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# A new marine ecosystem model for the University of Victoria Earth system climate model

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#### Abstract

Earth system climate models (ESCMs) are valuable tools that can be used to gain a better understanding of the climate system, global biogeochemical cycles, and how anthropogenically-driven changes may affect them. Here we describe improvements

- <sup>5</sup> made to the marine biogeochemical ecosystem component of the University of Victoria's ESCM (version 2.9). Major changes include corrections to the code and equations describing phytoplankton light limitation and zooplankton grazing, the implementation of a more realistic zooplankton growth and grazing model, and the implementation of an iron limitation scheme to constrain phytoplankton growth. The new model is evalu-
- ated after a 10 000-yr spin-up and compared to both the previous version and observations. For the majority of biogeochemical tracers and ecosystem processes the new model shows significant improvements when compared to the previous version and evaluated against observations. Many of the improvements are due to better simulation of seasonal changes in higher latitude ecosystems and the effect that this has on
- ocean biogeochemistry. This improved model is intended to provide a basic new ESCM model component, which can be used as is or expanded upon (i.e., the addition of new tracers), for climate change and biogeochemical cycling research.

#### 1 Introduction

The oceans have a large effect on the Earth's climate and play an important role in
global biogeochemical cycles. Although many of the interactions within the oceans and between the oceans and other systems (i.e., atmosphere, land, etc.) are driven purely by physical or chemical mechanisms, biological processes play an important, but often less understood, role. Marine ecosystems affect the climate primarily through the "carbonate" and "soft tissue" pumps (i.e., the "biological" pump) (Longhurst and Harrison, 1989; Volk and Hoffert, 1985). These pumps work through the biological uptake of carbon in the surface ocean and the subsequent transport (mainly by sinking



and zooplankton grazing dynamics) of the small fraction of it that is not recycled along the way (mostly through respiratory processes) to the deep ocean, where it is unable to affect the climate (i.e., as  $CO_2$ ) for hundreds to thousands of years. The biological pump has been estimated to export between 5 and 20 GtCyr<sup>-1</sup> out of the surface layer (Henson et al., 2011; Honjo et al., 2008; Laws et al., 2000). However, as indicated by the large range of estimates there is great uncertainty in our understanding of the magnitude of carbon export (Henson et al., 2011), and thus it's effect on the Earth's climate.

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Marine ecosystems effect global biogeochemical cycles in a number of ways. As mentioned above they play an important role in the carbon cycle through the uptake, recycling, and sequestration of some CO<sub>2</sub>. They also play a large role in the cycling of nitrogen, phosphorus, and oxygen. In surface waters nitrogen and phosphorus are major nutrients that are consumed by, and drive, primary production (PP) and thus are linked to the carbon cycle. Since these nutrients often limit PP their availability can thus

- influence the climate system by controlling the magnitude and location of biological pump processes. When heterotrophic members of the ecosystem (i.e., bacteria, zooplankton, etc.) consume the organic material that was fixed during PP, biogeochemical cycles are further effected as oxygen is consumed during respiration and new C-, N-, and P-containing chemical compounds are formed from the processed organic matter.
- Biological oxygen consumption can have large impact on the ocean's oxygen concentration when it takes place away from the surface and under some physical conditions may cause areas of the ocean to become hypoxic or anoxic. Most of the organic C, N, and P is eventually recycled by heterotrophs back into an inorganic form and once again becomes available for uptake by primary producers when it reaches the surface
- ocean (if it is not already there). However, some N may be returned to the atmosphere as N<sub>2</sub> or N<sub>2</sub>O when microorganisms in hypoxic/anoxic waters utilize N containing compounds. This loss of N to the atmosphere is not one-way as nitrogen can also be removed from the atmosphere by nitrogen fixing organisms. Of course these biotically driven cycles do not take place without interactions between chemically and physically



driven C, N, and P cycles and there are frequent exchanges of material between these cycles. Although the major biogeochemical pathways are known, there is much uncertainty in the magnitude and time-scale of each pathway and the biological communities that drive them.

- <sup>5</sup> Ocean biology and the effects that it has on important processes are changing because of naturally and anthropogenically driven environmental change. Increasing anthropogenic greenhouse gas emissions have, and are, altering the Earth's climate and warming the planet (IPCC, 2007; Sarmiento and Gruber, 2002). The oceans have taken up much of the heat that has accumulated in the Earth system and near-surface ocean
- <sup>10</sup> waters have already warmed between 0.5 and 0.7 degrees over the last 100 yr (Hansen et al., 2010; IPCC, 2007). Since most physical, chemical, and biological processes are temperature dependant these changes have had an effect on many marine ecosystem processes. This has in turn affected the oceans role in the Earth's climate system and global biogeochemical cycles (often through feedback effects). The magnitude of
- the biological pump is predicted to decline in response climate change, resulting in reduced carbon sequestration and higher atmospheric CO<sub>2</sub> levels (Sarmiento et al., 1998). Given the magnitude of this carbon sink and its effect on the climate, accurately quantifying and understanding how it may change due to anthropogenic activities is important. Furthermore, since manipulation of the biological pump is being considered
- as a means of climate engineering (Keith, 2000; Oschlies et al., 2010b), it essential to have a thorough understanding of potential effects and side effects. Temperature changes are also effecting the ocean's capacity to hold dissolved oxygen and causing the global volume of oxygen held by the ocean to decrease (i.e., ocean deoxygenation) (Gruber, 2011), which may effect biogeochemical cycles if more areas become
- <sup>25</sup> hypoxic or anoxic. The effects of anthropogenic activities on ocean biology are not limited to temperature since CO<sub>2</sub> emissions are having an effect on ocean chemistry and making marine waters more acidic (i.e., ocean acidification) (Hofmann and Schellnhuber, 2010), to the detriment of some species. Furthermore, anthropogenic activities are causing many coastal waters to become eutrophic and are increasing remote open



ocean nutrient concentrations (Doney, 2010), which has a fertilization effect on marine ecosystems. Since marine ecosystems are a key component of oceanic systems and play a role in other Earth system processes it is essential to understand how they function and how they may function in the future as the world changes.

- <sup>5</sup> Many approaches have been used to study the role of marine ecosystems in the Earth's climate system and global biogeochemical cycles. Models are one of the few tools available for understanding these dynamics on global basis and for predicting how they may change in the future as a result of anthropogenic influences. Here we describe improvements made to the marine ecosystem component of an Earth system of the future of the few to be an exact system component of an Earth system.
- tem model of intermediate complexity (EMICs). EMICs were developed roughly two decades ago for a broad spectrum of purposes (Claussen et al., 2002) and could still be considered as being in the early stages of development. Since these models are much more complex than ocean models without explicit atmospheric or terrestrial components the marine ecosystems described by them are usually much more simplistic
- <sup>15</sup> when compared to current state of the art marine ecosystem models that now include multiple "functional groups" or "biogeochemical guilds" of organisms and biogeochemical tracers (Hood et al., 2006). While adding multiple functional groups or additional biogeochemical tracers may not be computationally feasible, or likely to improve model skill (Friedrichs et al., 2007; Kriest et al., 2010), there are many improvements that
- <sup>20</sup> can be made to EMIC marine ecosystem models to improve their performance and realism. Our improvements to the marine ecosystem component of the University of Victoria Earth System Climate Model (UVic ESCM) are designed to correct previous errors, add more realistic processes, and provide a basic platform for future applications.

