the optimal temperature ( $T_{opt}$ ), maximum growth rate ( $\mu_{max}$ ) at  
 151  $T_{opt}$ , and the temperature interval ( $dT$ ):

152 
$$\mu^T = \mu_{max} \times \exp \left[ \frac{-(T - T_{opt})^2}{dT^2} \right] \quad (3)$$



153

154 *Figure 2. Maximum growth rates for the 11 PFTs as a function of temperature from observations (grey circles). The three-*  
 155 *parameter fit to the data is shown in green and the two-parameter fit is shown in blue, using the parameter values from*  
 156 *Table 2. For full PFT names see Table 1. The R<sup>2</sup> for both fits to the data are given in Appendix Table A2.*

**Table 2.** Parameters used to calculate PFT specific growth rate with three-parameter fit (Eq. 3) in PlankTOM11.

PFT	$\mu_{\max}$ (d <sup>-1</sup> )	T <sub>opt</sub> (°C)	dT (°C)
FIX	0.2	27.6	8.2
PIC	0.8	24.8	11.2
COC	1.0	20.4	7.4
MIX	1.1	34.0	20.0
PHA	1.4	15.6	13.0
DIA	1.3	23.2	17.2
BAC	0.4	18.8	20.0
PRO	0.4	22.0	20.0
MES	0.4	31.6	20.0
MAC	0.2	33.2	20.0
JEL	0.2	23.6	18.8

157

158 The available observations measure growth rate, but the model requires specification of the grazing rate (Eq. 1).

159 Growth of zooplankton and grazing ( $g^T$ ) are related through the gross growth efficiency (GGE):

160 
$$g^T = \frac{u^T}{GGE} \quad (4)$$

161 GGE is the portion of grazing that is converted to biomass. This was previously collated by Moriarty (2009) from  
 162 the literature for crustacean and gelatinous macrozooplankton for the development of PlankTOM10. We extracted  
 163 the data for jellyfish from this collation (all scyphomedusae) which gave an average GGE of  $0.29 \pm 0.27$ , n=126  
 164 (Moriarty, 2009).

165

166 **2.1.2 Jellyfish PFT Grazing**

167

168 The food web, and thus the trophic level of PFTs is determined through grazing preferences. The relative  
 169 preference of jellyfish zooplankton for the other PFTs was determined through a literature search (Colin et al.,  
 170 2005; Costello and Colin, 2002; Flynn and Gibbons, 2007; Malej et al., 2007; Purcell, 1992, 1997, 2003; Stoecker

171 et al., 1987; Uye and Shimauchi, 2005a; see Appendix Table A3 for further detail). The dominant food source  
 172 was mesozooplankton (specifically copepods), followed by proto-zooplankton (most often ciliates) and then  
 173 macrozooplankton (Table 3). There is little evidence in the literature for jellyfish actively consuming autotrophs.  
 174 One of the few pieces of evidence is a gut content analysis where ‘unidentified protists... some chlorophyll  
 175 bearing’ were found in a small medusa species (Colin et al., 2005). Another is a study by Boero et al. (2007)  
 176 which showed that very small medusae such as *Obelia* will consume bacteria and may consume phytoplankton.  
 177 Studies on the diet of the ephyrae life cycle stage are limited in comparison to those on medusa, but the literature  
 178 does show evidence for ephyrae consuming protists and phytoplankton (Båmstedt et al., 2001; Morais et al., 2015).  
 179 We assume that ephyrae are likely to have a higher preference for autotrophs, due to their smaller size as with the  
 180 small medusa, but that this will have a minimal effect on the overall preferences and the biomass consumed, so  
 181 preferences for autotrophs are kept low. Once the relative preference is established, the absolute value of the  
 182 preference is tuned to improve the biomass of the different PFTs, as in Le Quéré et al. (2016). Table 3 shows the  
 183 relative preference of jellyfish for its prey assigned in the model, along with the preferences of the other  
 184 zooplankton PFTs. The zooplankton relative preferences are based around a predator-prey size ratio, which by  
 185 design is set to 1 for zooplankton-diatom. Preferences to other PFTs and to particulate carbon are then set relative  
 186 to the preference for diatoms. The preference ratios are weighted using the global carbon biomass for each type  
 187 against a total food biomass weighted mean (sum of all the PFTs), calculated from the MAREDAT database,  
 188 following the methodology used for the other PFTs (Buitenhuis et al., 2013a; Le Quéré et al., 2016). Zooplankton  
 189 grazing is calculated using:

$$190 \quad g_{F_k}^{Z_j} = \mu^T \frac{p_{F_k}^{Z_j}}{K_{1/2}^{Z_j} + \sum p_{F_k}^{Z_j} F_k} \quad (5)$$

191 where  $g_{F_k}^{Z_j}$  is the grazing rate by zooplankton  $Z_j$  on food source  $F_k$  as shown in Eq. 1, where  $\mu^T$  is the growth rate  
 192 of zooplankton (Eq. 3),  $p_{F_k}^{Z_j}$  is the preference of the zooplankton for the food source (prey) and  $K_{1/2}^{Z_j}$  is the half  
 193 saturation constant of zooplankton grazing. The parameter values for grazing used in the model are given in Table  
 194 4.

195

### 196 2.1.3 Jellyfish PFT Respiration

197

198 Previous analysis of respiration rates of jellyfish found that temperature manipulation experiments with  $Q_{10}$  values  
 199 of  $>3$  were flawed because the temperature was changed too rapidly (Purcell, 2009; Purcell et al., 2010). In a  
 200 natural environment, jellyfish gradually acclimate to temperature changes which has a smaller effect on their  
 201 respiration rates. Purcell et al. (2010) instead collated values from experiments that measured respiration at  
 202 ambient temperatures, providing a range of temperature data across different studies. They found that  $Q_{10}$  for  
 203 respiration was 1.67 for *Aurelia* species (Purcell, 2009; Purcell et al., 2010). Moriarty (2009) collated a respiration  
 204 dataset for zooplankton, including gelatinous zooplankton, using a similar selectivity as Purcell et al. (2010) for  
 205 experimental temperature, feeding, time in captivity and activity levels. Jellyfish were extracted from the Moriarty

**Table 3.** Relative preference, expressed as a ratio, of zooplankton for food (grazing) used in PlankTOM11. For each zooplankton the preference ratio for diatoms is set to 1.

PFT	PRO	MES	MAC	JEL
<b>Autotrophs</b>				
FIX	2	0.1	0.1	0.1
PIC	3	0.75	0.5	0.1
COC	2	0.75	1	0.1
MIX	2	0.75	1	1
DIA	1	1	1	1
PHA	2	1	1	1
<b>Heterotrophs</b>				
BAC	4	0.1	0.1	0.1
PRO	0	2	1	7.5
MES	0	0	2	10
MAC	0	0	0	5
JEL	0	0	0.5	0
<b>Particulate matter</b>				
Small organic particles	0.1	0.1	0.1	0.1
Large organic particles	0.1	0.1	0.1	0.1

206

207 (2009) dataset, which also included experiments on non-adult and non-*Aurelia* species medusae, unlike the Purcell  
208 et al. (2010) dataset. The relationship between temperature and respiration is heavily skewed by body mass  
209 (Purcell et al., 2010). The data were thus normalised by fitting to a general linear model (GLM) using a least  
210 squares cost function, to reduce the effect of body mass on respiration rates (Ikeda, 1985; Le Quéré et al., 2016).

211  $GLM = \log_{10}RR = a + b \log_{10}BM + c T$  (6)  
212

213  $cost\ function = \sum \left( \frac{R_{GLM}^T - R_{obs}^T}{R_{obs}^T} \right)^2$  (7)

214 Where  $RR$  is the respiration rate,  $BM$  is the body mass, and  $T$  and  $R^T$  are the observed temperature and associated  
 215 respiration rate. The parameters values were then calculated using  $R_0 = e^a$ , and  $Q_{10} = (e^c)^{10}$ , where  $e$  is the  
 216 exponential function. The resulting fit to data is shown in Fig. 3. The parameter values for respiration used in the  
 217 model are given in Table 4. Macrozooplankton respiration values are also given in Fig. 3 and Table 4, to provide  
 218 a comparison to another zooplankton PFT of the most similar size available.

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**Table 4.** PlankTOM11 parameter values for macrozooplankton and jellyfish, with the associated equation.

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Parameters	JEL	MAC	Equation
Respiration			
$R_{0^\circ}^{Z_j}$ (d <sup>-1</sup> )	0.03	0.01	Eq. 1
$d_{Z_j}$	1.88	2.46	Eq. 1
Mortality			
$m_{0^\circ}^{Z_j}$ (d <sup>-1</sup> )	0.12	0.02	Eq. 1
$c_{Z_j}$	1.20	3.00	Eq. 1
$K^{Z_j}$ (μmol C L <sup>-1</sup> )	20.0e-6	20.0e-6	Eq. 1
GGE	0.29	0.30	Eq. 4
Grazing half saturation constant $K_{1/2}^{Z_j}$ (μmol C L <sup>-1</sup> )	10.0e-6	9.0e-6	Eq. 5

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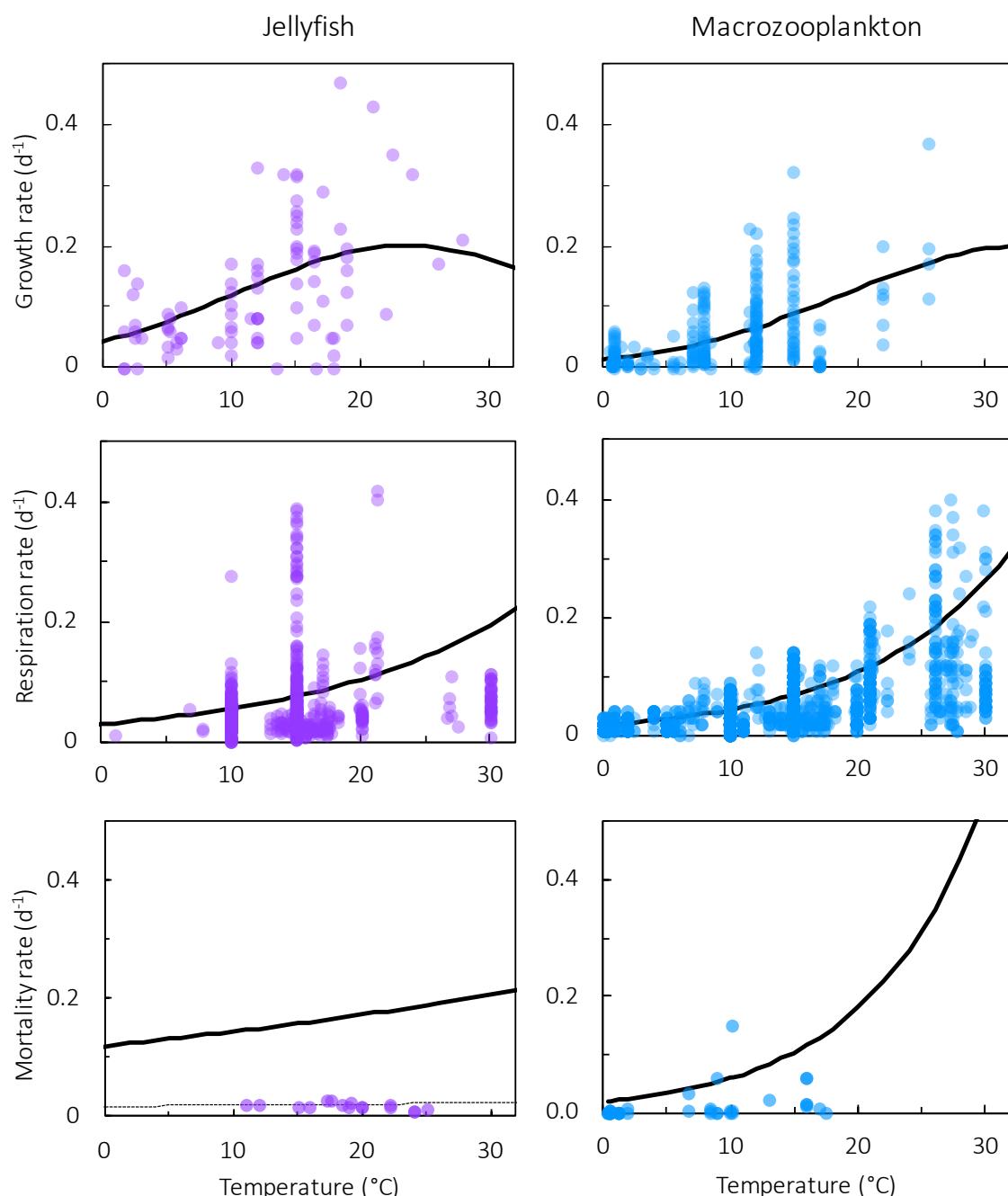
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220 **2.1.4 Jellyfish PFT Mortality**

