IRON IN THE SOUTHERN OCEAN
A model study of iron sources and their impact on the phytoplankton

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ERKLÄRUNG

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Ebenfalls erkläre ich hiermit, dass es sich bei den von mir abgegebenen Arbeiten um drei identische Exemplare handelt.

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(Unterschrift)
ABSTRACT

In the iron limited Southern Ocean, the iron sources have a large impact on the spatial distribution and magnitude of the phytoplankton blooms, and thereby also on the net primary and export production (NPP and EP). In this thesis, the Southern Ocean input of iron from the sediments and from vertical supply to the mixed layer is investigated in a new Ocean General Circulation Biogeochemical Model (OGCBM). Focus is especially on the model’s representation of the iron sources, how they affect the biological production in the model and what can be learned from this about the iron sources in nature.

The first part of the thesis documents the global performance of the Regulated Ecosystem Model (REcoM$^2$) coupled to the Finite Element Sea-ice Ocean Model (FESOM). It shows that the model does a reasonable job reproducing the large scale surface patterns of the biogeochemical fields when compared to observations and other non-eddy resolving OGCBMs. The spatial fields of chlorophyll $a$ and NPP are especially good. Further analysis of the Southern Ocean shows that the total net primary and export production here fits well with previous estimates.

The second part of the thesis assesses the impact of the Ocean General Circulation Model (OGCM) on the vertical iron supply to the mixed layer of the Southern Ocean. Two similar model runs were carried out with REcoM$^2$ coupled to two different ocean models; FESOM and MITgcm. The study shows that the modeled vertical iron fluxes, and thereby the NPP and EP, are strongly dependent on the ocean model. This is partly due to differences in the mixed layer dynamics, but is also affected by other mechanisms, such as the strength of the meridional overturning circulation. The study highlights the importance of considering the skill of the OGCM forcing the biogeochemistry. This is important for contemporary runs, but especially when predicting future changes to the biological pump using large scale OGCBMs.

The rate of iron release from the sediments in nature is not well constrained, leading to large variations in its strength between biogeochemical models. The third part of the thesis focuses on the implications of changes in the modeled strength of the sediment source, and accompanying higher scavenging rates, on the phytoplankton composition and the opal export. The study shows, that while variations in the strength of the benthic iron flux can be counteracted by an appropriate scavenging rate, thereby producing very similar values for the total NPP in the Southern Ocean, it has a large impact on the spatial distribution of the NPP, on the diatoms and on the opal export. In order to optimize the sedimentary iron input to large scale OGCBMs, it is thus important to look beyond the modeled NPP and assess the biological production by diatoms, as well as the vertical export of opal, especially for the Southern Ocean.

The general discussion focuses on the relative importance of the different iron sources in the Southern Ocean, how the different sources may change in the future and what factors modelers need consider when predicting these changes. Lastly, further directions for work regarding the iron sources in the Southern Ocean as well as the new FESOM-REcoM$^2$ model are described.
ZUSAMMENFASSUNG

Im eisenlimitierten Südpolarmeer haben Eisenquellen einen großen Einfluss auf die räumliche Verteilung und die Stärke von Phytoplanktonblüten, und dadurch auch auf die Nettprimär- und Exportproduktion (NPP und EP). In dieser Arbeit wird der Eiseneintrag in die Deckschicht des Südpolarmeers aus Sedimenten und durch vertikalen Eintrag mit einem neuartigen Ozeanzirkulations- und Biogeochemie-Modell (OGCBM) untersucht. Der Fokus liegt hierbei besonders auf der Repräsentation der Eisenquellen im Modell, wie diese die biologische Produktion im Modell beeinflussen und was wir daraus über die Eisenquellen in der Natur lernen können.


Die Rate, mit der Eisen in der Natur aus Sedimenten freigesetzt wird, ist kaum bekannt, was dazu führt, dass sie auch zwischen verschiedenen biogeochemischen Modellen stark variiert. Der dritte Teil der Arbeit beschäftigt sich mit den Auswirkungen unterschiedlich starker Eisenfreisetzung und den dazugehörigen größeren Adsorptionsraten auf Partikeleroberflächen auf die Phytoplanktonzusammensetzung und den Opalexport. Die Studie zeigt, dass Variationen in der Stärke des bentischen Eiseneintrags durch eine geeignete Adsorptionsrate auf Partikeleroberflächen teilweise kompensiert werden, so dass trotzdem ähnliche Werte für die Gesamt-NPP im Südpolarmeer erreicht werden. Der bentische Eiseneintrag hat einen großen Einfluss auf die räumliche Verteilung der NPP, auf Diatomeen und auf den Opalexport. Um den sedimentären Eiseneintrag in großskaligen OGCBMs zu verbessern, ist es wichtig, nicht nur die simulierte NPP zu beurteilen, sondern auch die biologische Produktivität von Diatomeen und den vertikalen Opalexport im Südpolarmeer zu analysieren.
Die allgemeine Diskussion am Ende dieser Arbeit konzentriert sich auf die relative Wichtigkeit der verschiedenen Eisenquellen im Südpolarmeer, in welcher Weise diese sich in der Zukunft ändern könnten, und welche Prozesse Ökosystemmodellierer für die Vorhersage dieser Änderungen berücksichtigen müssen. Darüber hinaus wird ein Ausblick auf weiterführende Arbeiten über die Eisenquellen im Südpolarmeer und das neuartige Modell FESOM-REcoM2 gegeben.
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Throughout my time at the AWI have been part of two working groups; Biogeo-science and Climate Dynamics. I would like to thank group members of both groups for welcoming me in them, allowing me to take advantage of both the diverse knowledge that the groups represent, but also the friendships and support that has come from it.

Christine Klaas and Dieter Wolf-Gladrow gave me the opportunity to join the RV Polarstern on a cruise in the Southern Ocean, giving me the opportunity to experience first hand how measurements of the marine biogeochemistry are obtained and how much work goes into it. It highly appreciate this opportunity and experience.

Last, but not least, my work at the AWI would never have been possible had it not been for the support and understanding of my family, and especially from my boyfriend Esben. Thank you for giving me the chance to follow my dreams by supporting our move to a new country and for putting in the extra work required to keep us all happy and healthy. Jeg elsker dig.
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<tr>
<td>ACC</td>
<td>Antarctic Circumpolar Current</td>
</tr>
<tr>
<td>AADW</td>
<td>Antarctic Deep Water</td>
</tr>
<tr>
<td>AAIW</td>
<td>Antarctic Intermediate Water</td>
</tr>
<tr>
<td>CDW</td>
<td>Circumpolar Deep Water</td>
</tr>
<tr>
<td>CORE</td>
<td>Coordinated Ocean-ice Reference Experiment</td>
</tr>
<tr>
<td>DFe</td>
<td>Dissolved Iron</td>
</tr>
<tr>
<td>DIC</td>
<td>Dissolved Inorganic Carbon</td>
</tr>
<tr>
<td>DIN</td>
<td>Dissolved Inorganic Nitrogen</td>
</tr>
<tr>
<td>DSi</td>
<td>Dissolved Inorganic Silicon</td>
</tr>
<tr>
<td>EP</td>
<td>Export Production</td>
</tr>
<tr>
<td>FESOM</td>
<td>Finite Element Sea-ice Ocean Model</td>
</tr>
<tr>
<td>HNLC</td>
<td>High Nutrient Low Chlorophyll</td>
</tr>
<tr>
<td>MITgcm</td>
<td>Massachusetts Institute of Technology general circulation model</td>
</tr>
<tr>
<td>MLD</td>
<td>Mixed Layer Depth</td>
</tr>
<tr>
<td>NADW</td>
<td>North Atlantic Deep Water</td>
</tr>
<tr>
<td>NPP</td>
<td>Net Primary Production</td>
</tr>
<tr>
<td>OGCBM</td>
<td>Ocean General Circulation Biogeochemical Model</td>
</tr>
<tr>
<td>OGCM</td>
<td>Ocean General Circulation Model</td>
</tr>
<tr>
<td>PF</td>
<td>Polar Front</td>
</tr>
<tr>
<td>REcoM</td>
<td>Regulated Ecosystem Model</td>
</tr>
<tr>
<td>REcoM2</td>
<td>Regulated Ecosystem Model with two phytoplankton classes</td>
</tr>
<tr>
<td>SACCF</td>
<td>Southern Antarctic Circumpolar Current Front</td>
</tr>
<tr>
<td>SAF</td>
<td>Subantarctic Front</td>
</tr>
<tr>
<td>SAM</td>
<td>Southern Annular Mode</td>
</tr>
<tr>
<td>TA</td>
<td>Total Alkalinity</td>
</tr>
<tr>
<td>TICZ</td>
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</table>
Part I

INTRODUCTION
1.1 Oceanographic Setting

The zonally unrestricted flow of the oceanic and atmospheric circulation is a unique property of the Southern Ocean (Fig. 1.1), allowing the strength of the Westerlies, approximately located between 45 and 55°S (Trenberth et al., 1990), to develop unhindered. This atmospheric flow drives the eastwards running Antarctic Circumpolar Current (ACC), which exists in geostrophic balance between the meridional pressure gradient and the Coriolis force. It encircles the Antarctica with an annual mean transport of approximately 130 Sv at Drake Passage (1 Sv = 1 × 10⁶ m³ s⁻¹, Cunningham et al., 2003). The ACC consists of a series of interwoven currents with multiple branches (Sokolov and Rintoul, 2009), making the flow highly complex. Three main fronts can nevertheless be identified: the Subantarctic Front (SAF), the Antarctic Polar Front (PF) and the Southern ACC Front (SACCF) (Orsi et al., 1995). Towards the north the ACC is limited by the SAF, which separates warmer saltier water in the north from the colder and fresher water towards the south (Olbers et al., 2004). The SACCF marks the southern boundary of the ACC (e.g. Orsi et al., 1995), and is frequently observed at Drake passage but does not encircle the Antarctic. The SAF and the PF, are characterized by strong zonal jets of elevated speed, as well as strong meridional gradients for temperature, density and other oceanographic properties.

The meridional gradients characterizing the area between the PF and SAF are brought on by the fact that the isopycnals rise steeply in the region (Rintoul and Trull, 2001). The southwards transported Circumpolar Deep Water (CDW) and North Atlantic Deep Water (NADW) consequently rises towards the surface south of the PF (Fig. 1.2), bringing the water properties of the deep water upwards here (Marshall and Speer, 2012). The shoaling water follows two paths dependent on density. The less dense water is transported northwards in a wind-driven Ekman flow, which is subducted north of the Antarctic PF, marking an important boundary for heat, salt and carbon fluxes. The denser shoaling water is converted to Antarctic Bottom Water (e.g. Gill, 1973; Rintoul, 1998), which is created along the ice shelves and is the densest water found in the world ocean (Orsi et al., 1999).

The steep density slopes between the PF and the SAF mean that the water column is weakly stratified, and thus easily destabilized. The area is therefore characterized by deep winter mixed layers and large seasonal variations of the mixed layer depth (MLD, de Boyer Montegut et al., 2004). The deep winter mixing, and associated subduction of Antarctic Intermediate Water (AAIW, Fig. 1.2), is key for the ventilation of the intermediate ocean, transferring physical and biogeochemical properties between the surface mixed layer and the water below (Rintoul and Trull, 2001).

The combination of the location of the Southern Ocean, the southwards shoaling pycnoclines and the deep mixed layers means that the Southern Ocean is an important area for transfer of properties such as heat, salt and biogeochemical tracers both between the southern part of the Atlantic, Indian and Pacific Oceans, and between the atmosphere and the ocean. It is therefore key to understand how the biogeochem-
Figure 1.1: Map of the ocean south of 30° S. Green arrows denote the strength and direction of the surface currents. The mean position of the Polar Front (PF) and Subantarctic Front (SAF) are marked in orange. The mean winter and summer ice extend is marked with black lines. The background color denotes depth as shown in the colorbar (Modified from Marshall and Speer, 2012)
Figure 1.2: Schematic of the southern limb of the meridional overturning circulation. The meridional direction of the water flow is marked with black and blue arrows. The direction of the ACC is marked with a bold yellow arrow and the location of the ACC with a dotted orange line. The main water masses are marked: Antarctic Deep Water (AADW), Lower Circumpolar Deep Water (LCDW), Upper Circumpolar Deep Water (CDW), Antarctic Intermediate Water (AAIW) and North Atlantic Deep Water (NADW). The direction of heat exchange between the atmosphere and the ocean is marked with vertical curly arrows. (From Olbers et al., 2012, Fig. 16.3)

1.2 THE SOUTHERN OCEAN; AN HNLC AREA

The southward shoaling of the isopycnals in the Southern Ocean (Fig. 1.2), brings nutrient-rich water towards the surface through isopycnal and diapycnal transport (Pollard et al., 2002). But despite of a relatively high supply of macronutrients to the surface water in the Southern Ocean, they are not fully utilized by phytoplankton growth. The combination of high macronutrient concentrations and low chlorophyll levels has led to the Southern Ocean being categorized as one of the world’s so-called High-Nutrient-Low-Chlorophyll (HNLC) areas (Martin et al., 1990a), a definition that it shares with the subarctic and equatorial Pacific (Martin and Fitzwater, 1988; Kolber et al., 1994). Of these areas, the Southern Ocean is by far the largest in area and in terms of the concentration of unused macronutrients (e.g. Sarmiento and Orr, 1991).

Since the early 1990’s it has been recognized that limitation by the micronutrient iron, which is needed in the photosynthesis as well as the respiration apparatus of phytoplankton (Geider and La Roche, 1994), plays a large role in the low productivity (e.g. Martin et al., 1990a; Martin et al., 1990b). As is the case for the macronutrients, iron is supplied to the the surface water through upwelling, but due to the complicated iron chemistry in the water (See section 1.4.1), the iron-to-macronutrient ratio is too low to sustain production in the Southern Ocean. A number of experiments, ranging from laboratory to large-scale iron fertilization, have later been carried out and demonstrated that iron does indeed induce growth in the HNLC areas (e.g. Smetacek...
et al., 2012). But other factors, such as light limitation (Mitchell et al., 1991) and grazing (Smetacek et al., 2004) also play a role.

Despite the overall low productivity in the Southern Ocean, recurring blooms are observed downstream of islands (e.g. Blain et al., 2001), in the vicinity of fronts (Moore and Abbott, 2002) and on the continental shelves (Fig. 1.3). This distribution largely mirrors the input of iron from external sources and indicates that they play an important role in the Southern Ocean.

### 1.3 Sources of iron

Traditionally, dust has been believed to be the most important source of iron to the Southern Ocean (e.g. Cassar et al., 2007), but this view has been challenged in the last decade. It is now known that other important sources include the shelf sediments (e.g. Westerlund and Öhman, 1991; Ardelan et al., 2010), melting sea-ice and icebergs (Smith et al., 2007; Raiswell et al., 2008) and hydrothermal vents (Tagliabue et al., 2010). Additionally, the iron supply from below into the mixed layer through upwelling (de Baar et al., 1995; Watson et al., 2000), entrainment (Tagliabue et al., 2014) and diffusion (Law et al., 2003) also plays a role. The magnitude of the input of iron to the Southern Ocean south of 35°S is not constrained, but a list of the current understanding of them can be found in Table (1.1).
Table 1.1: Previous estimates of iron sources to the Southern Ocean south of 35° S (From Tagliabue et al., 2010, Fig. 3)

<table>
<thead>
<tr>
<th>SOURCE</th>
<th>MAGNITUDE [10^7 mol Fe yr(^{-1})]</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dust</td>
<td>8 - 54</td>
</tr>
<tr>
<td>Sediment</td>
<td>460 - 1137</td>
</tr>
<tr>
<td>Ice</td>
<td>2 - 5</td>
</tr>
<tr>
<td>Hydrothermal vents</td>
<td>38</td>
</tr>
</tbody>
</table>

Quantifying the sources of the water’s dissolved iron is nearly impossible from traditional measurements, and large scale estimates of the iron input to the ocean thus mainly come from modeling studies. Recently, a new technique to pin-point the origin of dissolved iron has, however, been developed, based on the isotopic composition of iron in sea water (Lacan et al., 2008; John and Adkins, 2010). The isotopic composition is a signature that can be used to identify the origin of the iron to an anoxic sedimentary source (Homoky et al., 2009; Severmann et al., 2010), a sedimentary non-reductive source (Radic et al., 2011; Homoky et al., 2013), aeolian input (Beard et al., 2003) and a hydrothermal source (Beard et al., 2003; Rouxel et al., 2008). Measurements of the isotopic fractionation in seawater are however scarce (Radic et al., 2011; Conway and John, 2014), and most large-scale estimates of the iron sources still come from modeling studies.

### 1.3.1 Dust

The earth’s crust contains on average 6.3% iron based on weight (Taylor, 1964), which can be transported from arid land surfaces to the surface ocean in dust plumes. Knowledge about dust deposition is obtained locally through sampling (e.g. Wagener et al., 2008; Heimburger et al., 2012), and large scale estimates are provided by atmospheric transport models (e.g. Mahowald et al., 2003; Johnson et al., 2010).

Oceanic dust deposition predominantly takes place in the Northern Hemisphere due to the larger proportion of land sources here. Input of dust to the Southern Ocean is small (Jickells et al., 2005), but aeolian input of iron may nevertheless have a disproportional large impact on biological production in this area due to its HNLC status (Mahowald et al., 2005).

In the Southern Ocean, the aeolian iron input mainly takes place downwind of the continents (Fig. 1.3) (Cassar et al., 2007). The bulk of the dust originates in Patagonia, but studies suggest that this iron source only comprises about 10% of the total iron supply to the Atlantic sector of the Southern Ocean and that it has a small impact on biological productivity in the area (Meskhidze et al., 2007; Johnson et al., 2010). The aeolian iron deposition from Australia has likewise been shown to be of small significance for biological production in the Southern Ocean (Mackie et al., 2008). This can for example be explained by the highly episodical nature of the input and the Australian soil composition. A much smaller fraction of the total dust input to the Southern Ocean originates in South Africa, about half of which is deposited south of 70°S (Piketh et al., 2000).
The highly episodical nature of dust deposition makes the amount of iron delivered to the ocean highly dependent on the magnitude, duration and frequency of the deposition event. Dust sampling in the Southern Ocean is hindered by its remote location (Wagener et al., 2008), leading to higher uncertainties in the dust models in this area. Other uncertainties in the dust deposition models include the solubility of iron, which in reality ranges from 1 to 80% and the iron content of the dust (Mahowald et al., 2005).

1.3.2 Ice

During sea-ice formation, iron is incorporated from the ocean as well as from organic matter trapped into the forming ice (Lannuzel et al., 2010). The iron concentration within the sea-ice can thus be an order of magnitude higher than in the underlying water (Lannuzel et al., 2007). This iron is transported with the ice and released in the marginal ice zone during melting (Fig. 1.3), with the potential to significantly increase the water’s iron concentration locally (van der Merwe et al., 2011). In the Southern Ocean, icebergs tend to drift westwards in the so-called iceberg alleys (Fig. 1.3), and they can thus be transported relatively far while releasing dissolved iron in their trajectories as they melt (Raiswell et al., 2008). Icebergs thus have the potential to affect production in remote areas of the Southern Ocean where iron limitation tends to be stronger than close to land.

Model studies show that the amount of iron released from sea-ice is minor compared to the sediment source (Lancelot et al., 2009; Wadley et al., 2014), and that adding a sea-ice iron source to a biogeochemical model changed the phytoplankton composition towards diatoms, but not the total amount of primary production (Wang et al., 2014).

1.3.3 Sediment

Iron measurements have revealed that the water’s iron concentration is elevated in proximity to the coast (Johnson et al., 1997; Moore and Braucher, 2008), something that can be explained by release of iron from shelf sediments. Increased iron concentrations occur in the pore water of the sediment due to anoxic conditions brought on by remineralization of organic matter (e.g. Burdige, 1993). Transport of iron from the pore water to the water column through diffusion is too slow to fully explain the observed iron fluxes, and other explanations of the transport are for example enhanced flux through bio-irrigation (Elrod et al., 2004) and resuspension of particulate iron in turbulent environments (e.g. de Jong et al., 2012).

The sediment derived iron mainly increases the iron concentration in the surface water near the coast. This is because the iron release decreases with depth, as deeper sediments receive less degradable organic material (Elrod et al., 2004). The sediment derived iron can, however, also be carried far off shore. This has for example been shown in the North Pacific (Lam et al., 2006; Lam and Bishop, 2008) and in the Southern Ocean, where iron concentrations were elevated in the vicinity of the coast and the bottom of the Weddell Sea (Westerlund and Öhman, 1991; Klunder et al., 2014). The physical environment plays a large role for sediment derived iron being transported relatively far in the Southern Ocean. Several studies indicate that iron
from the Patagonian Shelf for instance is carried in the ACC over long distances (e.g. Lösch et al., 1997; de Baar et al., 1995).

In the Southern Ocean, persistent phytoplankton blooms are observed downstream of islands such as South Georgia (e.g. Korb and Whitehouse, 2004) and the Kerguelen (e.g. Blain et al., 2001; Blain et al., 2007) and Crozet (e.g. Bakker et al., 2007; Pollard et al., 2009) Islands, despite of the overall HNLC characterized ocean. This feature is known as the island mass effect, and is brought on by so-called natural iron fertilization. The mechanism behind it is not fully understood, but a number of mechanisms have been put forward as an explanation. First of all, the continental rise around the islands provides iron to the water from the sediments (e.g. Planquette et al., 2007). The iron flux from the sediments is further increased by the currents being forced upwards by the shallowing topography, gaining speed and turbulence and thereby also increasing the iron resuspension (e.g. de Jong et al., 2012). Secondly, iron has a nutrient-like vertical profile (Johnson et al., 1997). The deep water that is forced upwards by the topography in the Southern Ocean consequently already entails a higher concentration of iron than the surface water, and is hence also a potential iron source. Blain et al. (2007) did indeed show that this mechanism was important downstream of the Kerguelen Plateau.

Recent model studies have confirmed the importance of the sediment source of iron in the Southern Ocean, agreeing that it is on average more important than the dust source (Tagliabue et al., 2009a; Lancelot et al., 2009; Wadley et al., 2014).

1.3.4 Hydrothermal supply

Hydrothermal vents supply iron to the deep ocean (e.g. Field and Sherrell, 2000; German et al., 2002; Statham et al., 2005). While it has previously been assumed that this was a minor source of iron to the surface ocean due to rapid formation of particulates (Elderfield and Schultz, 1996), more recent studies suggest that the sig-
A significant amount of iron is protected by complexation with organic ligands (Bennett et al., 2008; Toner et al., 2009). Advection further from the source is thus possible, suggesting that hydrothermal vents supply a significant amount of iron to the deep ocean, though the exact amount is highly uncertain (Carazzo et al., 2013; Saito et al., 2013). Using the iron isotope fractionation technique, Conway and John (2014) demonstrated that hydrothermal vents added iron to the deep water of the Equatorial Atlantic Ocean which was detectable up to 1000 km away. They speculated that this iron source could serve as a buffer, continuously adding iron to the ocean also when other sources are intermittently low.

In the Southern Ocean, a model study suggests that hydrothermal vents play a role for the surface iron supply, especially in the iron depleted Pacific Ocean where the fast-spreading submarine ridges can be found (Tagliabue et al., 2010). Newer studies do, however, indicate that the hydrothermal iron supply may play a larger role in the Atlantic sector of the Southern Ocean than previously thought (Fig. 1.4) (Klunder et al., 2011; Saito et al., 2013).

1.3.5 Supply to the mixed layer from below

The fact that large scale wind-driven upwelling dominates south of the Polar Front means that a continuous iron flux into the mixed layer takes place here (e.g. de Baar et al., 1995; Watson et al., 2000; Hoppema et al., 2003). The deep winter mixing in the area between the Antarctic PF and SAF (de Boyer Montegut et al., 2004) transport iron into the surface water through entrainment (Tagliabue et al., 2014). Iron is also brought upwards through vertical diffusion (Law et al., 2003; Boyd et al., 2005), through horizontal advection across the base of the mixed layer (Ellwood et al., 2008) and through a transport induced by mesoscale eddies.

1.4 THE MARINE IRON CYCLE

1.4.1 The iron cycle in nature

Historically, advances in the knowledge of the iron cycle have been held back by the difficulties in measuring the water’s concentration of dissolved iron. This was mainly caused by the trace concentrations of iron in the water combined with on-board contamination on research vessels (Achterberg et al., 2001; Bowie et al., 2006). The first reliable measurements of iron profiles ranging the entire water column were presented in the 1980’ies (e.g. Martin and Gordon, 1988; Martin et al., 1989), showing a nutrient-like profile with lower concentrations in the surface water and higher below the mixed layer. Johnson et al. (1997) presented a compilation of iron profiles from different ocean basins, confirming the nutrient-like profile and later global compilations were presented by Parekh et al. (2005) and Moore and Braucher (2008). Due to its remoteness and extreme weather, the Southern Ocean is, however, the least represented area in these compilations. But as measurement techniques have gradually improved and sampling programs, such as GEOTRACES (www.geotraces.org), have been launched, the knowledge of the iron concentrations in the Southern Ocean have improved (e.g. Klunder et al., 2011; Tagliabue et al., 2012).

Like the macronutrients, iron has a nutrient-like profile as it is taken up in the surface water and then remineralized from organic material deeper in the water col-
The marine iron cycle is highly complex and is as of now not fully understood (Gledhill and Buck, 2012). The following is a summary of the main pathways within the iron cycle, but is not a full overview.