#### 25 2 A brief description of the UVic ESCM

The UVic Earth System Climate Model (Eby et al., 2009) version 2.9 consists of three components: a three dimensional general circulation model of the ocean, a terrestrial



model, and a simple one layer atmospheric energy-moisture balance model (Weaver et al., 2001). All model components use a common horizontal resolution of 1.8° latitude  $\times$  3.6° longitude and the oceanic component has nineteen levels in the vertical with thicknesses ranging from 50 m near the surface to 500 m in the deep ocean. The

- <sup>5</sup> oceanic circulation model (Modular Ocean Model 2) includes physical parameterizations for diffusive mixing along and across isopycnals, eddy induced tracer advection (Gent and McWilliams, 1990), and, in our configuration, a scheme for the computation of tidally induced diapycnal mixing over rough topography (Simmons et al., 2004). The atmospheric energy-moisture balance model interactively calculates heat and wa-
- <sup>10</sup> ter fluxes to the ocean, land, and sea ice. Wind velocities to calculate the momentum transfer to the ocean and to a dynamic-thermodynamic sea ice model, surface heat and water fluxes, and the advection of water vapor in the atmosphere are prescribed from NCAR/NCEP monthly climatology data. The terrestrial model of vegetation and carbon cycles (Meissner et al., 2003) is based on the Hadley Center model TRIFFID.
  <sup>15</sup> Continental ice sheets are assumed to remain constant. The simulation of sea ice has
- <sup>15</sup> Continental ice sheets are assumed to remain constant. The simulation of sea ice has been evaluated and found to be in good agreement with observations (Bitz et al., 2001; Saenko et al., 2002).

### 3 Rationale for improving the marine ecosystem model

The most recent UVic ESCM marine ecosystem module of Schmittner et al. (2008) has
been used extensively in biogeochemical studies of climate change (Oschlies et al., 2008; Schmittner et al., 2009b), paleooceanography (Schmittner and Galbraith, 2008), climate engineering (Oschlies, 2009; Oschlies et al., 2010a,b), and the nitrogen cycle (Somes et al., 2010a,b). These studies have provided valuable insight on a number of research topics and have help to answer questions about pressing issues of climate change. However, despite the usefulness of this model and the relatively good skill at which it simulates biogeochemical tracer distributions, there are some aspects of the



model that can be improved, and some errors that need to be corrected, for future applications.

A first set of errors in the original Schmittner et al. (2008) model had already been identified and corrected by Schmittner et al. (2009a). Additional analysis of various <sup>5</sup> aspects of the model results led to the identification of a few other previously overlooked model errors and aspects that could be improved. Originally, we had focused solely on correcting these errors and improving the equations and code describing light limitation of phytoplankton growth and zooplankton grazing. The improvements made to the equations/code describing light limitation of phytoplankton growth were made because light attenuation due to diazotroph biomass was not included in them. Since some species such as *Trichodesmium* can form dense blooms at the surface that can significantly attenuate light (Capone and Zehr, 1997; Sellner, 1997) it was viewed as important to simulate this effect on the phytoplankton community. In addition, there was

- error in the code (in both versions 2.8 and 2.9) describing light attenuation in the layers below the surface. This error occurred because the biomass of diazotrophs from the layers above was not used in calculations of light attenuation at depth. A second set of changes was made to the equations/code describing zooplankton grazing on phytoplankton and diazotrophs. The original equations were not formulated to provide a multiple-prey functional response, making grazing unrealistic when both types
- of plankton ("ordinary" phytoplankton and diazotrophs) were present (i.e., ingestion was incorrectly calculated). Correctly modeling grazing on multiple prey items is important for realistically simulating zooplankton growth and the different components of their diet (Anderson et al., 2010; Gentleman et al., 2003). While correcting zooplankton grazing we also decided to make their growth and grazing more realistic. However,
- these changes, along with conclusions drawn from recent studies on the role of iron in phytoplankton growth, required us to also modify phytoplankton growth and mortality in the model.

In Schmittner et al. (2008) zooplankton were limited to grazing on phytoplankton and diazotrophs, excretion was not a function of grazing, and their growth rate was



not a function of temperature. Since the zooplankton community grazes on itself and detritus, in addition to phytoplankton (Calbet and Saiz, 2005; Dilling and Brzezinski, 2004; Kleppel, 1993; Sherr and Sherr, 2002), we decided to represent these food web pathways in the model. Furthermore, since the excretion of inorganic nitrogen by zoo-

- <sup>5</sup> plankton is related to their ingestion and diet (Davidson et al., 1995; Glibert et al., 1992; Saba et al., 2009), we feel that it is important to allow excretion to be a function of grazing, even if our current formulation is simplistic due to Redfield stoichiometry and the desire to keep computational costs low (i.e., no complex formulations as in Anderson, 1992 or Pahlow and Prowe, 2010). Finally, since the growth of a major prey item, phyto-
- plankton, in the model is temperature dependent and zooplankton growth is also known to be dependent on temperature (Hirst and Bunker, 2010), we feel that it is important to allow zooplankton growth to change with temperature. However, making zooplankton growth and grazing a function of temperature had dramatic effects on phytoplankton growth as it was formulated in Schmittner et al. (2008).
- In Schmittner et al. (2008) phytoplankton growth at high latitudes in cold water, especially in the Southern Ocean, had been controlled by unrealistically high rates of grazing due to the fixed (temperature independent) growth rate of zooplankton (i.e., this fixed value while reasonable for warmer waters was too high for colder waters). When this top-down control was reduced by our modifications (i.e., making zooplankton growth
- a function of temperature), phytoplankton biomass and productivity became very high in these nutrient rich waters. Since phytoplankton biomass and productivity are known to be low in these regions due to iron limitation (Martin, 1992; Martin et al., 1991), it was necessary to include some form of iron limitation in the model. Additionally, by including the effects of iron limitation we hoped to improve the spatial distribution of di-
- <sup>25</sup> azotrophs, whose growth is also known to be limited by iron availability (Berman-Frank et al., 2001; Berman-Frank et al., 2007). Unfortunately, iron cycling and the interactions between iron and biology are quite complex and not fully understood, making it difficult to model (Galbraith et al., 2010; Moore and Braucher, 2008; Moore et al., 2004; Tagliabue et al., 2009). Somes et al. (2010a,b) successfully used an iron mask, based



on measured atmospheric dust deposition, in a nitrogen isotope study with the UVic 2.8 model to constrain diazotroph growth and achieve a more reasonable diazotroph spatial distribution than in Schmittner et al. (2008). Following this simple approach, we also elected to use an iron mask to constrain the growth of both diazotrophic and non-diazotrophic phytoplankton. However, unlike in Somes et al. (2010a,b) the mask that we use is of dissolved iron, and thus accounts for sources of iron from both aeolian dust sources and sedimentary efflux (Mahowald et al., 2005; Moore and Braucher, 2008).

#### 4 Model description

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### 4.1 Configuration of the circulation model

- The ocean circulation model described in Sect. 2 and the standard physical settings as set in the version 2.9 download (http://www.climate.uvic.ca/model/) have been modified slightly to have the similar physical dynamics to those in Schmittner et al. (2008). These modifications include turning off the Bryan-Lewis vertical mixing option, turning on the tidal mixing option, increasing the vertical diffusivity parameter in the Southern Ocean,
- <sup>15</sup> and implementing an anisotropic viscosity scheme in the tropics to improve the simulation of the equatorial currents. Based on the UVic 2.8 studies by Goes et al. (2010) and Schmittner et al. (2009b) the vertical background mixing parameter,  $K_{vb}$ , in the Southern Ocean (south of 40° S) was set to  $1.0 \text{ cm}^2 \text{ s}^{-1}$  in our implementation of UVic 2.9. The sinking of detritus is also different than in Schmittner et al. (2008) as it is not
- <sup>20</sup> constant below 1000 m but continues to increases linearly with depth (this is the standard formulation in the downloadable model version). An anisotropic viscosity scheme (Large et al., 2001) is implemented as in (Somes et al., 2010b) to improve equatorial circulation.