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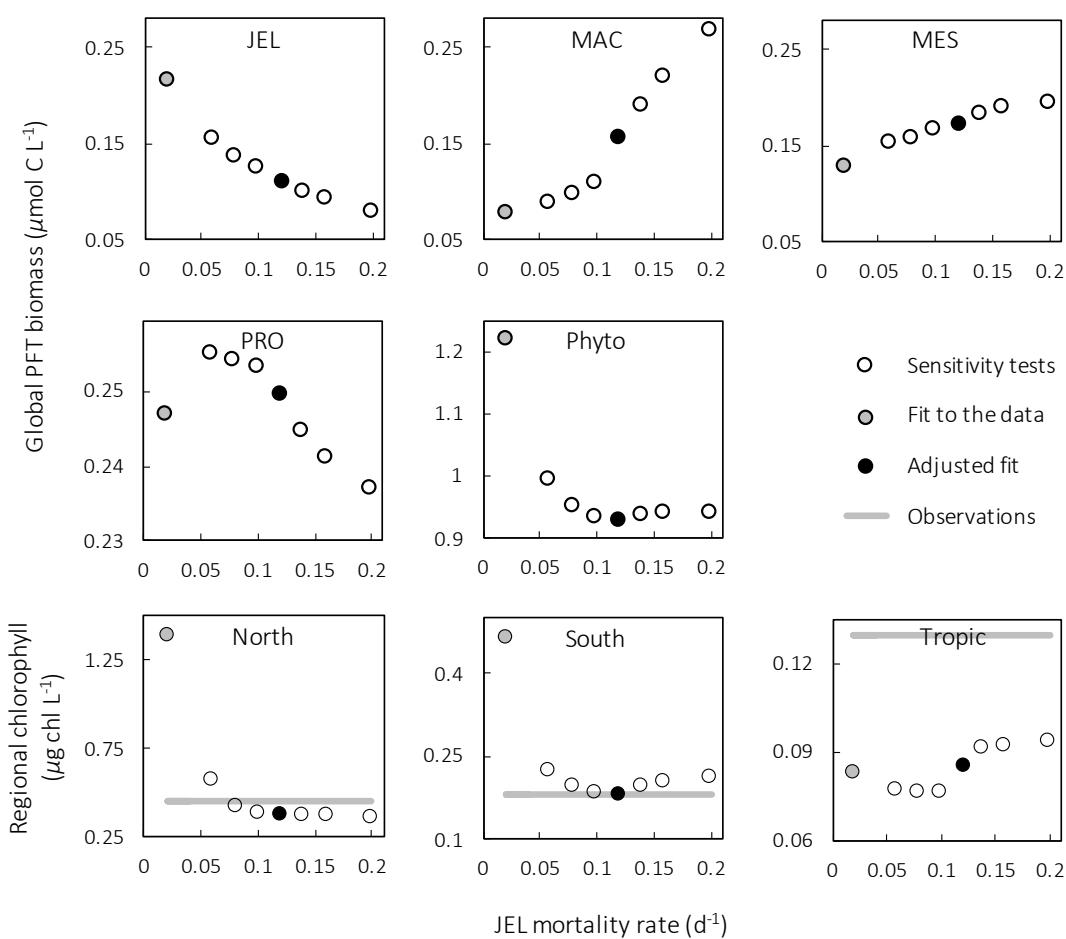
222 There is limited data on mortality rates for jellyfish and to use mortality data from the literature on any  
 223 zooplankton group some assumptions must be made (Acevedo et al., 2013; Almeda et al., 2013; Malej and Malej,  
 224 1992; Moriarty, 2009; Rosa et al., 2013). These assumptions are: that the population is in a steady state where  
 225 mortality equals recruitment, reproduction is constant and that mortality is independent of age (Moriarty, 2009).  
 226 All models with zooplankton mortality rates follow these assumptions. In reality the mortality of a zooplankton  
 227 population is highly variable. Steady states are balanced over a long period (if a population remains viable),  
 228 reproduction is restricted to certain times of year and the early stages of life cycles are many times more vulnerable

229 to mortality. Despite these assumptions, with the limited data on mortality rates, the larger uncertainty lies with  
 230 the data rather than the assumptions (Moriarty, 2009). The half saturation constant for mortality ( $K_{1/2}^{Z_j}$  in Eq. 1) is  
 231 set to  $20 \mu\text{mol C L}^{-1}$  the same as other zooplankton types, due to the lack of PFT specific data. In the small amount  
 232 of data available and suitable for use in the model (16 data points from two studies) mortality ranged from 0.006  
 233 – 0.026 per day (Acevedo et al., 2013; Malej and Malej, 1992). Applying the exponential fit to this data gave a  
 234 mortality rate at  $0^\circ\text{C}$  ( $m_0^{Z_j}$  in Eq. 1) of 0.018 per day. Sensitivity tests were carried out from this mortality rate  
 235 due to low confidence in the value.



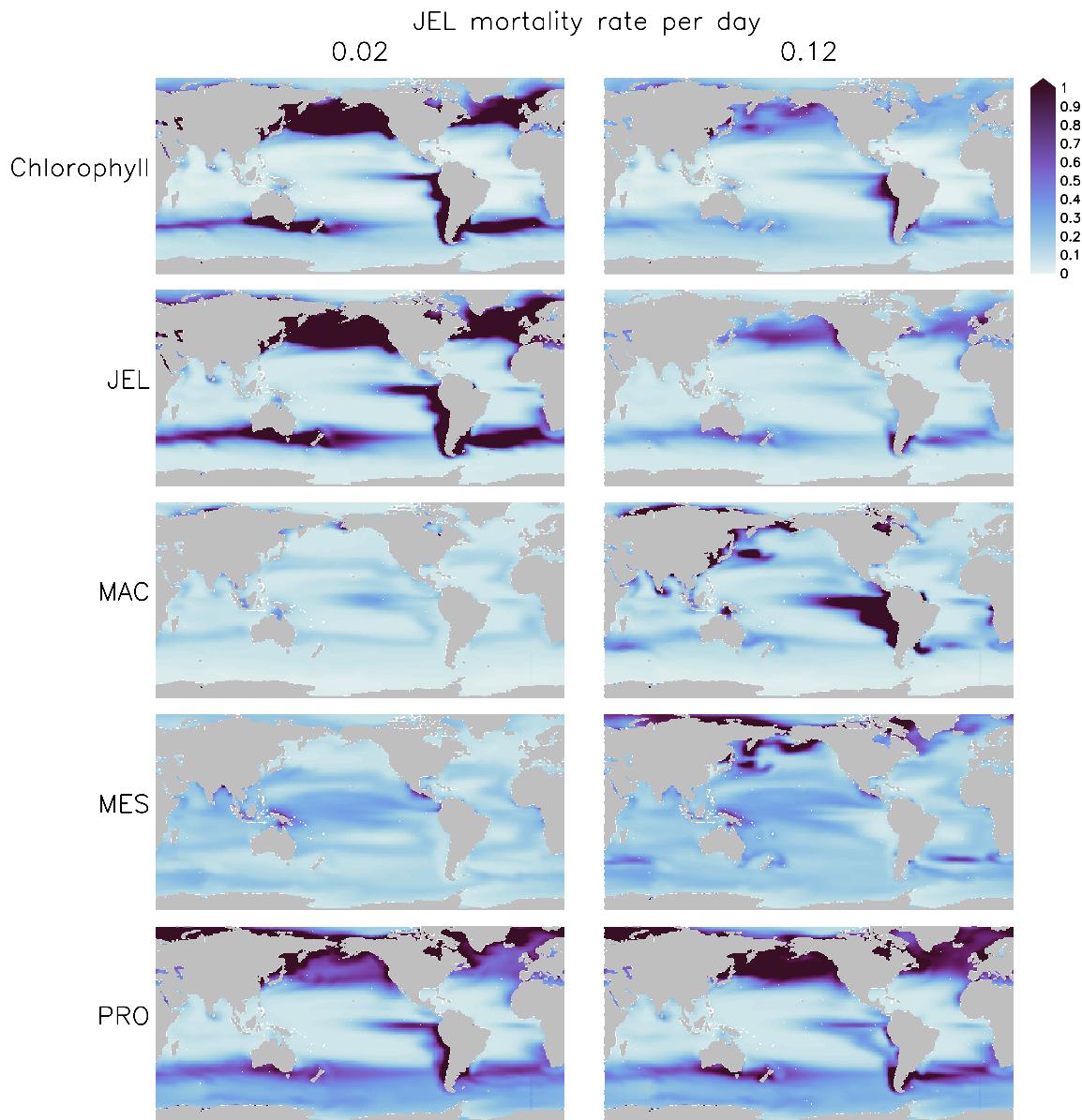
236  
 237 *Figure 3. Maximum growth rates (top), respiration rates (middle) and mortality rates (bottom) for jellyfish (left; purple)  
 238 and macrozooplankton (right; blue) PFTs as a function of temperature. The fit to the data is shown in black, using the parameter  
 239 values from Table 2 and Table 4. Growth rates are the same as shown in Fig. 2, on a different scale. For jellyfish mortality the  
 240 thin dashed line is the fit to data and the solid line is the adjusted fit (Table 4).*

241 Results from a subset of the sensitivity tests are shown in Fig. 4. The model was found to best represent a range  
 242 of observations when jellyfish mortality was increased to 0.12 per day. The fit to mortality for the data ( $\mu_0 =$   
 243 0.018) and the adjusted mortality ( $\mu_0 = 0.12$ ) is shown in Fig. 3. This value was chosen based on expert judgement  
 244 of the overall fit across multiple data streams. Whereas it was informed by the quantitative values in Table 6, the  
 245 final choice required the balance of positive and negative performance that required expert judgement rather than  
 246 a statistical number. Mortality rate values closer to 0.018 per day allowed jellyfish to dominate macro- and  
 247 mesozooplankton, greatly reducing their biomass (Fig. 4 and Fig. 5). Low jellyfish mortality also resulted in  
 248 higher chlorophyll concentrations than observed, especially in the high latitudes (Fig. 4 and Fig. 5; Bar-On et al.,  
 249 2018; Buitenhuis et al., 2013b). The adjusted mortality rate used for PlankTOM11 may be accounting for several  
 250 components missing from experimental data including the impact of higher trophic level grazing in the Avecedo  
 251 et al. (2013) study, which in copepods is 3-4 times higher than other sources of mortality (Hirst and Kiørboe,  
 252 2002), the greater vulnerability to mortality experienced during the early stages of the life cycle and mortality due  
 253 to parasites and viruses, especially during blooms (Pitt et al., 2014).



254

255 *Figure 4. Results from sensitivity tests on jellyfish mortality rates. The adjusted fit simulation used for PlankTOM11 is shown*  
 256 *by the black filled circle and the fit to the data simulation is shown by the grey filled circle; global mean PFT biomass ( $\mu\text{mol}$*   
 257  *$\text{C L}^{-1}$ ) for 0-200m depth (top - middle), regional mean surface chlorophyll concentration ( $\mu\text{g chl L}^{-1}$ ; bottom). For the regional*  
 258 *mean chlorophyll the observations are calculated from SeaWiFS. All data are averaged for 1985-2015, and between 30°*  
 259 *and 55° latitude in both hemispheres: 140-240°E in the north and 140-290°E in the south (see Fig. 8). Phyto is the sum of all*  
 260 *the phytoplankton PFTs.*



261

**Figure 5.** Annual mean surface chlorophyll ( $\mu\text{g chl L}^{-1}$ ) and zooplankton carbon biomasses ( $\mu\text{mol C L}^{-1}$ ) of JEL, MAC, MES and PRO for adjustment of JEL mortality for the simulation with 0.02 mortality/day $^{-1}$  (left) and the adjusted fit simulation with 0.12 mortality/day $^{-1}$  (right) used in PlankTOM11. Results are shown for the surface box (0-10 meters) and averaged for 1985-2015.

266 PlankTOM11 uses a mortality rate for jellyfish that is much higher than the limited observations (Fig. 4 and Fig.  
267 5). Lower jellyfish mortality is likely to be more representative of adult life stages, as jellyfish experience high  
268 mortality during juvenile life stages, especially as planula larvae and during settling (Lucas et al., 2012). The  
269 limited observations of jellyfish mortality are from mostly adult organisms, which may explain the dominance of  
270 jellyfish in the model when parameterised with the observed mortality fit. The higher mortality used for this study  
271 may be more representative of an average across all life stages. Experimental jellyfish mortality is also likely to  
272 be lower than *in situ* mortality due to factors such as senescence post-spawning and bloom conditions increasing  
273 the prevalence of disease and parasites and thus increasing mortality (Mills, 1993; Pitt et al., 2014). Using a higher  
274 mortality for this study is therefore deemed reasonable.

275

276 **2.1.5 Organic Carbon Cycling Through the Plankton Ecosystem**

277

278 In PlankTOM11, the growth of phytoplankton modifies dissolved inorganic carbon into DOC, which then  
279 aggregates into POC<sub>S</sub> and POC<sub>L</sub> (Fig. 1b). POC<sub>S</sub> is also generated from protozooplankton egestion and excretion  
280 and is consumed through grazing by all zooplankton. POC<sub>L</sub> is also generated by aggregation from POC<sub>S</sub>, egestion  
281 and excretion by all zooplankton, and from the mortality of mesozooplankton, macrozooplankton and jellyfish,  
282 and is consumed through grazing by all zooplankton. The portion of POC<sub>S</sub> and POC<sub>L</sub> which is not grazed, sinks  
283 through the water column and is counted as export production at 100m (Fig. 1b). The sinking speed of POC<sub>S</sub> is 3  
284 m/d<sup>-1</sup> and the sinking speed of POC<sub>L</sub> varies, depending on the concentration of ballast and the resulting particle  
285 density. Proto-, meso- and macrozooplankton excretion is largely in the form of particulate and solid faecal pellets,  
286 while this makes up very little of jellyfish excretion. Jellyfish instead produce and slough off mucus as part of  
287 their feeding mechanism (Pitt et al., 2009), which is represented in the model in the same way as the faecal pellet  
288 excretion, as a fraction of unassimilated grazing contributing to POC<sub>L</sub>.

289

290 **2.1.6 Additional Tuning**

291

292 Following the change to the growth rate formulation (from Eq. 2 to Eq. 3), all PFT growth rates are lower  
293 compared to the published version of PlankTOM10 (Le Quéré et al., 2016), but the change is largest for  
294 *Phaeocystis*, diatoms, bacteria and protozooplankton (Fig. 2). Further tuning is carried out to rebalance the total  
295 biomass among phytoplankton PFTs following the change in formulation. The tuning included increasing the  
296 grazing ratio preference of mesozooplankton for *Phaeocystis* and the grazing ratio preference of protozooplankton  
297 for picophytoplankton within the limits of observations. Tuning also included increasing the half saturation  
298 constant of the phytoplankton *Phaeocystis*, picophytoplankton and diatoms for iron. The tuning resulted in a  
299 reduction of *Phaeocystis* biomass and an increase in diatom biomass, without disrupting the rest of the ecosystem.  
300 Diatom respiration was also increased to reduce their biomass towards observations. Finally, bacterial biomass  
301 was increased closer to observations by reducing the half saturation constant of bacteria for dissolved organic  
302 carbon and reducing the maximum bacteria uptake rate. See Appendix Table A4 for the parameter values before  
303 and after tuning.