In the ocean, the total iron can be divided into a dissolved iron pool, including ligand-bound iron, and a pool of non-labile iron, with strong interactions between these pools, making their relative contribution highly variable. The dissolved iron consists of the redox pair Fe(II) and Fe(III) (Fig. 1.5). In the oxidized ocean, Fe(II) is rapidly converted to Fe(III), a process which is faster at higher temperatures (Millero et al., 1987). Reduction of ligand-bound Fe(III) to Fe(II) is dependent on photochemical processes and thus varies with the light level in the water (e.g. Barbeau et al., 2001; Rijkenberg et al., 2005). The pool of Fe(III) is highly insoluble (Liu and Millero, 2002), and more than 99% of the dissolved iron in the ocean exists in complexation with stable organic ligands (e.g. Rue and Bruland, 1995; van den Berg, 2006). Ligand-bound iron is partly bioavailable (e.g. Hutchins et al., 1999; Maldonado et al., 2005; Tagliabue et al., 2009b). The water’s concentration of ligand, which has been shown to vary between 0.2 and 10 μmol m$^{-3}$ (Gledhill and Buck, 2012), varies with depth (Ibisanmi et al., 2011) and in the horizontal domain (e.g. Thuróczy et al., 2011; Mohamed et al., 2011). Traditionally, two ligand classes have been classified, char-
Figure 1.6: Overview of the iron cycle model from Parekh et al. (2004). Bioavailable iron is marked in white boxes, and grey boxes represent non-available iron. Fe’: Free dissolved iron, bioavailable. Fe-S: Fe scavenged by organic material, non-bioavailable. Fe-L: Fe in complexation with ligands, non-bioavailable.

characterized by strong and weak binding to iron respectively (e.g. Rue and Bruland, 1995). The dissolved iron that does not form complexes with ligands is continuously removed from the water column through scavenging onto particles in the water such as detritus and phytoplankton (Fig. 1.5).

1.4.2 The iron cycle in models

When parameterizing a biological process in a biogeochemical model, a balance between the complexity of nature and the computational demand must be found. The relatively complex iron cycle is thus treated comparably simple in commonly used Ocean General Circulation Biogeochemical Models (OGCBMs) (e.g. Moore and Braucher, 2008; Yool et al., 2011), for example as the dissolved iron commonly is represented by a single tracer. The gradually increasing amount of iron measurements globally (Johnson et al., 1997; Parekh et al., 2005; Moore and Braucher, 2008) and in the Southern Ocean (Klunder et al., 2011; Tagliabue et al., 2012) has improved the understanding of the iron cycle and likewise its representation in models.

Johnson et al. (1997) presented an iron model in which scavenging only occurred when the iron concentration exceeded 0.6 μmol Fe m\(^{-3}\). This was based on the assumption that the iron concentration was close to 0.6 μmol Fe m\(^{-3}\) throughout the deep ocean, and that the dissolved iron was bound to strong ligands at lower concentrations. This assumption was incorporated in a few global models (e.g. Archer and Johnson, 2000; Aumont et al., 2003), in which the removal of iron through scavenging was balanced by aeolian iron input.

Parekh et al. (2004) were the first to introduce and interactive ligand concentration in a model, and were thereby able to reproduce the inter-ocean differences in deep water iron concentrations that newer measurements had revealed (Parekh et al., 2005); the North Atlantic had the highest and the Southern Ocean the lowest deep water iron concentrations. Newer global models adopted this approach and likewise
allowed scavenging at lower iron concentrations (e.g. Moore et al., 2004; Aumont and Bopp, 2006).

The framework for iron modeling presented by Parekh et al. (2004) and Parekh et al. (2005) is still used in a number of models (e.g. Aumont and Bopp, 2006; Dutkiewicz et al., 2009; Yool et al., 2011), including REcoM2 (Hauck et al., 2013). The iron model is relatively simple; the dissolved free iron is represented by a single iron pool and it has one ligand class of constant temporal and spatially concentration (Fig. 1.6). The pool of dissolved iron that is not bound to ligands (free iron) is continuously removed through a constant scavenging rate.

The effect of irradiance, temperature and pH on the iron cycle has been added to a global model by Tagliabue et al. (2009b), more complexity, including differentiation of Fe(II) and Fe(III) by Tagliabue and Völker (2011) and spatially varying ligand concentrations has been explored by Völker and Tagliabue (2014).

The continuous removal of dissolved iron from the water column through scavenging must be balanced by iron input in OGCBMs. The standard has been to add this through aeolian input (Archer and Johnson, 2000; Aumont et al., 2003; Parekh et al., 2005). But in the last decade, it has become clear that the sediment source of iron plays a large role, both on the global scale and especially in the Southern Ocean (e.g. Moore and Braucher, 2008; Tagliabue et al., 2009a). This source has therefore been implemented in models. Moore et al. (2004) employed a constant flux rate of 2 μmol m−2 day−1 in sediments shallower than 1100 meters. But due to the resolution of the model grid being lower than the width of the shelves in many places, the impact of this iron source was small in this model set-up. Aumont and Bopp (2006) therefore modulated their constant iron flux by an availability factor. This factor was controlled by the degree of anoxia in the sediments, decreasing with depth, and by the fraction of the cell that is on the shelf. Elrod et al. (2004) showed that the iron flux from the sediments is correlated with the release of carbon, and their model was incorporated into a global model by Moore and Braucher (2008). This parametrization improved the iron field in the model when it was compared to iron measurements on the global scales.

A model study by Tagliabue et al. (2010) showed that input of iron to the deep ocean by hydrothermal vents plays a role in the Pacific Ocean on long time scales, but this source has so far not been added to any other OGCBM. A couple of models have added an iron input from melting sea-ice and icebergs to the ocean (Lancelot et al., 2009; Wang et al., 2014). The mechanism behind sequestration of iron into the sea-ice is, however, complex and not fully understood.

As more knowledge is gained about the iron cycle and its effect on marine biological production it becomes possible to further add to the complexity of the iron cycle in OGCBMs. Regarding the iron cycle, the process understanding is, however, still so incomplete that adding more complexity to a model may not improve the model’s representation of reality. In model runs investigating the role of climate change on biological production, it may be sensible to use an iron model in which the role of temperature, light and pH is incorporated. As is often the case in ocean biogeochemical modeling, it is therefore necessary to judge the knowledge gained by adding complexity against the increased computational demand that this complexity imposes.
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Global Ocean General Circulation Biogeochemical Models (OGCBMs) are sophisticated tools, routinely used to investigate the dynamics of the earth system. They can be used to examine the mechanisms behind current trends, for example CO$_2$-uptake in the Southern Ocean (e.g. Hauck et al., 2013; Frölicher et al., 2015), or to evaluate the effect of future increasing temperatures on marine phytoplankton (e.g. Bopp et al., 2013; Moore et al., 2013b; Vanoppenolle et al., 2013).

The results of the biogeochemical components are sensitive to the representation of the ocean circulation, which for example influences the amount of nutrients brought to the surface water through upwelling and vertical mixing (e.g. Sinha et al., 2010). The Mixed Layer Depth (MLD) is an especially important feature because it affects the amount of radiation available for photosynthesis as well as the entrainment of nutrients to the surface water (e.g. Williams and Follows, 2003). This is particularly true for the Southern Ocean where models differ significantly regarding the mixed layer depth and biological production (e.g. Doney et al., 2004; Najjar et al., 2007).

Currently, global OGCBMs are solely applied on quasi-regular meshes, offering stable and well tested model environments (e.g. Moore et al., 2004; Aumont and Bopp, 2006). Large efforts are nevertheless continuously made to improve the representation of the ocean circulation in the models. This includes utilization of new modeling techniques. An example of this is the global isopycnal OGCBM from Bergen (Assmann et al., 2010), where the distribution of the $z$-coordinates depend on the density structure of the water column, allowing for a more natural vertical division than the traditionally fixed intervals (Griffies et al., 2009).

Another possibility is the usage of a model running on an unstructured grid (e.g. Danilov, 2013; Ringler et al., 2013). The core strength of a model employing an unstructured grid is the ability to highly resolve chosen areas, while providing seamless transition to areas of lower resolution (Pain et al., 2005). High resolution can for example be an advantage in places with strong ocean currents, such as the Antarctic Circumpolar Current (ACC) or around the equator. Resolving the shelf breaks could improve the description of bottom water formation around Antarctica (e.g. Matsumura and Hasumi, 2010; Haid and Timmermann, 2013). The flow through narrow straits impacts the circulation of the open ocean, and a high resolution here may therefore also be an advantage (e.g. Wekerle et al., 2013). The advantages of a multi-resolution grid makes it logical and intriguing to couple a biogeochemical model to an Ocean General Circulation Model (OGCM) employing an unstructured grid. This has already been done for coastal applications (e.g. Larsen et al., 2013), but so far not for a model covering the global domain. A reason for this is the increased computational demand an unstructured mesh model adds, especially when using the finite element method for discretization (Danilov, 2013). This is a practical feature that becomes all the more important in a global OGCBM as the biogeochemical component introduces a significant number of tracers, which makes it computationally
demanding from the outset. The ability to add high resolution locally does indeed save computer time, but it also adds the requirement of a smaller time step.

Another aspect of coupling a biogeochemical model to an ocean model based on the unstructured mesh technology is that a global OGCM with satisfying skill must exist for such a coupling. At the Alfred-Wegener-Institute, such a model exists in the Finite Element Sea-ice Ocean Model (FESOM).

2.2 THE FINITE ELEMENT SEA-ICE OCEAN MODEL

FESOM is a global 3-dimensional OGCM, coupled to a sea-ice model. Discretization is carried out employing the finite element methodology, and FESOM runs on a triangulated unstructured grid. The model has been developed in the Climate Dynamics group at the Alfred Wegener Institute in Bremerhaven, Germany. It is based on the Finite Element Model of the North Atlantic (Danilov et al., 2004), which has been further developed for the global setup and coupled to a sea-ice model (Timmermann et al., 2009). The newest version of FESOM (v. 1.4) is comprehensively described by Wang et al. (2014). FESOM is participating in the Coordinated Ocean-ice Reference Experiments (COREs) introduced by Griffies et al. (2009). Here, the performance of a number of coupled sea-ice ocean models are evaluated on the global scale (Griffies et al., 2009; Sidorenko et al., 2011) and in local ocean basins such as the North Atlantic (Danabasoglu et al., 2014) and the Southern Ocean (Downes et al., 2014).

FESOM solves the standard set of hydrostatic primitive equations under the Boussinesq approximations in a global setting. The governing equations are divided into a dynamical and a thermodynamical part, which are solved separately, divided by a lag of half a time step (Wang et al., 2008). Further information about the numerics of FESOM can be found in Danilov et al. (2004) and Wang et al. (2014), while the sea-ice component is described in detail by Timmermann et al. (2009).

FESOM is forced with the CORE-datasets, which entail the atmospheric fields needed to calculate air-sea fluxes of heat, fresh water and momentum (Large and Yeager, 2004; Large and Yeager, 2009). These datasets are based on the NCEP reanalysis data, which has been adjusted towards observations to reduce biases. The datasets were specifically developed for coupled sea-ice ocean models and therefore pay special
attention to the conservation of heat and freshwater on a global scale. The CORE-forcing used in the current thesis consists of a normal year climatology (CORE-I), which is based on the years 1948 to 2000 (Large and Yeager, 2004), and of an inter annually varying dataset (CORE-II) covering 1948 to 2008 (Large and Yeager, 2009). The forcing is provided on a grid with an approximate resolution of $1.875\degree \times 1.915\degree$. When running FESOM under the CORE forcing, it is interpolated onto the grid used by FESOM.

FESOM runs on a mesh that is vertically stratified and horizontally triangulated and unstructured. The mesh thus consist of prisms, which are further divided into tetrahedra (Fig. 2.2). This structure means that optimized meshes can be created for different studies. Recent applications of FESOM utilizing the unstructured mesh include a study of the importance of high resolution in the Canadian Arctic through-flow (Wekerle et al., 2013), a study of bottom water formation in the Weddell Sea (Haid and Timmermann, 2013) and of future melting of the Antarctic ice sheet (Hellmer et al., 2012).

### 2.3 The Regulated Ecosystem Model

Marine phytoplankton plays an important role in the global carbon cycle as it is responsible of up to half of the global primary production (e.g. Field et al., 1998). Understanding the mechanisms behind phytoplankton growth, and thus carbon fixation, is therefore important in order to predict the response of the global carbon cycle to climate change. One way to gain further understanding is through the use of OGCBMs, which can calculate the complex interplay between biogeochemical fluxes and the oceanic and atmospheric circulation.

Biogeochemical models describe biogeochemical processes through parameterizations; mathematical equations that describe observations from the laboratory or the field to a satisfying degree. Biogeochemical processes are, however, highly complex, and the parameterizations thus represent strongly simplified versions of reality.

Marine phytoplankton growth is constrained by the availability of light (e.g. Strzepek et al., 2012) and nutrients (e.g. Moore et al., 2013a), and by the presence of grazers (e.g. Cavender-Bares et al., 1999). These factors are therefore an integral part of biogeochemical models. To simplify the dependence of the phytoplankton growth rate
on the availability of various nutrients in global OGCBMs, they commonly rely on the Redfield ratio, namely the notion that the intracellular C:N:P molar ratio in phytoplankton can be described by a constant 106:16:1 ratio (Redfield, 1958). It implies that growth is limited by a single nutrient, thereby simplifying the models substantially both in the growth parameterization and in the computational efficiency (O’Neill et al., 1989). The latter because it means that the phytoplankton concentration in the water can be described by a single value, rather than a value for each of the elements contained in the phytoplankton. The Redfield ratio can be extended to also encompass other nutrients, such as silicon and iron, which are important on the global scale; silicon is needed by diatoms (e.g. Dugdale and Wilkerson, 1998), and iron limits production in large parts of the ocean, such as the Southern and the North and Equatorial Pacific Ocean, where the external iron input is relatively small (e.g. Martin et al., 1991; Kolber et al., 1994; Martin and Fitzwater, 1988). These nutrients have therefore also been implemented into global biogeochemical models (e.g. Moore et al., 2002; Aumont et al., 2003).

In reality, however, the intracellular nutrient stoichiometry is not constant. This was first proven by Ketchum (1939) and Kuenzler and Ketchum (1962) who showed that marine phytoplankton are able perform luxury uptake of phosphorous. In his original paper, Redfield (1958) likewise pointed out that the reported C:N:P ratio was an average and that it does in fact vary significant in nature. Variable stoichiometry was first described in a biochemical model when Droop (1968) pioneered the concept of the cell quota. He showed that the growth rate of a B12-limited phytoplankton species was dependent on the cell to B12 quota, rather than just on the B12 concentration (Fig. 2.3). The cell quota model has later been shown to hold for many different nutrients and organisms (e.g. Droop, 1983, Table 1), and has thus provided the basis for later biogeochemical models including variable stoichiometry. Geider et al. (1998) developed a model with variable intracellular stoichiometry, allowing for variable C:Chl and C:N ratios. This framework has been used in a number of cell models (e.g. Armstrong, 1999; Flynn, 2001), as well in a number of global biogeochemical ocean circulation models (e.g. Moore et al., 2002; Vichi et al., 2007). For global models, the varying stoichiometry has been extended to encompass other important nutrients such as silicon and iron (Aumont and Bopp, 2006; Vichi et al., 2007).

At the Alfred Wegener Institute, the Regulated Ecosystem Model (REcoM) was first developed to reproduce conditions in mesocosm experiments (Schartau et al., 2007). It is based on the model from Geider et al. (1998) and does thus entail variable stoichiometry. The original version (Schartau et al., 2007) was based on dissolved inorganic nitrogen as the only limiting nutrient, but has later been extended to also describe the silicon cycle (Hohn, 2009) and a simple iron cycle based on Parekh et al. (2005). REcoM was extended to cover the global ocean as it was coupled to the Massachusetts Institute of Technology general circulation model (MITgcm, Hohn, 2009). In this configuration it has mainly been used to investigate the carbon cycle in the Southern Ocean (e.g. Hauck et al., 2013; Taylor et al., 2013; Losch et al., 2014)

In the Southern Ocean, variable stoichiometry is especially important as the intracellular C:Si ratio in the phytoplankton is coupled to the availability of iron (e.g. Hutchins and Bruland, 1998; Takeda, 1998), which limits production in large parts of the Southern Ocean (e.g. Martin et al., 1990; Martin et al., 1991). Adaptation in the C:Si therefore has an impact on the export of silicon to the deep ocean in the area.
Figure 2.3: The cell quota model proposed by Droop (1968). The growth rate is dependent on the cell quota, with $Q_{\text{min}}$ being the minimum quota necessary for life. The maximum specific growth rate ($\mu_\infty$) is asymptotically approached as the cell quota increases (From Klausmeier et al., 2008).

In a global model, the choice between variable and constant stoichiometry is however always a balance between the advantage that the complexity provides in terms of reproducing the biogeochemistry and the complexity it adds to the calculations and thus the computational burden. The latter becomes even more important when the OGCM is computationally expensive to run, such as is the case in finite element model.

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In the Southern Ocean, the micronutrient iron plays a large role for the control of net primary production. Knowledge about the input of iron to the Southern Ocean mixed layer and their impact on biological production come from a wide variety of observations (e.g. Blain et al., 2001; Wagener et al., 2008; van der Merwe et al., 2011), but due to the inaccessibility and size of the Southern Ocean, Global Ocean General Circulation Biogeochemical Models (OGCBMs) are commonly used for large scale studies of the impact of the different iron sources on biological production in the Southern Ocean (e.g. Lancelot et al., 2009; Tagliabue et al., 2009; Wadley et al., 2014). But while these models are sophisticated tools that give an insight into the large scale dynamics that can not be gained through observations, they are heavily dependent on parameterization of processes that are not resolved in them as well as on tuning towards a realistic state. Understanding how the models affect the results is therefore important before drawing conclusions from the model results.

The aim of this thesis is to gain an understanding of the relative importance of the different iron sources to the Southern Ocean, and to what extend the state of the ocean model and the parameterization and tuning of the iron cycle in the biogeochemical model affects the representation of the iron sources.

To explore the significance of the input of different iron sources in a model on the phytoplankton composition, the Finite Element Sea-ice Ocean Model (FESOM) has been coupled to the Regulated Ecosystem Model (REcoM). Before using a newly coupled model for research, the skill of the coupled model must be assessed against observational data. This assessment is first of all to see how well the model performs, but also to gain an understanding of the weaker sides of the model. In publication I, the coupled model is presented, and a skill assessment is performed with a special focus on the iron cycle and the Southern Ocean.

Traditionally, most research focussing on the iron input to the Southern Ocean has concentrated on the external sources, such as the sediment and the dust source. But as biological production occurs in the surface mixed layer, the vertical supply of iron across the base of the mixed layer may likewise play a large role (e.g. Tagliabue et al., 2014). Southern Ocean net primary and export production (NPP and EP) calculated in OGCBMs is very dependent on the state of the ocean circulation model (Doney et al., 2004; Najjar et al., 2007). One factor that may play a role in this is the vertical iron supply. Publication II presents two model runs with REcoM2 coupled to FESOM and to the Massachusetts Institute of Technology general circulation model (MITgcm). The results of the model runs are analyzed regarding the vertical iron fluxes in the seasonal and spatial domain. It is further discussed how the ocean general circulation model also has an impact on the total NPP and EP as well as on the phytoplankton composition.

The sediment source of iron has repeatedly been argued to dominate the iron input to the Southern Ocean (Lancelot et al., 2009; Tagliabue et al., 2009; Wadley et al., 2014). In publication III it is investigated to what degree the strength of the sediment source affects the modeled phytoplankton composition and distribution.
As the strength of the sediment source varies widely between measurements, it is discussed how changes in modeled phytoplankton composition and opal export can be used to determine an appropriate strength of the sediment source for OGCBMs.

In the general discussion, the relative contribution of iron from the sediments and from below to the surface mixed layer of the Southern Ocean is debated based on publication II and III. It is further discussed how future increased temperatures are likely to change the strength of all of the iron sources, thereby affecting the biological production.

For the new FESOM-REcoM$_2$ model many future research directions are possible, both regarding the iron sources in the Southern Ocean, and regarding other topics. Some future directions are suggested in the end of the thesis.

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Part II

PUBLICATIONS
LIST OF PUBLICATIONS WITH DECLARATION OF OWN CONTRIBUTION

4.1 PUBLICATION I


A manual explaining how to run the coupled FESOM-REcoM2 model and how the exchange of information between the two models is performed, was published in the supplements of this paper along with the source code for the coupled FESOM-REcoM2 model. It can be found online at: http://www.geosci-model-dev.net/7/2769/2014/gmd-7-2769-2014-supplement.zip.

I have recoded the biogeochemical model REcoM2 and coupled it to FESOM under the guidance of Dmitry Sidorenko and Christoph Völker. During this process and the consecutive tuning of the model, results were especially discussed with Sergey Danilov and Martin Losch. I have drafted the manuscript and subsequently discussed it with the co-authors.

4.2 PUBLICATION II


The study was planned in cooperation with Christoph Völker. I carried out the model run with REcoM2 coupled to FESOM and Judith Hauck the model run with REcoM2 coupled to MITgcm. I have performed the analysis of the model output from both models, drafted the manuscript and discussed it with the co-authors.

4.3 PUBLICATION III

V. Schourup-Kristensen, D. A. Wolf-Gladrow and C. Völker: Sensitivity of Southern Ocean phytoplankton composition to changes in model iron scavenging and sediment supply. To be submitted.

I have coupled the biogeochemical model REcoM2 to FESOM 1.4 and performed the model runs for the study, which were planned in cooperation with Christoph Völker. I have performed the analysis of the model output from both models, drafted the manuscript and discussed it with the co-authors.
A skill assessment of the biogeochemical model REcoM2 coupled to the Finite Element Sea Ice–Ocean Model (FESOM 1.3)

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Abstract. In coupled biogeochemical–ocean models, the choice of numerical schemes in the ocean circulation component can have a large influence on the distribution of the biological tracers. Biogeochemical models are traditionally coupled to ocean general circulation models (OGCMs), which are based on dynamical cores employing quasi-regular meshes, and therefore utilize limited spatial resolution in a global setting. An alternative approach is to use an unstructured-mesh ocean model, which allows variable mesh resolution. Here, we present initial results of a coupling between the Finite Element Sea Ice–Ocean Model (FESOM) and the biogeochemical model REcoM2 (Regulated Ecosystem Model 2), with special focus on the Southern Ocean.

Surface fields of nutrients, chlorophyll $a$ and net primary production (NPP) were compared to available data sets with a focus on spatial distribution and seasonal cycle. The model produces realistic spatial distributions, especially regarding NPP and chlorophyll $a$, whereas the iron concentration becomes too low in the Pacific Ocean. The modelled NPP is 32.5 Pg C yr$^{-1}$ and the export production 6.1 Pg C yr$^{-1}$, which is lower than satellite-based estimates, mainly due to excessive iron limitation in the Pacific along with too little coastal production. The model performs well in the Southern Ocean, though the assessment here is hindered by the lower availability of observations. The modelled NPP is 3.1 Pg C yr$^{-1}$ in the Southern Ocean and the export production 1.1 Pg C yr$^{-1}$.

All in all, the combination of a circulation model on an unstructured grid with a biogeochemical–ocean model shows similar performance to other models at non-eddy-permitting resolution. It is well suited for studies of the Southern Ocean, but on the global scale deficiencies in the Pacific Ocean would have to be taken into account.

1 Introduction

Primary production plays a large role in ocean carbon cycling, and understanding the drivers behind primary production is therefore of paramount importance when it comes to understanding the changes that a future warmer climate will bring. Observations, as well as coupled biogeochemical–ocean models, indicate that climate change will decrease the oceanic net primary production (NPP) (Behrenfeld et al., 2006; Steinacher et al., 2010). This would have far-reaching implications, from changes of the carbon cycle to effects on fisheries.

Coupled biogeochemical–ocean models are important tools used to analyse the net primary production in the ocean and the effects of climate change on it (e.g. Le Quéré et al., 2003; Bopp et al., 2013). The biogeochemical results of such models are highly impacted by the mixing and circulation of the ocean model as it controls processes such as horizontal advection and nutrient supply to the surface layer (Doney et al., 2004). Supply of nutrients through upwelling is especially important when it comes to modelling the equatorial Pacific (Aumont et al., 1999) and the Southern Ocean, where production is iron limited and sensitive to new supply. Results from the second Ocean Carbon-Cycle Model Intercomparison Project (OCMIP-2) highlighted the importance of the ocean model; they showed how the representation of the
Ocean circulation in the Southern Ocean has a large impact on the calculations of present and future uptake of CO$_2$ (Doney et al., 2004), and reported that the global export production varied between 9 and 28 GtC yr$^{-1}$ when the same biogeochemical model was coupled to different OGCMs (Najjar et al., 2007).

Traditionally, global OGCMs employ structured grids with relatively uniform spatial resolution in the entire domain, and local refinement is done by utilizing nested models. The unstructured-mesh technology is emerging as an alternative to nesting in ocean models, and is gradually becoming more widespread within the ocean modelling community (e.g., Chen et al., 2003; Danilov et al., 2004; Piggott et al., 2008). As solutions for the global ocean state provided by models formulated on unstructured meshes have improved (e.g., Sidorenko et al., 2011), it has become feasible to exploit the advantages offered by such models in biogeochemical modelling by coupling a biogeochemical model to an unstructured-mesh ocean model (Hill et al., 2014). One may then benefit from the possibility of aligning the grid with the bathymetry, or refining it in areas of interest without the loss of accuracy that the nesting introduces at boundaries. This is especially relevant when it comes to modelling features such as mixed layer depth, upwelling and the presence of fronts and eddies that are of vital importance for realistic modelling of ecosystems.

A drawback of the unstructured-mesh technology is that, although computer time is saved by using high resolution in chosen areas only, it still uses a substantial amount of computer time as it is less efficient per degree of freedom as compared to structured models. Furthermore, extra care must be taken for models formulated using the continuous finite elements as their local conservation of volume and tracers is formulated in the cluster-weighted sense. This brings some ambiguity into analyzing fluxes between grid cells, while divergences are well defined (Sidorenko et al., 2009).

Before using a newly coupled biogeochemical–ocean model for scientific studies, the skill of the model must be assessed (e.g. Assmann et al., 2010). Performing a skill assessment is not a trivial exercise, considering both the lack of data, especially for parameters such as dissolved iron and export production, and also the inherent uncertainty of the biogeochemical models, in which complex biochemical processes are described by comparatively simple mathematical parameterizations. We have coupled the Regulated Ecosystem Model 2 (REcoM2) to the Finite Element Sea Ice–Ocean Model (FESOM), and in this paper a skill assessment of the coupled model is carried out with emphasis on the Southern Ocean. We show to what extent the results are comparable to observations and discuss how they compare to results from other models.