#### 4.2 New ecosystem model description

As discussed above, the marine ecosystem/biogeochemical model (Fig. 1) is a modified version of the NPZD model of Schmittner et al. (2008). As in the original model it consists of seven prognostic variables that are embedded within the ocean circulation

<sup>5</sup> model described above. The state variables include two phytoplankton classes (nitrogen fixers and other phytoplankton), zooplankton, particulate detritus, nitrate (NO<sub>3</sub>), phosphate (PO<sub>4</sub>), and oxygen (O<sub>2</sub>). Additional biogeochemical tracers include dissolved inorganic carbon (DIC) and alkalinity (ALK). All biological variables and particulate detritus are expressed in units of mmol N m<sup>-3</sup>. Constant (~ Redfield) stoichiometry
 relates the C, N, and P content of the biological variables and their exchanges with the inorganic variables (NO<sub>3</sub>, PO<sub>4</sub>, O<sub>2</sub>, ALK, and DIC). Parameters that are new or differ from those of Schmittner et al. (2008) are listed in Table 1.

Below the surface each variable changes its concentration C according to the following equation

$$_{15} \quad \frac{\partial C}{\partial t} = T + S, \tag{1}$$

where T represents all transport terms including advection, isopycnal and diapycnal diffusion, and convection. *S* denotes the source minus sink terms, which describe the

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following biogeochemical interactions:

$$S(PO_4) = (\mu_D D + \mu_P^* P_O + (\gamma - \varpi))(Graze_{P_O} + Graze_{P_D} + Graze_D + Graze_Z)$$
  
-  $J_O P_O - J_D P_D)R_{P:N}$ 

$$S(\text{NO}_3) = (\mu_D D + \mu_P^* P_O + (\gamma - \varpi))(\text{Graze}_{P_O} + \text{Graze}_{P_D} + \text{Graze}_D + \text{Graze}_Z)$$

$$= -J_O P_O - u_N J_D P_D)(1 - 0.8R_{O:N}r_{\text{sox}}^{NO_3})$$

$$S(P_{\rm O}) = J_{\rm O}P_{\rm O} - {\rm Graze}_{P_{\rm O}} - \mu_{\rm P}^*P_{\rm O} - m_{P_{\rm O}}P_{\rm O}$$

$$S(P_{\rm D}) = J_{\rm D}P_{\rm D} - \text{Graze}_{P_{\rm D}} - m_{P_{\rm D}}P_{\rm D}$$
<sup>(5)</sup>

$$S(Z) = \varpi(\text{Graze}_{P_0} + \text{Graze}_{P_D} + \text{Graze}_{D} + \text{Graze}_{Z}) - m_Z Z^2$$

$$S(D) = (1 - \gamma)(\text{Graze}_{P_{\text{O}}} + \text{Graze}_{P_{\text{D}}} + \text{Graze}_{\text{D}} + \text{Graze}_{\text{Z}}) + m_{P_{\text{O}}}P_{\text{O}} + m_{P_{\text{D}}}P_{\text{D}} + m_{\text{Z}}Z^{2}$$

$$(10) - \mu_{\text{D}}D - \text{Graze}_{\text{D}} - w_{\text{D}}\partial D\partial z$$

$$S(O_2) = F_{sfc} - S(PO_4)R_{O:P}r_{sox}^{O_2}$$
(8)

#### 4.2.1 Phytoplankton

The growth rate of "ordinary" non-diazotrophic phytoplankton,  $J_{\Omega}$ , is determined by irradiance (I) and the concentrations of NO<sub>3</sub> and PO<sub>4</sub>, while the growth rate of diazotrophs, 15



(2)

(7)

 $J_{\rm D}$ , is determined by irradiance (*I*) and the concentration of PO<sub>4</sub>.

$$J_{O} = \min\left(J_{OI}, J_{O}^{\max} \cdot \frac{NO_{3}}{k_{N} + NO_{3}}, J_{O}^{\max} \cdot \frac{PO_{4}}{k_{P} + PO_{4}}\right)$$
(9)  
$$J_{D} = \min\left(J_{DI}, J_{D}^{\max} \cdot \frac{PO_{4}}{k_{P} + PO_{4}}\right)$$
(10)

<sup>5</sup> For both types of phytoplankton light-limited growth,  $J_{(O \text{ or } D)I}$ , is basically calculated as in Schmittner et al. (2005, 2008) using

$$J_{(O \text{ or } D)/} = \frac{J_{(O \text{ or } D)}^{\max} \alpha I}{\left[ (J_{(O \text{ or } D)}^{\max})^2 + (\alpha I)^2 \right]^{1/2}}$$
(11)

However, these equations now account for the effects of diazotroph biomass on light
 attenuation. Modifications were also made in the code so that the integrated biomass of
 phytoplankton and diazotrophs from the layers above is now used in these calculations
 for depths below the surface. Thus, shortwave radiation at depth *z* is

$$I = I_{z=0} \text{PAR } e^{-k_{w}\tilde{z} - k_{c} \int_{0}^{\tilde{z}} (P_{O} + P_{D}) dz} \cdot \left[ 1 + a_{i} (e^{-k_{i}(h_{i} + h_{s})} - 1) \right]$$
(12)

<sup>15</sup> and light-limited growth, averaged over depth, with a triangular shaped diurnal cycle becomes

$$J_{I}^{\text{ave}} = \frac{1}{\Delta z \cdot 24 \text{ h}} \int_{z-\frac{\Delta z}{2}}^{z+\frac{\Delta z}{2}} \int_{(O \text{ or } D)/}^{z-\Delta t} J_{(O \text{ or } D)/} dz dt$$
$$= \frac{G_{\text{D}}}{k_{\text{w}} \Delta z} \left[ \Phi\left(\frac{2G_{I}}{G_{\text{d}}}\right) - \Phi\left(\frac{2G_{I}}{G_{\text{D}}}e^{-(k_{\text{w}}+k_{\text{c}}(P_{\text{O}}+P_{\text{D}}))\Delta z}\right) \right]$$
1146

(13)

with  $\tilde{z}$  and  $G_l$  calculated as in Schmittner et al. (2005, 2008) and

$$G_{\rm d} = J_{\rm (O \ or \ D)}^{\rm max} \cdot d$$

with d denoting the day length in a fraction of 24 h.

5 The function

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$$\Phi(u) = \ln\left(u + \sqrt{1 + u^2}\right) - \left(\sqrt{1 + u^2} - 1\right)/u$$

is approximated as

$$\Phi(u) = \frac{0.555588u + 0.004926u^2}{1 + 0.188721u} \tag{16}$$

as recommended by Evans and Garçon (1997). The condition for Eq. (16) being a very good approximation, 0 < u < 20, is always fulfilled.

The maximum potential growth rates of phytoplankton in the top three vertical layers are dependent on both temperature (T) and dissolved iron (Fe) (illustrated in Fig. 2a). For non-diazotrophic phytoplankton the maximum growth rate in these layers is

$$J_{\rm O}^{\rm max} = a \left( \frac{\rm Fe}{k_{\rm Fe}^{P} + \rm Fe} \right) \exp(T/T_{\rm b}) \tag{17}$$

and for diazotrophs it is

$$J_{\rm D}^{\rm max} = c_{\rm D} \max\left[0, a\left(\frac{{\rm Fe}}{k_{\rm Fe}^D + {\rm Fe}}\right) \left(\exp(T/T_{\rm b}) - 2.61\right)\right]$$
(18)

Temperature is calculated by model and the maximum growth rate of non-diazotrophic phytoplankton at 0  $^{\circ}$ C is set at 0.6 d<sup>-1</sup>, based on observations of mixed phytoplankton community growth rates (Le Quéré et al., 2005). The maximum growth rate of



(14)