304 As shown in Eq. 1, there is a component in the mortality of zooplankton to represent predation by organisms not  
305 included in the model. The jellyfish PFT is a significant grazer of macrozooplankton and mesozooplankton (Table  
306 3), to account for this additional grazing the mortality term for macrozooplankton and the respiration term for  
307 mesozooplankton were reduced compared to model versions where no jellyfish are present (Table 5). Respiration  
308 is reduced in place of mortality for mesozooplankton as their mortality term had already been reduced to zero to  
309 account for predation by macrozooplankton (Le Quéré et al., 2016). The jellyfish PFT is also a significant grazer  
310 of protozooplankton, however, following the adjustment of protozooplankton grazing on picophytoplankton to

311 account for changes to the growth rate formulation and the low sensitivity of protozooplankton to jellyfish  
312 mortality (Fig. 4) additional changes to protozooplankton parameters were found to be unnecessary.

---

**Table 5.** Changes to non-jellyfish PFT parameters across the PlankTOM simulations. PlankTOM10<sup>LQ16</sup> is the latest published version of PlankTOM with 10 PFTs (Le Quéré et al., 2016), while PlankTOM10 is the simulation from this study.

Parameters	PlankTOM10 <sup>LQ16</sup>	PlankTOM10	PlankTOM10.5	PlankTOM11
MAC mortality	0.020	0.012	0.005	0.005
MES respiration	0.014	0.014	0.001	0.001

313

### 314 2.1.7 Model Simulations

315

316 The PlankTOM11 simulations are run from 1920 to 2015, forced by meteorological data including daily wind  
317 stress, cloud cover, precipitation and freshwater riverine input from NCEP/NCAR reanalysed fields (Kalnay et  
318 al., 1996). The simulations start with a 28-year spin for 1920-1948 where the meteorological conditions for year  
319 1980 are used, looping over a single year. Year 1980 is used as a typical average year, as it has no strong El  
320 Nino/La Nina, as in Le Quéré et al. (2010). Furthermore, because of the greater availability of weather data  
321 (including by satellite) in 1980 compared to 1948, the dynamical fields are generally more representative of small-  
322 scale structures than the earlier years. There is a small shock to the system at the start of meteorological forcing,  
323 but this stabilises within a few years and decades before the model output is used for analysis. Tests of different  
324 spin-up years were carried out in Le Quéré et al. (2010), including both 1948 and 1980, with little impact on trends  
325 generally. The spin up is followed by interannually varying forcing for actual years from 1948-2015. All analysis  
326 is carried out on the average of the last 31-year period of 1985-2015. PlankTOM11 is initialised with observations  
327 of dissolved inorganic carbon (DIC) and alkalinity (Key et al., 2004) after removing the anthropogenic component  
328 for DIC (Le Quéré et al., 2010), NO<sub>3</sub>, PO<sub>4</sub>, SiO<sub>3</sub>, O<sub>2</sub>, temperature and salinity from the World Ocean Atlas  
329 (Antonov et al., 2010).

330 Two further model simulations were carried out in order to better understand the effect of adding the jellyfish  
331 PFT. The first simulation sets the jellyfish growth rate to 0, so that it replicates the model set up with 10 PFTs in  
332 Le Quéré et al. (2016), here called PlankTOM10<sup>LQ16</sup>, but it includes the updated growth formulation (Sect. 2.1.1)  
333 and additional tuning (Sect. 2.1.5). The simulation is labelled ‘PlankTOM10’ in the figures. This simulation is  
334 otherwise identical to PlankTOM11 except for the mortality term for macrozooplankton and the respiration term  
335 for mesozooplankton, which were initially returned to PlankTOM10<sup>LQ16</sup> values, to account for the lack of  
336 predation by jellyfish. Macrozooplankton mortality was then tuned down from the PlankTOM10<sup>LQ16</sup> value, from  
337 0.02 to 0.012, to account for the change to the growth calculation (Table 5). The second additional simulation is  
338 carried out to test the addition of an 11<sup>th</sup> PFT in comparison to the addition of jellyfish as the 11<sup>th</sup> PFT. This is  
339 done by parameterising the jellyfish PFT identically to the macrozooplankton PFT, so that there are 11 PFTs

340 active, with two identical macrozooplankton. This simulation is called PlankTOM10.5. Otherwise, these  
341 simulations were identical to PlankTOM11.

342

## 343 2.2 JELLYFISH BIOMASS OBSERVATIONS

344

345 MARine Ecosystem biomass DATa (MAREDAT) is a database of global ocean plankton abundance and biomass,  
346 harmonised to common units and is open source available online (Buitenhuis et al., 2013b). The MAREDAT  
347 database is designed to be used for the validation of global ocean biogeochemical models. MAREDAT contains  
348 global quantitative observations of jellyfish abundance and biomass as part of the generic macrozooplankton  
349 group (Moriarty et al., 2013). The jellyfish sub-set of data has not been analysed independently yet.

350 For this study, all MAREDAT records under the group Cnidaria medusae ('true' jellyfish) were extracted from  
351 the macrozooplankton group (Moriarty et al., 2013) and examined. The taxonomic level within the database varies  
352 from phylum down to species. The data covers the period from August 1930 to August 2008 and contains  
353 abundance (individuals/m<sup>3</sup>, n=107,156) and carbon biomass (µg carbon L<sup>-1</sup>, n=3,406). The carbon biomass data  
354 are used over the abundance data despite the fewer data available, as they can be directly compared to  
355 PlankTOM11 results. Carbon biomass is calculated from wet weight/dry weight conversion factors for species  
356 where data records are sufficient (Moriarty et al., 2013). The data were collected at depth ranging from 0 to 2442m.  
357 The majority of the data (97%) were collected in the top 200m with an average depth of 44m (± 32m). Data from  
358 the top 200m are included in the analysis. The original un-gridded biomass data were binned into 1°x1° degree  
359 boxes at monthly resolution, as in Moriarty et al. (2013), reducing the number of gridded biomass data points to  
360 849.

361 In MAREDAT, jellyfish biomass data are only present in the Northern Hemisphere, which is likely to skew the  
362 data. Another caveat to the data is that a substantially smaller frequency of zeros is reported for biomass than for  
363 abundance. Under-reporting of zero values will increase the average, regardless of the averaging method used.  
364 Biomass observations from other global studies (Bar-On et al., 2018; Lucas et al., 2014; Luo et al., 2020) are used  
365 conjunctly with the global jellyfish biomass calculated here because of the poor spatial coverage.

366 To compare to the other PFTs within the MAREDAT database, global jellyfish biomass was calculated according  
367 to the methods in Buitenhuis et al. (2013b). Buitenhuis et al. (2013b) calculate a biomass range, using the median  
368 as the minimum and the arithmetic mean (AM) as the maximum. The jellyfish zooplankton biomass range in  
369 MAREDAT was calculated as 0.46 – 3.11 PgC, with the median jellyfish biomass almost as high as the  
370 microzooplankton and higher than meso- and macrozooplankton (Buitenhuis et al., 2013b). The jellyfish biomass  
371 range calculated here is used to validate the new jellyfish component in the PlankTOM11 model.

## 372 3 RESULTS

### 373 3.1 JELLYFISH BIOMASS

374

375 The global jellyfish biomass estimated by various studies gives a range of results: 0.1 PgC (Bar-On et al., 2018),  
376  $0.32 \pm 0.49$  PgC (Lucas et al., 2014),  $0.29 \pm 0.56$  PgC (Luo et al., 2020, updated from Lucas et al.) and 0.46 –  
377 3.11 PgC calculated in this study (Sect. 2.2). Jellyfish biomass in PlankTOM11 is within the range but towards  
378 the lower end of observations at 0.13 PgC, with jellyfish accounting for 16% of the total zooplankton biomass  
379 (Table 6). When the modelled biomass was tuned to match the higher observed biomass by adjusting the mortality  
380 rate, jellyfish dominate the entire ecosystem significantly reducing levels of the other zooplankton and increasing  
381 chlorophyll above observations for the Northern and Southern Hemispheres (Fig. 4 and Fig. 5).

382 PlankTOM11 generally replicates the patterns of jellyfish biomass with observations. High biomass occurs at  
383 around 50–60°N across the oceans, with the highest biomass in the North Pacific. PlankTOM11 also replicates  
384 low biomass in the Indian Ocean, and the eastern half of the tropical Pacific shows higher biomass than other  
385 open ocean areas in agreement with patterns in observations (Fig. 6; Lucas et al., 2014; Luo et al., 2020). However,  
386 PlankTOM11 underestimates the high jellyfish biomass in the tropical Pacific (Fig. 6). Most of the data informing  
387 the jellyfish parameters is from temperate species, so the model will better represent higher latitudes than lower  
388 latitudes. This is likely responsible for some of the underestimation of biomass in this region. The competition of  
389 jellyfish with macrozooplankton also plays a role (see Sect. 3.3 for further discussion). The lack of biomass  
390 observations around 40°S makes it difficult to determine if the peak in jellyfish biomass in PlankTOM11 at this  
391 latitude is representative of reality. The maximum biomass in the southern hemisphere is mostly around coastal  
392 areas i.e. South America and southern Australia (Fig. 6). This is expected from reports and papers on jellyfish in  
393 these areas (Condon et al., 2013; Purcell et al., 2007 and references therein). A prevalence of jellyfish in coastal  
394 areas is apparent (Fig. 6), in line with observations (Lucas et al., 2014; Luo et al., 2020), even without any specific  
395 coastal advantages for jellyfish in the model (see macrozooplankton in Le Quéré et al., 2016).  
396 However, PlankTOM11 underestimates the range of observations in the top 200m (Fig. 6). PlankTOM11  
397 overestimates the minimum values and underestimates the maximum values. However, part of this discrepancy  
398 may be due to under-sampling in the observations. A key caveat in jellyfish data is that the data is not uniformly  
399 distributed spatially or temporally and not proportionally distributed between various biomes of the ocean, with  
400 collection efforts skewed to coastal regions and the Northern Hemisphere (MAREDAT; Lilley et al., 2011; Lucas  
401 et al., 2014; Luo et al., 2020). This sampling bias and sampling methods also tend to favour larger, less delicate  
402 species, which are often scyphomedusae with a meroplanktonic life cycle.

403 Jellyfish are characterised by their bloom and bust dynamic, resulting in patchy and ephemeral biomass. The  
404 mean:max biomass ratio of observations (MAREDAT) was compared to the same ratio for PlankTOM11 to assess  
405 the replication of this characteristic. The observations give a wide range of ratios depending on the type of mean  
406 used. The PlankTOM11 ratio falls within this range, but towards the lower end (Table 7). PlankTOM11 replicates  
407 some of the patchy and ephemeral biomass of jellyfish.

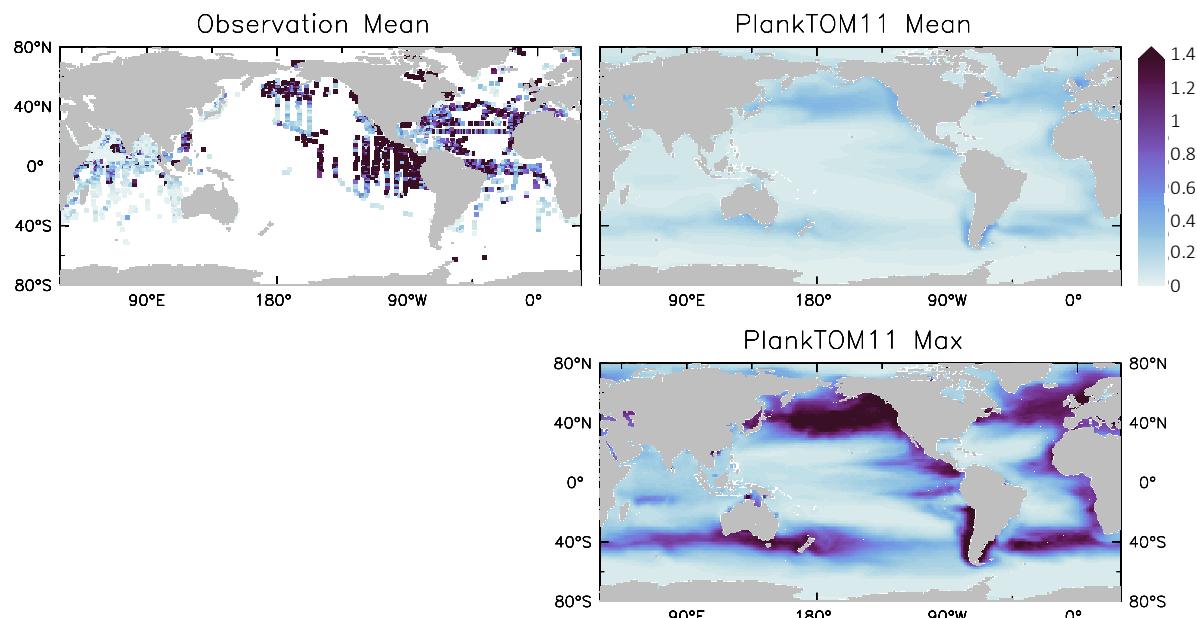
408 Jellyfish biomass in MAREDAT has poor global spatial coverage. The region around the coast of Alaska has the  
409 highest density of observations and is used here to evaluate the mean, range and seasonality of the carbon biomass  
410 of jellyfish as represented in PlankTOM11. The gridded jellyfish observations from Luo et al., (2020; see Fig. 6)  
411 are available as a mean over time and depth, so cannot be used to evaluate range or seasonality. Spatially, the  
412 observations peak around the north coast of Alaska while PlankTOM11 peaks around the south coast (Fig. 7).  
413 This difference is likely due to the lack of small-scale physical processes in the model due to the relatively coarse

**Table 6.** Global mean values for rates and biomass from observations and the PlankTOM11 and PlankTOM10 models averaged over 1985–2015. In parenthesis is the percentage share of the plankton type of the total phytoplankton or zooplankton biomass. The percentage share of mixed-phytoplankton is not included, as there are no mixed-phytoplankton observations, therefore, the phytoplankton percentages are of total phytoplankton minus mixed-phytoplankton. References for observations are given in Appendix Table A5.