Figure 1. Horizontal resolution of FESOM’s unstructured grid.

2 Method

2.1 Ocean model

The ocean component of FESOM solves the hydrostatic primitive equations under the Boussinesq approximations (Danilov et al., 2004; Wang et al., 2008). Elastic viscous plastic (EVP) sea ice dynamics is used together with the thermodynamics adopted from Parkinson and Washington (1979) as described in detail by Timmermann et al. (2009). Currently, FESOM is used for simulation of the three-dimensional global ocean with special focus on the Arctic and the Antarctic (Haid and Timmermann, 2013; Wekerle et al., 2013). The latest FESOM version is comprehensively described in Q. Wang et al. (2014).

FESOM operates on unstructured meshes that permit the main feature of the model: the capability of local grid refinement in an otherwise global set-up without nesting. The model domain is discretized by a horizontally triangulated and unstructured, but vertically stratified, mesh with tetrahedral volumes. Integration is carried out on an Arakawa A-grid, which uses vertical $z$ coordinates for simplicity. The mesh used in this study (Fig. 1) is similar to the one used by Sidorenko et al. (2011), in which the horizontal resolution ranges from 15 km in the polar regions to 180 km in the subtropical gyres. In the vertical it has 32 layers, nine of which are located in the upper 100 m.

The bottom topography of FESOM’s grid is constructed using a combination of different data products; the bulk of FESOMS bathymetry, from 60° S to 64° N, is based on topography data from the General Bathymetric Chart of the Oceans (GEBCO, 1 min resolution), south of 60° S, the bottom topography from Timmermann et al. (2010) with a resolution of 1 min (Rtopo-1) is used and north of 69° N it is based on data from the International Bathymetric Chart of the Arctic Oceans with 2 km resolution (IBCAO, version 2; Jakobsson et al., 2008). Between 64 and 69° N, a combination of the GEBCO and IBCAO data sets is used. FESOM’s bottom topography is created using bilinear interpo-
The version of FESOM used here utilizes a linear representation on triangles (in 2-D) and tetrahedrals (in 3-D) for all model variables. The same is true for the biological tracers, which are treated similar to temperature and salinity. The temporal discretization is implicit for sea surface tracers, which are treated similar to temperature and salinity. The vertical mixing is calculated using the PP-scheme first described by Pacanowski and Philander (1981) with a background vertical diffusivity of $1 \times 10^{-4}$ m$^2$s$^{-1}$ for momentum and $1 \times 10^{-5}$ m$^2$s$^{-1}$ for tracers. Redi diffusion (Redi, 1982) and Gent and McWilliams parameterization of the eddy mixing (Gent et al., 1990) are applied with a critical slope of 0.04.

The skill of FESOM has been assessed within the CORE framework (Griffies et al., 2009; Sidorenko et al., 2011; Downes et al., 2014), where several sea ice–ocean models were forced with the normal year (CORE-I) and interannually varying (CORE-II) atmospheric states (Large and Yeager, 2004, 2009) and results compared. In these assessments, the full flexibility of FESOM’s unstructured mesh was not utilized, but the results from FESOM were still within the spread of the other models, and it was consequently concluded that FESOM is capable of simulating the large-scale ocean circulation to a satisfactory degree.

### 2.2 Biogeochemical model

The Regulated Ecosystem Model 2 (REcoM2) belongs to the class of so-called quota models (Geider et al., 1996, 1998), in which the internal stoichiometry of the phytoplankton cells varies depending on light, temperature and nutrient conditions. Uptake of macronutrients is controlled by internal concentrations as well as the external nutrient concentrations, and the growth depends only on the internal nutrient concentrations (Droop, 1983). Iron uptake is controlled by Michaelis–Menten kinetics. An overview of the compartments and fluxes in REcoM2 can be seen in Fig. 2.

The model simulates the carbon cycle, including calcium carbonate as well as the nutrient elements nitrogen, silicon, and iron. It has two classes of phytoplankton: nanophytoplankton and diatoms, and additionally describes zooplankton and detritus. The model’s carbon chemistry follows the guidelines provided by the Ocean Carbon Model Intercomparison Project (Orr et al., 1999), and the air–sea flux calculations for CO$_2$ are performed using the parameterizations suggested by Wanninkhof (1992).

We do not add external sources to the macronutrient pools since the timescale of the runs is short compared to the residence time of the macronutrients in the ocean (Broecker et al., 1982).

Iron has a much shorter residence time (Moore and Braucher, 2008) and is strongly controlled by external sources as well as scavenging. Dissolved iron is taken up and remineralized by phytoplankton, it reacts with ligands and it is scavenged by detritus in the water column (Parck et al., 2005). New iron is supplied to the ocean by dust and sedimentary input. For dust input, REcoM2 uses monthly averages (Mahowald et al., 2003; Luo et al., 2003), which have been modified to fit better to the observations from Wagenet et al. (2008) (N. M. Mahowald, personal communication, 2011). The model assumes that 3.5% of the dust field consists of iron and that 1.5% of this iron dissolves when deposited in the surface ocean. This gives a total aeolian input of $2.65 \times 10^8$ mol DFe yr$^{-1}$ (DFe – dissolved iron) to the ocean on average. A flux of iron from the sediment has been added accounting for an input of $2.67 \times 10^8$ mol DFe yr$^{-1}$ on average. It is incorporated following Elrod et al. (2004) with the magnitude of the iron concentration released by the sediment being dependent on the rate of carbon remineralization in the sediment.

The model has 1 zooplankton class, which is the model’s highest trophic level. Grazing is calculated by a sigmoidal Holling type 3 model with fixed preferences on both phytoplankton classes (Gentleman et al., 2003).

The sinking speed of detritus increases with depth, from $20$ m day$^{-1}$ at the surface, to $192$ m day$^{-1}$ at $6000$ m depth (Kriest and Oschlies, 2008). Sinking detritus is subject to remineralization.

REcoM2 has sediment compartments for nitrogen, silicon, carbon and calcium carbonate, which consist of one layer into which the detritus sinks when reaching the lower-most ocean layer. Remineralization of the sunken material subsequently occurs in the benthos, and the nutrients are returned to the water column in dissolved state.

---

**Figure 2.** The pathways in the biogeochemical model REcoM2.
REcoM1 and 2 have previously been used for large-scale simulations with focus on the Southern Ocean in set-ups with the MITgcm (MIT general circulation model) (Hohn, 2009; Taylor et al., 2013; Hauck et al., 2013), and the purpose of the current coupling between REcoM2 and FESOM is likewise studies of the Southern Ocean.

A full description of the model equations can be found in Appendix A along with lists of parameters used in the current run.

2.3 Model experiment

We present a numerical hindcast experiment with a newly coupled biogeochemical–ocean general circulation model. The run was forced using the CORE-II data set, which was developed for the use of coupled sea–ocean models and gives an interannually varying forcing for the years 1949 to 2008 (Large and Yeager, 2009). As focus here is on evaluating the biological surface processes of a newly coupled model, we follow the example of Vichi and Masina (2009) and Yool et al. (2011) and let the coupled model run for a total of 38 years, from 1971 to 2008. The first 33 years are considered spin-up and we present the results for the years 2004 to 2008. Prior to activating the biogeochemical module, the ocean model had been spun up for 300 years, which is sufficient to reach a quasi-equilibrium state (Fig. 8 in Sidorenko et al., 2011). The length of the time step used throughout the run was 1800 s.

In REcoM2, the tracers for dissolved inorganic nitrogen (DIN) and dissolved silicon (DSi) were initialized with values from the Levitus World Ocean Atlas climatology of 2005 (Garcia et al., 2006), and the dissolved inorganic carbon (DIC) and total alkalinity (TA) tracers were initialized with contemporary values from the Global Ocean Data Analysis Project (GLODAP) data set (Key et al., 2004). Due to scarcity of observations for DFe, the iron field was initialized with an output from the Pelagic Interaction Scheme for Carbon and Ecosystem Studies (PISCES) model (Aumont et al., 2003), which has been modified south of 45° S with mean observed profiles from Tagliabue et al. (2012). All other tracers were initialized with arbitrary small values.

We used a constant value for the atmospheric CO2 during the simulation. Because of the duration of the simulation, the carbon cycle is not in equilibrium at the end of the run, and we do not focus on this part of the model here.

2.4 Data and skill metrics

The focus of this skill assessment is on the key parameters of the physical, chemical and biological surface fields, for which we examine the model behaviour on the global scale and in the ocean regions shown in Fig. 3. We have a special interest in the Southern Ocean and therefore also look further into the production and its drivers there. On the temporal scale we primarily focus on annual climatologies of the modelled fields, but also show the seasonal development for certain parameters.

The performance of the model regarding sea surface temperature (SST), sea surface salinity (SSS), mixed layer depth (MLD), DIN, DSi, chlorophyll a (Chl), NPP and export production (EP) is summarized in Taylor diagrams, displaying the spatial agreement between the modelled surface multi-year means and the observations, and one also taking the seasonal variations into account. Due to a lack of monthly data, EP is only plotted in the former Taylor plot. The spatial distribution of the modelled and observed surface climatologies of these same fields, except SST and SSS, have additionally been plotted to show the bias which is not captured by the Taylor diagrams. SST and SSS were omitted as they have been evaluated elsewhere (Sidorenko et al., 2011). Global climatologies of dissolved iron concentrations do not exist, making it impossible to evaluate the modelled iron fields in the manner described above. Instead, the modelled surface climatology of the iron concentration is plotted on its own, and mean values of the ocean basins are compared to observation compilations in table form. NPP is the model’s end result, and we have therefore additionally plotted the mean seasonal cycle of the NPP for each of the ocean basins defined in Fig. 3, for the model result as well as satellite-based observations. Due to the large effect the modelled mixed layer has on the prediction of the NPP, we have also chosen to illustrate the seasonal cycle of the modelled and observed mean MLD in these same ocean basins.

The Taylor diagrams (Taylor, 2001) show the correlation ($r$), the normalized root mean square error (RMSE) and the normalized standard deviation (SD) between the model results and the observations. The correlation between the model and the observations show whether the two data sets increase and decrease simultaneously, the SDs tells us about the magnitude of the changes in the data, but not when these changes occur and the centered RMSE reflects differences in the overall patterns of the two fields after the bias has been removed. The perfect fit between model and observations will have a correlation and a SD of 1 and a RMSE of 0.

A full list of the observations used can be seen in Table 1. For the NPP we use the Vertically Generalized Productivity Model (VGPM) product from the ocean productivity web page (http://www.science.oregonstate.edu/ocean/productivity/index.php), which is based on the Sea-viewing Wide Field-of-view Sensor (SeaWIFS, 2012) chlorophyll measurements and the VGPM NPP model (Behrenfeld and Falkowski, 1997). We have downloaded monthly values from the web page, and from these calculated the spatial and seasonal means. The EP fields from Schlitzer (2004), Siegel et al. (2014) and Laws et al. (2000) were provided as climatologies. The field from Laws et al. (2000) and Siegel et al. (2014) are satellite based, whereas the field from Schlitzer (2004) comes from an inverse model.

Satellite-based estimates of chlorophyll a, NPP and EP provide detailed spatial and temporal data, but obtaining...
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Table 1. List of the observational data sets used for the skill assessment.

<table>
<thead>
<tr>
<th>Data set</th>
<th>Variable name</th>
<th>Temporal coverage</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sea surface temperature</td>
<td>SST</td>
<td>Monthly climatology</td>
<td>Garcia et al. (2010)</td>
</tr>
<tr>
<td>Sea surface salinity</td>
<td>SSS</td>
<td>Monthly climatology</td>
<td>Garcia et al. (2010)</td>
</tr>
<tr>
<td>Mixed layer depth</td>
<td>MLD</td>
<td>Monthly climatology</td>
<td>de Boyer Montegut et al. (2004)</td>
</tr>
<tr>
<td>Dissolved inorganic nitrogen</td>
<td>DIN</td>
<td>Monthly climatology</td>
<td>Garcia et al. (2010)</td>
</tr>
<tr>
<td>Dissolved inorganic silicon</td>
<td>DSi</td>
<td>Monthly climatology</td>
<td>Garcia et al. (2010)</td>
</tr>
</tbody>
</table>

them is not trivial. Remotely sensed global ocean colour values are first converted to chlorophyll \(a\), and under a number of assumptions about, for instance, mixed layer depth, temperature and light, NPP (Behrenfeld and Falkowski, 1997) and finally EP (e.g. Laws et al., 2000; Siegel et al., 2014) can be estimated. Increasing uncertainty is introduced during the process, and the satellite-based estimates are not as such observations, but rather another way of modelling the chlorophyll \(a\), NPP and EP. The spread between the different satellite-based estimates of NPP is large. Carr et al. (2006) showed that estimates of the global NPP differed by a factor 2 between 24 models, with the largest discrepancies occurring in the high-nutrient low-chlorophyll and extreme temperature areas. The SeaWIFS (2012) algorithms have further been shown to significantly underestimate chlorophyll \(a\) concentrations in the Southern Ocean (Gregg and Casey, 2004), and one must consequently be aware of this when using satellite-based estimates from the Southern Ocean.

The Arctic Ocean is likewise an area in which observations are scarce, and for the seasonal Taylor diagrams the modelled results have consequently been removed in both of the polar regions when comparable observational values did not exist. As is the case in the Southern Ocean, it is especially the satellite-based observation of NPP that is affected by this during the winter months. The missing seasonal data has likewise led us to cut off the northern basins at 70° N in Fig. 3.

In addition to comparing our results to available observations, we discuss them in relation to those of other biogeochemical models. For DFe, Chl and NPP we have plotted our bias plots to the scale used by Schneider et al. (2008). They present the results of the IPSL (Institut Pierre Simon Laplace) model, the MPI (Max Planck Institute) model and the NCAR (National Center for Atmospheric Research) model, providing a range of results to hold our model against. We additionally compare our model to the results from Assmann et al. (2010), who present the results of another non-traditional ocean model coupled to a biogeochemical model, and to Yool et al. (2011). For the export production, Moore et al. (2004) provided a thorough discussion of the export of particulate organic matter as well as opal, and our spatial plots of export production is consequently plotted to the scale used by them.

3 Model results

3.1 Physics: mixed layer depth, salinity and temperature

The fit between the spatial distributions of modelled and observed surface temperature and salinity is very good for both spatial (Fig. 4a) and monthly spatial fields (Fig. 4b), with the correlations being higher than 0.99 and the normalized SDs close to 1 for both fields. As is general practice in ocean-only models (Griffies et al., 2009), FESOM’s surface salinity is weakly restored towards the Polar Science Center Hydrographic Climatology (PHC) (Steele et al., 2001) with a piston velocity of 20 m yr\(^{-1}\).

In both FESOM and the observations (de Boyer Montegut et al., 2004), the mixed layer depth is defined as the depth at which the difference between the potential density at 10 m depth and the MLD is greater than 0.03 kg m\(^{-3}\). The spatial distribution of the mean MLD has a correlation of 0.68 and a normalized SD of 0.85 when compared to the data-based estimates (Fig. 4).
Figure 4. Taylor diagrams (Taylor, 2001) showing correlation, normalized SD and the normalized root mean square error between values of the model results and observations (Table 1), weighted by area. (a) Spatial distribution; (b) spatial–seasonal distribution. All values are surface values, except the mixed layer depth and the vertically integrated NPP. (a) Uses the yearly mean calculated over 2004–2008 and (b) uses the monthly means of the same years. All fields have been interpolated to a 1° × 1° grid, using linear interpolation.

Figure 5. Mean spatial distribution of the MLD\textsubscript{max} over the years 2004–2008. (a) Observation based (de Boyer Montegut et al., 2004). (b) Modelled. (c) Residual: modelled − observation based.

The seasonal variability of the MLD leads to entrainment of water with high nutrient concentrations to the surface water during winter, and the maximum depth of the mixed layer during the year (MLD\textsubscript{max}) is therefore especially important from a biological point of view. Overall, the modelled MLD\textsubscript{max} fits well with the observations, but it is generally too shallow in the Southern Ocean (Fig. 5), with the consequence that limiting nutrients are not adequately replenished during winter. This may lead to a too small NPP in the area as well as a dominance of nanophytoplankton over diatoms, as the former needs a lower iron concentration for production. This will be further discussed in Sect. 3.5.

For the monthly fields, the correlation between the modelled MLD and the observations is above 0.6 and the SD equals 1 (Fig. 4b). We investigate this further by plotting the mean depth of the mixed layer in different ocean regions defined in Fig. 3. All basins have correlations above 0.9, except the northern Indian and equatorial basins (Fig. 6), leading us to conclude that the seasonal change in the MLD is well predicted by FESOM.

### 3.2 Nutrients and nutrient limitation

The annual mean surface distribution of DIN and DSi have correlations between model results and observations of 0.91 and 0.86, respectively (Fig. 4a). In the Southern Ocean, the surface DIN concentrations have a negative bias for DIN (Fig. 7) and a positive for DSi (Fig. 8) when the spatial distribution of modelled and observed values are compared. The DIN concentration additionally becomes too high in the sub-
tropical Pacific Ocean (Fig. 7), something we will later argue happens due to a strong iron limitation in the area.

The correlation between model results and observations for the spatial–seasonal distribution of DIN and DSi is close to 0.75 for both fields (Fig. 4b). For both nutrients, the seasonal cycle has the best agreement with the observations in the polar regions (not shown).

Iron has been shown to play a large role as limiting nutrients for phytoplankton in the Southern Ocean, as well as the equatorial and subarctic Pacific (Martin et al., 1991), and is therefore a key parameter in the model. We compare the model’s surface iron concentration to compilations of observations (Moore and Braucher, 2008; Tagliabue et al., 2012) and to other biogeochemical models (i.e. Schneider et al., 2008). It must be mentioned here that the model is not independent of the observations from Tagliabue et al. (2012) as they are also used for initialization of dissolved iron. But as we only compare surface values, and the residence time of iron in the Southern Ocean is much shorter than the model run, the surface iron concentrations at the end of the model run should not be affected by the initialized values.

The pattern of the surface iron concentration in the Atlantic Ocean (Fig. 9) fits well with the observations in Table 2 as well as with the results from the MPI and NCAR models in Schneider et al. (2008), with relative high concentrations in the equatorial region fed by the dust plume from the Sahara, and concentrations decreasing towards the poles.

Iron concentrations in the IPSL model and Assmann et al. (2010) are somewhat lower in the equatorial and southern part of the Atlantic Ocean than our result.

In the Indian Ocean, our surface iron concentrations agree well with the IPSL and NCAR models as well as with the results from Yool et al. (2011), with values higher than 1 nM in the Arabian Sea and falling towards 0.3 nM in the main

<table>
<thead>
<tr>
<th>Basin</th>
<th>Latitudinal extent</th>
<th>Model [nM]</th>
<th>Obs [nM]</th>
</tr>
</thead>
<tbody>
<tr>
<td>North Atlantic</td>
<td>45–70°N</td>
<td>0.34</td>
<td>0.68</td>
</tr>
<tr>
<td>Northern central Atlantic</td>
<td>10–45°N</td>
<td>1.03</td>
<td>0.68</td>
</tr>
<tr>
<td>Southern central Atlantic</td>
<td>45–10°S</td>
<td>0.28</td>
<td>0.44</td>
</tr>
<tr>
<td>Northern Indian</td>
<td>45–10°S</td>
<td>1.10</td>
<td>1.21</td>
</tr>
<tr>
<td>North Pacific</td>
<td>45–70°N</td>
<td>0.14</td>
<td>0.31</td>
</tr>
<tr>
<td>Equatorial Pacific</td>
<td>10°S–10°N</td>
<td>0.02</td>
<td>0.84</td>
</tr>
<tr>
<td>Southern central Pacific</td>
<td>45–10°S</td>
<td>0.02</td>
<td>0.31</td>
</tr>
<tr>
<td>Atlantic Southern Ocean</td>
<td>90–45°S</td>
<td>0.14</td>
<td>0.39*</td>
</tr>
<tr>
<td>Indian Southern Ocean</td>
<td>90–45°S</td>
<td>0.08</td>
<td>0.33*</td>
</tr>
<tr>
<td>Pacific Southern Ocean</td>
<td>90–45°S</td>
<td>0.09</td>
<td>0.15*</td>
</tr>
</tbody>
</table>
Our values also fit well with the observations in the northern Indian Ocean (Table 2), and this, along with the agreement between the models using varying magnitudes of sedimentary iron input, indicates that the coastal upwelling in the Arabian Sea is well captured in these models, and that this upwelling is responsible for the high surface iron concentrations in the area. The lower surface iron concentration in South East Asia is on the other hand evident in all of these models with the exception of the IPSL model, indicating that the sediment source plays a larger role in this area. Here, unfortunately we do not have observations to validate the models. In the Pacific Ocean, our result is closest to the one from Assmann et al. (2010), though they have a higher iron concentration along the North and South American west coast, indicating a stronger coastal upwelling in their ocean model. We have a lower surface iron concentration than all models presented by Schneider et al. (2008), even though they all have low concentrations locally. The observations in Table 2 indicate that all models underestimate the surface iron concentrations in the Pacific, especially in the equatorial region where the upwelling plays the largest role. In the Southern Ocean, Table 2 shows that our surface iron concentration is too low, but the spatial distribution of the surface iron fits well with observed values, with the highest values found in the vicinity of the Antarctic and east of the
Patagonian shelf (Tagliabue et al., 2012). This distribution can mainly be attributed to the sediment and dust sources of iron and the seasonal ice coverage impeding iron uptake by phytoplankton near the Antarctic continent. These factors are also responsible for maintaining the relatively high surface iron concentration in the Arctic, which becomes iron limited in the absence of the sediment source of iron. Assmann et al. (2010) and Moore and Braucher (2008) also experienced this, with the latter mentioning that the missing sediment source has a modest impact on productivity and iron concentrations away from the Arctic. In the Southern Ocean, our result agrees the best with the IPSL model, with a relatively high coastal concentration, which then falls towards the north. Both the MPI and NCAR models have relatively constant higher values in the area south of 40° S of around 0.5 and 0.3 nM, respectively.

The comparison to observations is, however, hindered by the different definitions of the ocean basins. The value of the North Atlantic from Moore and Braucher (2008), for instance, roughly covers the North Atlantic as well as the northern central Atlantic of our definition (Fig. 3). For the equatorial Pacific, Moore and Braucher (2008) report the value 0.84 nM for the whole ocean, and 0.11 nM for the open ocean, where our value is closer to the latter due to the missing coastal processes in the model.

Nutrient uptake limitation is described by Michaelis–Menten kinetics in the model. The Michaelis–Menten coefficient (MM) is computed as MM = [Nut]/([Nut] + K_{Nut}), with [Nut] being the nutrient concentration, and K_{Nut} a nutrient and phytoplankton dependent half-saturation constant.

To plot the distribution of the mean surface limitation we follow the example of Schneider et al. (2008), where the nutrient with the lowest MM in a given place is seen as limiting and it is assumed that other factors, such as temperature and light, are limiting when all Michaelis–Menten coefficients are above 0.7 (Fig. 10).

When looking at the yearly mean, iron limits nutrient uptake for both nanophytoplankton and diatoms up to 45° S and in most of the Pacific. Nanophytoplankton is mainly nitrogen limited in the Atlantic and Indian oceans, concuring with the result by Assmann et al. (2010), Yool et al. (2011) and the IPSL model in Schneider et al. (2008). For diatoms, silicon is limited in the Atlantic and Indian oceans as well as in the Arctic, a feature that we only share with Yool et al. (2011).

In the high latitudes, the modelled nanophytoplankton become light limited during the respective winter months. For the Arctic this is most pronounced in February, where the light limitation reaches down to 45° N in the Atlantic Ocean. For the Southern Ocean, the highest degree of light limitation occurs in August when the area south of 55° S is affected. The higher nutrient demand by diatoms means that they are co-limited by iron and light during winter (not shown).

Figure 10. Following the example of Schneider et al. (2008), the spatial distribution of the annually mean limiting nutrients in the model’s surface water has been calculated. (a) Nanophytoplankton (Fe = 56.6%, DIN = 40.1% of total area). (b) Diatoms (Fe = 53.6%, DIN = 2.5% and DSi = 43.8% of total area).

### 3.3 Chlorophyll and net primary production

Global NPP sums up to 32.5 PgCyr⁻¹ in the model (Table 3), which is lower than the satellite-based estimate of 47.3 PgCyr⁻¹ (SeaWIFS, 2012; Behrenfeld and Falkowski, 1997) and also slightly below the estimate range of 35–70 PgCyr⁻¹ given by Carr et al. (2006), but higher than the modelled values ranging from 23.7 to 30.7 PgCyr⁻¹ reported by Schneider et al. (2008).

On a global scale, diatoms account for 25.9% of all production in the model. In the subtropical gyres, we see close to zero percent of NPP from diatoms, whereas it constitutes close to 100% in the Arctic Ocean (not shown).

The correlations between the spatial distribution of modelled results and satellite data are 0.75 for both chlorophyll a and NPP (Fig. 4a), with the mean surface chlorophyll a concentration being somewhat overestimated as compared to the satellite-based estimates in the high latitudes, while the equatorial regions have concentrations that are too low (Fig. 11) and the extent of the subtropical gyres is too large. Yool et al. (2011) have a higher equatorial chlorophyll a concentration in the equatorial regions of the Atlantic and Pacific oceans as compared to our model, but their concentration in the Southern Ocean is even higher than ours. Furthermore, when we compare our model to the IPSL model (Schneider et al., 2008), we again see that our equatorial chlorophyll a concentrations are lower, whereas the concentrations in the North Atlantic and Southern Ocean are fairly similar to our result.
Table 3. Net primary and export production for the global domain and the Southern Ocean south of 50°S, for REcoM2 and from literature.