(15)

diazotrophs is set to  $c_D = 0.4$  times that of non-diazotrophic phytoplankton. For the e-folding temperature of biological rates,  $T_b = 15.65$ °C (Schmittner et al., 2008), diazotrophs can grow only at temperatures higher than 15°C. The dissolved iron concentration is determined by a three-vertical layer, global mask (Fig. 3) of mean monthly

- dissolved iron concentration outputs from the BLING model (Galbraith et al., 2010). To facilitate the use of monthly data, daily dissolved iron concentrations are calculated using linear interpolation routines from Press et al. (1992). In addition, since the BLING model was run at a different global resolution than the one here, these differences were accounted for when creating the mask by linearly interpolating BLING output to
- the UVic ESCM grid. Below the top three vertical layers (i.e., below 240 m depth) where light is likely the most limiting factor of phytoplankton growth, iron limitation is assumed to negligible and the maximum potential growth rate is dependent only on temperature as in Schmittner et al. (2008). Since this masking approach does not resolve the iron cycle or even allow phytoplankton to take up iron the half-saturation constants that determine a butchlose the limitation is assumed to resolve the statement of the saturation of the saturation constants that determine a butchlose the saturation constants that the saturation const
- determine phytoplankton iron limitation were set during the tuning process to achieve a reasonable spatial distribution of phytoplankton and do not reflect any real phytoplankton affinities for this micronutrient.

Non-diazotrophic phytoplankton mortality is no longer set as a quadratic function as in Schmittner et al. (2008). Instead these phytoplankton now die at linear rate of  $0.03 d^{-1}$  which in combination with the additional loss term representing the temperature-dependent fast remineralization process, sets their total mortality at  $\geq 0.045 d^{-1}$ .

#### 4.2.2 Zooplankton

In contrast to the earlier model of Schmittner et al. (2008), zooplankton are now allowed to graze on detritus and other zooplankton (self-predation), in addition phytoplankton and diazotrophs. Grazing varies with prey density and is characterized by a multipleprey Holling II functional response that assigns preferences,  $\psi_i$ , for different types of



	$\text{Graze}_{P_{\text{O}}} = \mu_{\text{Z}}^{\text{max}} Z \theta_{P_{\text{O}}} P_{\text{O}}$
	$\text{Graze}_{P_{\text{D}}} = \mu_{\text{Z}}^{\text{max}} Z \theta_{P_{\text{D}}} P_{\text{D}}$
	$\text{Graze}_{\text{D}} = \mu_{\text{Z}}^{\text{max}} Z \theta_{\text{D}} D$
5	$\text{Graze}_{\text{Z}} = \mu_{\text{Z}}^{\text{max}} \theta_{\text{Z}} Z^2$
	where
	$\theta_{P_{\rm O}} = \psi_{P_{\rm O}}/\varphi$
	$\theta_{P_{\rm D}} = \psi_{P_{\rm D}}/\varphi$
10	$\theta_{\rm D} = \psi_{\rm D}/\varphi$
	$\theta_{\rm Z} = \psi_{\rm Z}/\varphi$
	$\varphi = P_{\rm O} + P_{\rm D} + D + Z + K_{\rm Graze}$
	and

(i) The rate of grazing on each type of prev is

<sup>15</sup> 
$$\mu_{Z}^{\max} = \mu_{Z}^{\theta} \cdot \max\left\{0, [0.5(\tanh(O_2 - 8\mu M) + 1)]b^{c \cdot \min(20^{\circ}C,T)}\right\}$$

In these equations the maximum potential growth rate of zooplankton (Eg. 28) is dependent on temperature (T) up to 20  $^{\circ}$ C, at which point it is capped, as in other models (Anderson et al., 2010), to avoid inordinately high grazing rates in the tropics. Since zooplankton are represented as a single state variable in the model, a maximum grazing rate of 0.4  $d^{-1}$  at 0 °C was chosen because it is approximately mid-way between the 20 measured growth rates of micro- and meso-zooplankton at 0  $^{\circ}$ C (0.6 and 0.24 d<sup>-1</sup>, respectively) (Le Quéré et al., 2005). Figure 2b illustrates the resulting annually averaged maximum potential zooplankton growth rate in the surface layer.

In these equations zooplankton grazing is also inhibited in hypoxic waters ( $\sim < 10 \mu M$  $O_2$ ) with grazing ceasing when  $O_2$  decreases below 5  $\mu$ M. Inhibition of zooplankton 25

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(28)

grazing in hypoxic and anoxic waters was based on observations that there is little effect of low  $O_2$  levels on the biomass of zooplankton down to  $\sim 10\,\mu\text{M}$ , at which point there are pronounced effects on species distributions and biomass (Seibel, 2011). However, since some species can survive and maintain their filtering activity at levels as low as

 $_{5}$  6  $\mu$ M O<sub>2</sub> and most euphausiids and copepods avoid only the core of oxygen minimum zones with less than 4.5  $\mu$ M O<sub>2</sub> (Ekau et al., 2010), our formulation (Eq. 28) allows some grazing to occur between 5 and 10  $\mu$ M O<sub>2</sub>.

The growth of zooplankton (1st term in Eq. 6) is determined by a growth efficiency term,  $\varpi$ , that was set at 0.4 based on measured (Kiørboe, 1989; Rivkin and Legendre, 2001) and theoretical (Landry and Calbet, 2004; Mitra, 2006) studies of zooplankton

- <sup>10</sup> 2001) and theoretical (Landry and Calbet, 2004; Mitra, 2006) studies of zooplankton growth efficiencies and production. The amount of C that they respire, and N and P that they excrete, is the difference between their assimilation and growth efficiencies ( $\gamma - \omega$ ). Their assimilation efficiency (1 –  $\gamma$ ), where  $\gamma$  has been set to 0.7 based on measured (Fenchel, 1982; Geider and Leadbeater, 1988; Hasegawa et al., 2001; Landry et al.,
- <sup>15</sup> 1984; Pelegrí et al., 1999) and theoretical (Anderson, 1994) studies of zooplankton assimilation and production, also determines the production of detritus from sloppy feeding, egestion, and fecal pellet production.

#### 4.2.3 Other modifications

Formulations of air-sea gas exchange and carbon chemistry are as described in Schmittner et al. (2008), except for a modification to the equation describing the production of  $CaCO_3$  so that it is

$$Pr(CaCO_3) = (1 - \gamma)(Graze_{P_0} + Graze_Z) + m_{P_0}P_0 + m_Z Z^2)R_{CaCO_3/POC}R_{C:P}$$
(29)

#### 5 Model evaluation

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For evaluation, and to generate a starting point for future modeling research, a model spin up of more than 10 000 yr has been performed using preindustrial boundary con-



ditions for insolation and a fixed atmospheric  $CO_2$  concentration of 283 ppm. Since the other model components have previously been evaluated in detail and published elsewhere (Meissner et al., 2003; Schmittner et al., 2005, 2008; Somes et al., 2010b; Weaver et al., 2001) the evaluation below focuses primarily on marine ecosystem dy-

- <sup>5</sup> namics and biogeochemical cycles. To evaluate how our modifications have changed the model dynamics comparisons are made with a 13 000 yr spin up of the Schmittner et al. (2008) model (this spin up is used as the control run in Oschlies et al., 2008), which is hereafter referred to as the "old model". Making these comparisons to the old model is valuable because in addition to evaluating whether the changes have
- <sup>10</sup> improved the model results, they also provide us with an idea of how the full Earth system model with the new ecosystem model component may compare to other models since the ESCM with the previous ecosystem formulation has been evaluated in several model inter-comparison projects (Archer et al., 2009; Cao et al., 2009). Comparisons are also made with observations from the World Ocean Atlas 2009 (WOA09)
- <sup>15</sup> database (Antonov et al., 2010; Garcia et al., 2010a,b; Locarini et al., 2010), the Global Data Analysis Project (GLODAP) (Key et al., 2004), and other observational studies. To make the statistical comparisons and model-observation difference plots with WOA09 and GLODAP data the simulated values were regridded onto the observed data grids in Ferret (software version 6.72) using the nearest function (@NRST). Note that since
- <sup>20</sup> both model spin ups have been performed using preindustrial conditions, comparisons with WOA09 and GLODAP data are expected to show some differences as preindustrial forcing will not account for any recent changes in global biogeochemical cycles or climate (i.e., the addition of anthropogenic CO<sub>2</sub> to the atmosphere). However, since there is little preindustrial data available for comparison and the largest anthropogeni-
- cally caused changes are likely to have occurred in the surface ocean we feel that it is still valuable to make comparisons to these data sets. Moreover, since most Earth system climate models use preindustrial spin ups as a starting point for experiments we feel that it is important to make these comparisons here before climate change or other experiments are conducted with the model.