	PlankTOM11	PlankTOM10	Observations
<b>Rates</b>			
Primary production (PgC y <sup>-1</sup> )	41.6	43.4	51-65
Export production at 100m (PgC y <sup>-1</sup> )	7.1	7.0	5-13
CaCO <sub>3</sub> export at 100m (PgC y <sup>-1</sup> )	1.3	1.2	0.6-1.1
N <sub>2</sub> fixation (TgN y <sup>-1</sup> )	97.2	95.9	60-200
<b>Phytoplankton biomass 0-200m (PgC)</b>			
N <sub>2</sub> -fixers	0.065 (8%)	0.075 (10%)	0.008-0.12 (2-8%)
Picophytoplankton	0.141 (17%)	0.153 (20%)	0.28-0.52 (35-68%)
Coccolithophores	0.248 (30%)	0.212 (27%)	0.001-0.032 (0.2-2%)
Mixed-phytoplankton	0.263	0.268	-
Phaeocystis	0.177 (22%)	0.170 (22%)	0.11-0.69 (27-46%)
Diatoms	0.183 (22%)	0.167 (21%)	0.013-0.75 (3-50%)
Total phytoplankton biomass	1.077	1.046	0.412 – 2.112
<b>Heterotrophs biomass 0-200m (PgC)</b>			
Bacteria	0.041	0.046	0.25-0.26
Protozooplankton	0.295 (36%)	0.330 (32.7%)	0.10-0.37 (27-31%)
Mesozooplankton	0.193 (23%)	0.218 (21.6%)	0.21-0.34 (25-66%)
Macrozooplankton	0.205 (25%)	0.460 (45.6%)	0.01-0.64 (3-47%)
Jellyfish zooplankton	0.129 (16%)	-	0.10-3.11
Total zooplankton biomass	0.823	1.008	0.42 – 4.46

414 model resolution. PlankTOM11 reproduces the observed mean jellyfish biomass around the coast of Alaska (0.16  
 415 compared to 0.13  $\mu\text{mol C L}^{-1}$ ), but it underestimates the maximum and spread of the observations (Table 8). The

416 spatial patchiness is somewhat replicated in PlankTOM11, although with a smaller variation (Fig. 7).  
 417 PlankTOM11 replicates the mean seasonal shape and biomass of jellyfish with a small peak over the summer  
 418 followed by a large peak in September in the observations and in October in PlankTOM11 (Fig. 7). Overall,  
 419 PlankTOM11 replicates the mean but underestimates the maximum biomass and temporal patchiness of the  
 420 observations (Fig. 7 and Table 8).



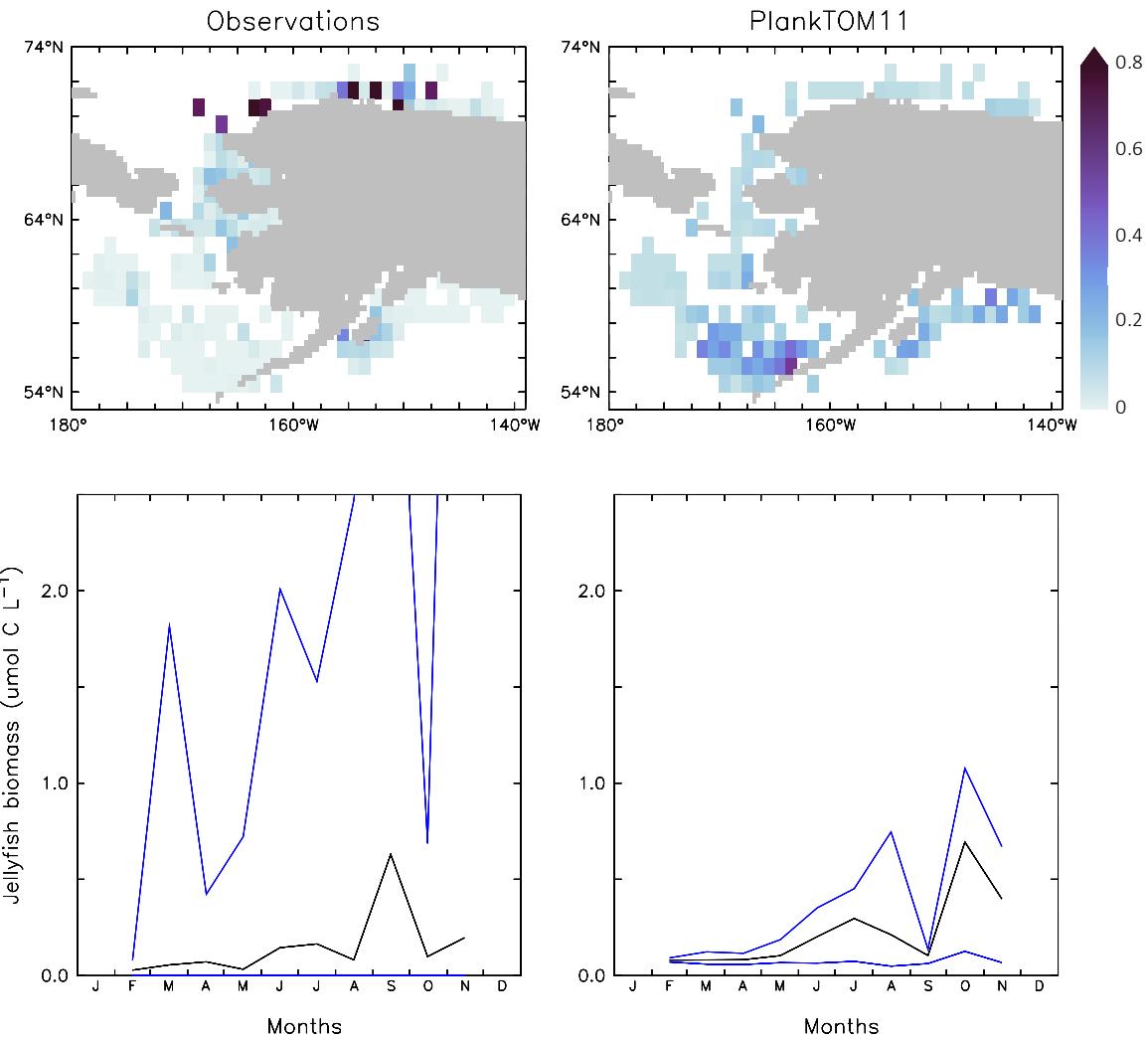
421

422 *Figure 6. Jellyfish carbon biomass ( $\mu\text{mol C L}^{-1}$ ) in PlankTOM11 and in observations from the Jellyfish Database Initiative (Luo  
 423 et al., 2020). PlankTOM11 results (left) are the mean and maximum biomass from monthly climatologies. Observations (right)  
 424 are the mean biomass, areas with no observations are in white. Observations are on a  $1 \times 1^\circ$  grid and are plotted using a  
 425 three-cell averaging filler for visual clarity. All data is for 0–200m. The gridded observation data is only available as a mean  
 426 over time and depth (Luo et al., 2020). Due to the patchy nature of the observations in depth and time, the mean may be  
 427 skewed high or low, while the model is sampled across the full time and depth.*

428

**Table 7.** Jellyfish biomass globally from observations (MAREDAT) and PlankTOM11. Three types of mean are given for the observations; Med is the median, AM is the arithmetic mean and GM is the geometric mean. The ratios are all scaled to mean = 1. All units are  $\mu\text{g C L}^{-1}$ .

		Mean	Max	Ratio
Observations	AM	3.61	156.0	1 : 43
	GM	0.95	156.0	1 : 165
	Med	0.29	156.0	1 : 538
PlankTOM11	AM	1.18	98.9	1 : 84



429

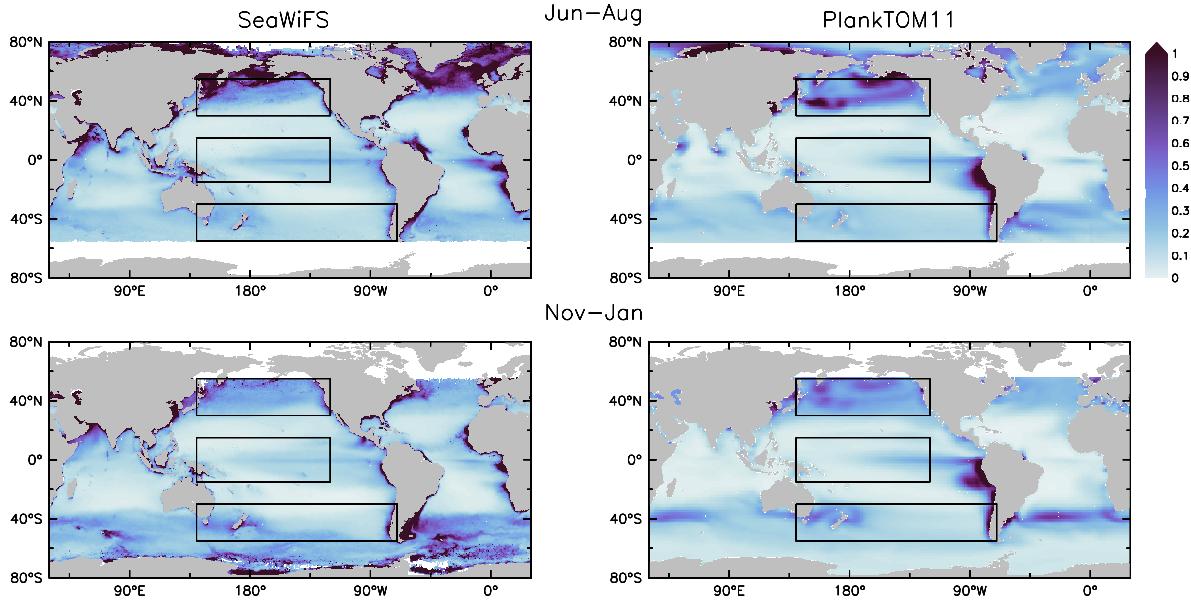
430 *Figure 7. Carbon biomass of jellyfish ( $\mu\text{mol C L}^{-1}$ ) from MAREDAT observations (left) and PlankTOM11 (right) for the coast of*  
 431 *Alaska (the region with the highest density of observations). The top panels show the mean jellyfish biomass and the bottom*  
 432 *panels show the seasonal jellyfish biomass, with the monthly mean in black and the monthly minimum and maximum in blue.*  
 433 *Observations and PlankTOM11 results are for 0-150m, as the depth range where >90% of the observations occur. No*  
 434 *observations were available for January or December.*

435

### 436 3.2 ECOSYSTEM PROPERTIES OF PLANKTOM11

437

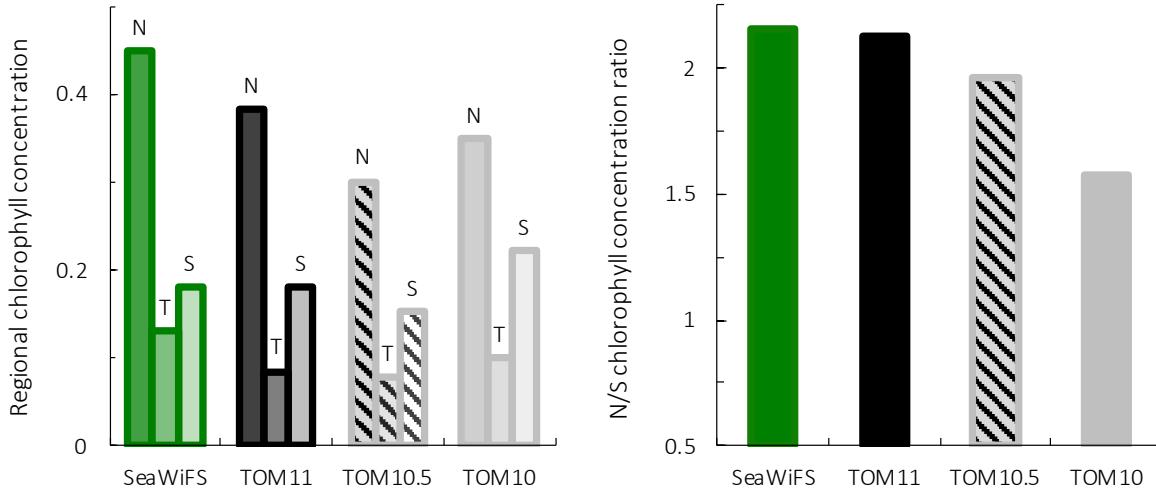
438 PlankTOM11 reproduces the main characteristics of surface chlorophyll observations, with high chlorophyll  
 439 concentration in the high latitudes, low concentration in the subtropics and elevated concentrations around the  
 440 equator (Fig. 8). PlankTOM11 also reproduces higher chlorophyll concentrations in the Northern Pacific than the  
 441 Southern (Fig. 9), and higher concentrations in the southern Atlantic than the southern Pacific Ocean (Fig. 8).  
 442 Overall the model underestimates chlorophyll concentrations, as is standard with models of this type (Le Quéré  
 443 et al., 2016) particularly in the central and northern Atlantic. PlankTOM11 also captures the seasonality of  
 444 chlorophyll, with concentrations increasing in summer compared to the winter for each hemisphere (Fig. 8).