<table>
<thead>
<tr>
<th>Units</th>
<th>FESOM–REcoM2</th>
<th>Previous studies</th>
</tr>
</thead>
<tbody>
<tr>
<td>NPP&lt;sub&gt;glo&lt;/sub&gt; [Pg C yr&lt;sup&gt;-1&lt;/sup&gt;]</td>
<td>32.5</td>
<td>35–70 (Carr et al., 2006)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>23.7–30.7 (Schneider et al., 2008)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>47.3 (Behrenfeld and Falkowski, 1997)</td>
</tr>
<tr>
<td>EPG&lt;sub&gt;glo&lt;/sub&gt; [Pg C yr&lt;sup&gt;-1&lt;/sup&gt;]</td>
<td>6.1</td>
<td>5.8–13.0 (Dunne et al., 2007)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>6 (Siegel et al., 2014)</td>
</tr>
<tr>
<td>Opal&lt;sub&gt;glo&lt;/sub&gt; [Tmol Si yr&lt;sup&gt;-1&lt;/sup&gt;]</td>
<td>74.5</td>
<td>69–185 (Dunne et al., 2007)</td>
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<tr>
<td>NPP&lt;sub&gt;SO&lt;/sub&gt; [Pg C yr&lt;sup&gt;-1&lt;/sup&gt;]</td>
<td>3.1</td>
<td>1.1–4.9 (Carr et al., 2006)</td>
</tr>
<tr>
<td>EPS&lt;sub&gt;SO&lt;/sub&gt; [Pg C yr&lt;sup&gt;-1&lt;/sup&gt;]</td>
<td>1.1</td>
<td>1 (Schlitzer, 2002; Nevison et al., 2012)</td>
</tr>
<tr>
<td>Opal&lt;sub&gt;SO&lt;/sub&gt; [Tmol Si yr&lt;sup&gt;-1&lt;/sup&gt;]</td>
<td>21.5</td>
<td>21–54 (Dunne et al., 2007)</td>
</tr>
</tbody>
</table>

Figure 11. Spatial distribution of mean surface concentrations of chlorophyll a plotted to the scale used by Schneider et al. (2008). (a) Satellite-based estimate (www.globcolour.info). (b) Modelled. (c) Residual: modelled – satellite-based estimate.

The spatial distribution of NPP (Fig. 12), follows the same pattern as chlorophyll a, with low production in the oligotrophic gyres along with a higher production in the temperate regions. Production in the gyres is on the low side compared to the satellite-based estimate (Fig. 13), and as they are known to underestimate production here (Friedrichs et al., 2009), our result is most likely much too low here. An explanation may be that the nanophytoplankton in the model does not represent the smallest phytoplankton classes, such as Prochlorococcus and Synechococcus, which are important in the gyres. Even though adoption of the modelled intracellular N:C ratio is possible, this is not enough to increase production here to the level seen in satellite-based estimates.

The missing coastal primary production along the west coast of Africa and South America (Fig. 12) along with a positive temperature bias in these areas (not shown) indicate that the upwelling is too weak here. FESOM has a coastal resolution of 40 km, which is relatively high, but this resolution only covers a narrow path along the coast (Fig. 1), which may not be sufficient for the upwelling zones to be resolved properly; additionally, the low resolution further out in the subtropical gyres could play a role. Furthermore, the driving force for the upwelling is the coastal winds, and the missing upwelling may partially result from a too low resolution of the atmospheric forcing; moreover, higher resolution allows strong surface winds closer to the coast, thereby increasing the strength of the upwelling (Gent et al., 2010).

Another explanation for the low coastal NPP is the missing riverine input of macronutrients, which at least in the case of silicon plays a role locally in places like Amazonas and the Arctic (Bernard et al., 2011). Yool et al. (2011) deal with the missing riverine nutrient input by restoring the nutrient fields along the coasts. Although Yool et al. (2011) have a larger coastal production in their model (especially along the coast of West Africa), they show that the nutrient restoring only has a small influence on this.

When comparing the mean spatial distribution of NPP with other models, our result is the closest to the NCAR model presented by Schneider et al. (2008), with a relatively high production rate in the North and equatorial Atlantic as well as the Indian Ocean. Moderate production in the area of the polar front in the Southern Ocean is a feature that
our model shares with the satellite-based estimate (Fig. 12) and with the NCAR and IPSL models from Schneider et al. (2008). Despite our model’s strong iron limitation in the Pacific Ocean, our results fit well with the IPSL model, though we have a smaller production rate in the southern Pacific.

Taking seasonal variations into account, NPP and surface chlorophyll $a$ have correlations of 0.66 and 0.57, respectively, when comparing the model and the satellite-based estimate (Fig. 4b), and the normalized SDs are equal to 1.47 and 1.94 for chlorophyll $a$ and NPP, respectively. Both are on the same order as the values presented by Doney et al. (2009).

The timing of the seasonal cycle of NPP is well captured in the majority of the ocean regions defined in Fig. 3 (Fig. 13). The Pearson’s correlation coefficients ($R$) range from 0.31 in the equatorial Atlantic to 0.93 in the southern central Atlantic (Fig. 13), with significant correlations in eight of the fourteen basins. In general, the modelled seasonal cycle is closest to the satellite-based estimate between 10 and 45° N and S, where the modelled NPP is low, but the magnitude of the seasonal variations fits well with the satellite-based estimate.

This is the case for all basins in the mentioned area, except the northern Indian basin.

The Southern Ocean stands out as it has a modelled NPP of the same magnitude as the satellite-based estimate, but the spring bloom occurs too early here, compared to the satellite-based estimate. This will be further discussed in Sect. 3.5.

3.4 Export production

The export of organic carbon out of the euphotic zone (export production – EP), is calculated at a reference depth, which in our case is set to the standard 100 m (e.g. Schneider et al., 2008; Doney et al., 2009). Here, we regard EP as the organic matter that sinks due to the effect of gravity, whereas the total EP also entails the vertical movement of POC by advection and diffusion plus a contribution from semi-labile DOC.

The global export production sums up to $6.1 \text{ Pg C yr}^{-1}$ in the model (Table 3), close to the satellite-based estimate of $6 \text{ Pg C yr}^{-1}$ from Siegel et al. (2014). It is also within the range of estimates presented by Dunne et al. (2007), but on the low side and closer to modelled estimates than to estimates based on observations or inverse models.

The modelled EP constitutes 20% of NPP on a global scale, which is similar to the ratio predicted by Laws et al. (2000). The EP field presented by Laws et al. (2000) is calculated at 100 m depth and is based on satellite observations of ocean colour, whereas the EP field calculated by Schlitzer (2004) is based on an inverse model and is calculated at 133 m. Comparing our model to these fields can consequently be argued to be more of a model–model comparison than a model–observation comparison.

The correlation is 0.28 and 0.48, when comparing the spatial distribution of EP in our model to the fields by Schlitzer (2004) and Laws et al. (2000), respectively (Fig. 4a), indicating that our spatial distribution is closer to the field from Laws et al. (2000), and the normalized SDs are 0.90 and 0.60, respectively.

The EP fields from Schlitzer (2004) and Laws et al. (2000) both have high export along the Equator, in the upwelling regions and along 45° N and S (Fig. 14a and b). In the Southern Ocean, Schlitzer (2004) has a comparably higher export in the Indian and Pacific sector and in the North Atlantic Schlitzer (2004) has less than Laws et al. (2000).

REcoM2 captures the overall pattern with high EP around 45° N and S and along the Equator (Fig. 14c and d), and the elevated EP in the North Atlantic is a feature that REcoM2 shares with the field from Laws et al. (2000). Turning to the differences, our EP is lower in the North Atlantic, slightly lower in the gyres and higher south of 45° S when compared to the field by Laws et al. (2000) (Fig. 14e). Compared to the climatology presented by Schlitzer (2004) (Fig. 14b), our EP is generally lower in the Pacific and in the upwelling regions along West Africa and western North and South America, whereas it is higher in the North and South Atlantic (Fig. 14f). In the Southern Ocean, the differences between
Figure 13. Mean NPP over the year in the ocean basins depicted in Fig. 3. The correlation coefficient is written in each plot, and the statistically significant correlations ($p < 0.05$) are marked with $\ast$.

Figure 14. Spatial distribution of export of particulate organic matter plotted to the scale used by Moore et al. (2004). (a) Laws. (b) Schlitzer. (c) and (d) modelled (same figure). (e) Residual: modelled — Laws. (f) Residual: modelled — Schlitzer.
the fields are especially visible in the Indian and Pacific sectors, where Laws et al. (2000) have very low export. Schlitzer has a rather high export and REcoM2’s export lies in between the two. Schlitzer argues that the satellites do not capture the deep blooms that occur in the area, thereby explaining the lack of EP in the satellite-based estimate.

Both the spatial distribution and magnitude of EP in our model are very similar to what was found by Moore et al. (2004).

Vertical export of opal is similarly calculated across a reference depth of 100 m. On a global scale we have a total opal export of 74.5 Tmol yr$^{-1}$. Previous estimates of global export of opal vary widely (Table 3), and our value is in the lower end of the estimates, as are our global values for NPP and EP.

### 3.5 The Southern Ocean

The coupled model FESOM–REcoM2, as a first step, will be used for studies regarding biogeochemical processes in the Southern Ocean south of 50° S, and we are therefore especially interested in its performance here. The reasons for this focus are further discussed in Sect. 4.

The model’s surface salinity and temperature as well as the nutrient fields are well represented in the spatial domain of the Southern Ocean, with all of them having correlation coefficients above 0.9 when compared to observations (Fig. 15). The chlorophyll $a$ and NPP fields both have somewhat lower correlations, with the correlation for NPP and chlorophyll $a$ being equal to 0.75 and 0.48, respectively.

For the MLD, the correlation between the model results and the observational-based estimate (de Boyer Montegut et al., 2004) is 0.63 in the Southern Ocean (Fig. 15). The MLD$_{max}$ is too shallow in the Indian and Pacific sectors of the Southern Ocean, especially in the area of the polar front (Fig. 5), causing this low value. Furthermore, FESOM simulates a too deep MLD$_{max}$ in the convection area of the Weddell Sea associated with deep-water formation. This is a common feature in sea ice–ocean models (e.g. Griffies et al., 2009) and should in itself not have a large impact on the production in the Southern Ocean.

The modelled NPP south of 50° S sums up to 3.1 Pg C yr$^{-1}$ (Table 3). Carr et al. (2006) summarize previous studies of NPP based on ocean colour and report an average NPP of 2.6 Pg C yr$^{-1}$ for the Southern Ocean. They also show that the largest uncertainties in satellite-based estimates regarding NPP are found in the Southern Ocean and that biogeochemical models generally predict higher NPP in the area than satellites.

The model’s export production equals 1.1 Pg C yr$^{-1}$ in the Southern Ocean, close to the 1 Pg C yr$^{-1}$ found by both Schlitzer (2002) and Nevison et al. (2012), and the EP : NPP ratio equals 36% in the Southern Ocean, similar to what was found by Nevison et al. (2012).

Considering the spatial distribution of EP in the Southern Ocean (Fig. 14), the model is closer to the estimate from Laws et al. (2000) with the highest export fluxes occurring in the northern part of the Atlantic and eastern Indian sectors of the Southern Ocean. REcoM2 does however have a larger EP closer to the Antarctic.

The fraction of the total biomass comprised by diatoms in the Southern Ocean defers between studies (Alvain et al., 2005; Hirata et al., 2011). In the present study, the diatoms are responsible for 25% of the NPP south of 50° S, varying from 0% in the very iron limited waters of the South Pacific to 100% in the iron replete regions of the Weddell Sea and on the Patagonian shelf (Fig. 16). Vogt et al. (2013) compared the results of four ecosystem models and showed that the percentage of diatom biomass in the Southern Ocean differed significantly between them, ranging from 20 to 100%. Our diatom percentage is accordingly within the spread of other models.

Production of the silicon-containing diatoms creates a sinking flux of biogenic silica that sums up to 21.5 Tmol Si yr$^{-1}$ south of 50° S in the model (Table 3), which is close to the satellite-based estimate of 25 ± 4 Tmol Si yr$^{-1}$ calculated south of 45° S (Dunne et al., 2007), but estimates vary significantly between studies (e.g. Moore et al., 2004; Jin et al., 2006; Holzer et al., 2014).

In REcoM2, the opal export in the Southern Ocean accounts for 29% of the global opal export (Table 3). This number similarly varies widely between studies, with ours being lower than the 70% suggested by Jin et al. (2006) and...
Holzer et al. (2014), but considering the area of the Southern Ocean, its contribution to the global opal flux is still large in our model.

High export fluxes of biogenic silica (Fig. 17) naturally occur in places with a corresponding high percentage of diatom production (Fig. 16). The largest values are found in the temporary ice zone between 60 and 70°S, as well as in the area east of Patagonia (Fig. 17), where dust and sediments supply iron to the surface water. A band of relatively high opal export is also present in the polar front in the Atlantic and Indian sectors of the Southern Ocean (Fig. 17). In most of the Southern Ocean, the modelled opal flux falls within a range of 0.4–2.5 mol Si m⁻² yr⁻¹. This is slightly higher than the values given by Moore et al. (2004) and lower than the values of up to 9 mol Si m⁻² yr⁻¹ in Jin et al. (2006).

The absence of diatom production in the Pacific sector of the Southern Ocean (Fig. 16), leading to a low opal export in the area (Fig. 17), is notable and can be explained by the pronounced iron limitation of the Pacific, which also reaches into the Southern Ocean and limits production here.*

**Control of bloom in the Southern Ocean**

We will now examine the roles of MLD and iron concentration in explaining the seasonal variability of NPP. For this purpose we define $R^{2*}$ as the temporal coefficient of determination multiplied by the sign of the regression slope. $R^{2*}$ is calculated for each spatial point in the domain south of 30°S between NPP and the MLD (Fig. 18a) and between NPP and DFe (Fig. 18b), using the average iron concentration over the top 100 m of the ocean. The $R^{2*}$ values show that the Southern Ocean is roughly divided into two zonal bands; one north of 60°S, in the area of the polar front (Moore et al., 1999), and one south of 60°S (Fig. 18).
The general picture north of 60° S is that the concentration of dissolved iron and the mixed layer depth both correlate positively with NPP (Fig. 18), indicating that production in the area mainly is iron controlled, and that production starts when the mixed layer deepens and brings iron and other nutrients to the surface. For the mean seasonal cycle of MLD and NPP north of 60° S (Fig. 19a), the magnitude of the modelled bloom fits nicely with the one from the satellite-based estimate, but the maximum occurs two months earlier. The mean MLD is well predicted by FESOM in the area (Fig. 19a), but it is consistently shallower than what is observed. This has the effect that the modelled phytoplankton receives a larger light intensity than is the case in the ocean, something that may affect the timing of the bloom. The mean iron concentration in the surface water is highly correlated with the depth of the mixed layer north of 60° S (Fig. 20a). The phytoplankton concentration starts increasing in July, when the iron concentration is high, and reaches a maximum in October, after which a combination of high grazer concentration and decreasing iron concentrations most likely causes the bloom to decline. Under nutrient and light replete conditions, the increase in biomass is a result of the balance between phytoplankton’s maximum growth rate and the grazing (Behrenfeld, 2010; Hashioka et al., 2013). On the one hand, this indicates that the model’s timing of the bloom could be changed by a smaller maximum growth rate, something that would change the phytoplankton dynamics on a global scale. On the other hand, the modelled zooplankton concentration is tightly coupled to the increase in phytoplankton concentration (Fig. 20a), and increasing the maximum grazing rate is another way of keeping the growth in biomass down. As modelled grazers are set to prefer nanophytoplankton over diatoms, this may further increase the diatom percentage in the Southern Ocean (Hashioka et al., 2013).

The NPP and MLD fields are negatively correlated south of 60° S, whereas correlation between NPP and DFe is close to zero here (Fig. 18). This indicates that light is the main limiting factor in this area and that iron is less important as a controlling factor. The intensity of the incoming light decreases with latitude, and is further decreased or blocked by the presence of sea ice during parts of the year, south of 60° S. The role of the sea ice for the timing of the spring bloom was highlighted by Taylor et al. (2013), who argued that the sea ice melting induces a shallower and more stable mixed layer, increasing the average light intensity received by the phytoplankton, thereby instigating growth. In our study, the modelled bloom is larger than what is estimated by the satellites south of 60° S, but the timing fits well (Fig. 19b). The difference can be explained by the aforementioned underestimation of NPP by the satellites. NPP starts increasing when the iron concentration is high and decreasing again when the iron concentration is low and the grazer concentration high (Fig. 20b).

It is worth noticing that the increase in production begins at the correct time in both areas, but that the rate of biomass increase is too high.

The sparse observations make it difficult to assess the validity of the modelled seasonal cycles of iron and zooplankton. Tagliabue et al. (2012) presented a seasonal cycle of DFe from the SR3 transect south of Tasmania. Their results indicate that the highest iron concentrations occur in January and February suggesting that our seasonal change in iron concentration, which peaks in September, is off. Our results, however, fit well with the model result from Hoppema et al. (2003), who also see a peak in the iron concentration in September.

4 Discussion

We have presented a skill assessment of an initial coupling between the biogeochemical REcoM2 and the Finite Element Sea Ice–Ocean Model (FESOM). FESOM’s capability of locally increased resolution has not been fully utilized in the current study, with the smallest distance between neighbouring grid points being only a factor of about 10 smaller than the largest, and as with most biogeochemical models (e.g. Yool et al., 2011) we do not resolve eddies. The model run presented here can thus be regarded as a baseline run, from which future work that further explores the capabilities of the new coupling can proceed.
Using the current resolution, the advantage of the new FESOM–REcoM2 coupling over the older MITgcm–REcoM2 coupling is not obvious, especially taking FESOM’s larger demand for computer time into account. The strength of the new coupling will become clearer when new studies on specialized meshes have been carried out. We are currently looking into the effect of the ocean model on the biogeochemistry of the Southern Ocean by comparing two model runs between which only the ocean model differs. We assess the differences of the supply of iron to the surface mixed layer from below, and how it affects the NPP in the area (Schourup-Kristensen et al., 2014).

In regional models of the Southern Ocean, fixed boundary conditions must be added along the northern boundary where the Southern Ocean is connected to the Atlantic, Indian and Pacific oceans, introducing an extra potential for error in such models. Running the model in a global configuration has the advantage that the boundary conditions can be omitted.

As the dynamics of the iron supply is something we plan to study further in the future, and it is a feature we would like to improve, especially in the Pacific Ocean, we will now discuss how changes to the iron cycle may change the results of the model.

In the Southern Ocean, the spatial distribution of iron in the model is reasonable, but it tends towards low values (Table 2), something that may be explained by the crudely constrained external iron sources in the area. The strength of the sediment source varies widely between models (e.g. Moore and Braucher, 2008; Aumont and Bopp, 2006), and in REcoM2 we have an input of iron from the dust and the sediment source, which is in the smaller end of the range. Increasing the strength of the sediment source would especially impact the iron concentration locally in the Southern Ocean, though it has been shown that iron from the sediments can be carried far from the source region (Lam and Bishop, 2008). The Atlantic sector of the Southern Ocean would have an especially large input due to the presence of the Patagonian shelf and the Antarctic Peninsula (e.g. Lancelot et al., 2009), but it would most likely not change the supply to the pelagic areas in the Indian and Pacific Southern Ocean substantially. For the Southern and Pacific Oceans it would be especially important to further explore the influence of the aeolian and sedimentary iron sources as well as the input from ice in the polar areas.

In the remote parts of the Southern Ocean, the input of iron to the mixed layer from below plays a large role (De Baar et al., 1995; Löscher et al., 1997), and Tagliabue et al. (2014) showed that the entrainment of iron during deepening of the mixed layer was especially important. FESOM’s MLDmax is too shallow in the Southern Ocean, especially in the region of the polar front in the Indian and Pacific sectors (Fig. 5), something likely affecting the degree of iron limitation in these two areas. The tight coupling between iron concentrations, NPP and MLD in the polar frontal area (Fig. 20a), further confirms the importance of entrainment as a supply mechanism of iron.

The lower iron input favors the smaller nanophytoplankton, which have a lower iron half-saturation constant, and thereby a lower requirement for iron. A larger input of iron would probably change the species composition towards more diatoms, but would not necessarily increase primary production (e.g. S. Wang et al., 2014). A higher percentage of diatoms would also possibly decrease the models surface silicate concentration, which tends towards high values in the Southern Ocean (Fig. 8). The effect on the silicon concentration is however complicated by the fact that the model’s carbon and silicate cycles are decoupled under iron limitation, leading to a higher silicate uptake when the phytoplankton is iron stressed (Hohn, 2009).

In the Pacific Ocean, the surface iron concentration is very low (Fig. 9), inducing iron limitation (Fig. 10), which leads to a build-up of DIN in the surface water (Fig. 7) and low NPP (Fig. 12). The external input of iron from dust and sediment in the model is marginal in the equatorial and southern part of the Pacific and input from upwelling is consequently important here. FESOM produces a reasonable upwelling of 40 Sv along the Equator (Johnson et al., 2001), whereas upwelling is small along the west coast of South America. We have a low iron flux in the upwelled water along the equatorial Pacific in our model (∼10 μmol m⁻² yr⁻¹) compared to the values suggested by Gordon et al. (1997) and Aumont et al. (2003), who reported 44 and 68 μmol m⁻² yr⁻¹, respectively. Our result is however higher.

![Figure 20. Seasonal change in the modelled NPP, MLD, DFe and zooplankton concentration for (a) the Southern Ocean from 40 to 60°S and (b) south of 60°S. DFe and zooplankton concentrations are averaged over the top 100 m of the ocean. NPP and MLD are normalized by the maximum of the monthly values.](image-url)
than the 5.1 μmol m$^{-2}$ yr$^{-1}$ suggested by Fung et al. (2000). The Fe:Cratio in the upwelled water in the equatorial area is 0.0015 μmol Fe mmol C$^{-1}$ (6.66 $\times$ 10$^6$ mmol C mmol Fe$^{-1}$), which is significantly lower than the prescribed constant intracellular ratio of 0.005 μmol Fe mmol C$^{-1}$ (2 $\times$ 10$^6$ mmol C mmol Fe$^{-1}$) in the modelled phytoplankton. The upwelled water consequently contains too little Fe to sustain growth, explaining why biological production is not able to utilize the upwelled DIN.

Allowing the model’s phytoplankton to adapt to the conditions in the water with a varying intracellular Fe:C ratio would be a possible way to increase production, as the ratio would then decrease in areas with low iron concentrations. Variable intracellular Fe:C stoichiometry, as found by Sunda and Huntsman (1995) and Wilhelm et al. (2013), is used in other models (e.g. Moore et al., 2002; Aumont and Bopp, 2006). The intracellular Fe:C ratio in diatoms ranged from 0.002 μmol Fe mmol C$^{-2}$ in the equatorial Pacific to 0.007 μmol Fe mmol C$^{-2}$ in the subtropical gyres in Moore et al. (2002) and the former value fits remarkably well with our Fe:C ratio in the upwelled water in the Pacific. A lower intracellular Fe:C ratio in our model would lower the intracellular Fe:N ratio and bring it closer to the observed nutrient ratio in the upwelled water, indicating that implementing varying ratios would indeed improve the model’s performance in the Pacific.

Other features that could potentially improve the iron cycle are spatially varying solubility of iron in the water, spatially varying ligand concentration and scavenging of iron onto dust particles in the water. The latter is present in the iron cycle used by Moore and Braucher (2008) and would likely counter the relatively high iron concentrations in the equatorial Atlantic and Indian oceans that are present in our model (Fig. 9).

### 5 Conclusions

In the current study we show that the newly coupled model FESOM–REcoM2 reproduces the large-scale productivity and surface nutrient patterns, with the main deficiency being the strongly iron limited Pacific Ocean. The totals NPP and EP are within the range of previous estimates, but in the lower end, mainly due to the low productivity in the Pacific. The ratio between EP and NPP is 20%, similar to the result from Laws et al. (2000).

In the Southern Ocean, the modelled spatial mean fields are likewise reasonable, though the comparison here is hindered by the scarcity of observed data. South of 50° S, the totals NPP and EP agree well with previous estimates, as does the EP:NPP ratio of 36%. Production is iron and light limited in the Southern Ocean making the external input of iron important as a controller of production.

On a global scale, the model provides reasonable seasonal variations of the NPP, but the main deficiency in the Southern Ocean is the early onset of the spring bloom in the area between 40 and 60° S.

Overall, the model results at the present resolution are comparable to those of other non-eddy-resolving biogeochemical models and it is well suited for studies of surface processes in the Southern Ocean on a timescale similar to the one used here.
Appendix A: Equations

In biogeochemical models, the biological state variables are subject to change by the ocean circulation through advection and turbulent mixing as well as by biological processes. Detritus further sinks vertically through the water column due to gravity, and exchange occurs across the surface and bottom boundaries for certain variables.

For a given volume of water, the change in concentration of a given biological state variable $C$ can be expressed as follows:

$$\frac{\partial C}{\partial t} = -\mathbf{U} \cdot \nabla C + \nabla \cdot (\kappa \nabla C) + \text{SMS}(C). \quad (A1)$$

Here, the term $-\mathbf{U} \cdot \nabla C$ represents the change in $C$ due to advection, and $\mathbf{U} = (u, v, w)$ denotes the velocity of the water in the $x$, $y$, and $z$ directions, respectively.

For sinking state variables, the speed of vertical sinking ($U_{\text{det}} = (0, 0, u_{\text{det}})$) is added to water’s velocity in the advection term.

The turbulent motion is taken into account through the term $\nabla \cdot (\kappa \nabla C)$ where $\kappa$ is the diffusivity tensor.

The term SMS($C$), where SMS stands for sources minus sinks, represents the changes due to biological processes. This is the term that comprises the main body of biogeochemical models.

Certain state variables are subject to fluxes across the boundaries of the ocean model. For these, the flux between the ocean and the benthos ($\text{BenF}$) is calculated at the bottom of the ocean and the flux between the ocean and the atmosphere ($\text{AtmF}$) is calculated for the surface layer of the ocean.

In the following, the equations that make up the source minus sink code in the biogeochemical model REcoM2 are described.

The quota approach makes it necessary to have more tracers than in a model based on fixed ratios, as we need to know the intracellular concentration of each of the modelled elements. REcoM2 has a total of 21 oceanic state variables (Table A1) and four benthos compartments (Table A2).