In addition to making comparisons to annual observations we also examine how well the model simulates the seasonal cycles of plankton and certain biogeochemical tracers. As far as we are aware this type of analysis has not been conducted for any of the previous UVic ESCM marine biogeochemical models. However, it is valuable to conduct these analyses because analyzing only annual averages, as is commonly done 5 for EMICs, may hide systematic deficiencies in simulating the seasonal cycle. Since the seasonal cycle is driven by changes in environmental forcing (light, stratification, temperature, etc.), correctly simulating of the sensitivity of the marine biogeochemistry to this seasonal forcing might be viewed as an indication of whether a model can correctly simulate the sensitivity to other environmental forcing such as those driven by anthropogenic CO<sub>2</sub> emissions.

#### 5.1 **Biogeochemical tracer evaluation**

#### 5.1.1 Annual results

10

The model simulates the observed present-day distributions of multiple tracers fairly well (Figs. 4–8) with most tracer distributions similar to the old model. However, when 15 compared to the old model there are some notable differences with the new formulation slightly overestimating nutrients (phosphate and nitrate) and alkalinity in the deep ocean but simulating oxygen, apparent oxygen utilization (AOU), DIC, and C14 better in many places. Some of these differences are highlighted (Fig. 4 bottom panels) at different depths with a model-data misfit evaluation (Kriest et al., 2010 equations 20 32 and 34) that shows the absolute difference between the observations and model

- results. Simulated surface nutrient concentrations compare reasonably well with observed annual surface nutrient concentrations at lower latitudes and the new model resolves major features such as the Eastern Pacific equatorial upwelling region as well
- as the old model (Fig. 9, PO<sub>4</sub> data not shown). At higher latitudes the simulated surface 25 nutrient concentrations are often too high or too low when compared to observations. Simulated oxygen concentrations at ~ 300 m with the new model are slightly better in



the Eastern Equatorial and North Pacific regions than with the old model (Fig. 10) but similar discrepancies exist in other regions when compared to observations.

#### 5.1.2 Seasonal results

For surface nitrate and phosphate concentrations, the new model produces a seasonal
pattern that is similar, especially in terms of magnitude, to WOA09 data (Figs. 11 and 12, panels a and c). However, there is often a slight temporal mismatch between the occurrences of seasonal changes in these nutrients. Nonetheless, the new model does a much better job of simulating these seasonal changes when compared to the old model (Figs. 11 and 12, panels a and b; Fig. 13). Since surface nutrient concentrations are strongly influenced by biological activities these results, and the differences between the model formulations, are mostly due to differences in simulated seasonal biological activity at higher latitudes (see below).

#### 5.2 Ecosystem evaluation

#### 5.2.1 Annual results

- Rates of annual global net primary production (NPP) and export production with the new model are almost the same as with the old model, while the rate of new production (the fraction of integrated NPP that is not supported by nutrients from the fast reminer-alization process, zooplankton excretion, and the remineralization of detritus) is lower and nitrogen fixation and denitrification rates are a few percent higher (Table 2). These
   simulated annual global NPP rates compare well to present day estimates of annual solution of annual solution.
- global NPP (44–67 Gt C yr<sup>-1</sup>) derived from satellite measurements (Behrenfeld et al., 2005; Carr et al., 2006; Westberry et al., 2008). However, the spatial patterns of NPP in the models are noticeably different (Fig. 14), although major features such as low NPP in the oligotrophic gyres are resolved in both. With the new model (Fig. 14a) well-
- <sup>25</sup> defined areas of intense NPP are present in the Equatorial Eastern Pacific, Atlantic,



and Indian Oceans and the Western South Atlantic Ocean, while with the old model NPP is highest in broad swaths of the Equatorial Eastern Pacific and Indian Oceans (Fig. 14b). Since PP at higher latitudes varies significantly on a seasonal basis we have not made further comparisons to other estimates of PP here and instead refer

- you to Sects. 5.2.2 and 6.2 for further analysis of PP on a seasonal basis. However, we should note that aside from seasonal light limitation at high latitudes and the constraints on growth imposed by temperature and our growth-limiting iron mask (Fig. 3a), PP in both old and new models is primarily nitrogen limited with very few locations where phosphorus limits PP (data not shown).
- <sup>10</sup> The model rates of N<sub>2</sub> fixation (Table 2) are within the wide range (~ 100– 200 TgNyr<sup>-1</sup>) of global N<sub>2</sub> fixation estimates (Codispoti, 2007; Deutsch et al., 2007; Gruber and Sarmiento, 1997; Karl et al., 2002). However, as with NPP the spatial patterns of N<sub>2</sub> fixation in the models are quite different (Fig. 15). With the new model (Fig. 15 a) N<sub>2</sub> fixation is high in the Tropical/Subtropical North Pacific, the Western Trop-
- ical/Subtropical South Pacific, the Western Tropical/Subtropical South Atlantic Ocean and the Indian Ocean, while with the old model (Fig. 15b) N<sub>2</sub> fixation was high in the Central and Eastern Tropical/Subtropical Pacific. Although the distribution of diazotrophs is poorly known in many areas of the ocean due to a lack of data, the patterns of N<sub>2</sub> fixation with new model are much more consistent with observations (Sohm et al.,
- $_{20}$  2011) than with the old model. However, as discussed by Somes et al. (2010a), who produced a similar distribution of  $N_2$  fixation with the old model when they constrained diazotroph growth with an Fe mask,  $N_2$  fixation does not extend as far north as it should in the North Pacific. Moreover, there is too much  $N_2$  fixation in the South Atlantic and not enough in the North Atlantic, where some of the highest rates of  $N_2$  fixation have
- <sup>25</sup> consistently been measured (Capone et al., 2005; Sohm et al., 2011). In the Indian Ocean the model also simulates too much N<sub>2</sub> fixation in the Bay of Bengal and not enough in the Arabian Sea.