445

446 **Figure 8.** Surface chlorophyll ( $\mu\text{g chl L}^{-1}$ ) averaged for June to August (top) and November to January (bottom). Panels show  
 447 observations from SeaWiFS (left) satellite and results from PlankTOM11 (right). Observations and model are averaged for  
 448 1997–2006. The black boxes show the Pacific north, tropic and south regions used in Fig. 4 and Fig. 9.

449



450

451 **Figure 9.** Surface chlorophyll for observations from SeaWiFS satellite, PlankTOM11, PlankTOM10.5 and PlankTOM10.  
 452 Regional chlorophyll concentration in  $\mu\text{g chl L}^{-1}$  (right) for the north (N), tropic (T) and south (S) Pacific Ocean regions shown  
 453 in Fig. 8 and the N/S chlorophyll concentration ratio (left). Observations and model are averaged for 1997–2006.

454

455 To assess the effect of adding jellyfish to PlankTOM, two additional simulations were conducted: PlankTOM10  
 456 where jellyfish growth is set to zero and PlankTOM10.5 where all jellyfish parameters are set equal to  
 457 macrozooplankton parameters (Sect. 2.1.6). The two simulations show similar spatial patterns of surface  
 458 chlorophyll to PlankTOM11, but different concentration levels. PlankTOM11 closely replicates the chlorophyll  
 459 ratio between the north and south Pacific with a ratio of 2.12, compared to the observed ratio of 2.16 (Fig. 9).

460 PlankTOM10 and PlankTOM10.5 underestimate the observed ratio with ratios of 1.57 and 1.96 respectively (Fig.  
461 9). Adding an 11<sup>th</sup> PFT improves the chlorophyll ratio, however, the regional chlorophyll concentrations for  
462 PlankTOM10.5 are a poorer match to the observations than PlankTOM11, especially in the north (Fig. 9).  
463 PlankTOM10 overestimates the observed chlorophyll concentration in the south (0.22 and 0.18 respectively; Fig.  
464 9). All three simulations underestimate chlorophyll concentration in the tropics compared to observations (Fig.  
465 9). The north/south chlorophyll ratio metric was developed by Le Quéré et al. (2016) as a simple method to  
466 quantify model performance for emergent properties, focussing on the Pacific Ocean as the area where this ratio  
467 is most pronounced in the observations. These simulations further support the suggestion by Le Quéré et al. (2016)  
468 that the observed distribution of chlorophyll in the north and south is a consequence of trophic balances between  
469 the PFTs and improves with increasing plankton complexity.

470 PlankTOM11 underestimates primary production by 10 PgC y<sup>-1</sup>, which is similar to the underestimation in  
471 PlankTOM10<sup>LQ16</sup> of 9 PgC y<sup>-1</sup>. As suggested by Le Quéré et al. (2016) this may be due to the model only  
472 representing highly active bacteria, which is unchanged between the model versions, while observed biomass is  
473 also from low activity bacteria and ghost cells. Export production and N<sub>2</sub> fixation are within the observational  
474 range, and CaCO<sub>3</sub> export is slightly overestimated (Table 6).

475 In PlankTOM11 each PFT shows unique spatial distribution in carbon biomass (Fig. 5). The total biomass of  
476 phytoplankton is within the range of observations, but the partitioning of this biomass between phytoplankton  
477 types differs from observations (Table 6). PlankTOM11 is dominated by mixed-phytoplankton and  
478 coccolithophores, together making up 47% of the total phytoplankton biomass. Diatoms and *Phaeocystis* are the  
479 next most abundant and fall within the observed range, followed by picophytoplankton with around half the  
480 observed biomass (Table 6). The observations are dominated by picophytoplankton, followed by *Phaeocystis* and  
481 Diatoms (Table 6). The modelled mixed-phytoplankton is likely taking up the ecosystem niche of  
482 picophytoplankton. Coccolithophores are overestimated by a factor of 10 and may also be filling the ecosystem  
483 niche of picophytoplankton in the model (Table 6). The phytoplankton community composition changed from  
484 PlankTOM10<sup>LQ16</sup> to PlankTOM11, with some phytoplankton types moving closer to observations and some  
485 moving further away. For example, for N<sub>2</sub>-fixers PlankTOM11 is in line with the upper end of observations at  
486 8%, while PlankTOM10 and PlankTOM10<sup>LQ16</sup> overestimate N<sub>2</sub>-fixers (10% and 11% respectively). For  
487 picophytoplankton, PlankTOM10<sup>LQ16</sup> is within the range of observations at 38%, while PlankTOM11 and  
488 PlankTOM10 underestimate the community share of picophytoplankton (17% and 20% respectively). For  
489 *Phaeocystis*, all three simulations underestimate the community share, but PlankTOM11 and PlankTOM10 (both  
490 22%) are closer to the lower end of observations (27%) than PlankTOM10<sup>LQ16</sup> (15%; Table 6; Le Quéré et al.,  
491 2016). Overall, the difference between PlankTOM10<sup>LQ16</sup> and PlankTOM11 is greater than the difference between  
492 PlankTOM10 and PlankTOM11, suggesting that the change to growth of PFT's had a larger effect on  
493 phytoplankton community composition than the addition of jellyfish. This is expected, as the growth change  
494 directly effects each PFT and model results are sensitive to PFT growth rates (Buitenhuis et al., 2006, 2010).  
495 Jellyfish affect phytoplankton community composition, but the effect is small.

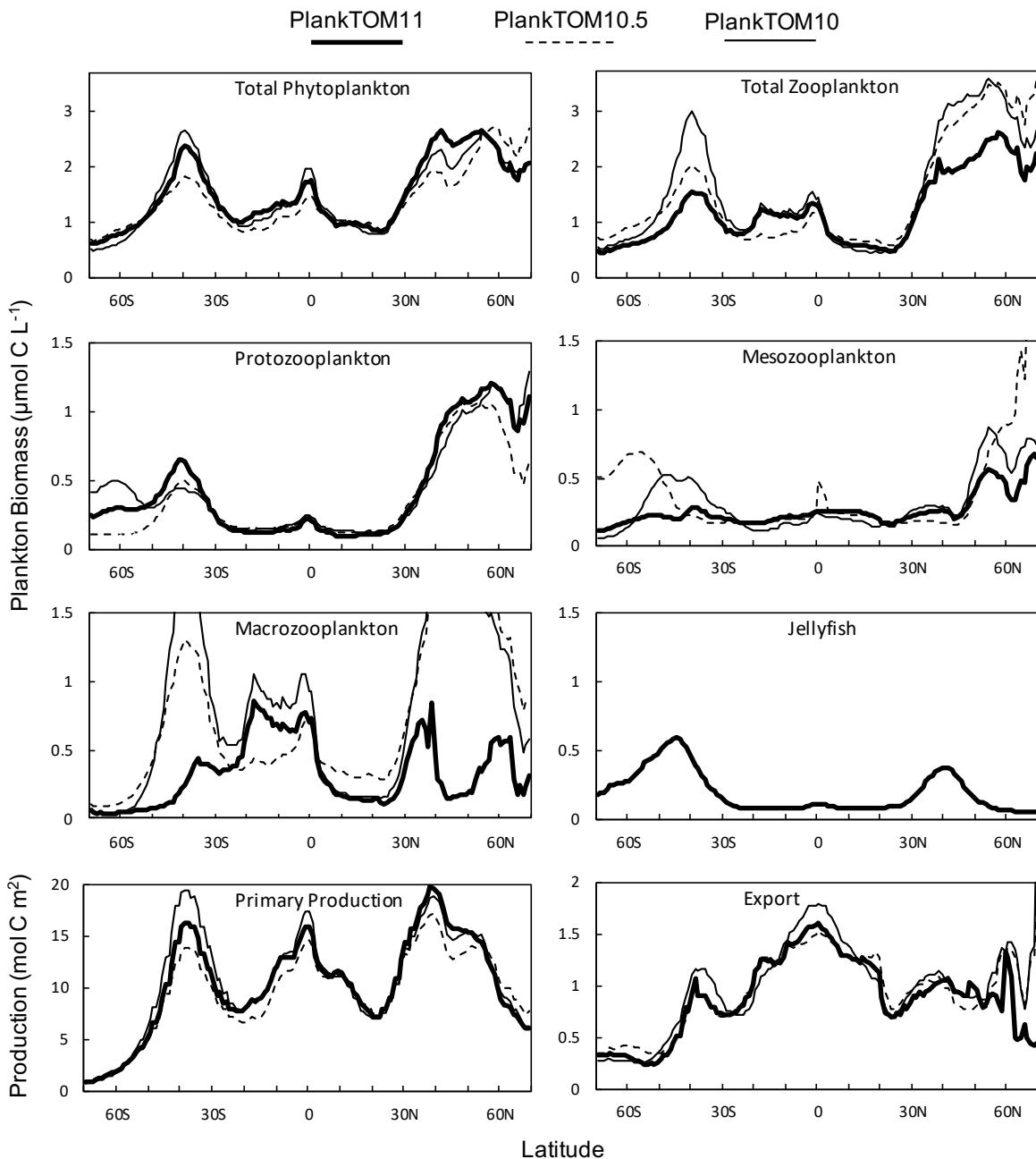
496

497 3.3 ROLE OF JELLYFISH IN THE PLANKTON ECOSYSTEM

498

499 Macrozooplankton exhibit the largest change in biomass between the three simulations, followed by  
500 mesozooplankton (Fig. 10). This is despite the higher preference of jellyfish grazing on mesozooplankton (ratio  
501 of 10) than on macrozooplankton (ratio of 5; Table 3). The central competition for resources between jellyfish  
502 and macrozooplankton is that they both preferentially graze on mesozooplankton, then on protozooplankton,  
503 although macrozooplankton have a lower preference ratio for zooplankton than jellyfish, as more of their diet is  
504 made up by phytoplankton (Table 3). In simple terms this means that for two equally sized populations of jellyfish  
505 and macrozooplankton, jellyfish would consume more meso- and protozooplankton than would be consumed by  
506 macrozooplankton. However, predator biomass, prey biomass and the temperature dependence of grazing interact  
507 to affect the rate of consumption (Eq. 5). The greatest difference in PFT biomass, especially macrozooplankton  
508 biomass, between simulations occurs in latitudes higher than 30° where jellyfish biomass is highest (Fig. 10). In  
509 the tropics, jellyfish have a low impact on the ecosystem due to their low biomass in this region (Fig. 6 and Fig.  
510 10).

511 The seasonality of the PFTs in each simulation is shown in Fig. 11 for 30-70° north and south, as the regions with  
512 the greatest differences between simulations (Fig. 10). In PlankTOM10 macrozooplankton represent the highest  
513 trophic level. The addition of another PFT at the same or at a higher trophic level (PlankTOM10.5 and  
514 PlankTOM11 respectively) reduces the biomass of the macrozooplankton, through a combination of competition  
515 and low-level predation (Fig. 10 and Fig. 11). For PlankTOM10.5 results, macrozooplankton is summed with the  
516 11<sup>th</sup> PFT (identical to macrozooplankton in this simulation). The addition of this 11th PFT at the same trophic  
517 level reduces the biomass of the macrozooplankton (Fig. 10 and Fig. 11), despite the macrozooplankton mortality  
518 being reduced from PlankTOM10 to PlankTOM10.5 (Table 5) which would be expected to increase  
519 macrozooplankton biomass. However, the low level of mutual predation between the two macrozooplankton PFTs  
520 slightly reduces their overall biomass. This reduction in biomass mostly occurs during the autumn  
521 macrozooplankton bloom, where the peak is reduced from PlankTOM10 to PlankTOM10.5, while the winter –  
522 spring biomass is similar across the two simulations (Fig. 11). The drop in mesozooplankton respiration from  
523 PlankTOM10 to PlankTOM10.5 (Table 5) lowers the rate of respiration, especially at lower temperatures. This  
524 likely accounts for the increase in PlankTOM10.5 mesozooplankton biomass at higher latitudes (Fig. 10). The  
525 addition of jellyfish changes the zooplankton with the highest biomass from macrozooplankton to  
526 protozooplankton and reduces the biomass of mesozooplankton, in both the north and south (Fig. 11). However,  
527 the impact on the biomass of mesozooplankton and protozooplankton is small, despite mesozooplankton being  
528 the preferential prey of jellyfish, followed by protozooplankton. The small impact of jellyfish on mesozooplankton  
529 and protozooplankton biomass may be due to trophic cascade effects where jellyfish reduce the biomass of  
530 macrozooplankton, which reduces the predation pressure of macrozooplankton on meso- and protozooplankton,  
531 whilst jellyfish simultaneously provide an additional predation pressure on meso- and protozooplankton. The  
532 decrease in predation by macrozooplankton may be compensated for by the increase in predation by jellyfish,  
533 resulting in only a small change to the overall biomass of mesozooplankton and protozooplankton.