### Table A1. State variables; ocean

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description and unit</th>
</tr>
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<tr>
<td>DIN</td>
<td>Dissolved inorganic nitrogen [mmol N m$^{-3}$]</td>
</tr>
<tr>
<td>DSi</td>
<td>Dissolved inorganic silicon [mmol N m$^{-3}$]</td>
</tr>
<tr>
<td>DFe</td>
<td>Dissolved inorganic iron [mmol Fe m$^{-3}$]</td>
</tr>
<tr>
<td>DIC</td>
<td>Dissolved inorganic carbon [mmol C m$^{-3}$]</td>
</tr>
<tr>
<td>Alk</td>
<td>Alkalinity [mol C m$^{-3}$]</td>
</tr>
<tr>
<td>PhyN$_{nano}$</td>
<td>Intracellular nitrogen concentration in nanophytoplankton [mmol C m$^{-3}$]</td>
</tr>
<tr>
<td>PhyC$_{nano}$</td>
<td>Intracellular carbon concentration in nanophytoplankton [mmol C m$^{-3}$]</td>
</tr>
<tr>
<td>PhyCalc</td>
<td>Intracellular calcite concentration in nanophytoplankton [mmol CaCO$_3$ m$^{-3}$]</td>
</tr>
<tr>
<td>PhyChl$_{nano}$</td>
<td>Intracellular chl a concentration in nanophytoplankton [mg Chl m$^{-3}$]</td>
</tr>
<tr>
<td>PhyN$_{dia}$</td>
<td>Intracellular nitrogen concentration in diatoms [mmol N m$^{-3}$]</td>
</tr>
<tr>
<td>PhyC$_{dia}$</td>
<td>Intracellular carbon concentration in diatoms [mmol C m$^{-3}$]</td>
</tr>
<tr>
<td>PhySi</td>
<td>Intracellular silicon concentration in diatoms [mmol Si m$^{-3}$]</td>
</tr>
<tr>
<td>PhyChl$_{dia}$</td>
<td>Intracellular chl a concentration in diatoms [mg Chl m$^{-3}$]</td>
</tr>
<tr>
<td>ZooN</td>
<td>Zooplankton nitrogen concentration [mmol N m$^{-3}$]</td>
</tr>
<tr>
<td>ZooC</td>
<td>Zooplankton carbon concentration [mmol C m$^{-3}$]</td>
</tr>
<tr>
<td>DetN</td>
<td>Detritus nitrogen concentration [mmol N m$^{-3}$]</td>
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<td>DetC</td>
<td>Detritus carbon concentration [mmol C m$^{-3}$]</td>
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<tr>
<td>DetSi</td>
<td>Detritus silicon concentration [mmol Si m$^{-3}$]</td>
</tr>
<tr>
<td>DON</td>
<td>Extracellular dissolved organic nitrogen [mmol N m$^{-3}$]</td>
</tr>
<tr>
<td>DOC</td>
<td>Extracellular dissolved organic carbon [mmol C m$^{-3}$]</td>
</tr>
</tbody>
</table>

#### A1 Sources minus sinks

##### A1.1 DIN and DSi

$$\text{SMS(DIN)} = \rho_N \cdot f_T \cdot \text{DON} - \frac{V_{\text{N}}^N \cdot \text{PhyC}_{\text{nano}}}{\text{N assimilation, nanophytoplankton}} - \frac{V_{\text{dia}} \cdot \text{PhyC}_{\text{dia}}}{\text{N assimilation, diatoms}} \quad (A2)$$

$$\text{SMS(DSi)} = \rho_{Si} \cdot f_T \cdot \text{DetSi} - \frac{V_{\text{Si}} \cdot \text{PhyC}_{\text{dia}}}{\text{Si assimilation, diatoms}} \quad (A3)$$

The state variables DON, PhyC$_{nano}$, PhyC$_{dia}$ and DetSi are listed in Table A1. The value of the remineralization rate ($\rho_N$) is listed in Table A3. The temperature dependency of remineralization ($f_T$) is calculated in Eq. (A54) and the nitrogen and silicon assimilation rates ($V_{\text{N}}^N$, $V_{\text{dia}}$, $V_{\text{Si}}$, Table A4).
are calculated in Eqs. (A48) and (A49), respectively. $\rho_{\text{Si}}^T$ will now be explained.

**Silicon remineralization:** the temperature dependent remineralization rate of silicon ($\rho_{\text{Si}}^T$, Table A4) is calculated following Kamatani (1982) up until a set maximum value:

$$\rho_{\text{Si}}^T = \min \left( 1.32 \times 10^{16} \text{day}^{-1} \cdot \exp \left( \frac{-11200}{T} \right), \rho_{\text{Si}} \cdot f_T \right). \tag{A4}$$

$T$ is the local temperature (Table A4). The remineralization rate ($\rho_{\text{Si}}$) is listed in Table A3 and the temperature dependency ($f_T$) is calculated in Eq. (A54).

**Input from benthos:** the bottom grid point of the water further receives remineralized inorganic matter from the benthos:

BenFDIN = $\rho_{\text{ben}}^N \cdot \text{BenthosN}$  \tag{A5}

BenFDSi = $\rho_{\text{ben}}^\text{Si} \cdot \text{BenthosSi}$.  \tag{A6}

BenFDIN and BenFDSi (Table A5) denote the fluxes of DIN and DSI into the bottom layer of the ocean. $\rho_{\text{ben}}^N$ and $\rho_{\text{ben}}^\text{Si}$ (Table A3) are constant remineralization rates. BenthosN and BenthosSi denote the vertically integrated benthos concentration of dissolved nitrogen and silicate, respectively (Table A2).

### A1.2 DFe

The intracellular iron concentration is connected to the intracellular carbon concentration through an assumed constant ratio ($q_{\text{Fe:C}}^{}$, Table A6). Biological uptake and release of iron is likewise connected to uptake and release of carbon.

$$\text{SMS(DFe)} = -q_{\text{Fe:C}}^{} \cdot (P_{\text{nano}} - r_{\text{nano}}) \cdot \text{PhyC}_{\text{nano}}$$

$$- \rho_{\text{Fe:C}} \cdot (P_{\text{dia}} - r_{\text{dia}}) \cdot \text{PhyC}_{\text{dia}}$$

$$\text{Nanophytoplankton net growth}$$

$$\text{Diatom net growth}$$

$$+ q_{\text{Fe:C}}^{} \cdot \epsilon_{\text{phy}}^{} \cdot f_{\lim,nano}^{} \cdot \epsilon_{\text{C}}^{\text{phy}} \cdot \text{PhyC}_{\text{nano}}$$

$$\text{Excretion from nanophytoplankton}$$

$$+ q_{\text{Fe:C}}^{} \cdot \epsilon_{\text{phy}}^{} \cdot f_{\lim,\text{dia}}^{} \cdot \epsilon_{\text{C}}^{\text{phy}} \cdot \text{PhyC}_{\text{dia}}$$

$$\text{Excretion from diatoms}$$

$$+ q_{\text{Fe:C}}^{} \cdot \rho_{\text{DetC}} \cdot f_T \cdot \text{DetC}$$

$$\text{Remineralization of detritus}$$

$$+ q_{\text{Fe:C}}^{} \cdot \epsilon_{\text{ZooC}}^{} \cdot \epsilon_{\text{ZooC}}^{} + r_{\text{Zoo}} \cdot \epsilon_{\text{ZooC}}$$

$$\text{Zooplankton excretion and respiration}$$

$$- q_{\text{Fe:C}}^{} \cdot \text{DetC} \cdot \epsilon_{\text{Fe:C}}'$$

$$\text{Scavenging} \tag{A7}$$

The state variables PhyC$_{\text{nano}}$, PhyC$_{\text{dia}}$, DetC and ZooC are listed in Table A1. The value for the constant Fe:C ratio ($q_{\text{Fe:C}}^{}$) is listed in Table A6 and the DOC excretion rates from phyto- and zooplankton ($\epsilon_{\text{phy}}^{}$ and $\epsilon_{\text{ZooC}}^{}$) and the degradation rate for detritus C ($\rho_{\text{DetC}}$) are listed in Table A3. The
### Table A4. Model variables.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description and unit</th>
</tr>
</thead>
<tbody>
<tr>
<td>Agg</td>
<td>Aggregation rate ([\text{day}^{-1}])</td>
</tr>
<tr>
<td>DissCalc</td>
<td>Rate of calcium carbonate dissolution ([\text{day}^{-1}])</td>
</tr>
<tr>
<td>(f_r)</td>
<td>Temperature dependence of rates, dimensionless</td>
</tr>
<tr>
<td>(G^*)</td>
<td>Phytoplankton available for food intake ([\text{mmol N m}^{-3}])</td>
</tr>
<tr>
<td>(G_{tot})</td>
<td>Total zooplankton grazing rate ([\text{mmol N m}^{-3} \text{day}^{-1}])</td>
</tr>
<tr>
<td>(G_{nano})</td>
<td>Nanophytoplankton-specific zooplankton grazing rate ([\text{mmol N m}^{-3} \text{day}^{-1}])</td>
</tr>
<tr>
<td>(G_{dia})</td>
<td>Diatom-specific zooplankton grazing rate ([\text{mmol N m}^{-3} \text{day}^{-1}])</td>
</tr>
<tr>
<td>PAR</td>
<td>Photosynthetically available radiation ([\text{W m}^{-2}])</td>
</tr>
<tr>
<td>(P_{nano}, P_{dia})</td>
<td>C-specific actual rate of photosynthesis ([\text{day}^{-1}])</td>
</tr>
<tr>
<td>(P_{max})</td>
<td>C-specific light saturated rate of photosynthesis ([\text{day}^{-1}])</td>
</tr>
<tr>
<td>(r_{nano}, r_{dia})</td>
<td>Phytoplankton respiration rate ([\text{day}^{-1}])</td>
</tr>
<tr>
<td>(r_{zoopl})</td>
<td>Zooplankton respiration rate ([\text{day}^{-1}])</td>
</tr>
<tr>
<td>(\rho_{CH}_T)</td>
<td>Temperature dependent remineralization rate of Si ([\text{day}^{-1}])</td>
</tr>
<tr>
<td>(\rho_{chl})</td>
<td>Rate of chlorophyll (a) synthesis ([\text{mg Chl mmol C}^{-1} \text{day}^{-1}])</td>
</tr>
<tr>
<td>(T)</td>
<td>Local temperature ([\text{K}])</td>
</tr>
<tr>
<td>(V_{N_{nano}}, V_{dia})</td>
<td>N assimilation rate for nanophytoplankton and diatoms, respectively ([\text{mmol N mmol C}^{-1} \text{day}^{-1}])</td>
</tr>
<tr>
<td>(V_{Si})</td>
<td>Diatom Si assimilation rate ([\text{mmol Si mmol C}^{-1} \text{day}^{-1}])</td>
</tr>
<tr>
<td>(w_{det})</td>
<td>Sinking velocity of detritus ([\text{m day}^{-1}])</td>
</tr>
</tbody>
</table>

### Table A5. Benthos variables.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description and unit</th>
</tr>
</thead>
<tbody>
<tr>
<td>BenFAlk</td>
<td>Flux of alkalinity from benthos to bottom water ([\text{mmol m}^{-2} \text{day}^{-1}])</td>
</tr>
<tr>
<td>BenFDIC</td>
<td>Flux of C from benthos to bottom water ([\text{mmol C m}^{-2} \text{day}^{-1}])</td>
</tr>
<tr>
<td>BenFDIN</td>
<td>Flux of N from benthos to bottom water ([\text{mmol N m}^{-2} \text{day}^{-1}])</td>
</tr>
<tr>
<td>BenFDSi</td>
<td>Flux of Si from benthos to bottom water ([\text{mmol Si m}^{-2} \text{day}^{-1}])</td>
</tr>
<tr>
<td>BenFFe</td>
<td>Flux of Fe from benthos to bottom water ([\text{mol Fe m}^{-2} \text{day}^{-1}])</td>
</tr>
<tr>
<td>BenFDetCalc</td>
<td>Flux of detritus calcite from the water to the benthos ([\text{mmol CaCO}_3 \text{m}^{-2} \text{day}^{-1}])</td>
</tr>
<tr>
<td>BenFDetC</td>
<td>Flux of detritus C from the water to the benthos ([\text{mmol C m}^{-2} \text{day}^{-1}])</td>
</tr>
<tr>
<td>BenFDetN</td>
<td>Flux of detritus N from the water to the benthos ([\text{mmol N m}^{-2} \text{day}^{-1}])</td>
</tr>
<tr>
<td>BenFDetSi</td>
<td>Flux of detritus Si from the water to the benthos ([\text{mmol Si m}^{-2} \text{day}^{-1}])</td>
</tr>
</tbody>
</table>

### Table A6. Parameters for iron calculations.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
<th>Description and unit</th>
</tr>
</thead>
<tbody>
<tr>
<td>(q^{Fe:C})</td>
<td>0.005</td>
<td>Intracellular Fe : C ratio ([\text{mmol Fe mmol C}^{-1}])</td>
</tr>
<tr>
<td>(K_{Fe})</td>
<td>100.0</td>
<td>Iron stability constant ([\text{m}^{-3} \text{mmol}])</td>
</tr>
<tr>
<td>(L_T)</td>
<td>1.0</td>
<td>Total ligand concentration ([\text{mmol m}^{-3}])</td>
</tr>
<tr>
<td>(k_{Fe})</td>
<td>0.0312</td>
<td>Scavenging rate of iron ([\text{m}^3 \text{mmol C}^{-1} \text{day}^{-1}])</td>
</tr>
<tr>
<td>(q_{Fe:C}^{sed})</td>
<td>0.011</td>
<td>Fe : C ratio for remineralization of Fe from benthos ([\text{mmol Fe mmol C}^{-1}])</td>
</tr>
</tbody>
</table>
phytoplankton respiration \((r_{\text{nano}}\text{ and } r_{\text{dia}})\) is calculated in Eq. (A45), the photosynthesis \((P_{\text{nano}}\text{ and } P_{\text{dia}})\) in Eq. (A44), the limitation by intracellular nitrogen \((f_{\text{lim},C_{\text{max}}})\) is described in Sect. A6.1, and the temperature dependency \((f_T)\) is calculated in Eq. (A54). The respiration by zooplankton \((r_{\text{zoo}})\) is calculated in Eq. (A46) and the scavenging will now be explained.

Scavenging: the calculation of the scavenging in REcoM2 is based on Parekh et al. (2004), case III. Here, the total concentration of dissolved iron \((Fe_T)\) is divided into iron bound to ligands \((Fe_L)\) and free iron \((Fe', Table A4):\)

\[
Fe_T = Fe_L + Fe'.
\]  

(A8)

Iron complexed with organic ligands is protected from scavenging. The total ligand concentration \((L_T)\) can be written:

\[
L_T = Fe_L + L'.
\]  

(A9)

Here \(L'\) denotes the free ligands.

We assume that the reaction between free iron and free ligand \((L' + Fe' = Fe_L)\) is fast enough to be in equilibrium:

\[
K_{FeL} = \frac{[Fe_L]}{[Fe']\cdot[L']}. 
\]  

(A10)

By prescribing the value of the conditional stability constant \((K_{FeL})\) as well as the assumed constant total ligand concentration \((L_T)\) and combining Eqs. (A8), (A9) and (A10), we can calculate the concentration of free iron \((Fe')\). This is then used to calculate the scavenging of \(Fe'\), which is assumed to be correlated with the concentration of detritus carbon (Eq. A7). The values for \(K_{FeL}\) and \(L_T\) are listed in Table A6.

The value for the scavenging rate \((k_{Fe})\) is an important controller of deep-water iron concentrations.

Iron input from dust: the surface layer of the ocean receives an input of iron from aeolian dust deposition. Dust is assumed to contain 3.5% iron of which 1.5% is instantaneously dissolved in the ocean. Sea ice blocks dust, and the dust falling here is lost from the system.

Iron input from the benthos: the release of iron to the bottom layer of water is assumed to be proportional to the release of inorganic carbon. This parameterization is based on the work by Elrod et al. (2004). It is calculated as follows:

\[
\text{BenFe} = \rho_{\text{Fe}}^{\text{ben}} \cdot \text{BenthosC} \cdot q_{\text{Fe}}^{\text{sed}}. 
\]  

(A11)

Here \(\text{BenFe} (Table A5)\) is the flux of iron into the bottom layer of the ocean. \(\rho_{\text{Fe}}^{\text{ben}} (Table A3)\) is the remineralization rate for the benthos carbon and \(q_{\text{Fe}}^{\text{sed}} (Table A6)\) is the iron : carbon ratio for the flux. \(\text{BenthosC} (Table A2)\) denotes the vertically integrated carbon concentration in the benthos compartment.

A1.3 DIC

\[
\text{SMS(DIC)} = (r_{\text{nano}} - P_{\text{nano}}) \cdot \text{PhyC}_{\text{nano}} + (r_{\text{dia}} - P_{\text{dia}}) \cdot \text{PhyC}_{\text{dia}} + \rho_{\text{C}} \cdot f_T \cdot \text{DOC} + r_{\text{zoo}} \cdot \text{ZooC} + \text{DOC} \cdot \text{DissCalc} \cdot \text{DetCalc} 
\]

(A12)

The state variables \(\text{PhyC}_{\text{nano}}, \text{PhyC}_{\text{dia}}, \text{DOC}, \text{ZooC} \text{and DetCalc} \) are described in Table A1. Respiration by nanophytoplankton \((r_{\text{nano}}), \text{diatoms} (r_{\text{dia}}) \text{and zooplankton} (r_{\text{zoo}})\) is calculated in Eqs. (A45) and (A46) and the photosynthesis terms \((P_{\text{nano}} \text{and } P_{\text{dia}})\) in Eq. (A44).

The value of the remineralization rate \(\rho_{\text{C}}\) is listed in Table A5. The dissolution of calcite from detritus \((\text{DissCalc})\) is calculated in Eq. (A34), and the value of the calcite production ratio \((\psi)\) is listed in Table A7. \(\psi\) denotes the percentage of the nanophytoplankton that are calcifiers, and their PIC : POC ratio.

Atmospheric input: the DIC concentration of the surface grid point is affected by the air–sea flux of CO2. It is calculated according to the guidelines provided by the Ocean Carbon Model Intercomparison Project (Orr et al., 1999). In the calculations the surface water CO2 concentration, alkalinity, temperature and salinity are taken into account.

Input from benthos: the bottom grid point of the water further receives remineralized inorganic carbon from the benthos:

\[
\text{BenDIC} = \rho_{\text{C}}^{\text{ben}} \cdot \text{BenthosC} + \text{DissCalc} \cdot \text{BenthosCalc}. 
\]  

(A13)

\(\text{BenDIC} (Table A5)\) denotes the flux of DIC into the bottom layer of the ocean and \(\rho_{\text{C}}^{\text{ben}} (Table A3)\) is a constant remineralization rate. The calcite dissolution rate \((\text{DissCalc})\) is calculated in Eq. (A34). \(\text{BenthosC} \text{and } \text{BenthosCalc} (Table A2)\) denote the vertically integrated carbon and calcium carbonate concentration in the benthos compartment.

A1.4 Total alkalinity

The model’s total alkalinity is changed by phytoplankton uptake of nutrients (nitrate and phosphate), precipitation and dissolution of calcium carbonate and remineralization of organic matter (Wolf-Gladrow et al., 2007). Phosphorous is not described in the model, but is taken into account using the
Table A7. Parameters for sources minus sinks equations.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
<th>Description and unit</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\psi$</td>
<td>0.1</td>
<td>Calcite production ratio, dimensionless</td>
</tr>
<tr>
<td>$\gamma$</td>
<td>0.3</td>
<td>Fraction of grazing flux to zooplankton pool, dimensionless</td>
</tr>
<tr>
<td>$m_{ZO}$</td>
<td>0.05</td>
<td>Zooplankton mortality rate $[m^3 \text{mmolN}^{-1} \text{day}^{-1}]$</td>
</tr>
<tr>
<td>$\phi_{phy}$</td>
<td>0.02</td>
<td>Max aggregation loss parameter for phytoplankton $N$ $[m^3 \text{mmolN}^{-1} \text{day}^{-1}]$</td>
</tr>
<tr>
<td>$\phi_{det}$</td>
<td>0.22</td>
<td>Max aggregation loss parameter for detritus $N$ $[m^3 \text{mmolN}^{-1} \text{day}^{-1}]$</td>
</tr>
<tr>
<td>$u_0$</td>
<td>20.0</td>
<td>Detritus sinking speed at surface $[m/day^{-1}]$</td>
</tr>
</tbody>
</table>

The state variables $\text{PhyC}_{nano}$ and $\text{PhyC}_{dia}$ are calculated in Eq. (A54). The remineralization rate can be found in Table A3 and the temperature dependency is calculated in Eq. (A34). Dissolution of calcium carbonate from detritus adds $\text{CO}_3^{2-}$ to the water and thereby increases the alkalinity with two moles for each dissolved mole calcium carbonate. $\text{Disscalc}$ is calculated in Eq. (A34).

The parameter $\psi$, specifying the calcifying fraction of the nanophytoplankton, is listed in Table A7 and the photosynthesis $(P_{nano})$ is calculated in Eq. (A44). The remineralization rate $(\rho_N)$ can be found in Table A3 and the temperature dependency $(f_T)$ is calculated in Eq. (A54).

Dissolution of calcium carbonate from detritus adds $\text{CO}_3^{2-}$ to the water and thereby increases the alkalinity with two moles for each dissolved mole calcium carbonate. $\text{Disscalc}$ is calculated in Eq. (A34). The state variables $\text{PhyC}_{nano}$, $\text{PhyC}_{dia}$, $\text{DON}$ and $\text{DetCaCO}_3$ are described in Table A1.

The nitrogen assimilation $(\text{V}_{nano})$ is calculated in Eq. (A48). The remineralization rate $(\rho_N)$ can be found in Table A3 and the temperature dependency $(f_T)$ is calculated in Eq. (A54).

The state variables $\text{PhyC}_{nano}$, $\text{PhyC}_{dia}$ and $\text{PhyN}_{dia}$ are described in Table A1.

The nitrogen assimilation $(\text{V}_{nano})$ is calculated in Eq. (A48) and the constant excretion rate $(\epsilon_{\text{phy}})$ is listed in Table A3. When the $N:C$ ratio becomes too high, excretion of DOC is downregulated by the limiter function $(f_{\text{lim}}^{N:C_{\text{max}}})$ described in Sect. A6.1. A further loss term is phytoplankton aggregation (Agg), which transfers $N$ to the detritus pools (Eq. A27).

The grazing loss $(G_{nano}$ and $G_{dia}$) is calculated in Eqs. (A52) and (A53), respectively.

### A1.6 Phytoplankton C

The state variables $\text{PhyC}_{nano}$ and $\text{PhyC}_{dia}$ are described in Table A1. The photosynthesis $(P_{nano}$ and $P_{dia}$) is calculated
in Eq. (A44) and the respiration ($r_{nano}$ and $r_{dia}$) in Eq. (A45). The constant DOC excretion rate ($\epsilon_{DOC}^{phy}$, Table A3) is downregulated by the limitation factor $f_{lim,nano}^{N:C_{max}}$ (Sect. A6.1) when the $N:C$ ratio becomes too high.

Aggregation of phytoplankton ($Agg$) is calculated in Eq. (A27) and grazing ($G_{nano}$ and $G_{dia}$) in Eqs. (A52) and (A53), respectively. $q_{C:N}^{nano} = PhyC_{nano}/PhyN_{nano}$ is used to convert the grazing units from mmolN to mmolC.

A1.7 Phytoplankton CaCO₃

Calciﬁers are assumed to comprise a certain fraction of the total nanophytoplankton concentration, speciﬁed by the parameter $\psi$ (Table A7), tying the calcite production of calciﬁers to the growth of nanophytoplankton.

$$SMS(PhyCalc) = \psi \cdot P_{nano} \cdot PhyC_{nano} - r_{nano} \cdot PhyCalc$$  \hspace{2cm} (A20)

The state variables $PhyC_{nano}$ and $PhyCalc$ are described in Table A1. The value for the calciﬁer fraction ($\psi$) is listed in Table A7 and the excretion parameter ($\epsilon_{Calc}^{phy}$) in Table A3. The excretion loss is downregulated by the limiter function $f_{lim,nano}^{N:C_{max}}$ (Sect. A6.1) when the intracellular $N:C$ ratio approaches a maximum value. The photosynthesis ($P_{nano}$) is calculated in Eq. (A44), the respiration ($r_{nano}$) in Eq. (A45), the grazing on nanophytoplankton ($G_{nano}$) in Eq. (A52) and the aggregation rate ($Agg$) is calculated in Eq. (A27). $q_{C:N}^{nano} = PhyCaCO₃/PhyN_{nano}$.

A1.8 Diatom silicon

$$SMS(PhySi) = \frac{V_{Si}^{PhyC_{dia}} - \epsilon_{Si}^{phy} \cdot f_{lim,dia}^{N:C_{max}} \cdot PhySi}{Aggregation loss}$$  \hspace{2cm} (A21)

The state variables $PhyC_{dia}$ and $PhySi$ are described in Table A1. The silicon assimilation rate ($V_{Si}^{Phy}$) is calculated in Eq. (A49), the aggregation rate ($Agg$) in Eq. (A27) and the grazing on diatoms ($G_{dia}$) is calculated in Eq. (A53). The limiter function ($f_{lim,dia}^{N:C_{max}}$) is described in Sect. A6.1. The value of the excretion parameter ($\epsilon_{Si}^{phy}$) is listed in Table A3 and the intracellular ratio between diatom silicon and carbon is defined as $q_{Si:N}^{ae} = PhySi/PhyN_{dia}$.