Recent global estimates of denitrification (defined here as the ensemble of biological processes that can convert fixed-N to  $N_2$ ) vary widely, are poorly constrained, and the



subject of considerable debate (Codispoti, 2007; Gruber, 2004), which makes it difficult to validate this aspect of the model results. Furthermore, since the model does not include benthic denitrification and, in the steady state essentially reached at the end of the spin up, N<sub>2</sub> fixation must be approximately equal to denitrification even in the absence of benthic denitrification. Thus, our model estimate of water-column denitrification (Table 2) are the same as the 150 TgNyr<sup>-1</sup> suggest by Codispoti et al. (2001) and Codispoti (2007), but are higher than the 50(±20) and 80(±20) TgNyr<sup>-1</sup> suggested by Gruber and Sarmiento (2002) and Gruber (2004). As with other model results, the location of water column denitrification with the new model is somewhat different than with the old model (Fig. 16). The new model simulates distinct areas of denitrification in the Eastern Tropical North and South Pacific with the most intense denitrification in the Eastern Tropical North and South Pacific occurred farther offshore in a broad area

- <sup>15</sup> with the most intense denitrification occurring around the Galapagos Islands (Fig. 16b). In both model versions denitrification also occurs in the Bay of Bengal and off the coast of Namibia, Africa. Additionally, with the new model denitrification occurs in the Gulf of Aden as well. Validating whether the model simulates water column denitrification in the correct locations is difficult since observations of denitrification are sparse and the
- <sup>20</sup> spatial extent of denitrification zones is often estimated by using many assumptions, and an often limited set of phosphate and nitrate observations, to calculate if a nitrate deficit occurs (Codispoti, 2007; Kamykowski and Zentara, 1990; Paulmier and Ruiz-Pino, 2009). However, it is generally agreed that the three major quasi-permanent sites of water column denitrification are within the suboxic portions of the thermocline in the
- Arabian Sea, the Eastern Tropical North Pacific, and the Eastern Tropical South Pacific off the coast of Chile (Codispoti, 2007). Of these major denitrification sites the model is only able to reasonably simulate the Eastern Tropical North Pacific one. In the Eastern Tropical South Pacific modeled denitrification is too close to the equator and does not occur off the coast of Chile where it should. In the Indian Ocean the model does not



simulate denitrification in the Arabian Sea and instead has it occurring in the Bay of Bengal and the Gulf of Aden because the simulated Indian Ocean oxygen minimum zone is in the wrong location (Fig. 10). While some denitrification has been suggest to occur in the Bay of Bengal and the Gulf of Aden (Paulmier and Ruiz-Pino, 2009),
 <sup>5</sup> minor amounts of denitrification occur in these areas when compared to the Arabian Sea. An overestimate of denitrification in the Bay of Bengal and an underestimate in the Arabian Sea has also been found in other coarse-resolution global biogeochemical models (Moore and Doney, 2007).

#### 5.2.2 Seasonal results

- <sup>10</sup> The new model formulation simulates much stronger seasonal ecosystem cycles when compared to the previous one (Figs. 17, 18a, b, and 19). Spring phytoplankton blooms, both in terms of biomass and productivity, are quite evident in both the Northern and Southern Hemispheres with the new model formulation (see Sect. 6.2 for more discussion on productivity). The new model does especially well in simulating the North Atlantic spring phytoplankton bloom (Fig. 19), which was almost inexistent with the old model, and which is known to be the most pronounced spring bloom of any open ocean region (Yoder et al., 1993). The seasonal cycle of zooplankton, in terms of
- biomass, is also more realistic with the new model formulation simulating a clear "classic" lag between the peak in phytoplankton biomass and that of zooplankton during
- the spring/summer at mid to high latitudes (Fig. 17). Diazotroph biomass also varies seasonally with the new model formulation, which is in contrast to the fairly constant presence of diazotrophs with the old model (Fig. 17c, d). Although data on the seasonal abundance of diazotrophs is limited, the available data sets suggest that diazotroph biomass does vary seasonally in many regions (Sohm et al., 2011), as it does in our new model.



#### 6 Discussion

#### 6.1 Annual and seasonal particulate fluxes

Since the export flux of particulate matter out of the euphotic zone is a primary driver of marine biogeochemical cycles and the biological pump we now discuss how well it is simulated by the models. For both the old and new model configuration, annual 5 estimates of export production and CaCO<sub>3</sub> export out of the euphotic zone (Table 2) are within the observational estimates (5.8–13 and 0.38–1.64 Gt C  $yr^{-1}$ , respectively) of a number of studies (listed in Table 3 in Dunne et al., 2007) The spatial patterns of our annual simulated fluxes (data not shown) are, for both models, similar to those of primary production (Fig. 14) with the highest export occurring below or near areas 10 of high productivity. To evaluate how well the model simulates carbon sequestration in the deep ocean and understand and how much upper layer export reaches it versus what is remineralized, simulated annual particulate organic carbon (POC) and CaCO<sub>3</sub> fluxes are compared (Figs. 20 and 21, Table 3) to the sediment trap data of Honjo et al. (2008), which is normalized to evaluate these fluxes at ~ 2 km. Although neither model 15 formulation performs particularly well, the new model is slightly better at simulating the flux of POC to the deep ocean (Fig. 20 RMSE value). Globally, the simulated flux of POC at about 2 km depth with the new model is lower than the observationally based estimates, while the flux of CaCO<sub>3</sub> (i.e., PIC) is higher (Table 3). Additionally, our simulated rain ratio at 2 km is lower than observations suggest. The simulated transfer 20 efficiency (the ratio of the POC flux at 2 km to export production) is also lower than estimated by Honjo et al. (2008), while the change in POC (the difference between export production and the POC flux at 2 km) is higher. However, Honjo et al. (2008) calculated export production using the model of Laws et al. (2000) and obtained a value that is

<sup>25</sup> lower than most other estimates (Dunne et al., 2007). If a higher export production value of 9.6 Gt C yr<sup>-1</sup> (from Dunne et al., 2007) is used to recalculate the transfer efficiency and the change in POC with the same deep ocean POC flux data of Honjo



et al. (2008), then our transfer efficiency is higher and our change in POC is lower than the deep ocean flux data suggests (Table 3). Uncertainties in the observations are relatively large because there exists only a limited amount of flux data, with possible biases in the data, and uncertainty in extrapolating sparse sediment trap data to

the global ocean, as well as uncertainty in estimates of export production derived from various models. The fact that our simulated annual global flux values are close to observational estimates, together with the relatively good agreement with mean profiles of biogeochemical tracers, suggests that the model does a reasonable job of simulating the global biological carbon pump with the associated impacts on nutrient and oxygen distributions, despite some regional biases and discrepancies.

Although both the new and old models have very similar annual global POC fluxes at 2 km (Fig. 20 and Table 2) their simulated seasonal fluxes of POC differ considerably (Fig. 22) due to the differences in seasonal ecosystem cycles. However, despite these differences in both models the peaks in POC flux occur approximately one to two

- <sup>15</sup> months after the seasonal peaks in primary production (Fig. 18a, b), which is consistent with observations that maximum flux timings are generally synchronous with the timing of blooms in temperate and polar waters and have a production-to-flux lag that is typically between 40 and 80 days (Lutz et al., 2007). The formulation of the downward CaCO<sub>3</sub> flux assumes instantaneous export and remineralization in both model
- <sup>20</sup> configurations. The seasonal fluxes of CaCO<sub>3</sub> are thus, at any depth, tightly and with zero phase shift linked to CaCO<sub>3</sub> production in the surface ocean, which is in turn proportional to the production of detritus. As a consequence, deep CaCO<sub>3</sub> fluxes with the new model (Fig. 23a) occur approximately a month earlier than the highest POC fluxes, which results in a seasonally variable rain ratio at depth, particularly at high latitudes
- <sup>25</sup> (Fig. 23b). While this is consistent with the fact that calcite and aragonite are some of the heaviest biominerals found in ocean water and thus sink more rapidly than POC, it is doubtful that in reality  $CaCO_3$  fluxes and lag times are as uniform as in the simulations since the size of the source material (i.e., coccoliths versus foraminiferals tests or pteropod shells) and its production, which is determined by the composition of the



community and thus not accounted for in the model, will play a large role in determining its sinking speed and export to depth (Honjo et al., 2008).