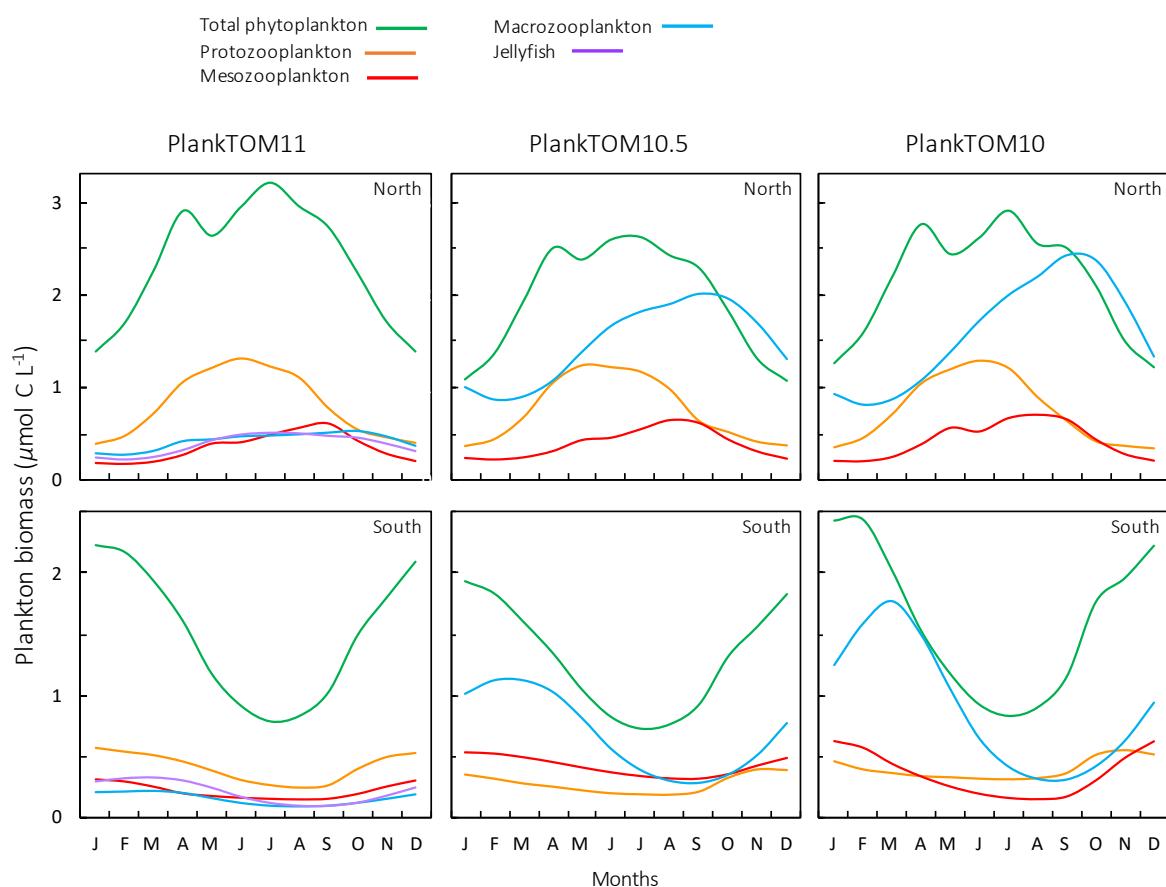
534

535 **Figure 10.** Zonal mean distribution for the PlankTOM11, PlankTOM10.5 and PlankTOM10 simulations. All plankton biomass  
 536 data are for the surface box (0-10m). For PlankTOM10.5 the MAC PFT has been summed with the 11<sup>th</sup> PFT that duplicates  
 537 MAC. The bottom panels are the zonal mean distribution of primary production, integrated over the top 100m, and export  
 538 production at 100m. All data are averaged for 1985-2015.

539 In PlankTOM11 there is a clear distinction between the biomass in the north and south, with higher biomass for  
 540 each PFT in the north compared to the south (Fig. 10 and Fig. 11). Plankton types have higher concentrations in  
 541 the respective hemisphere's summer, and a double peak in phytoplankton in the north (Fig. 11). PlankTOM10  
 542 also has a higher biomass of each PFT in the north compared to the south, but the difference is smaller than that  
 543 in PlankTOM11 (Fig. 10 and Fig. 11). The key difference between the two models is the biomass of  
 544 macrozooplankton. In PlankTOM10 macrozooplankton are the dominant zooplankton, especially in late summer  
 545 and autumn where their biomass matches and even exceeds the biomass of phytoplankton in the region (Fig. 11).  
 546 In PlankTOM11 neither macrozooplankton, nor any other zooplankton, come close to matching the biomass of

547 phytoplankton. The largest direct influence of jellyfish in these regions is its role in controlling macrozooplankton  
 548 biomass, through competition for prey resources, particularly mesozooplankton and protozooplankton, and  
 549 through the predation of jellyfish on macrozooplankton.

550 In PlankTOM11 in the north, phytoplankton display a double peak in seasonal biomass, with a smaller peak in  
 551 April of  $2.9 \mu\text{mol C L}^{-1}$ , followed by a larger peak in July of  $3.2 \mu\text{mol C L}^{-1}$  (Fig. 11). The addition of jellyfish  
 552 amplifies these peaks from PlankTOM10 and PlankTOM10.5 (Fig. 11) and from PlankTOM10 (Le Quéré et al.,  
 553 2016). Observations (MAREDAT) show two peaks in phytoplankton biomass although the peaks are offset in  
 554 timing from all three PlankTOM simulations. The amplitude of the full seasonal cycle in observations is  $0.78 - 2.67 \mu\text{mol C/L}$  (median – mean) with all three PlankTOM simulations falling well within this range (Table A6).  
 555 Removing the winter months, where there is less variability, gives a non-winter observational amplitude of  $0.7 - 2.12 \mu\text{mol C/L}$ . PlankTOM11 is the highest, with a non-winter amplitude of  $0.97 \mu\text{mol C/L}$ , with the other two  
 556 simulations lower at  $0.8 \mu\text{mol C/L}$  (PlankTOM10.5) and  $0.81 \mu\text{mol C/L}$  (PlankTOM10; Table A6).  
 557 PlankTOM10<sup>LQ16</sup> has a lower seasonal amplitude than PlankTOM11, although a slight higher non-winter  
 558 amplitude by  $0.05 \mu\text{mol C/L}$  (Table A6). The changes to phytoplankton seasonal biomass are not evenly  
 559 distributed across the PFT's, with coccolithophores and Phaeocystis exhibiting the largest changes (Fig. A1).  
 560



562

563 **Figure 11.** Seasonal surface carbon biomass ( $\mu\text{mol C L}^{-1}$ ) of total phytoplankton PFTs, protozooplankton,  
 564 mesozooplankton and jellyfish. For PlankTOM10.5 the MAC PFT has been summed with the 11<sup>th</sup> PFT that duplicates MAC.  
 565 Panels shown PFT biomass for PlankTOM11 (left), PlankTOM10.5 (middle) and PlankTOM10 (right), for two regions; the north  
 566  $30^\circ\text{N} - 70^\circ\text{N}$  (top) and the south  $30^\circ\text{S} - 70^\circ\text{S}$  (bottom) across all longitudes. All data are averaged for 1985-2015.

567 Primary production follows a similar pattern to total phytoplankton biomass across the three simulations, with  
568 higher biomass across more latitudes in the north compared to the south, although primary production differs from  
569 phytoplankton at the equator where it reaches a similar magnitude peak as in the south (Fig. 10). Export production  
570 has a markedly different zonal mean distribution across latitudes than PFT biomass and primary production, with  
571 the highest production in the tropics for all three simulations. The large variation in zooplankton biomass in the  
572 north and south between the three simulations is not reflected in export production, as would be expected (Fig.  
573 10). Around 40°S and 0° PlankTOM10 primary production peaks and is the highest of the three simulations. This  
574 is reflected in PlankTOM10 export peaking at the same latitudes. Around 30-55°N PlankTOM11 primary  
575 production peaks and is the highest of the three simulations, but this is not reflected in PlankTOM11 export  
576 peaking over the same latitudes (Fig. 10). Due to the lower total zooplankton biomass in PlankTOM11 compared  
577 to the other two simulations, mostly due to the reduced macrozooplankton, driven by the peak in jellyfish biomass.  
578 primary production peaks as there is reduced grazing on phytoplankton, but due to lower zooplankton biomass  
579 and therefore less zooplankton egestion, excretion and mortality there is less production of POC<sub>L</sub>.

580 Globally primary production is higher in PlankTOM10, than in PlankTOM11, but export is slightly lower, as are  
581 POCs and POC<sub>L</sub> (Table 6; Fig. A2), indicating that more of the carbon is retained and circulated in the plankton  
582 ecosystem in PlankTOM10 than in PlankTOM11. This is not just due to an additional top PFT, as in  
583 PlankTOM10.5, primary production and export are the lowest (Table 6; Fig. A2). However, as mentioned  
584 previously, the changes to export are smaller than expected given the large changes to zooplankton biomass and  
585 ecosystem structure. This is likely due to a bottle neck effect in the model structure, where, for example, mortality  
586 from three zooplankton PFTs, enters a single pool (Fig. 1b).

587

## 588 4 DISCUSSION

589

590 Model results suggest high competition between macrozooplankton (crustaceans) and jellyfish, with a key role of  
591 jellyfish being its control on macrozooplankton biomass, which via trophic cascades influences the rest of the  
592 plankton ecosystem. The growth rate of jellyfish is higher than that of macrozooplankton for the majority of the  
593 ocean (where the temperature is less than ~25°C) but the mortality of jellyfish is also significantly higher than  
594 macrozooplankton, again for the majority of the ocean. The combination of high growth and mortality means that  
595 jellyfish have a high turnover rate in temperate waters. In situations where jellyfish mortality is reduced (but still  
596 higher than macrozooplankton mortality), jellyfish outcompete macrozooplankton for grazing. Below 20°C  
597 jellyfish and macrozooplankton respiration is almost the same, so will have minimal influence on their relative  
598 biomass. Biomass is not linearly related to the growth, respiration and mortality rates, with biomass also dependent  
599 on prey availability, total PFT biomass and other variables. Because jellyfish also prey directly on  
600 macrozooplankton, the biomass of macrozooplankton can rapidly decrease in a positive feedback mechanism.  
601 Within oligotrophic regions both jellyfish and macrozooplankton biomass is low, as expected due to limited  
602 nutrients limiting phytoplankton growth in these regions. Around equatorial upwelling regions, macrozooplankton  
603 outcompete jellyfish. Macrozooplankton also outcompete jellyfish in many coastal areas including around

604 northern Eurasia because they have a built-in coastal and under-ice advantage to represent enhanced recruitment  
605 in these environments which likely tips the balance in their favour (Le Quéré et al., 2016). Around 40°S and 40-  
606 50°N jellyfish mostly outcompete macrozooplankton, water temperature here is around 10-17°C which is a  
607 temperature where jellyfish growth is the most above macrozooplankton growth and macrozooplankton mortality  
608 nearing jellyfish mortality, which combined together favour jellyfish over macrozooplankton. This sensitivity of  
609 the composition of the zooplankton community to the mortality of jellyfish could help explain why jellyfish are  
610 seen as increasing globally. A reduction in jellyfish mortality during early life-stages i.e. through reduced  
611 predation on ephyrae and juveniles by fish (Duarte et al., 2013; Lucas et al., 2012), could quickly allow jellyfish  
612 to outcompete other zooplankton, especially macrozooplankton.

613 The high patchiness of jellyfish in the observations is partly but not fully captured in PlankTOM11 (Fig. 7 and  
614 Table 7). The reasons for limited patchiness include the model resolution of  $\sim 2^\circ \times 1^\circ$  which doesn't allow for the  
615 representation of small-scale physical mixing such as eddies and frontal regions, which have been shown to  
616 influence bloom formation (Benedetti-Cecchi et al., 2015; Graham et al., 2001). Physical processes are likely to  
617 be more responsible for jellyfish patchiness than behaviours, due to their simplistic locomotion. For example,  
618 many jellyfish blooms occur around fronts, upwelling regions, tidal and estuarine regions, and shelf-breaks where  
619 currents can aggregate and retain organisms (Graham et al., 2001). A few large individuals of the species  
620 *Rhizostoma octopus* (barrel jellyfish) have been found to have the capacity to actively swim counter current that  
621 could aim to orientate themselves with currents, with the potential to aid bloom formation and retention (Fossette  
622 et al., 2015). However, this active swimming behaviour is not representative across the group and would only  
623 move the jellyfish within an area less than the resolution of the model. Furthermore, there is currently insufficient  
624 data and an incomplete understanding of such swimming behaviours to include it in a global model.

625 The maximum biomass of jellyfish in PlankTOM11 is  $98.9 \mu\text{g C L}^{-1}$ , compared to the observed maximum biomass  
626 of  $156 \mu\text{g C L}^{-1}$  and the mean:max ratio is within the range of observations although towards the lower end (Table  
627 7). This demonstrates that even without replication of high patchiness, PlankTOM11 still achieved some  
628 ephemeral blooms where jellyfish achieved a high biomass.

629 A key limitation of the representation of jellyfish in the model is the exclusion of the full life cycle. Most jellyfish  
630 display metagenesis, alternating between a polyp phase that reproduces asexually and a medusa phase that  
631 reproduces sexually (Lucas and Dawson, 2014). PlankTOM11 currently only characterises the pelagic phase of  
632 the jellyfish life cycle, with parameters based on data from the medusae and ephyrae. The biomass of jellyfish is  
633 maximal during the pelagic medusa stage, as medusae are generally several orders of magnitude larger than polyps  
634 and one polyp can release multiple ephyrae into the water column (Lucas and Dawson, 2014). Although most  
635 hydromedusae persist in the plankton for short periods of time, larger scyphomedusae can live for 4-8 months and  
636 individuals in some populations can survive for more than a year by overwintering; something that may be  
637 facilitated by global climate change (Boero et al., 2016). Polyps develop from planula larvae within 5 weeks of  
638 settlement, and can persist far longer than medusae owing to their asexual mode of reproduction and the fact that  
639 they can encyst, which allows them to remain dormant until environmental conditions are favourable for budding  
640 (Lucas and Dawson, 2014). Unusually, mature medusae of *Turritopsis dohrnii* can revert back to the polyp stage  
641 and repeat the life cycle, which effectively confers immortality (Martell et al., 2016). Our understanding of polyp

642 ecology is almost entirely based on laboratory reared specimens of common, eurytolerant species, with the  
643 patterns observed being locale- and species-dependent. We know that temperature changes can trigger the budding  
644 of ephyrae by scyphopolyps, which may lead to an increase in the medusa population (Han and Uye, 2010; Lucas  
645 and Dawson, 2014), but the number of species whose polyps have been located and studied in situ is minuscule  
646 and so estimates of polyp abundance or biomass are impossible even to estimate.