A1.9 Phytoplankton Chl $a$

$$SMS(PhyChl_{nano}) = \frac{q_{Chl}^{nano} \cdot PhyC_{nano} - G_{nano} \cdot q_{Chl:C_{nano}}^{nano}}{Chl\ a\ synthesis} - \frac{deg_{Chl} \cdot PhyChl_{nano}}{Degradation\ loss} - \frac{Agg \cdot PhyChl_{nano}}{Aggregation\ loss}$$  \hspace{2cm} (A22)

$$SMS(PhyChl_{dia}) = \frac{q_{Chl}^{dia} \cdot PhyC_{dia} - G_{dia} \cdot q_{Chl:C_{dia}}^{dia}}{Chl\ a\ synthesis} - \frac{deg_{Chl} \cdot PhyChl_{dia}}{Degradation\ loss} - \frac{Agg \cdot PhyChl_{dia}}{Aggregation\ loss}$$  \hspace{2cm} (A23)

The state variables $PhyC_{nano}$, $PhyC_{dia}$, $PhyChl_{nano}$ and $PhyChl_{dia}$ are described in Table A1. The chlorophyll $a$ synthesis ($\psi_{Chl}$) is calculated in Eq. (A47), the aggregation ($Agg$) in Eq. (A27) and the degradation parameter ($deg_{Chl}$) is listed in Table A3. The grazing fluxes ($G_{nano}$ and $G_{dia}$) are calculated in Eqs. (A52) and (A53), respectively. The conversion factor from mmolN to mgChl $a$ is defined as $q_{Chl:C_{dia}}^{ae} = PhyChl/PhyN_{dia}$.

A1.10 Zooplankton

$$SMS(ZooN) = \gamma \cdot (G_{nano} + G_{dia}) - m_{zoop} \cdot ZooN^2$$  \hspace{2cm} (A24)

$$SMS(ZooC) = \gamma \cdot (G_{nano} \cdot q_{C:N}^{nano} + G_{dia} \cdot q_{C:N}^{dia})$$  \hspace{2cm} (A25)

The state variables $ZooN$ and $ZooC$ are described in Table A1. It is assumed, due to sloppiness feeding, that only a fraction of the grazed phytoplankton ($\gamma$, Table A7) enters the zooplankton pool. The rest is transferred to detritus.

Grazing by nanophytoplankton and diatoms ($G_{nano}$ and $G_{dia}$) is calculated in Eqs. (A52) and (A53), respectively. The respiration by zooplankton ($r_{zoop}$) is calculated in Eq. (A46). The value of the mortality parameter ($m_{zoop}$) is listed in Table A7 and the DON and DOC excretion ($\epsilon_{DOC}^{zoop}$ and $\epsilon_{DOC}^{ZOOP}$) in Table A3.
The quotas $q_{\text{nano}}^{\text{C:N}}$ and $q_{\text{dia}}^{\text{C:N}}$ convert the units of the grazing from mmol N to mmol C and are defined as follows: $q_{\text{nano}}^{\text{C:N}} = \text{PhyC}_{\text{nano}} / \text{PhyN}_{\text{nano}}$ and $q_{\text{dia}}^{\text{C:N}} = \text{PhyC}_{\text{dia}} / \text{PhyN}_{\text{dia}}$.

### A1.11 Detritus N and C

\[
\text{SMS(DetN)} = (G_{\text{nano}} + G_{\text{dia}}) \cdot (1 - \gamma) + m_{\text{zoo}} \cdot \text{ZooN}^2 \quad \text{Sloppy feeding Zoolankton mortality} \\
+ \text{Agg} \cdot (\text{PhyN}_{\text{nano}} + \text{PhyN}_{\text{dia}}) \\
- \rho_{\text{DetN}} \cdot f_T \cdot \text{DetN} \quad \text{Degradation to DON} \\
\text{SMS(DetC)} = (G_{\text{nano}} \cdot q_{\text{nano}}^{\text{C:N}} + G_{\text{dia}} \cdot q_{\text{dia}}^{\text{C:N}}) \cdot (1 - \gamma) \quad \text{Sloppy feeding Zoolankton mortality} \\
+ m_{\text{zoo}} \cdot \text{ZooN}^2 \cdot q_{\text{zoo}}^{\text{C:N}} \\
+ \text{Agg} \cdot (\text{PhyC}_{\text{nano}} + \text{PhyC}_{\text{dia}}) \quad \text{Phytoplankton aggregation} \\
- \rho_{\text{DetC}} \cdot f_T \cdot \text{DetC} \quad \text{Degradation to DOC} \quad (A26)
\]

The state variables ZooN, PhyN$_{\text{nano}}$, PhyN$_{\text{dia}}$, DetN and DetC are described in Table A1. Due to sloppy feeding, the grazed phytoplankton partly goes to the zooplankton pool and partly to the detritus pool, depending on the grazing efficiency $\gamma$ (Table A7). The grazing ($G_{\text{nano}}$ and $G_{\text{dia}}$) is calculated in Eqs. (A52) and (A53), respectively.

The quadratic mortality loss from zooplankton ($m_{\text{zoo}}$) is listed in Table A7. The temperature dependent degradation of detritus to DOM ($\rho_{\text{DetN}}$ and $\rho_{\text{DetC}}$) is calculated in Table A3. The state variables PhySi and DetSi are described in Table A1. The state variables PhyCalc and DetCalc are described in Sect. A6.1. The aggregation rate ($\text{Agg}$) is calculated in Eq. (A27), the respiration rate ($f_T$) is calculated in Eq. (A28), the remineralization rate ($\rho_{\text{DetN}}$) is listed in Table A3. The intracellular ratio between diatoms silicon and carbon is defined as $q_{\text{Si:N}}^{\text{C:N}} = \text{PhySi} / \text{PhyN}_{\text{dia}}$.

**Loss to benthos**: when the sinking detritus reaches the bottom grid point it is assumed that it continues sinking into the benthic compartment with the speed $w_{\text{det}}$ (Eq. A28). This leads to a detrital flux (BenFDetN and BenFDetC, Table A5) from the water column to the benthos:

\[
\begin{align*}
\text{BenF}_{\text{DetN}} &= - w_{\text{det}} \cdot \text{DetN} \\
\text{BenF}_{\text{DetC}} &= - w_{\text{det}} \cdot \text{DetC}. 
\end{align*}
\]

The state variables DetN and DetC are described in Table A1.

### A1.12 Detritus Si

\[
\text{SMS(DetSi)} = \left(\frac{\text{PhyN}_{\text{Si}}^{\text{N:Cmax}} + \text{Agg}}{\text{PhySi}}\right) \quad \text{Diatom excretion} \\
\text{Diatom excretion} \text{Aggregation} \\
+ G_{\text{dia}} \cdot q_{\text{Si:N}}^{\text{C:N}} - f_T \cdot f_T \cdot \text{DetSi} \quad \text{Sloppy feeding Remineralization to DSI} \quad (A31)
\]

The state variables PhySi and DetSi are described in Table A1. The limiter function ($f_T$) is calculated in Eq. (A28), the remineralization rate ($\rho_{\text{DetSi}}^w$) in Eq. (A4) and the temperature dependency of remineralization ($f_T$) is calculated in Eq. (A54). The value of the excretion parameter ($\epsilon_{\text{N}}^w$) is listed in Table A3. The intracellular ratio between diatom silicon and carbon is defined as $q_{\text{Si:N}}^{\text{C:N}} = \text{PhySi} / \text{PhyN}_{\text{dia}}$.

**Loss to benthos**: when the sinking detritus reaches the bottom grid point it is assumed that it continues sinking into the benthic compartment with the speed $w_{\text{det}}$ (Eq. A28). This leads to a detrital flux (BenFDetSi, Table A5) from the water column to the benthos:

\[
\begin{align*}
\text{BenF}_{\text{DetSi}} &= - w_{\text{det}} \cdot \text{DetSi}. 
\end{align*}
\]

The state variable DetSi is described in Table A1.

### A1.13 Detritus CaCO$_3$

\[
\text{SMS(DetCalc)} = \epsilon_{\text{C}}^w \cdot f_{\lim,\text{nano}} \cdot \text{PhyCalc} \quad \text{Nanophytoplankton excretion} \\
\text{Nanophytoplankton excretion} \text{Aggregation} \\
+ \left(\frac{\text{Agg}}{\text{PhyCalc}} + r_{\text{nano}}\right) \cdot \text{PhyCalc} \quad \text{Respiration} \\
\text{Respiration} \text{Grazing loss} \\
+ G_{\text{nano}} \cdot q_{\text{nano}}^{\text{C:CaCO$_3$}} \cdot \text{PhyCalc} \quad \text{Diss$_{calc}$} \cdot \text{DetCalc} \quad (A33)
\]

The state variables PhyCalc and DetCalc are described in Table A1. The limiter function ($f_{\lim,\text{nano}}$) is described in Sect. A6.1. The aggregation rate ($\text{Agg}$) is calculated in Eq. (A27), the respiration rate ($r_{\text{nano}}$) in Eq. (A45) and...
the nanophytoplankton grazing rate \(G_{nano}\) in Eq. (A52). The excretion rate \(e_{nano}^{phy}\) is listed in Table A3, the ratio \(q_{nano}^{CaCO3}\) = \(PhyCalc / PhyN_{nano}\), and the calcite dissolution rate will now be explained.

**Calcite dissolution:** as the detritus calcite sinks through the water column it is subject to dissolution (\(Disscalc\), Table A4) occurring on a length scale of 3500 m (Yamanaka and Tajika, 1996).

\[
Disscalc = \frac{w_z}{3500 m} \quad \text{(A34)}
\]

\(w_z\) denotes the sinking speed at depth \(z\) and is calculated in Eq. (A28).

**Loss to benthos:** when the sinking detritus reaches the bottom grid point it is assumed that it continues sinking into the benthic compartment with the speed \(w_{det}\) (Eq. A28). This leads to a detrital flux (\(BenFDetCalc\), Table A5) from the water column to the benthos:

\[
BenFDetCalc = -w_{det} \cdot DetCaCO3. \quad \text{(A35)}
\]

The state variable \(DetCalc\) is described in Table A1.

### A1.14 Dissolved organic material

\[
SMS(DON) = \frac{\epsilon^{phy}_{N} \cdot f_{lim,nano} \cdot PhyN_{nano}}{N:C_{max,nano}} \quad \text{Nanophytoplankton excretion}
\]

\[
+ \frac{\epsilon^{phy}_{N} \cdot f_{lim, dia} \cdot PhyN_{dia}}{N:C_{max, dia}} \quad \text{Diatom excretion}
\]

\[
+ \frac{\epsilon^{zoo}_{N} \cdot ZooN}{N:C_{min}} \quad \text{Zooplankton excretion}
\]

\[
- \rho_{N} \cdot f_{T} \cdot DON \quad \text{Remineralization}
\]

\[
SMS(DOC) = \frac{\epsilon^{phy}_{C} \cdot f_{lim,nano} \cdot PhyC_{nano}}{N:C_{max,nano}} \quad \text{Nanophytoplankton excretion}
\]

\[
+ \frac{\epsilon^{phy}_{C} \cdot f_{lim, dia} \cdot PhyC_{dia}}{N:C_{max, dia}} \quad \text{Diatom excretion}
\]

\[
+ \frac{\epsilon^{zoo}_{C} \cdot ZooC}{N:C_{min}} \quad \text{Zooplankton excretion}
\]

\[
- \rho_{C} \cdot f_{T} \cdot DOC \quad \text{Remineralization}
\]

The state variables are described in Table A1. The values for excretion of nitrogen and carbon from phyto- and zooplankton (\(\epsilon_{N}^{phy}, \epsilon_{C}^{phy}, \epsilon_{N}^{zoo}\) and \(\epsilon_{C}^{zoo}\)) are listed in Table A3 along with the degradation rates for detritus (\(\rho_{det}\) and \(\rho_{detC}\)) and remineralization rates of DON and DOC (\(\rho_{N}\) and \(\rho_{C}\)). The limitation factors \(f_{lim,nano}^{N:C_{max,nano}}\) and \(f_{lim, dia}^{N:C_{max, dia}}\) are described in Sect. A6.1 and the temperature dependency \(f_{T}\) is calculated in Eq. (A54).

### A2 Sources minus sinks, benthos

The model has a benthos compartment which consists of one layer. Matter is supplied to this layer through sinking detritus, and it hence has pools of nitrogen, silicon, carbon and calcite. When sinking detritus reaches the bottom it continues sinking into the benthos with the speed \(w_{det}\), calculated by Eq. (A28) and is thus lost from the water column. In the benthos, the detritus is assumed to be remineralized to dissolved inorganic matter. This is then re-released to the water’s pools of dissolved inorganic matter (DIN, DIC, Alk and DSI):

\[
SMS(BenthosN) = w_{det} \cdot DetN - \rho_{ben}N \quad \text{(A38)}
\]

\[
SMS(BenthosSi) = w_{det} \cdot DetSi - \rho_{ben}Si \quad \text{(A39)}
\]

\[
SMS(BenthosC) = w_{det} \cdot DetC - \rho_{ben}C \quad \text{(A40)}
\]

\[
SMS(BenthosCalc) = w_{det} \cdot DetCalc - Disscalc \cdot BenthosCalc. \quad \text{(A41)}
\]

The state variables are described in Table A1 (\(DetN, DetSi, DetC\) and \(DetCalc\)) and Table A2 (\(BenthosN, BenthosSi, BenthosC\) and \(BenthosCalc\)). The remineralization rates (\(\rho_{ben}N, \rho_{ben}Si\) and \(\rho_{ben}C\)) are listed in Table A3 and the calcite dissolution rate (\(Disscalc\)) is calculated in Eq. (A34).

### A3 Phytoplankton growth

#### A3.1 Photosynthesis

The rate of the C-specific photosynthesis is calculated for both nanophytoplankton and diatoms (\(P_{nano}\) and \(P_{dia}\), Table A4).

The calculation is based on the work by Geider et al. (1998) and differs between nanophytoplankton and diatoms in the nutrient limitation; nanophytoplankton is limited by iron and nitrogen while diatoms are additionally limited by silicon.

\[
P_{nano}^{max} = \mu_{C}^{max} \cdot \min \left( f_{lim,nano}^{N:C_{min}}, f_{lim,nano}^{N:C_{max}} \right) \cdot f_{T} \quad \text{(A42)}
\]

\[
P_{dia}^{max} = \mu_{C}^{max} \cdot \min \left( f_{lim, dia}^{Fe:C_{min}}, f_{lim, dia}^{N:C_{min}}, f_{lim, dia}^{Si:C_{min}} \right) \cdot f_{T} \quad \text{(A43)}
\]

The actual C-specific photosynthesis rate depends on how much photosynthetically available radiation (PAR, Table A4) the cell can harvest. This is controlled by the light harvesting efficiency (\(\epsilon\)) and the intracellular Chl : C ratio as well as the
A3.3 Chlorophyll \(a\) synthesis

Chlorophyll \(a\) synthesis is coupled to N uptake in REcoM2. The uptake of nitrogen by the phytoplankton (\(V_N\)) is converted to chlorophyll units with a maximum Chl : N ratio (\(q_{\text{Chl:N}}^{\text{max}}\)). This highest possible chlorophyll synthesis rate can then be downregulated by the ratio between the actual photosynthesis and the light absorption, leading to a smaller rate when photosynthesis is small.

The chlorophyll \(a\) synthesis is calculated for both nanophytoplankton and diatoms (\(S_{\text{Chl,nan}}\) and \(S_{\text{Chl,dia}}\), Table A4).

\[
S_{\text{Chl}} = V_N \cdot q_{\text{Chl:N}}^{\text{max}} \cdot \min \left( 1, \frac{P}{\alpha \cdot q_{\text{Chl:C}} \cdot \text{PAR}} \right) \tag{A47}
\]

The nitrogen assimilation (\(V_{N,\text{nan}}\) and \(V_{N,\text{dia}}\)) is calculated in Eq. (A48), and the value for the maximum Chl : N ratio (\(q_{\text{Chl:N}}^{\text{max}}\)) can be seen in Table A8. The C-specific photosynthesis (\(P_{\text{nan}}\) and \(P_{\text{dia}}\)) is calculated in Eq. (A44), and the values for \(\sigma_{\text{nano}}\) and \(\sigma_{\text{dia}}\) are listed in Table A8. PAR denotes the photosynthetically available radiation and we define \(q_{\text{Chl:C}} = \text{PhyChl/PhyC.}\)

A3.4 Nitrogen and silicon assimilation

Nitrogen; the carbon-specific N assimilation rate is based on the maximum rate of carbon-specific photosynthesis (\(P_{\text{max}}\)), which is then modulated by the maximum N : C uptake ratio (\(\sigma_{\text{N:C}}\)) and by the intracellular quota between N and C (\(f_{\text{N:C,Max}}\)), which downregulates N assimilation when the N : C ratio approaches a maximum value (Sect. A6.1). The concentration of DIN in the surrounding water modifies the N assimilation through Michaelis–Menten kinetics.

\[
V_N = V_{\text{cm}} \cdot P_{\text{max}} \cdot \sigma_{\text{N:C}} \cdot f_{\text{N:C,Max}} \frac{\text{DIN}}{K_N + \text{DIN}} \tag{A48}
\]

Table A8. Parameters for phytoplankton growth.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
<th>Description and unit</th>
</tr>
</thead>
<tbody>
<tr>
<td>(\sigma_{\text{nu}})</td>
<td>0.19</td>
<td>Light harvesting efficiency for nanophytoplankton [mmol C m(^{-2}) (mg Chl W day(^{-1})]</td>
</tr>
<tr>
<td>(\sigma_{\text{dia}})</td>
<td>0.23</td>
<td>Light harvesting efficiency for diatoms [mmol C m(^{-2}) (mg Chl W day(^{-1})]</td>
</tr>
<tr>
<td>(K_{\text{N,nan}})</td>
<td>0.55</td>
<td>Half-saturation constant for nanophyto N uptake [mmol N m(^{-3})]</td>
</tr>
<tr>
<td>(K_{\text{N,dia}})</td>
<td>1.00</td>
<td>Half-saturation constant for diatom N uptake [mmol N m(^{-3})]</td>
</tr>
<tr>
<td>(K_{\text{Si,nan}})</td>
<td>4.00</td>
<td>Half-saturation constant for diatom Si uptake [mmol Si m(^{-3})]</td>
</tr>
<tr>
<td>(\alpha_{\text{max}})</td>
<td>3.0</td>
<td>Rate of C-specific photosynthesis [day(^{-1})]</td>
</tr>
<tr>
<td>(\alpha_{\text{res}})</td>
<td>0.01</td>
<td>Maintenance respiration rate constant [day(^{-1})]</td>
</tr>
<tr>
<td>(\sigma_{\text{N:C,}\text{max}})</td>
<td>0.2</td>
<td>Maximum uptake ratio N : C [mmol N mmol C(^{-1})]</td>
</tr>
<tr>
<td>(\sigma_{\text{Si:C}},\text{max})</td>
<td>0.2</td>
<td>Maximum uptake ratio Si : C [mmol Si mmol C(^{-1})]</td>
</tr>
<tr>
<td>(\tau)</td>
<td>0.01</td>
<td>Timescale for zooplankton respiration [day(^{-1})]</td>
</tr>
<tr>
<td>(V_{\text{cm}})</td>
<td>0.7</td>
<td>Scaling factor for C-specific N uptake, dimensionless</td>
</tr>
<tr>
<td>(\xi)</td>
<td>2.33</td>
<td>Cost of biosynthesis of N [mmol C mmol N(^{-1})]</td>
</tr>
</tbody>
</table>
The values of the parameters for $V_{cm}, \sigma_{Si:C}, K_{N}^{nano}$ and $K_{Si}^{dia}$ are listed in Table A8. The maximum rates of photosynthesis ($\mu_{C}^{max}$ and $\mu_{C}^{max}$) are calculated in Eqs. (A42) and (A43), respectively, and $f_{lim,N}^{C}C_{max}$ is described in Sect. A6.1. DIN denotes the surrounding water’s concentration of dissolved inorganic nitrogen.

**Silicon**: silicon assimilation ($V_{Si}$, Table A4) only occurs in diatoms. The maximum silicon uptake rate is calculated as the maximum photosynthetic rate ($\mu_{C}^{max}$) multiplied by the maximum Si : C ratio $(\sigma_{Si:C})$ and the scaling factor for the maximum nitrogen uptake ($V_{cm}$). The actual uptake depends on the surrounding water’s silicon concentration through Michaelis–Menten kinetics and the temperature dependency ($f_{T}$). It is additionally downregulated when the N : C or Si : C ratios become too high ($f_{Si}^{lim,}\sigma_{Si:C}$ and $f_{lim,N}^{C}C_{max}$). The N : C ratio is taken into account, as a too high ratio indicates that the intracellular concentration of energy rich carbon molecules becomes too low to use energy on silicon uptake.

$$V_{Si} = \frac{V_{cm} \cdot \mu_{C}^{max} \cdot f_{Si}^{lim}}{K_{Si} + D Si} \cdot \frac{\sigma_{Si:C} \cdot f_{lim,N}^{C}C_{max} \cdot f_{N}^{C_{max}}}{f_{lim,dia}}$$

(A49)

The scaling factor for the N uptake ($V_{cm}$) and the C-specific photosynthesis rate ($\mu_{C}^{max}$) are listed in Table A8 along with the maximum Si : C uptake ratio $(\sigma_{Si:C})$. The temperature dependency ($f_{T}$) is calculated in Eq. (A54), and the limitation by the intracellular ratios Si : C and N : C ($f_{lim}^{Si}C_{max}$ and $f_{lim,dia}$) are described in Sects. A6.3 and A6.1, respectively. $K_{Si}$ is listed in Table, and DSI denotes the surrounding water’s concentration of dissolved inorganic silicon.

**A4 Grazing**

REcoM2 has a single zooplankton class, which is also the highest trophic level in the model. Grazing on both nanophytoplankton and diatoms is calculated using a Type 3 Sigmoidal model as described by Gentleman et al. (2003).

The maximum grazing rate ($G_{max}$) is modulated by the temperature through the Arrhenius function ($f_{T}$) and by the prey availability through a quadratic Michaelis–Menten function. The model has fixed preferences for both phytoplankton classes and $G'$ (Table A4) is the phytoplankton concentration available for food intake, in our case:

$$G' = PhyN_{nano} + PhyN_{dia} \cdot f_{Z}^{dia}.$$  

(A50)

Here, the parameter $f_{Z}^{dia}$ (Table A9) specifies the grazing preference for diatoms. The total grazing ($G_{tot}$, Table A4) is calculated as follows:

$$G_{tot} = G_{max} \cdot \frac{G'}{K_{G} + G'} \cdot f_{T} \cdot ZooN.$$  

(A51)

This total grazing can be divided into the grazing on nanophytoplankton and diatoms ($G_{nano}$ and $G_{dia}$, Table A4):

$$G_{nano} = G_{tot} \cdot \frac{PhyN_{nano}}{G'}.$$  

(A52)

$$G_{dia} = G_{tot} \cdot \frac{PhyN_{dia} \cdot f_{Z}^{dia}}{G'}.$$  

(A53)

The values for the maximum grazing rate ($G_{max}$), the half-saturation constant ($K_{G}$) and the fraction of diatoms available for grazing ($f_{Z}^{dia}$) are listed in Table A9. The temperature dependency ($f_{T}$) is calculated in Eq. (A54) and the state variables are described in Table A1.

**A5 Temperature dependence of rates**

Temperature dependence of metabolic rates ($f_{T}$, Table A4) is taken into account through an Arrhenius function:

$$f_{T} = \exp \left( -\frac{4500 K \cdot \left( T^{-1} - T_{ref}^{-1} \right)}{1} \right).$$  

(A54)

Here, $T$ is the local temperature in Kelvin and $T_{ref}$ is a reference temperature (Table A10). The value $4500 K$ is the slope of the linear part of the Arrhenius function. Figure A1 illustrates how the metabolic rates decrease for $T < T_{ref}$ and increase for $T > T_{ref}$.

**A6 Nutrient limitation**

One factor controlling the metabolic processes in the phytoplankton is the intracellular ratios between nutrients and carbon.

Five different limiters are used for this; one that downregulates uptake of N and release of C when the N : C ratio becomes too high ($f_{lim,N}^{C_{max}}$, Sect. A6.1), three that downregulate photosynthesis when the nutrient : C ratios become too low ($f_{lim}^{Si}C_{min}$, $f_{lim}^{Si}C_{min}$ and $f_{lim}^{Fe}$, Sect. A6.2) and one that downregulates silicon assimilation when the Si : C ratio becomes too high ($f_{lim}^{Si}C_{max}$, Sect. A6.3).

These limiters will now be described after a general explanation of the function.

The way the intracellular ratios between nutrients and carbon limit uptake in the model is based on the work by Geider et al. (1998), but has been modified to the non-linear function, which is calculated as follows:

$$f_{lim}(slope, q_{1}, q_{2}) = 1 - \exp(-slope(|\Delta q| - \Delta q)^2).$$  

(A55)

Here, $\Delta q = q_{1} - q_{2}$ is the difference between the intracellular nutrient : C quota and a prescribed max or min quota, which is chosen depending on the situation.

The dimensionless constant slope regulates the degree of limitation for a given $\Delta q$.