### 6.2 Primary production

A closer examination of PP with the new model shows that while seasonal PP at higher
latitudes compares well with some estimates (i.e., Fig. 19), simulated PP is too high in the tropical upwelling regions along the equator and the Northern Indian Ocean with both the new and old models when compared to satellite-based estimates (Fig. 18) and observations (data not shown) of PP such as those made by Balch et al. (2011) in the Equatorial Pacific and Marañón et al. (2000) in the Equatorial Atlantic. Furthermore, with either model there is too little productivity in oligotrophic regions when compared to the satellite-based models (Fig. 18) and observations (i.e., such as those made by Marañón et al., 2000; Tilstone et al., 2009 in the Atlantic and by Van Wambeke et al., 2008 in the South Pacific). However, it is worth noting that the satellite-based estimates of PP do not agree very well with each other (Figs. 18c, d, 19 right panels) and

- thus caution must be taken when making comparisons to these models. Furthermore, satellite-based estimates of PP are inherently based on models, so that this type of assessment essentially makes model-to-model comparisons, which cannot be considered true validation. Nonetheless, there are systematic differences between our model and estimates of PP and a number of reasons for these differences. First, the model
- <sup>20</sup> parameters were set so that the total annual global PP was within the range of estimates (i.e., 44–67 Gt C yr<sup>-1</sup>; from Table 2 in Westberry et al., 2008). Setting the model parameters to achieve this level of annual global PP may be problematic because these satellite-based estimates include coastal PP, which can be quite high and accounts for a significant amount of the estimated total. Since the UVic ESCM model has a course
- $_{25}$  grid resolution (1.8 × 3.6°) the coastal regions are poorly represented, if at all in many places, in the marine ecosystem model. Thus, simulated open ocean productivity becomes much too high when the model is parameterized to simulate levels of annual global PP that include coastal PP. This problem does not appear to be limited to the



UVic ESCM as a number of other global marine ecosystem models (Moore et al., 2004; Vichi and Masina, 2009; Yool et al., 2011) also simulate higher than observed PP in many regions (especially the equatorial upwelling regions) to achieve estimated annual PP within the range of 40 to  $67 \text{ PgCyr}^{-1}$ . Furthermore, this problem is not limited to estimates of global PP as Dunne et al. (2007) noted that a similar problem exists in 5 calculations of particle export and burial which use satellite-based PP data that is regridded onto coarser grids excluding coastal areas. The second reason for model and observational differences is the use of invariant half-saturation constants for nutrient uptake, which prevent phytoplankton from adapting to low nutrient concentrations in oligotrophic regions and thus severely limits their growth in these areas (e.g., Smith 10 et al., 2009). Finally, the dissolved iron mask that we use to constrain phytoplankton growth may be inaccurate as it is generated by another model (i.e., from the BLING model, Galbraith et al., 2010) and not well validated due to a lack of observations in many ocean regions. Moreover, we do not explicitly model the cycling of iron or it's uptake by phytoplankton, which may be necessary to correctly simulate the effects of

<sup>15</sup> uptake by phytoplankton, which may be necessary to correctly simulate the effec iron limitation on phytoplankton growth.

### 6.3 Grazing

Grazing related processes are quite different in the two model versions (Fig. 24). The annual percentage of primary production that is grazed daily is much higher in most regions with new model when compared to the old one (Fig. 24a, b). The notable exception to this trend is in the Southern Ocean south of 60° S where it is not surprising that grazing is lower in the new model since the maximum potential grazing rate is now a function of temperature and much lower in this region (Fig. 3b) than in with the old model where it was set globally at 1.5 d<sup>-1</sup>. The differences between the model versions are also particularly striking in the oligotrophic gyres where almost no grazing occurred with the old model because there were essentially no zooplankton present in these regions (data not shown). Validating how much primary production is lost to grazing is difficult to do because there are few observations to constrain this



trophic transfer on a global scale and it can vary significantly regionally and seasonally. Nonetheless, global estimates suggest that mesozooplankton consume between ~ 10–15% of oceanic PP (Behrenfeld and Falkowski, 1997; Calbet, 2001) and that microzooplankton consume 59–75% of it (Calbet and Landry, 2004). Grazing with the new model, where an average of 71% of PP is consumed by zooplankton, falls within the range of these estimates, while with the old model an average of only 52% of PP was consumed which is below the lowest estimate for microzooplankton grazing alone. Furthermore, with the new model the ratio of zooplankton to phytoplankton biomass (integrated over the upper 500 m), which is strongly influenced by grazing, is much higher than with the old model (Table 2) and more consistent with observations of high heterotrophic to autotrophic biomass ratios in the open ocean (Gasol et al., 1997).

#### 6.4 Nutrient regeneration

The annually averaged remineralization rates of inorganic nutrients by zooplankton (i.e., excretion) are quite different in the two model versions (Fig. 24c, d). Most of the remineralization with the old model occurs in the tropics, which is not surprising since with this model formulation zooplankton excretion is dependent only on biomass and temperature, whereas ingestion is temperature independent. Remineralization with the new model, where it is now a function of grazing, occurs primarily in areas where grazing is high with the spatial pattern of remineralization in Fig. 24c looking much like the

- pattern of primary production in Fig. 14a. The differences between the model versions are also due to the effects of the new grazing formulation on the pathways of nutrient remineralization. With the new grazing formulation more nutrients are remineralized through excretion than with the old model formulation where the fast recycling loop (i.e., implicit remineralization of organic matter by bacteria) was a more important path-
- way (Fig. 25). The relative importance of one pathway versus another in the euphotic zone is difficult to quantify on a global scale since it depends on many factors that vary both temporally and spatially. Furthermore, neither model explicitly includes bacteria or dissolved organic matter, which may be necessary to effectively simulate bacterial



remineralization. Nonetheless, there is considerable evidence that zooplankton play a more important role in nutrient regeneration than the old model formulation allows for (Banse, 1995; Ferrier-Pages and Rassoulzadegan, 1994; Glibert, 1998).

#### 7 Summary and conclusions

- Overall, the new ecosystem model formulation has achieved our goals of improving the realism of simulated biological processes and biogeochemical cycles. When evaluated against observational data the majority of the results show an improvement over the old model. Many of these improvements can be attributed to a better simulation of higher latitude ecosystem seasonality and the effect that it has on biogeochemical cycles. The
   important modifications include:
  - The inclusion of diazotroph biomass in calculations of light attenuation and phytoplankton light limitation.
  - Modifications to the zooplankton growth and grazing formulation to add more realistic dynamics.
- The use of a dissolved iron mask to constrain phytoplankton growth in iron-limited regions of the ocean.
  - Changes to the parameters governing phytoplankton and zooplankton growth (i.e., maximum growth rates) to achieve more realistic rates and better responses to seasonal forcing.
- Despite these modifications there are several aspects of the model that might be improved in the future. One of the most important issues that needs further consideration is the discrepancy between modeled PP and observational estimates. A number of factors would need to be addressed to improve simulated PP. First, there is the decision of what levels of annual PP should be accounted for in a model that does not resolve



coastal waters and the significant amount of PP that occurs in them. Perhaps models should simulate lower levels of global primary productivity instead of trying to simulated global rates that are as high as satellite estimates. Including coastal areas or accounting for the flux of material from them may also be necessary, although this would involve

- a significant modification of the model grid, physics, and boundary conditions. Perhaps 5 including variable half-saturation constants for nutrient uptake or introducing more phytoplankton types may also be necessary to better simulate PP (e.g., Smith et al., 2009). However, any of these options introduces other problems, which have been discussed extensively by others (Anderson, 2005; Flynn, 2005; Hood et al., 2006), and may not
- necessarily improve the simulation of biogeochemical cycles. Furthermore, the mask of dissolved iron concentrations that we use to constrain phytoplankton growth is derived from a model and could be improved to help address issues of where and how much primary productivity occurs. This limiting factor could be improved either by including a dynamic iron cycling model within the ecosystem model or by obtaining a better mask
- as more measurements of dissolved iron become available. Regardless, these issues are beyond the goals we set out to achieve with the improvements made here and can be addressed in the future. Our modifications have significantly improved the marine ecosystem model and provided a better platform for future research with the UVic Earth System Climate Model.
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#### Table 1. Model parameters.