647 Models that include the full jellyfish life cycle are still relatively new, and their focus has been locale- and species-  
648 dependant (e.g. Henschke et al., 2018; Schnedler-Meyer et al., 2018). The aim of this study was not to reproduce  
649 small-scale blooms, but rather to assess at the large and global scale the influence of jellyfish on the plankton  
650 ecosystem and biogeochemistry. We consider it enough to note that higher temperature within PlankTOM11  
651 increases the growth rate, which translates into increased biomass if there is sufficient food, thus providing a  
652 representation of an increasing medusa population. The inclusion of jellyfish life cycles into PlankTOM11 would  
653 introduce huge uncertainties due to the lack of clear in situ life cycle data and is beyond the scope of the exercise.

654 There is currently no coastal advantage for jellyfish included in the model, as there is for macrozooplankton,  
655 which have a coastal and under-ice advantage for increased recruitment (Le Quéré et al., 2016). Introducing a  
656 similar coastal advantage for jellyfish could introduce an element of life cycle benefits i.e. the increased  
657 recruitment and settlement of planula larvae onto hard substrate in coastal regions and also ephyrae released from  
658 nearshore systems may benefit from being in nearshore waters (restricted there by mobility and current-closure  
659 systems) in much the same way as for other neritic planktonic taxa (Lucas et al., 2012). Alternatively, a deep-  
660 water disadvantage could be introduced for jellyfish to introduce an element of their life cycle dependencies in  
661 that the polyps require benthic substrate for settlement and development into the next life stage and are dependent  
662 on plankton for food, which are more abundant in shallower coastal waters. Future work on PlankTOM11 could  
663 investigate the strengths and weaknesses of these two avenues (coastal advantage and deep-water disadvantage)  
664 for introducing a jellyfish lifecycle element.

665 Jellyfish in PlankTOM11 are parameterised using data largely from temperate species, because this is the majority  
666 of the data available. This may explain some of the prevalence of jellyfish in PlankTOM11 at mid- to high-  
667 latitudes and the lower biomass in the tropics. Experimental rate data for a wider range of jellyfish species from  
668 a wider range of latitudes is required to address this bias. Another limitation of jellyfish representation in the  
669 model is the lack of body size representation. Generally smaller individuals have greater biological activity, while  
670 larger individuals have greater biomass. Depending on the time of year and life history strategy the dominant  
671 source of biomass will shift between smaller and larger individuals. The size distribution of body mass in jellyfish  
672 is particularly wide compared to other PFTs (Table 1), so representing jellyfish activity by an average sized  
673 individual could well skew the results.

674 Trophic interactions explain the improvement of spatial chlorophyll with the introduction of jellyfish to the model  
675 (PlankTOM10 to PlankTOM10.5 to PlankTOM11), especially the North/South ratio. The three simulations have  
676 identical physical environments, with the influence of jellyfish as the only alteration, so any differences between  
677 the three can be attributed to the ecosystem structure. Jellyfish are the highest trophic level represented in  
678 PlankTOM11, with preference for meso-, followed by proto-, and then macrozooplankton. However, the largest  
679 influence of jellyfish is on the macrozooplankton, because the grazing pressure on mesozooplankton from

680 macrozooplankton is reduced, and the grazing on protozooplankton by macro- and mesozooplankton is reduced,  
681 while the grazing pressure from jellyfish on both meso- and protozooplankton is increased. The combined changes  
682 to macrozooplankton and jellyfish grazing pressure counteract to reduce the overall change in grazing pressure.  
683 The top down trophic cascade from jellyfish on the other zooplankton also changes some of the grazing pressures  
684 on the phytoplankton, which translates into regional and seasonal effects on chlorophyll. Jellyfish increase  
685 chlorophyll in the northern pacific and reduce it in the southern pacific, relative to PlankTOM10 (Fig. 9).  
686 Seasonally, in the global north jellyfish increase phytoplankton biomass most during the summer and in the global  
687 south jellyfish decrease phytoplankton biomass most during the summer, relative to PlankTOM10 (Fig. 11). In  
688 the north, most of this summer increase in phytoplankton comes from coccolithophores and *Phaeocystis*, while in  
689 the south most of the summer decrease comes from coccolithophores, picophytoplankton and mixed  
690 phytoplankton (Fig. A1).

691 The complexity of zooplankton has been increased, however, the complexity of particulate organic carbon has  
692 not, resulting in a bottleneck in carbon export. The low sensitivity of the modelled export to changes in  
693 zooplankton composition is likely due to the small number of particulate organic carbon pools. For example,  
694  $POC_L$  would export the same carbon particulate whether mesozooplankton, macrozooplankton or jellyfish  
695 dominate. There is variety built into the zooplankton contribution to  $POC_L$  as the amount entering is dependent  
696 on the grazing rate, growth, biomass etc. of each zooplankton, but it all becomes one type of particulate matter  
697 once it enters the pool.

698 The two pools of particulate organic carbon in PlankTOM11 are insufficient to represent the variety of particulate  
699 organic carbon generated by the increased variety of zooplankton as the model has been developed. The  
700 contribution of mortality to  $POC_L$  is orders of magnitude different between mesozooplankton and jellyfish  
701 carcasses. The composition of the carcasses is also very different, with the high water-content of jellyfish compared  
702 to other zooplankton, which effects the carcass sinking behaviour (Lebrato et al., 2013a). Mass deposition events  
703 of jellyfish carcasses (jelly-falls), at depths where the carbon is unlikely to be recycled back into surface waters at  
704 short to medium time scales, are known to contain significant amounts of carbon and can contain in excess of a  
705 magnitude more carbon than the annual carbon flux (Billett et al., 2006; Yamamoto et al., 2008). PlankTOM11  
706 likely substantially underestimates jellyfish contribution from mortality (Luo et al., 2020). Through rapidly  
707 sinking jelly-falls, jellyfish cause a large pulse in export (Lebrato et al., 2012, 2013a, 2013b), not yet accounted  
708 for in PlankTOM11. The global export in PlankTOM11 (7.11 PgC/y) is within global estimates of 5 - 12 PgC/y.  
709 The main reason for export being towards the lower end of observations is that the global primary production in  
710 PlankTOM11 is lower than the observed rate. Another potential explanation which may enhance the low export  
711 is that within the model jellyfish have a high turnover rate, due to their high growth, grazing and mortality rates,  
712 thus taking in a high proportion of carbon, but they are not then acting as a direct rapid source of sinking carbon  
713 through their mortality.

714 The contribution of egestion and excretion (see Fig. 1b and Fig. A2) to  $POC_L$  is also very different between  
715 mesozooplankton, macrozooplankton and jellyfish, most particularly that the main contribution from meso- and  
716 macrozooplankton is in the form of solid faecal pellets, while for jellyfish the main contribution is from mucus  
717 (Hansson and Norrman, 1995). The composition and sinking behaviour of faecal pellets and mucus will be

718 substantially different, with mucus sinking more slowly and more likely to act as a nucleus for enhanced  
719 aggregation with other particles, forming a large low-density mass (Condon et al., 2011; Pitt et al., 2009).

720 Work is currently underway on PlankTOM to increase the size partitioning of particulate organic carbon through  
721 introducing a size-resolving spectral model with a spectrum of particle size and size-dependent sinking velocity  
722 (Kriest and Oschlies, 2008). This method has the advantage of improving the representation of particulate organic  
723 carbon production from all PFTs but is substantially more computer expensive. Another role of jellyfish may be  
724 that they act as significant vectors for carbon export, but with the current POC partitioning in PlankTOM11 this  
725 role has not been elucidated here. The potential influence of introducing increased size partitioning on carbon  
726 export could be significant, with peaks in jellyfish biomass being followed by a pulse in carbon export as there is  
727 rapid sinking of large carcasses (Lebrato et al., 2012; Luo et al., 2020).

728 Jellyfish have been included in a range of regional models, the majority are fisheries-based ecosystem models,  
729 namely ECOPATH and ECOPATH with ECOSIM (Pauly et al., 2009). These include regional models of the  
730 Northern Humboldt Current system (Chiaverano et al., 2018), the Benguela Upwelling System (Roux et al., 2013;  
731 Roux and Shannon, 2004; Shannon et al., 2009) and an end-to-end model of the Northern California Current  
732 system, based on ECOPATH (Ruzicka et al., 2012). Jellyfish have also been included in regional Nutrient  
733 Phytoplankton Zooplankton Detritus (NPZD) models, representing small-scale coastal temperate ecosystems with  
734 simple communities, for example, Schnedler-Meyer et al. (2018) and Ramirez-Romero et al. (2018). These models  
735 have provided valuable insight into jellyfish in the regions studied, but the focus on coastal ecosystems and either  
736 a top-down approach (ECOPATH) or highly simplified ecosystem (NPZD) limits their scope. PlankTOM11 offers  
737 the first insight into the role of jellyfish using a global biogeochemical model that represents multiple plankton  
738 functional types.

739

### 740 3.5 CONCLUSION

741

742 Jellyfish have been included as a PFT in a global ocean biogeochemical model for the first time as far as we can  
743 tell at the time of writing. The PlankTOM11 model provides reasonable overall replication of global ecosystem  
744 properties and improved surface chlorophyll, particularly the north/south ratio. The replication of global mean  
745 jellyfish biomass, 0.13 PgC, is within the observational range, and in the region with the highest density of  
746 observations PlankTOM11 closely replicates the mean and seasonal jellyfish biomass. There is a deficit of data  
747 on jellyfish carbon biomass observations and physiological rates. Monitoring and data collection efforts have  
748 increased over recent years; we recommend a further increase especially focussing in less-surveyed regions and  
749 on non-temperate species.

750 The central role of jellyfish is to exert control over the other zooplankton, with the greatest influence on  
751 macrozooplankton. Through trophic cascade mechanisms jellyfish also influence the biomass and spatial  
752 distribution of phytoplankton. PlankTOM11 is a successful first step in the inclusion of jellyfish in global ocean  
753 biogeochemical modelling. The model raises interesting questions about the sensitivity of the zooplankton

754 community to changes in jellyfish mortality and calls for a further investigation in interactions between  
755 macrozooplankton and jellyfish. Future model development, alongside POC improvements, could include an  
756 exploration of the life cycle, coastal advantages, and higher resolution ocean physical processes to enhance  
757 patchiness.

758

**Table A1:** Sources and metadata for jellyfish growth rates, including references with associated number of data, species and life stage used to inform the growth parameter of jellyfish in PlankTOM11.

Reference	n	Species	Life Stage
Båmstedt et al., (1997)	3	<i>Cynea capillata</i>	Ephyrae
Daan (1986)	8	<i>Sarsia tubulosa</i>	Medusae
Frandsen & Riisgård (1997)	5	<i>Aurelia aurita</i>	Medusae
Hansson (1997)	20	<i>Aurelia aurita</i>	Medusae
Møller & Riisgård (2007a)	34	<i>Sarsia tubulosa, Aurelia aurita, Aequorea vitrina</i>	Medusae, ephyrae
Møller & Riisgård (2007b)	10	<i>Aurelia aurita</i>	Medusae, ephyrae
Olesen (1994)	8	<i>Aurelia aurita, Chrysaora quinquecirrha</i>	Medusae, ephyrae
Widmer (2005)	10	<i>Aurelia labiata</i>	Ephyrae

**Table A2:** The fit to the growth data for PFT's for the new three-parameter fit used in this study (see Eq. 3 and Fig. 2) and the two-parameter fit (see Eq. 2 and Fig. 2).

PFT	R <sup>2</sup>		n
	Two-parameter	Three-parameter	
CNI	9.58	11.36	98
MAC	36.57	36.76	253
MES	0.32	0.34	2742
PRO	0.00	7.81	1300
BAC	1.66	1.66	1429
DIA	9.59	9.58	439
PHA	6.29	37.07	67
MIX	21.25	19.17	95
COC	33.91	36.01	322
PIC	20.17	20.29	150
FIX	2.67	10.62	32

**Table A3:** Sources and metadata for jellyfish grazing preferences, including references with associated species, life stage and preference for prey (categorised into PFTs) with any notable phrases used to inform the grazing of jellyfish in PlankTOM11.

Reference	Species/Class/Genera	Life Stage	PFT preference
Båmstedt et al. (2001)	<i>Aurelia aurita</i>	Ephyrae	Mixed-phytoplankton, mesozooplankton and particulate organic material
Colin et al. (2005)	<i>Aglaura hemistoma</i>	Medusa	“microplanktontic omnivores”; protozooplankton and some phytoplankton
Flynn and Gibbons (2007)	<i>Chrysaora hysoscella</i>	Medusa	Wide variety ranging in size from protozooplankton to macrozooplankton, with the “numerically dominant” prey as mesozooplankton
Malej et al. (2007)	<i>Aurelia</i> sp.	Medusa	Mesozooplankton and protozooplankton
Morais et al. (2015)	<i>Blackfordia virginica</i>	Medusa	Mesozooplankton and diatoms
Purcell (1992)	<i>Chrysaora quinquecirrha</i>	Medusa	Mesozooplankton (upto 71% of diet)
Purcell (1997)	Hydromedusa		“mostly generalist feeders”, mesozooplankton as a preference
Purcell (2003)	<i>Aurelia labiata</i> , <i>Cyanea capillata</i> , <i>Aequorea aequorea</i>		Mainly mesozooplankton
Stoecker et al. (1987)	<i>Aurelia aurita</i>	Medusa	Protozooplankton and mesozooplankton preferentially removed from “natural microzooplankton” assemblage. In cultured prey assemblage, larger protozooplankton were selected.
Uye and Shimauchi (2005b)	<i>Aurelia aurita</i>	Medusa	Mostly mesozooplankton, some protozooplankton
Costello and Colin (2002)	<i>Aglantha digitale</i> , <i>Sarsia tubulosa</i> , <i>Proboscidactyla flaviderrata</i> , <i>Aequorea victoria</i> , <i>Mitrocoma cellularia</i> , <i>Phialidium gregarium</i>	Medusa	Mesozooplankton (crustacean) and protozooplankton (ciliates)

**Table A4:** Additional tuning parameter values for PlankTOM11 (see Sect.2.1.5) following the change to the growth rate formulation. ‘Before growth change’ values are those used in PlankTOM10<sup>LQ16</sup> and ‘after growth change’ values are used in simulations for this study (PlankTOM11, PlankTOM10.5 and PlankTOM10).