**A6.1 Intracellular regulation of N uptake and C release**

Here we take a closer look at the limiter $f_{lim,N}^{C_{max}}$, which downregulates the metabolic processes listed in Table A11 when the intracellular N : C ratio becomes too high.
Table A9. Parameters for grazing.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
<th>Description and unit</th>
</tr>
</thead>
<tbody>
<tr>
<td>$f_{dia}^2$</td>
<td>0.50</td>
<td>Relative grazing preference for diatoms, dimensionless</td>
</tr>
<tr>
<td>$G_{max}$</td>
<td>2.40</td>
<td>Maximum grazing rate at 0°C [day$^{-1}$]</td>
</tr>
<tr>
<td>$K_G$</td>
<td>0.35</td>
<td>Half-saturation constant for grazing loss [(mmol N m$^{-3}$)$^2$]</td>
</tr>
</tbody>
</table>

Table A10. Parameters for limitation functions.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
<th>Description and unit</th>
</tr>
</thead>
<tbody>
<tr>
<td>$K_{Fe}^{nano}$</td>
<td>0.04</td>
<td>Half-saturation constant for nanophytoplankton Fe uptake [μmol Fe m$^{-3}$]</td>
</tr>
<tr>
<td>$K_{dia}^{Fe}$</td>
<td>0.12</td>
<td>Half-saturation constant for diatom Fe uptake [μmol Fe m$^{-3}$]</td>
</tr>
<tr>
<td>$q_{N:Cmin}^{max}$</td>
<td>0.04</td>
<td>Min intracellular N : C ratio for nanophytoplankton [mmol N mmol C$^{-1}$]</td>
</tr>
<tr>
<td>$q_{Si:Cmin}^{max}$</td>
<td>0.04</td>
<td>Min intracellular Si : C ratio for diatoms [mmol Si mmol C$^{-1}$]</td>
</tr>
<tr>
<td>$q_{N:Cmax}$</td>
<td>0.8</td>
<td>Max intracellular N : C ratio for diatoms [mmol N mmol C$^{-1}$]</td>
</tr>
<tr>
<td>slope$^N_{min}$</td>
<td>50</td>
<td>Minimum limiter regulator for N, [mmol C mmol N$^{-1}$]</td>
</tr>
<tr>
<td>slope$^N_{max}$</td>
<td>1000</td>
<td>Maximum limiter regulator for N, [mmol C mmol N$^{-1}$]</td>
</tr>
<tr>
<td>slope$^Si_{min}$</td>
<td>1000</td>
<td>Minimum limiter regulator for Si, [mmol C mmol N$^{-1}$]</td>
</tr>
<tr>
<td>slope$^Si_{max}$</td>
<td>1000</td>
<td>Maximum limiter regulator for Si, [mmol C mmol N$^{-1}$]</td>
</tr>
<tr>
<td>$T_{ref}$</td>
<td>288.15</td>
<td>Reference temperature for Arrhenius function [K]</td>
</tr>
</tbody>
</table>

Figure A1. The Arrhenius function plotted with the parameters used in REcoM2.

It is calculated with Eq. (A55) using the following parameters:

slope = slope$^N_{max}$, $q_1 = q_{N:Cmax}^{max}$, $q_2 = q_{N:C}$,

where slope$^N_{max}$ is listed in Table A10 along with the prescribed maximum N : C quota ($q_{N:Cmax}^{max}$). $q_{N:C}$ is the current intracellular quota.

In Fig. A2 it is illustrated how the limiter function changes with changing intracellular N : C quota; when the intracellular concentration of nitrogen increases as compared to carbon, the rate of the processes that are affected by the limiter will be downregulated. Total limitation ($f_{lim}^{N:Cmax} = 0$) occurs when the quota becomes equal to or higher than 0.2, or the equivalent of 21.2N : 106C.

A6.2 Intracellular regulation of C uptake

Photosynthesis is limited by the nutrients iron, nitrogen and in the case of diatoms also silicon.

Nitrogen and silicon limitation: in the case of N and Si, the regulation is controlled by the intracellular ratios N : C ($f_{lim}^{N:Cmin}$) and Si : C ($f_{lim}^{Si:Cmin}$). These limiters downregulate the rate of photosynthesis when the intracellular nutrient : C ratios become too low (Fig. A3).
Figure A2. Change in limiter function $f_{N:C_{\text{max}}\lim}$ with N:C quota.

Figure A3. Change in the limiter function $f_{N:C_{\text{min}}\lim}$ with N:C quota.

They are calculated with Eq. (A55) using the following parameters (Table A10):

- slope $= \text{slope}_N$, $q_1 = q_{N:C_{\text{min}}}$, $q_2 = q_{N:C}$
- slope $= \text{slope}_S$, $q_1 = q_{S:C_{\text{min}}}$, $q_2 = q_{S:C}$

Iron: for iron, the water's concentration is used to calculate the limitation. This is done using Michaelis–Menten kinetics:

$$f_{\text{Fe}} = \frac{\text{Fe}}{K_{\text{Fe}} + \text{Fe}} \quad (A56)$$

The half-saturation constants ($K_{\text{Fe}}^{\text{nano}}$ and $K_{\text{Fe}}^{\text{dia}}$) are listed in Table A10.

A6.3 Intracellular regulation of Si uptake

Diatom uptake of silicon is downregulated by the function $f_{\text{Si}:C_{\text{max}}\lim}$ (Fig. A4) when the intracellular Si:C ratio exceeds a set limit. The limiter function is described in Eq. (A55) and is calculated using the following variables (Table A10):

- slope $= \text{slope}_{\text{Si}}^{\text{max}}$, $q_1 = q_{\text{Si}:C}$, $q_2 = q_{\text{Si}:C_{\text{max}}}$
Supplementary material

The supplements contain the full code for REcoM2 along with a manual for FESOM–REcoM2, containing an overview of the code structure and instructions for running the coupled model.

The Supplement related to this article is available online at doi:10.5194/gmd-7-2769-2014-supplement.

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V. Schourup-Kristensen et al.: A skill assessment of FESOM–REcoM2


Vertical supply of iron to the mixed layer of the Southern Ocean: The ocean model effect

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Submitted

Abstract In the Southern Ocean, the micronutrient iron plays a key role as a limiting nutrient for phytoplankton growth. Studies of the iron supply to the surface mixed layer have traditionally focused on the contribution from aeolian and sedimentary input, but recent work has highlighted the importance of the supply from below. We performed a model study in which the biogeochemical model REcoM² was coupled to two different ocean models, the Finite Element Sea-ice Ocean Model (FESOM) and the MIT general circulation model (MITgcm). We analyzed the magnitude of the iron sources from below in the two models and assessed how the choice of ocean model impacts the iron fluxes across the base of the mixed layer and thereby also the modeled net primary and export production. Our results show a remarkable difference in terms of magnitude and mode of iron transport; the mean iron supply is four times higher in MITgcm than in FESOM, and the dominant pathway is through entrainment, whereas diffusion dominates in FESOM. This difference also has the effect that NPP is 48 % higher in FESOM than in MITgcm and that the export efficiency, the ratio between net primary production and export production, is smaller. In conclusion our study shows that the ocean model, and especially the mixed layer depth, has a key influence on the modeled iron fluxes into the surface mixed layer and thereby also on net primary and export production in the Southern Ocean.
6.1 INTRODUCTION

The Southern Ocean is known as one of the so-called High-Nutrient Low-Chlorophyll (HNLC) areas, characterized by low levels of chlorophyll $a$ despite of high concentrations of macronutrients. Since the early 1990’ies it has been recognized that iron, which for instance is needed for phytoplankton photosynthesis and respiration, is an important limiting factor for primary production in the Southern Ocean (e.g. Martin, 1990; Martin et al., 1990; Martin et al., 1991).

Dissolved iron is supplied to the surface mixed layer by two mechanisms; remineralization of organic material (Boyd et al., 2010) and so-called new supply. The latter is provided by external sources such as dust deposition (e.g. Wagener et al., 2008), remineralization from the sediments (e.g. Blain et al., 2001), release from icebergs (Lin et al., 2011) and it is supplied from the deeper water by physical mechanisms such as upwelling, entrainment and diffusion (e.g. Tagliabue et al., 2014). The iron supply from the deep water plays an especially large role in the remote waters in the Southern Ocean (de Baar et al., 1995; Löscher et al., 1997) because the other sources tend to be localized in their distribution. The amount of iron supplied through the different sources is, however, not well quantified. Recent modeling studies have shown that the sediment source plays a larger role than the other external iron sources (Lancelot et al., 2009; Tagliabue et al., 2009), but studies regarding the impact of the supply from below are sparse.

In the Southern Ocean, large scale upwelling occurs south of the polar front (e.g. Speer et al., 2000), and it has been suggested that this contributes significantly to the total iron input to the surface water, with estimated average values ranging from 8 to 50 μmol m$^{-2}$ yr$^{-1}$ (Watson et al., 2000; Hoppema et al., 2003; de Baar et al., 1995). But Tagliabue et al. (2014) argued that entrainment of iron through deepening of the mixed layer in the fall dominate the supply of iron from below, and that upwelling is of little importance when averaged over the entire Southern Ocean as it is cancelled out by the removal of iron through downwelling. The significance of diffusion likewise differs between studies; consistent with the findings of Tagliabue et al. (2014), Croot et al. (2007) showed that diffusion was of minor importance east of New Zealand, whereas Law et al. (2003) found that vertical eddy diffusion played a large role in the Antarctic Circumpolar Current (ACC) south of New Zealand. The effect of horizontal iron advection across the base of the mixed layer has been argued to play a role south of Australia (Ellwood et al., 2008), something that is induced by the combination of a southwards shoaling pycnocline and northwards wind-driven Ekman transport.

The flux of iron across the base of the mixed layer is dependent on the ocean mixing and circulation, which for example controls the depth of the mixed layer, the strength of the upwelling and the volume of entrained water. But it is likewise affected by the iron concentration just below the mixed layer and the iron gradient across the base of the mixed layer. The ferricline can be defined as the depth at which the iron gradient ($\partial Fe/\partial z$) has its highest value. The relative position of the ferricline to the base of the mixed layer has a large impact on the iron flux into the mixed layer; when the mixed layer depth (MLD) is shallower than the ferricline, only little iron can be supplied to the surface layer, and when the MLD comes below the ferricline a much larger supply can take place (Tagliabue et al., 2014).
In Ocean General Circulation Biogeochemical Models, a realistic ocean circulation and mixing is especially difficult to reproduce in the Southern Ocean (e.g. Downes et al., 2014), and this has the consequence that the export production (EP) estimated by the biogeochemical components varies widely in this region (Doney et al., 2004; Najjar et al., 2007). Models are, despite of this ambiguity, a tool routinely used to investigate the dynamics of export and primary production in the Southern Ocean (e.g. Taylor et al., 2013; Marinov et al., 2006), and although biogeochemical models are inherently simplistic and require strict validation (e.g. Assmann et al., 2010; Schourup-Kristensen et al., 2014), a large uncertainty regarding the vertical fluxes of iron is additionally introduced by the representation of ocean mixing and advection. Knowledge about the way the ocean model impacts the vertical iron fluxes is therefore crucial in order to interpret results regarding the complex interplay between physics and biology in the present, but also in future perturbations of the Southern Ocean.

In this paper we examine the impact of the Ocean General Circulation Model (OGCM) on the input of iron to the mixed layer of the Southern Ocean from below. We do this by comparing the results of two global OGCMs, the Finite Element Sea-ice Ocean Model (FESOM) and the Massachusetts Institute of Technology general circulation model (MITgcm), coupled to the same biogeochemical model, (The Regulated Ecosystem Model, REcoM). We assess how the ocean model impacts the magnitude and spatial distribution of net primary production (NPP) as well as the amount of carbon that is exported to the deep ocean.

6.2 Methods

6.2.1 Experiment

Two similar experiments were carried out with the coupled FESOM-REcoM (Schourup-Kristensen et al., 2014) and MITgcm-REcoM (Hauck et al., 2013) models. The ocean models were spun-up for 300 years after which the coupled REcoM-OGCM runs were started and run for a total of 50 years. The CORE-I normal year forcing (Large and Yeager, 2009) was used throughout the runs, and the analysis carried out here is based on the last five years of the coupled runs.

6.2.2 Ocean model I: FESOM

The Finite Element Sea-Ice Ocean Model (FESOM) is a global ocean circulation model, solving the primitive equations using the hydrostatic and boussinesq approximations. It is coupled to a dynamic-thermodynamic sea-ice model, which operates on the same surface mesh. Discretization is carried out using the finite element method (Sidorenko et al., 2011; Wang et al., 2014).

The sub-grid scale processes of vertical mixing are parameterized using a Pacanowski-Philander type scheme, which ensures increased mixing as stratification weakens (Pacanowski and Philander, 1981). This is combined with additional mixing near the surface over a depth defined by the Monin-Obukhov length, following the suggestion of Timmermann and Beckmann (2004) to avoid unrealistically shallow summer mixed layers. The background vertical eddy diffusion is set to $1 \cdot 10^{-4} \text{ m}^2 \text{s}^{-1}$ for momentum and $1 \cdot 10^{-5} \text{ m}^2 \text{s}^{-1}$ for tracers, with a maximum value of $2 \cdot 10^{-2} \text{ m}^2 \text{s}^{-1}$.
Horizontal mixing along neutral density slopes is incorporated following Redi (1982) and the eddy induced advection is implemented through the Gent and McWilliams scheme (Gent and Mcwilliams, 1990) with a critical neutral slope of 0.004.

For the spin-up, temperature and salinity fields were initialized with values from the PHC3.0 climatology (Steele et al., 2001) and the sea-ice was taken from a previous simulation. The sea surface salinity is restored towards the PHC3.0 climatology with a piston velocity of 10 m per 60 days. The version of FESOM used in the current run is presented by Sidorenko et al. (2011).

FESOM uses a prismatic mesh which is horizontally triangulated and unstructured, but vertically aligned, with the prisms divided into tetrahedra. Calculations are carried out on an A-grid (Arakawa and Lamb, 1977), with all fluxes and concentrations being calculated at nodes located in the corners of the prisms (Fig. 6.1). The mesh used here is similar to the one used by Schourup-Kristensen et al. (2014), with the horizontal resolution ranging from 15 km in the polar regions to 180 km in the central subtropical gyres. For a better description of the equatorial undercurrent and coastal processes, a resolution of 40 km was applied around the Equator and along the coasts. In the Southern Ocean the resolution ranges from 15 km close to the Antarctic to 150 km at 50°S. In the vertical, FESOM uses z-coordinates and the mesh has 32 levels. The resolution of the mesh increases towards the surface with 9 levels located in the upper 100 meters of the water column.

6.2.3 Ocean model II: MITgcm

As second general circulation model we use the Massachusetts Institute of Technology general circulation model (MITgcm, Marshall et al., 1997), which also solves the primitive equations under the Boussinesq approximation, in its hydrostatic mode. The model is discretized on a regular latitude-longitude-depth grid using the Arakawa C grid (Fig. 6.1). Longitudinal resolution is constant at 2°, while the latitudinal resolution varies from 0.38° to 2°, with higher resolution around the equator to better resolve the equatorial current systems, and increasing resolution southwards of the equator. In the vertical we use 30 layers, with 6 layers within the uppermost 100 m and layer thickness increasing with depth up to 500 m below a depth of 3700 m. The setup used here excludes the Arctic north of 80°N and we use the bathymetry of Timmermann et al. (2010).

Our setup includes a dynamic sea-ice model (Losch et al., 2010), a parameterization for sub-grid-scale mixing (Gent and Mcwilliams, 1990) with variable coefficient
and for density-driven downslope flows (Campin and Goosse, 1999). Vertical mixing
within the mixed layer is parameterized using the $k$-profile parameterization (Large
et al., 1994). The setup is described further by Hauck et al. (2013).

6.2.4 Biogeochemical module: RECOm2

The Regulated Ecosystem Model with two phytoplankton classes (RECOm2) belongs
to the class of so-called quota models, in which the intracellular nutrient ratios are al-
lowed to differ from the Redfield ratios within certain limits. Coupled to the MITgcm,
different versions of RECOm have been used for carbon cycle research in the Southern
Ocean (Hauck et al., 2013; Taylor et al., 2013; Losch et al., 2014), and it has recently
been coupled to the Finite Element Sea-ice Ocean Model, which similarly operates
in the global domain (Schourup-Kristensen et al., 2014).

RECOm2 has two phytoplankton classes, nanophytoplankton and diatoms, and one
group of zooplankton. It describes the nutrients nitrogen, silicon and iron. The iron
cycle is based on the work by Parekh et al. (2005). The concentration of dissolved iron
is changed by uptake and release from phyto- and zooplankton, remineralization of
detritus and by scavenging. The intracellular iron to carbon ratio is constant, and the
biological uptake and release of iron is hence tied to the carbon cycle.

The aeolian input of iron is prescribed as monthly climatologies (Luo et al., 2003;
Mahowald et al., 2003). We assume that the dust contains 3.5% of iron, of which 1.5 %
is instantly released in the surface ocean and the rest is lost. Iron is further added to
the water from the sediments, where remineralization of iron is calculated based on
the rate of carbon remineralization from the sediments to the ocean following Elrod
et al. (2004). A full description of the version of RECOm2 used in this study, including
parameter settings, can be found in Schourup-Kristensen et al. (2014).

RECOm2 is initialized with arbitrarily small values in all tracers except for dissolved
iron (DFe), dissolved inorganic nitrogen (DIN), dissolved inorganic carbon (DIC), dis-
solved inorganic silicon (DSi) and Total Alkalinity (TA). TA and DIC are initialized with
contemporary values from the GLODAP dataset (Key et al., 2004) and DSi and DIN
with values from the Levitus World Ocean Atlas from 2005 (Garcia et al., 2006). So
far, the measurements of iron profiles are too sparse to interpolate to a global field.
We therefore initialize the field of dissolved iron with output from the earth system
model PISCES (Aumont and Bopp, 2006), which has been modified in the Southern
Ocean to the mean profiles presented by Tagliabue et al. (2012).

6.2.5 Calculations

We assess the mean flux of dissolved iron across the base of the mixed layer in the
Southern Ocean south of $50^\circ$S, through vertical advection (up- and downwelling),
horizontal advection, en- and detrainment and vertical eddy diffusion (Fig. 6.2). We
define the mixed layer using the density threshold criteria of 0.03 kg m$^{-3}$ following
de Boyer Montegut et al. (2004).
Figure 6.2: Schematics of the main physical processes responsible for bringing or removing iron to or from the mixed layer of the ocean (After Levy et al., 2013, Fig. 1b).

The flux of dissolved iron across the base of the mixed layer ($F(\text{DFe})$) is calculated following Levy et al. (2013).

$$F(\text{DFe}) = -\text{DFe}_h \left( \text{w}_h + \text{u}_h \cdot \nabla H_h + \frac{\partial h}{\partial t} \right) + \left( K_z \cdot \frac{\partial \text{DFe}}{\partial z} \right)_h + \int_0^h \text{E}_{\text{DFe}} \, dz \quad (6.1)$$

Here, the subscript $h$ denotes the depth of the base of the mixed layer. The subduction term (i.e. Cushman-Roisin, 1987) entails the supply of iron from up-/downwelling ($\text{DFe}_h \cdot \text{w}_h$), from lateral advection across the sloping base of the mixed layer ($\text{DFe}_h \cdot \text{u}_h \cdot \nabla H_h$) and from entrainment through the deepening/shoaling of the mixed layer over time ($\text{DFe}_h \cdot \partial h / \partial t$). The variable $\text{DFe}_h$ denotes the iron concentration at the base of the mixed layer and $\text{w}_h$ and $\text{u}_h = (u_h, v_h)$ are the vertical and horizontal velocity at the base of the mixed layer, respectively. The slope of the base of the mixed layer is defined as $\nabla H_h = (\partial h / \partial x, \partial h / \partial y)$ and $\partial h / \partial t$ is the change in the MLD with time.

The vertical eddy diffusion term covers the mixing from sub-grid scale turbulence, causing a transport of iron across the base of the mixed layer and is calculated online in both models. Here, the variable $K_z$ denotes the diffusivity coefficient and $\partial \text{DFe} / \partial z$ is the iron gradient, both at depth $h$.

The Eddy mixing term denotes the eddy mediated isopycnal transport of iron across the sloping base of the mixed layer. It is parameterized and is added to the advection term during online calculation in both models. We accordingly do not show explicit numbers of the iron supply through eddy mediated transport in the current study.

To obtain the yearly input of iron to the mixed layer from below, Eq. (6.1) is integrated over time. The model output has been saved with monthly intervals in both models, and the offline calculated iron fluxes are thus calculated for this time interval.
In models, the calculation of vertical fluxes is restricted by the finite nature of the model grids. Consequently, differences in the calculations of iron transport through up-/downwelling and diffusion occur due to the formulation of FESOM and MITgcm on an A-grid and C-grid respectively (Fig. 6.1).

For the up- and downwelling, the velocity and the iron concentration are defined in the same grid point in FESOM, whereas the iron concentration below or above the flux depth is used in the MITgcm, dependent on the direction of the vertical velocity.

For the vertical eddy diffusivity, the iron gradient at the point of the vertical eddy diffusion coefficient is needed. In FESOM, we used the gradient above the $K_z$-node due to the MLD definition. In MITgcm, the gradient is calculated as the difference between the concentration depth above and below the $K_z$, which is defined at the flux-points (Fig. 6.1).

For the horizontal advection, the result from FESOM is interpolated onto a regular $1^\circ \times 1^\circ$ grid, and calculations are performed using the centered difference gradient for both models.

### 6.3 Results

#### 6.3.1 Yearly mean iron supply and biological production

The mean supply of iron across the base of the mixed layer south of $50^\circ$S varies widely between the two models (Fig. 6.3), summing up to $20 \mu$mol Fe m$^{-2}$ yr$^{-1}$ in FESOM, and to $84 \mu$mol Fe m$^{-2}$ yr$^{-1}$ in MITgcm.

En- and detrainment transports iron into the surface layer during mixed layer deepening, and removes it during shallowing. It is important in both models, supplying $4.3 \mu$mol Fe m$^{-2}$ yr$^{-1}$, or $22\%$ of the mean iron supply from below, in FESOM, and in the MITgcm it is the largest of the iron sources with a total mean of $44 \mu$mol Fe m$^{-2}$ yr$^{-1}$, or $53\%$ of the iron supply (Fig. 6.3).

Vertical eddy diffusion across the base of the mixed layer occurs due to small-scale turbulent mixing brought on by energy cascading from for instance surface wind mixing. It is the dominant mechanism for vertical iron transport in FESOM, summing up to $13 \mu$mol Fe m$^{-2}$ yr$^{-1}$ and comprising $63\%$ of the total iron supply from below. In the MITgcm, the vertical eddy diffusion is the smallest of the sources taken into account, summing up to $2 \mu$mol Fe m$^{-2}$ yr$^{-1}$, or $2.5\%$ of the total mean supply (Fig. 6.3).

South of the Antarctic Polar Front upwelling dominates over downwelling, and in our study the mean upwelling south of $50^\circ$S sums up to $29$ Sv in FESOM and $27$ Sv in the MITgcm ($1$ Sv = $10^6$ m$^3$ s$^{-1}$), both slightly lower than the $34$ Sv estimated by Sloyan and Rintoul (2001). The net supply of iron to the mixed layer through upwelling is relatively small in both models, summing up to $0.9 \mu$mol Fe m$^{-2}$ yr$^{-1}$, or $4\%$ of the total supply from below in FESOM and to $5.3 \mu$mol Fe m$^{-2}$ yr$^{-1}$, or $6\%$ in MITgcm (Fig. 6.3).

The pycnoclines, and hence also the MLD, shoal towards the south in the Southern Ocean, while the meridional flow of the surface water is in a northerly direction due to Ekman flow induced by the westerlies. This has the consequence that water is transported across the sloping base into the mixed layer, significantly contributing to the inflow of dissolved iron in both models (Fig. 6.3); in FESOM it comprises
2.1 μmol Fe m$^{-2}$ yr$^{-1}$, or 11% of the mean supply, and in MITgcm 33 μmol Fe m$^{-2}$ yr$^{-1}$ (39%).

New iron is further supplied to the surface water through aeolian and sedimentary input in REcoM2. The input from dust south of 50°S comprises 1.29 μmol Fe m$^{-2}$ yr$^{-1}$ in both models, and the total supply from the sediment averages 0.43 and 0.45 μmol Fe m$^{-2}$ yr$^{-1}$ in FESOM and MITgcm, respectively.

The annual average NPP south of 50°S sums up to 3.1 Pg C yr$^{-1}$ in FESOM, and in MITgcm it equals 2.1 Pg C yr$^{-1}$ (Table 6.1). Despite of the difference in NPP between the models, both results are within the range of previous NPP estimates (e.g. Carr et al., 2006). The annual EP, calculated at a reference depth of 100 m, sums up to 1.12 and 1.22 Pg C yr$^{-1}$ in FESOM and MITgcm, respectively, which is close to values from previous studies (Table 6.1).

For a model run in equilibrium, new production equals the export production, and we can consequently define the export efficiency as the ratio between the modeled EP and NPP. In FESOM, the mean export efficiency is 0.36 and in MITgcm it is 0.58 (Table 6.1). Previous studies (Boyd et al., 2005; Bowie et al., 2009; de Jong et al., 2012) show that the ratio between new and total NPP is highly variable, depending on the availability of iron, with measurements ranging from 0.05 to 0.6 EP/NPP (Table 6.1). The export efficiency in FESOM of 0.36 is thus within this range, while the MITgcm value of 0.58 is in the higher end, consistent with less iron limitation (e.g. Bowie et al., 2009).

### 6.3.2 Seasonal cycle

**Mixed layer depth and ferricline**

Consistent with observations (Sallée et al., 2010), the mean seasonal cycle of the mixed layer in the two models reaches the shallowest point in January and the deepest in August and September (Fig. 6.4). The MLD in FESOM is consistently shallower than in the observations, with the mean MLD$_{max}$ being 21 m too shallow, whereas the MLD in MITgcm is too shallow from January to May, and too deep from June to
Table 6.1: Yearly NPP, EP and export efficiency south of 50°C for the two models as well as results from previous studies.

<table>
<thead>
<tr>
<th>Value</th>
<th>Unit</th>
<th>FESOM</th>
<th>MITgcm</th>
<th>Others</th>
</tr>
</thead>
<tbody>
<tr>
<td>NPP</td>
<td>[Pg C yr⁻¹]</td>
<td>3.1</td>
<td>2.1</td>
<td>1.1 - 4.9ᵃ</td>
</tr>
<tr>
<td>EP</td>
<td>[Pg C yr⁻¹]</td>
<td>1.1</td>
<td>1.22</td>
<td>1ᵇ</td>
</tr>
<tr>
<td>Export efficiency</td>
<td></td>
<td>0.36</td>
<td>0.58</td>
<td>0.05 - 0.6ᶜ</td>
</tr>
</tbody>
</table>

ᵃ(Carr et al., 2006).
ᵇ(Schlitzer, 2002; Nevison et al., 2012).
ᶜ(Bowie et al., 2009; Boyd et al., 2005; de Jong et al., 2012).