Parameter	Symbol	Value	Units
Phytoplankton ( $P_{\Omega}$ , $P_{D}$ ) coefficients			
Maximum growth rate at 0 °C	а	0.6	d <sup>-1</sup>
Other phytoplankton mortality rate	m <sub>Po</sub>	0.03	d <sup>-1</sup>
Diazotroph mortality rate	m <sub>Pp</sub>	0.015	d <sup>-1</sup>
Fast recycling term (microbial loop)	$\mu_{P}^{*}$	0.015	d <sup>-1</sup>
Diazotrophs' handicap	$c_{D}$	0.4	Dimensionless
Half-sat. constant for $P_{O}$ Fe limitation	$K_{\rm Fe}^P$	0.1	nmol Fe $\mathrm{m}^{-3}$
Half-sat. constant for $P_{\rm D}$ Fe limitation	$K_{Fe}^{D}$	0.1	nmol Fe m $^{-3}$
Zooplankton (Z) coefficients			
Maximum growth rate at 0 °C	$\mu_{z}^{\theta}$	0.4	d <sup>-1</sup>
Maximum growth rate parameters	b	1.066	Dimensionless
	С	1.0	(°C) <sup>-1</sup>
Growth efficiency	arpi	0.4	Dimensionless
Assimilation efficiency	γ	0.7	Dimensionless
Mortality rate	mz	0.06	$d^{-1}$
Preference for grazing on $P_{O}$	$\psi_{P_{ m O}}$	0.3	Dimensionless
Preference for grazing on $P_{\rm D}$	$\psi_{P_{ m D}}$	0.1	Dimensionless
Preference for grazing on D	$\psi_{D}$	0.3	Dimensionless
Preference for self-predation	$\psi_{Z}$	0.3	Dimensionless
Half-sat. constant for grazing	K <sub>Graze</sub>	0.15	mmol N m <sup>-s</sup>
Other coefficients			
Detritus sinking speed at surface	w <sub>D0</sub>	14	$m d^{-1}$
CaCO <sub>3</sub> over nonphotosynthetical POC production ratio	$R_{CaCO_3:POC}$	0.03	Dimensionless
CaCO <sub>3</sub> remineralization e-folding depth	$D_{CaCO_3}$	6500	m
Molar C : N ratio of organic matter	R <sub>C:N</sub>	6.625	Dimensionless
Molar $O_2$ : N ratio of aerobic remineralization	R <sub>O:N</sub>	10.0	Dimensionless
Remineralization rate at 0°C	$\mu_{D0}$	0.055	d-'



Property	New model	Old model	Units
Primary production	52	54	Gt C yr <sup>-1</sup>
Nitrogen fixation	150	141	Tg N yr <sup>-1</sup>
Denitrification	150	141	Tg N yr <sup>−1</sup>
New production	7.48	8.66	Gt C yr <sup>-1</sup>
Export production at 130 m	7.03	7.12	Gt C yr <sup>-1</sup>
CaCO <sub>3</sub> export at 130 m	0.83	0.72*	Gt C yr <sup>-1</sup>
Rain ratio at 130 m	0.12	0.10	Molar ratio
POC flux at 2 km	0.36	0.35	Gt C yr <sup>-1</sup>
$CaCO_3$ flux at 2 km	0.49	0.47*	Gt C yr <sup>-1</sup>
Rain ratio at 2 km	0.73	0.74	Molar ratio
Zoo : phytoplankton biomass (upper 50 m)	0.63	0.75	Molar ratio
Zoo : phytoplankton biomass (upper 500 m)	0.94	0.67	Molar ratio

 $^{*}$   $D_{CaCO_3}$  was set at 4500 m which differs from the value of 3500 m used in Schmittner et al. (2008).

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Table 3. Comparison of critical properties that drive the global biological pump.

Model	Honjo et al. (2008)	Units
7.03	5.73	Gt C yr <sup>-1</sup>
0.36	$0.43 \pm 0.05$	Gt C yr <sup>-1</sup>
0.49	$0.41 \pm 0.05$	Gt C yr <sup>-1</sup>
5.1	7.6; 4.4*	%
6.67	5.3; 9.17*	Gt C yr <sup>-1</sup>
0.73	1.05	Molar ratio
99.92	120	mmol C m <sup>-2</sup> yr
136	112	mmol C m <sup>-2</sup> yr
	Model 7.03 0.36 0.49 5.1 6.67 0.73 99.92 136	ModelHonjo et al. (2008) $7.03$ $5.73$ $0.36$ $0.43 \pm 0.05$ $0.49$ $0.41 \pm 0.05$ $5.1$ $7.6; 4.4^*$ $6.67$ $5.3; 9.17^*$ $0.73$ $1.05$ $99.92$ $120$ $136$ $112$

\* Recalculated using an export production value of  $9.6 \,\mathrm{GtCyr^{-1}}$  from Dunne et al. (2007).



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**Fig. 1.** Ecosystem model schematic which illustrates the flux (arrows) of material between model variables (squares). See text for a detailed description of these fluxes.











Fig. 3. Examples of the seasonably variable dissolved iron concentrations used to constrain phytoplankton growth. Examples shown are for the surface layer in (a) January and (b) June.





**Fig. 4.** Top panels: Global and basin-wide averaged vertical profiles of various model tracers (solid lines) compared with observations (squares) and the previous model formulation (dotted lines, global comparisons only). Bottom panel: The misfits between the globally averaged vertical model profiles (solid line: new formulation, dotted line: previous formulation) and observations.







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Fig. 5. Zonally averaged ocean basin comparisons of PO<sub>4</sub> between the new and old models and the World Ocean Atlas 2009 data set.



























Fig. 10. Annual oxygen concentrations at ~ 300 m simulated with the new (a) and old (b) models and from the WOA09 database (c).



**Fig. 11.** Hovmöller diagrams of the zonally averaged seasonal changes in surface nitrate (monthly climatology minus annual climatology) with the new **(a)** and old **(b)** models and from the WOA09 database **(c)**.





**Fig. 12.** Hovmöller diagrams of the zonally averaged seasonal changes in surface phosphate (monthly climatology minus annual climatology) with the new **(a)** and old **(b)** models and from the WOA09 database **(c)**.





**Fig. 13.** Root mean squared error (RMSE) comparisons between the new and old models and WOA09 monthly (a) nitrate and (b) phosphate data. Zonal statistics were calculated only for grid cells where observations were present and then averaged longitudinally to give a global value.































**Fig. 18.** Hovmöller diagrams of the zonally averaged seasonal changes in integrated primary production as simulated by the **(a)** new and **(b)** old models and from **(c)** Behrenfeld and Falkowski's (1997) vertically generalized production model (VGPM) and **(d)** Westberry et al.'s (2008) carbon-based productivity model (CbPM).



**Fig. 19.** Spring-time primary production comparisons in the North Atlantic between the new and old model simulations and estimates from Behrenfeld and Falkowski's (1997) vertically generalized production model (VGPM) and Westberry et al.'s (2008) carbon-based productivity model (CbPM).





**Fig. 20.** Simulated fluxes of POC at 2 km with the **(a)** new and **(b)** old models. Observed fluxes from Honjo et al. (2008) are plotted as filled circles. Statistical comparisons with the observations are shown in **(c)** and **(d)** for the new and old model versions, respectively.





**Fig. 21.** The simulated fluxes of **(a)** PIC and the **(b)** rain ratios at 2 km. Observed PIC fluxes and rain ratios (Honjo et al., 2008) are plotted as filled circles. Statistical comparisons with the observations are shown in **(c)** and **(d)**.

















**Fig. 24.** Annual vertically integrated percentage of primary production grazed per day (top panels) and zooplankton excretion (bottom panels) with the (**a** and **c**) new and (**b** and **d**) old model formulations.





**Fig. 25.** Comparison between model versions of the pathways of nitrate regeneration. Values are from globally integrated data.