Parameter	Before growth change	After growth change
Grazing preference ratio of mesozooplankton for <i>Phaeocystis</i>	0.75	1
Grazing preference ratio of protozooplankton for picophytoplankton	2	3
Half saturation constant of phytoplankton grazing on iron		
Diatoms	40.0e-9	80.0e-9
Picophytoplankton	10.0e-9	25.0e-9
<i>Phaeocystis</i>	25.0e-9	80.0e-9
Half saturation constant of bacteria for dissolved organic carbon	10.0e-6	8.0e-7
Maximum bacteria uptake rate	3.15	1.90
Diatom respiration	0.012	0.12

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**Table A5.** Global mean values for rates and biomass from observations with the associated references. In parenthesis is the percentage share of the plankton type of the total Phytoplankton or Zooplankton biomass.

	Observations	Reference for the data
<b>Rates</b>		
Primary production (PgC $y^{-1}$ )	51-65	Buitenhuis et al. (2013b)
Export production at 100m (PgC $y^{-1}$ )	5-13	Henson et al. (2011), Palevsky et al. (2018)
CaCO <sub>3</sub> export at 100m (PgC $y^{-1}$ )	0.6-1.1	Lee (2001), Sarmiento et al. (2002)
N <sub>2</sub> fixation (TgN $y^{-1}$ )	60-200	Gruber (2008)
<b>Phytoplankton biomass 0-200m (PgC)</b>		
N <sub>2</sub> -fixers	0.008-0.12 (2-8%)	Luo et al. (2012)
Picophytoplankton	0.28-0.52 (35-68%)	Buitenhuis et al. (2012b)
Coccolithophores	0.001-0.032 (0.2-2%)	O'Brien et al. (2013)
Mixed-phytoplankton	-	-
<i>Phaeocystis</i>	0.11-0.69 (27-46%)	Vogt et al. (2012)
Diatoms	0.013-0.75 (3-50%)	Leblanc et al. (2012)
<b>Heterotrophs biomass 0-200m (PgC)</b>		
Bacteria	0.25-0.26	Buitenhuis et al. (2012a)
Protozooplankton	0.10-0.37 (27-31%)	Buitenhuis et al. (2010)
Mesozooplankton	0.21-0.34 (25-66%)	Moriarty and O'Brien (2013)
Macrozooplankton	0.01-0.64 (3-47%)	Moriarty et al. (2013)
Jellyfish zooplankton	0.10-3.11	Bar-On et al. (2018), Lucas et al. (2014), Buitenhuis et al. (2013b)

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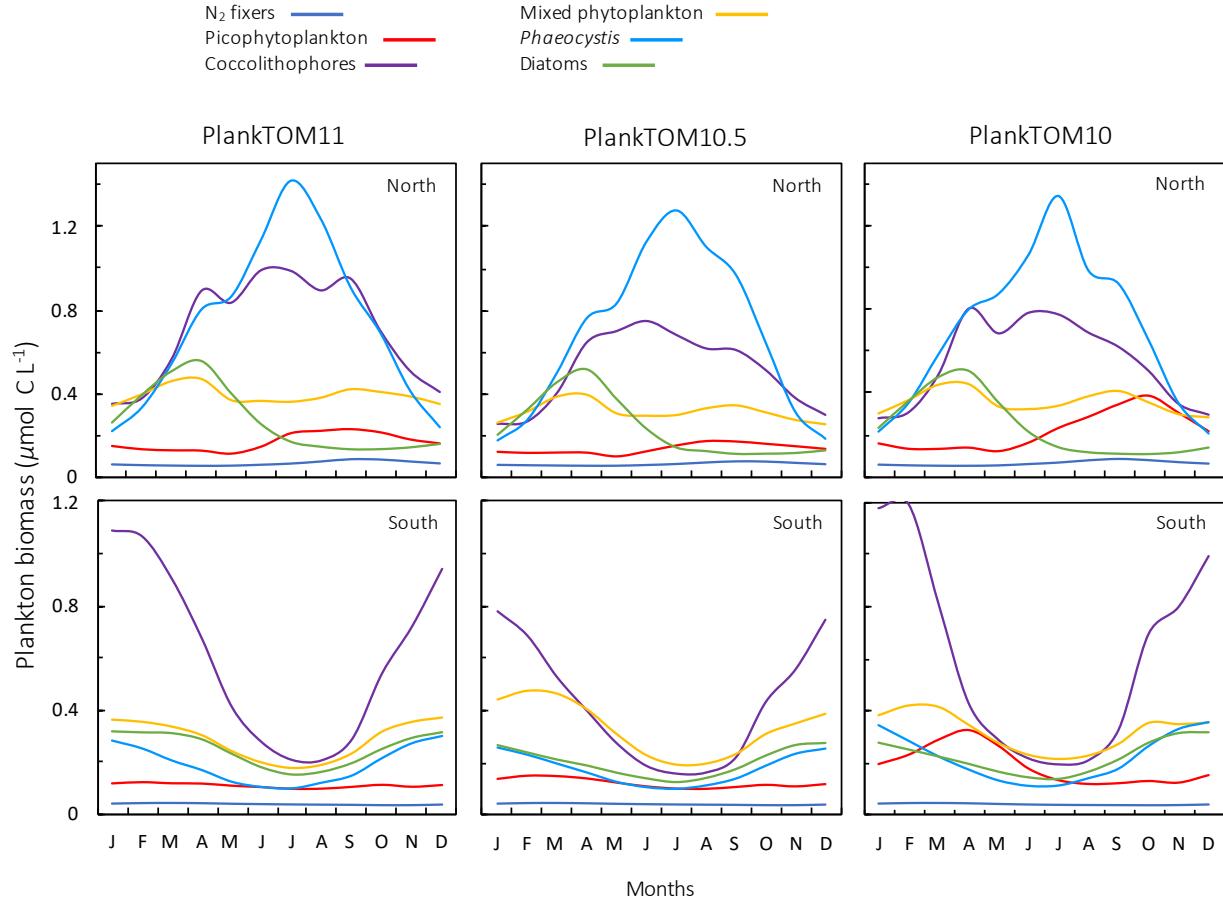
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**Table A6:** Total phytoplankton biomass ( $\mu\text{mol C L}^{-1}$ ) for  $30^\circ\text{N} - 70^\circ\text{N}$  across all longitudes. Observations are from gridded MAREDAT, all data are for the surface ocean (0-10 meters). Phytoplankton types include picophytoplankton, *Phaeocystis*, diatoms, nitrogen-fixers and coccolithophores. The seasonal amplitude is the amplitude for the full seasonal cycle (January – December) and the non-winter amplitude is the amplitude for March – October.

	Seasonal Amplitude	Non-winter Amplitude
Observations (median – mean)	0.78 – 2.67	0.70 – 2.12
PlankTOM11	1.82	0.97
PlankTOM10.5	1.54	0.80
PlankTOM10	1.69	0.81
PlankTOM10 <sup>LQ16</sup>	1.68	1.02

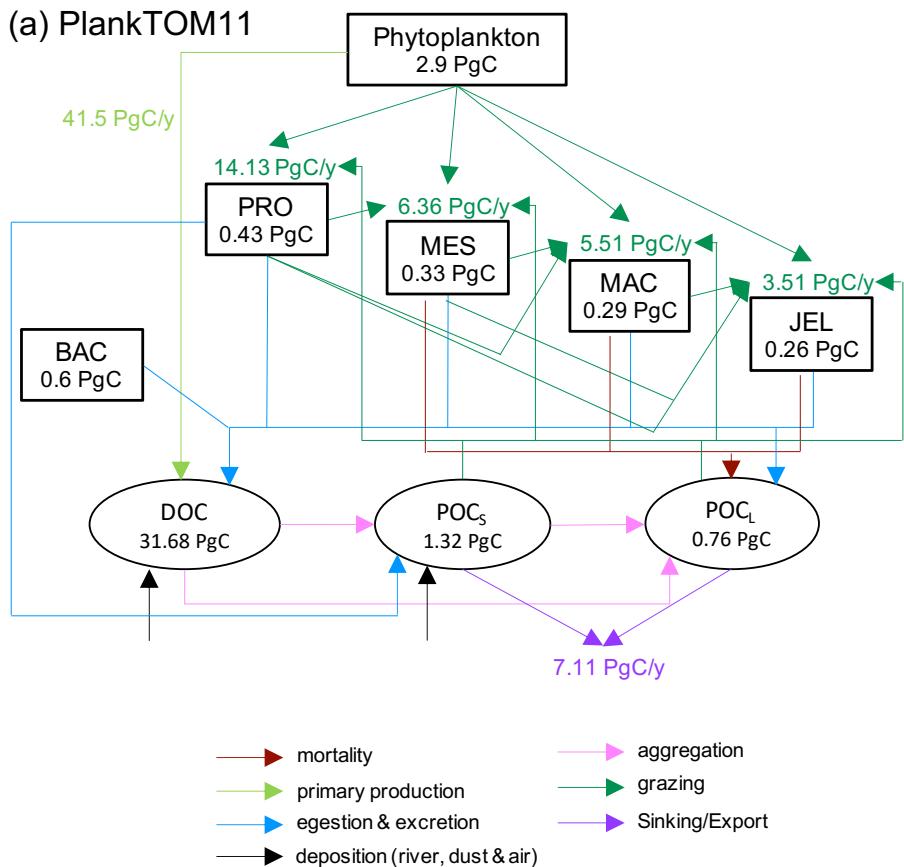
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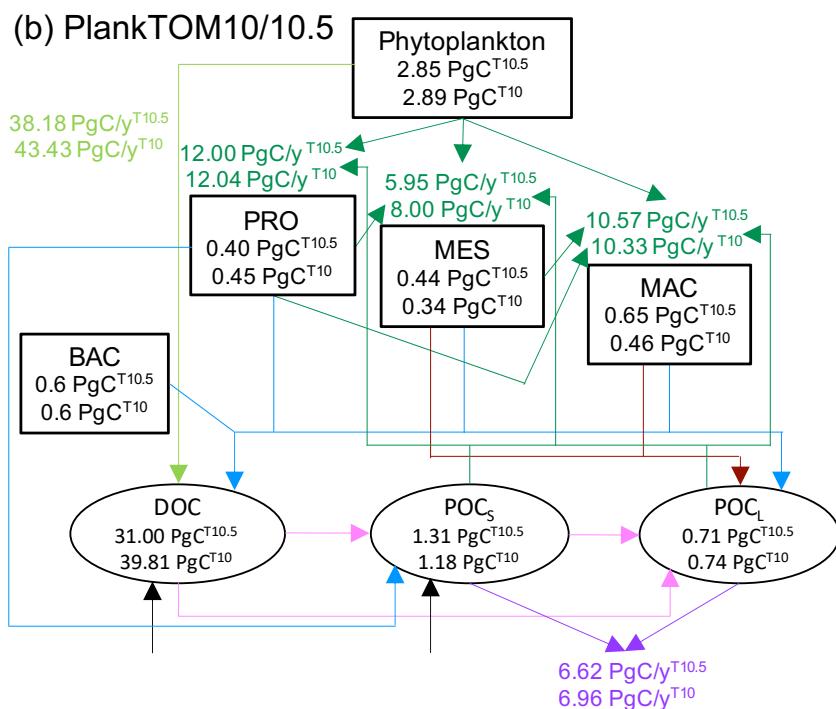


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**Figure A1.** Seasonal surface carbon biomass ( $\mu\text{mol C L}^{-1}$ ) of phytoplankton PFTs;  $N_2$  fixers, picophytoplankton, coccolithophores, mixed phytoplankton, *Phaeocystis* and diatoms. Panels shown PFT biomass for PlankTOM11 (left), PlankTOM10.5 (middle) and PlankTOM10 (right), for two regions; the north  $30^\circ\text{N} - 70^\circ\text{N}$  (top) and the south  $30^\circ\text{S} - 70^\circ\text{S}$  (bottom) across all longitudes. All data are averaged for 1985–2015.



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**Figure A2.** Schematic representation of global carbon biomass and rates in the PlankTOM marine ecosystem model including sources and sinks for dissolved organic carbon (DOC) and small ( $POC_s$ ) and large ( $POC_l$ ) particulate organic carbon. (a) PlankTOM11 and (b) PlankTOM10 and PlankTOM10.5. Carbon biomass (PgC) of PFT's and organic carbon pools are given within boxes and ovals, carbon rates (PgC/y) of primary production (light green), grazing (dark green) and export production (purple) are given next to the corresponding arrows. All data are averaged for 1985 to 2015.

800 **Author Contribution**

801 RMW, CLQ, ETB and SP conceptualized the research goals and aims. RMW carried out the formal analysis  
802 with contributions from CLQ and ETB. RW developed the model code with significant contributions from ETB,  
803 and RMW performed the simulations. RMW prepared the manuscript with contributions from all co-authors.

804 The authors declare that they have no conflict of interest.

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