Figure 6.4: The seasonal development of the modeled mean iron profile south of 50°C in the upper 200 meter. The depth of the modeled and observed mixed layer and the ferricline is likewise plotted. Observations of MLD from Sallée et al. (2010). a) FESOM b) MITgcm

November. The MITgcm’s mean MLDₘᵢₙ is 20 m shallower than in the observations, and the MLDₘₐₓ is 32 m deeper.

The depth of the mixed layer has a large impact on the vertical iron profile in the two models and hence also on the relative position of the MLD and the ferricline (Fig. 6.4). Following Tagliabue et al. (2014), the ferricline is defined as the depth where the maximum iron gradient occurs. In FESOM, the mean MLD is located well above the ferricline from January to April (Fig. 6.4a), and the iron concentration at the base of the mixed layer is consequently very low during these months. In MITgcm, the mean MLD is situated at approximately the same depth as the ferricline from January to May, and below it from June to November (Fig. 6.4b), and the mean iron concentration at the base of the mixed layer is consequently higher throughout the year in this model as compared to FESOM.

Iron supply

The iron supply through entrainment shows a clear seasonal cycle in both models, being large and positive during mixed layer deepening in the fall and large and negative in spring (Fig. 6.5a and b). The seasonal deepening of the MLD starts earlier in the MITgcm (Fig. 6.4), leading to a longer period of net entrainment in this model. Despite of the large detrainment in the spring in both models (Fig. 6.5a and b), the cumulative iron flux from entrainment is positive until October in FESOM (not shown),
Figure 6.5: The monthly mean of iron supply to the mixed layer through entrainment, vertical eddy diffusion, horizontal advection and upwelling in a) FESOM and b) MITgcm. Notice the different axes and scales. The monthly mean of the iron concentration in the mixed layer, the total NPP as well as the contributions from nanophytoplankton and diatoms is plotted for c) FESOM and d) MITgcm. The seasonal cycle of the total NPP is calculated from the MODIS satellite and the VGPM algorithm (Behrenfeld and Falkowski, 1997).
with possible implications for the spring bloom as discussed below. In MITgcm, the positive cumulative iron flux from entrainment is positive from February onwards (not shown).

The magnitude of the iron flux from vertical eddy diffusivity differs greatly between the models (Fig. 6.3), and so does the timing of the supply; in FESOM the highest rate of supply occurs from June to November when the iron concentration at the base of the mixed layer is relatively high, whereas the major part of the supply from vertical eddy diffusion in MITgcm takes place from January to June during deepening of the mixed layer, as deepening leads to a higher diffusivity coefficient at the base of the mixed layer.

The maximum supply of iron through horizontal advection occurs in August in FESOM and in September in MITgcm, but the inflow starts two months earlier in MITgcm than in FESOM, due to the earlier deepening of the mixed layer (Fig. 6.5a and b).

The iron supply through upwelling is small in both models (Fig. 6.3), with the largest input occurring during winter and spring in both models (Fig. 6.5a and b). As the strength of the upwelling is similar in the two models, the larger mean flux in MITgcm (Fig. 6.3) can be attributed to the higher iron concentration at the base of the mixed layers (Fig. 6.7a and b).

**Biological production**

In FESOM, the timing of the spring bloom initialization is correct, but the maximum is reached two months earlier than in the observations (Fig. 6.5c). In MITgcm, the bloom is initiated slightly late, but the timing of the maximum fits well with the satellite-based estimates (Fig. 6.5d). FESOM thus has a better date of initialization and MITgcm a better peak timing, but both models have too steep increases in NPP during the spring bloom. The difference between the models shows that the circulation and mixing plays a role regarding the timing of the biological processes. One factor that may impact the timing of the spring bloom is the MLD, which is deeper in MITgcm than in FESOM north of $60^\circ$S. Less light thus reaches the phytoplankton in the mixed layer on average in MITgcm, pointing towards a larger role of light limitation in this model. This is further confirmed by MITgcm’s higher iron concentration in the surface water (Fig. 6.4).

The iron concentration in the mixed layer is relatively low in FESOM (Fig. 6.5c). As the nanophytoplankton have a smaller demand for iron through a smaller half saturation constant, they dominate production in this model (Fig. 6.5c). The larger amount of nanophytoplankton in FESOM further has the effect that a larger degree of remineralization takes place in the mixed layer, sustaining NPP in November and December (Fig. 6.5c). In MITgcm, the deeper mixed layer keeps the production light limited longer, and the iron concentration in the mixed layer higher. In August and September production is dominated by diatoms, which have a higher maximum growth rate in the model. When the iron surface concentration decreases towards the end of the year, the fraction of diatoms likewise decrease due to their higher half saturation constant (Fig. 6.5d).

The average composition of the NPP differs significantly between the two models; nanophytoplankton is responsible for 79% of the Southern Ocean NPP in FESOM, whereas they comprise 33% of the NPP in MITgcm. In a study based on satellite observations, Hirata et al. (2011) show that diatoms comprise less than 20% of the biomass in the Southern Ocean (Fig. 1 in Vogt et al., 2013), in agreement with FESOM. Bracher et al. (2009) conversely show that diatoms dominate production in the South-
ern Ocean, which is in line with Hoffmann et al. (2008) who argued that diatoms are responsible for up to 75% of the NPP, and fits well with the result from MITgcm. The amount of diatoms in the Southern Ocean varies widely between model studies as shown by Vogt et al. (2013). Our study suggests that part of the explanation for this large difference could lie in the OGCMs and their representation of the vertical iron supply.

6.3.3 Spatial distribution

Mixed layer depth and ferricline
In FESOM, the yearly averaged MLD is shallower than in the observations in the whole Southern Ocean (Fig. 6.6a and b), though the spatial distribution with deeper mixed layers near the Polar Front and shallower in the temporarily ice-covered zone (TICZ) is captured. In MITgcm, the yearly averaged MLD is slightly shallow in the area of the TICZ, approximately south of 60°S (Fig. 6.6c). The deeper MLD in the area between 50 and 60°S is also captured in this model, though it becomes too deep in the Pacific and especially in the Atlantic sector of the Southern Ocean.

The seasonal amplitude of the MLD especially affects the amount of the entrained iron (Eq. 6.1). In FESOM, the amplitude is captured, and is even slightly high compared to observations, north of 60°S (Fig. 6.6d and e). But it is smaller than in the observations in the TICZ south of 60°S. The amplitude is closer to the observations in MITgcm (Fig. 6.6d and f), though it again becomes too large in the Atlantic and eastern Pacific Southern sectors of the Southern Ocean.
The mean iron concentration at the base of the mixed layer shows how much iron is available for supply to the surface mixed layer. In FESOM, the mean iron concentration at the base of the mixed layer is below 0.1 μmol Fe m$^{-3}$ in large areas of the Southern Ocean (Fig. 6.7a). The highest concentrations, ranging from 0.1 to 0.3 μmol Fe m$^{-3}$, occur in the Weddell and Ross Gyres as well as in the areas of deep winter mixed layers in the eastern Pacific and Indian Oceans. The mean iron concentration at the base of the mixed layer in the MITgcm range from 0.1 to 0.3 μmol Fe m$^{-3}$ in the majority of the Southern Ocean, and reaches as high as 0.4 μmol Fe m$^{-3}$ in the upwelling region south of 60°S (Fig. 6.7b).

The distance between the ferricline and the MLD highly affects the amount of iron that is supplied to the mixed layer (Tagliabue et al., 2014), and the spatial distribution of this field confirms that the MLD on average is located above the ferricline in FESOM (Fig. 6.7c). It is only in the area between 50 and 60°S and in the coastal convection zones, the places characterized by deep mixed layers (Fig. 6.6b) and high mean iron concentrations (Fig. 6.7a), that the mean MLD is located deeper than the ferricline in this model, leading to the relatively low iron concentration at the base of the mixed
layer. In MITgcm, the MLD is located below the ferricline north of 60°S (Fig. 6.7d), and south of 60°S the MLD and the ferricline are much closer to each other than in FESOM. The iron concentration at the base of the mixed layer is consequently higher in MITgcm (Fig. 6.7b).

Iron supply
In FESOM, entrainment mainly occurs north of 60°S (Fig. 6.8a). It is highly correlated with the seasonal amplitude of the MLD (Fig. 6.6e), with the higher rate of entrainment occurring in the areas of a larger MLD amplitude. This happens as the deep mixing has the effect that the MLD is located closer to or below the ferricline (Fig. 6.7c), leading to a higher iron concentration at the base of the mixed layer (Fig. 6.7a), thereby increasing the magnitude of the iron entrainment.

In MITgcm, a large degree of entrainment takes place north of 60°S (Fig. 6.8b). South of 60°S, MITgcm likewise has positive entrainment in large areas (Fig. 6.8b), brought on by the deeper mixed layer (Fig. 6.6c), the larger seasonal amplitude of the MLD (Fig. 6.6f) and the smaller distance between the MLD and the ferricline in MITgcm than in FESOM (Fig. 6.7c and d). In this model, there is no close relationship between the dynamics of the mixed layer (Fig. 6.6c and f) and the strength of the entrainment. This is due to the overall larger seasonal amplitude and larger depth of the mixed layer, causing the MLD to be below or close to the ferricline on average in the majority of the Southern Ocean (Fig. 6.7c and d).

The input of iron to the mixed layer through horizontal advection mainly occurs north of 60°S (Fig. 6.8c and d), where the meridional slope of the mixed layer depth is largest (Fig. 6.6b and e). Both models have a large input in the eastern Pacific part of the Southern Ocean and south of Tasmania (Fig. 6.8c and d). But where MITgcm additionally has input in the Indian sector, FESOM has a flux of iron out of the mixed layer here, with the consequence that the net mean supply through this mechanism becomes smaller. These negative fluxes are caused by the zonal flow of water, and thereby also iron, out of the mixed layer in the affected areas (not shown).

In FESOM, the higher iron fluxes from vertical eddy diffusion take place in a zone around Antarctica, providing the largest iron input to the TICZ, as well as to the areas of deep winter mixing (Fig. 6.8e). In this model, a high degree of vertical eddy diffusion occurs intermittently when the water column is unstable and when the wind mixing induces deeper mixed layers, and this intermittent nature is also visible in the spatial distribution. In MITgcm, the average input from vertical eddy diffusion can be disregarded due to its small contribution (Fig. 6.3), but the vertical eddy diffusion rate is higher in the temporary ice zone, especially in the Indian and Pacific sectors of the Southern Ocean. The supply north of 60°S is very small (Fig. 6.8f).

The largest input through upwelling occurs south of 60°S in both models (Fig. 6.8g and h), where the iron concentration is highest (Fig. 6.7a and b).

Net primary production
In FESOM the highest rates of NPP take place in the area between 60 and 50°S, where the strength of the entrainment is highest (Fig. 6.8a) and the mixed layer is deepest (Fig. 6.6b). South of 60°S the highest production occurs in the areas where the vertical eddy diffusion is highest.

The total NPP is higher in FESOM than in MITgcm (Table 6.1), and the differences in the rate of NPP are also evident in the spatial distribution (Fig. 6.9), especially where
Figure 6.8: Spatial distribution of the input of iron through entrainment (a and b), horizontal advection (c and d), vertical eddy diffusion (e and f) and upwelling (g and h) in FESOM and MITgcm, respectively. Notice the different scales.
Figure 6.9: Spatial distribution of the mean rate of summer (October to March) NPP in the Southern Ocean. a) FESOM, b) MITgcm.

FESOM has a larger production in the area of the deep winter mixed layers (Fig. 6.6c). In MITgcm the connection between the iron supply and the NPP is much less obvious. The pattern of the distribution does, however, show similarities between the two models; the main areas of elevated production occur north of 60°S and especially east of the Patagonian shelf in the Atlantic sector of the Southern Ocean in both models, as is also the case in reality (e.g. Arrigo et al., 2008).

6.4 DISCUSSION

6.4.1 Mixed layer depth

The differences we observe in the iron dynamics of our two model runs shows that the state of the ocean model, including the representation of the mixed layer, has a large impact on the biogeochemical results.

The mixed layer in the Southern Ocean is affected by a number of interconnected processes, such as wind mixing and buoyancy forcing, and it is particularly difficult to constrain in models (e.g. Sallée et al., 2013; Downes et al., 2014). The parameterization of the vertical mixing is one component that has a large impact on the MLD. The KPP scheme, which is used in MITgcm, is known to allow for deeper mixing when a stable thermocline is present (Large et al., 1994). This explains some of the difference between the models. Additionally, the MLD in FESOM has been improved in version 1.4, in which the KPP scheme is used (Downes et al., 2014). The slope of the pycnoclines, and thereby the strength of the stratification, is affected by the GM-parameterization of the eddy induced transport in both models. The strength of this transport may however differ between the two models due to tuning. A steeper slope of the pycnoclines in MITgcm likewise allows for deeper mixing than in FESOM. Lastly the difference in spatial resolution between the two models may also impact the MLD.

Another important physical factor is the strength of the large scale overturning circulation in the OGCMs, which largely controls the southward transport of iron into the Southern Ocean, and thereby the iron supply to and the iron fluxes across the
6.4 Discussion

The impact of the ocean circulation and mixing on the biogeochemical results is, however, highly complex, and made even more complicated by the internal feedbacks of the biogeochemical model. This study can thus not point to the exact reason for the differences in the biogeochemical results, but it highlights that the OGCM has a large impact on the biogeochemical results, and that the vertical iron supply can explain a large part of these differences.

6.4.2 Iron supply from below

Due to the large seasonal amplitude of the mixed layer (Fig. 6.5b), the iron supply through entrainment dominates the supply in MITgcm. As the average seasonal amplitude is overestimated compared to the observations, it is likely that the entrainment likewise is somewhat overestimated in this model. For FESOM, the mean seasonal amplitude is well represented (Fig. 6.5a), but the MLD is too shallow throughout the season, leading to a lower iron concentration at the base of the mixed layer and thereby less input through entrainment. It is thus likely that the entrainment is underestimated in this model and entrainment does most likely play a significant role in the Southern Ocean in reality, especially in the areas of deep wind mixing. This fits well with the study by Tagliabue et al. (2014), where they concluded that the entrainment contributed the most to the flux of iron to the mixed layer in the Southern Ocean. In the their analyzed observations, the flux of iron supplied through entrainment was $21 \text{ μmol Fe m}^{-2} \text{ yr}^{-1}$ (Tagliabue et al., 2014), a value between the results in this study (Fig. 6.3). The range provided by Tagliabue et al. (2014) are closest to the results from FESOM in the area north of $60^\circ$S, despite of the better fit regarding iron concentration and MLD in MITgcm. It must however be noted that the observed profiles in Tagliabue et al. (2014) are sparse and do not give a complete picture of the Southern Ocean.

Tagliabue et al. (2014) found that the supply of iron through vertical eddy diffusion was very small across the observation sites, similar to MITgcm, whereas Law et al. (2003) showed that vertical eddy diffusion played a significant role in the highly iron depleted waters south of New Zealand, agreeing more with the results from FESOM. The fact that vertical eddy diffusion dominates the supply of iron from below in FESOM also fits well with the results from an eddy-permitting model study by Dufour et al. (2013) who found that vertical eddy diffusion dominated the supply of DIC to the mixed layer of the Southern Ocean.

Previous studies focussing on the supply of iron through upwelling have usually calculated the flux by using the magnitude of the water transport multiplied by a constant iron concentration, such as 0.3 to 0.4 μmol Fe m$^{-3}$ (e.g. Watson et al., 2000; Hoppema et al., 2003). This results in a higher supply through upwelling than what is the case in the present study, and points to the important role the mixed layer plays in determining the vertical fluxes. Tagliabue et al. (2014) argued that the upwelling of iron on average is of minor importance in the Southern Ocean as it is cancelled out by downwelling north of the Polar Front, something that is true in both of our models. But despite of the insignificance of the upwelling for the iron flux into the mixed layer, it must be noted that this mechanism plays a large role regarding the supply of iron to the base of the mixed layer, from where it is brought upwards through other mechanisms. As the the iron concentration at the base of the mixed layer varies significantly between our models, this mechanism may be very important.
6.4.3 Other iron sources

In the Southern Ocean iron is supplied through several other sources including aeolian deposition, sedimentary input and release from icebergs and sea-ice. In FESOM, the average iron input from the sediment and the aeolian sources, is smaller than, or on the order of, the vertical sources considered in the current study, but likely important locally. Regarding MITgcm, the external sources comprises a much smaller fraction of the sources considered here and their impact on production is thus likewise much smaller. The size of the sediment source is, however, not well constrained and varies widely between models, with the models used in the current study having a source that is in the smaller end of the range normally used (Aumont and Bopp, 2006; Moore and Braucher, 2008). It is thus possible that the sediment source in reality plays a larger role than what our study indicates. Models are routinely used to quantify the magnitude of the external iron sources in the Southern Ocean (e.g. Lancelot et al., 2009; Wadley et al., 2014). The sediment source has been shown to play a significant role regarding the iron supply, especially in the Atlantic sector of the Southern Ocean and locally in proximity to islands (e.g. Lancelot et al., 2009; Tagliabue et al., 2009; Pollard et al., 2007).

In previous studies the supply of iron from below is normally disregarded. The current study does however highlight how the supply from below affects the total NPP and the phytoplankton composition in the Southern Ocean. Our study thus suggests that the performance of the ocean model, and thereby the iron supply from below, must be evaluated before assessing the importance of the external sources.

6.4.4 Biological production

Remote sensing products of Southern Ocean ocean color (e.g. Arrigo et al., 2008) have revealed that high primary production occurs in certain places. This includes areas downwind of the continents (e.g. Jickells et al., 2005), downstream of islands, continental shelves and topographic features (e.g. Blain et al., 2001; de Jong et al., 2012) and along the receding ice edge (Arrigo et al., 2008). This distribution of NPP indicates that the input of iron from external sources such as dust and sediments indeed play a large role in the Southern Ocean, concurring with model studies (e.g. Lancelot et al., 2009; Tagliabue et al., 2009). While we recognize the role of external iron sources locally in the Southern Ocean, we do argue that the vertical iron fluxes are important for maintaining the iron concentration in the Southern Ocean, especially in remote areas and below sea-ice.

6.4.5 Export efficiency

The value of the export efficiency denotes the fraction of exported NPP, and is therefore a key number regarding the model’s carbon cycle. For a model in equilibrium, the amount of nutrients supplied to the mixed layer must be balanced by the amount exported through sinking of organic material. The export ratio is consequently expected to be larger in a system with a larger input of nutrients.

After deriving export efficiencies (fe-ratios) from iron measurements in the strongly iron limited area south-west of New Zealand, Boyd et al. (2005) argued that a realistic range for the export efficiency in the area was 0.05 to 0.48. The higher ratios here
were connected to higher iron inputs. In line with this, de Jong et al. (2012) found an even higher export ratio of 0.59 in the Atlantic sector of the Southern Ocean where iron limitation is less severe. This connection between iron availability and export efficiency is also observed in our study, where MITgcm has the larger efficiency (Table 6.1).

In a measurement-based study, Bowie et al. (2009) found low export efficiencies (fe-ratios) in the severely iron depleted area south of Australia and remarked that regenerated iron must be important for production in this region. In FESOM, where production is dominated by the less iron demanding nanophytoplankton (Fig. 6.5a), the continued production in the summer is most likely likewise fueled by regenerated iron. As the turnover rate is higher for nanophytoplankton this iron “source” is most likely higher in FESOM than in MITgcm where diatoms dominate production (Fig. 6.5b).

Maiti et al. (2013) presented a collection of previous estimates of the export efficiency from the Southern Ocean, showing that they vary widely but that the export efficiency generally decreases with increasing NPP. This latter feature is also seen in our study, where FESOM has a higher NPP, but lower export efficiency than MITgcm. In MITgcm, the ratio of diatom production is higher than in FESOM, and as diatoms have a smaller remineralization rate this leads to the larger export efficiency in this model. This may indicate that models with diatom dominance have a higher export efficiency, something that again must be kept in mind when evaluating model runs regarding the marine carbon-cycle.

6.4.6 Future iron supply

The response of the Southern Ocean mixed layer depth to climate change is not fully understood. Two counteracting mechanisms can be predicted; shallowing due to stronger stratification (e.g. Sallée et al., 2013) and deepening due to stronger winds (e.g. Meijers, 2014). The current study shows that the direction of change could greatly impact the iron supply to the mixed layer, and thereby the biological production and export in the area. As the reproduction of a realistic mixed layer depth is already a challenge for ocean models under current forcing (e.g. Meijers, 2014; Downes et al., 2014), the modeled response of marine NPP and EP to climate change could be strongly influenced by unidentified biases in the ocean model, with large implications for the predicted carbon fluxes in the model.

6.5 Conclusion

Our study clearly shows that special care must be taken to adequately represent the ocean circulation and mixing in coupled biogeochemical ocean models, both in contemporary and in future runs, as it has a large effect on the modeled iron fluxes across the base of the mixed layer, and thereby the net primary production and the export efficiency.

Differences in the ocean state led to the sources of iron from below being very different in our two models, both in magnitude and in terms of what source dominates the iron supply. In FESOM, the supply of iron was lower than in MITgcm and the vertical eddy diffusion dominated the overall supply, while entrainment was important north of 60°S and during the spring bloom. Entrainment dominated the iron supply
in MITgcm due to a larger variability of the mixed layer in this model. Horizontal advection also plays a role in MITgcm.

The net primary production in both models was reasonable compared to satellite-based observations, but the species composition differed significantly between them, with diatoms being responsible for 67% of the NPP in MITgcm and 21% in FESOM. MITgcm’s larger iron supply from below, and thereby also higher mixed layer iron concentrations, caused the larger and more iron demanding diatoms to thrive in this model. The larger supply of iron in MITgcm also led to a higher export efficiency of 0.58, as opposed to 0.36 in FESOM. The ocean state, and thereby the mixed layer dynamics, of the ocean general circulation model is thus of major importance for studies of the carbon cycle in the Southern Ocean.

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Sensitivity of Southern Ocean phytoplankton composition to changes in model iron scavenging and sediment supply

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Abstract One way to study the impact of the Southern Ocean iron sources on biological production is by using ocean general circulation biogeochemical models (OGCBMs). Most of such studies agree that the sediment source is the most important of the external iron sources to the Southern Ocean. Neither this benthic iron flux nor the removal through scavenging of iron onto organic particles are, however, well constrained by measurements, and these fluxes thus vary widely between models. While many OGCBMs produce relatively similar numbers for the total net primary production (NPP) in the Southern Ocean, it is not clear to what extend the large differences in the iron supply and iron removal affects the modeled phytoplankton composition and opal export in the Southern Ocean.

We are presenting a sensitivity study, in which the strength of the benthic iron flux was varied simultaneously with the counteracting scavenging of iron in the water. We show that while changing the benthic iron flux substantially, it is possible to recreate both the mean iron profiles and the total NPP in the Southern Ocean by choosing an appropriate scavenging rate. The changed iron input does, however, change the spatial distribution of the NPP along with the amount of NPP from diatoms. A benthic shelf iron flux of 1.2\,\mu mol\,Fe\,m^{-2}\,day^{-1}, leading to a total input of $2.14 \times 10^9$\,mol\,Fe\,yr^{-1} to the Southern Ocean, leads to the best results when evaluating against knowledge of iron fluxes, as well as diatom distributions and opal export. Given the large degree of freedom regarding the modeled strength of the benthic iron flux and the scavenging rate, this study highlights the importance of evaluating the total NPP against observations, but in particular also against observations of diatom abundance and opal export. Data for both of these are, however, scarce, and the study thus shows how additional data collection regarding the Southern Ocean phytoplankton composition is necessary to further improve the evaluation of OGCBMs.
Iron is thought to be the main limiting factor of biological production in 30 to 50% of the world’s surface ocean (Moore et al., 2002; Aumont et al., 2003; Dutkiewicz et al., 2005). The main known areas are the so-called High-Nutrient Low-Chlorophyll (HNLC) regions: The Southern Ocean (Martin et al., 1990; Boyd et al., 2000), the Equatorial Pacific (Kolber et al., 1994; Behrenfeld et al., 1996) and the Northern Pacific (Martin and Fitzwater, 1988).

Traditionally, the most important external iron source to the ocean has been thought to be aeolian dust input (e.g. Jickells et al., 2005), but several studies have shown that the sediment source likewise plays an important role. An increase in the water’s concentration of dissolved iron towards the coast and shelf areas was first documented by Johnson et al. (1997) and has later been widely confirmed (e.g. Johnson et al., 2003; Moore and Braucher, 2008). Studies showing that coastally derived iron was carried far into the Pacific Ocean from the Aleutians (Lam et al., 2006) and Kamchatka (Lam and Bishop, 2008) further indicated that this source can not be ignored in large scale iron budgets.

Determining the origin of the dissolved iron in the water is, however, difficult. A new method for this, based on the stable isotopes of iron in seawater, has recently been developed (John and Adkins, 2010; Conway et al., 2013). The stable isotope technique has confirmed the importance of the sediment source of iron; Conway and John (2014) showed that sediment derived iron was advected far from its source region in the Equatorial Atlantic, and Radic et al. (2011) found that iron in the eastern equatorial Pacific originated in the sediments of Papua New Guinea.

In the Southern Ocean, the aeolian iron input supplies a significant amount of new iron in the areas downwind of Patagonia and Australia (e.g. Cassar et al., 2007), but since half of the atmospheric supply is lost on a length scale of 1500 km (Prospero et al., 1989) much less reaches the remoter areas of this ocean leading to widespread iron limitation. Satellite-based measurements of Southern Ocean chlorophyll nevertheless show consistent patches of high concentrations downstream of islands such as South Georgia (e.g. Korb and Whitehouse, 2004), the Kerguelen (e.g. Blain et al., 2001) and Crozet Island (e.g. Pollard et al., 2009). One explanation for these blooms is that the sediment source of iron also plays a role in the Southern Ocean, where released iron can travel relatively far when it is transported in the fast flowing jet streams of the Antarctic Circumpolar Current (de Baar et al., 1995; Croot et al., 2004a). Several model studies have confirmed the importance of the sediment source of iron, both in the Southern Ocean (Tagliabue et al., 2009; Lancelot et al., 2009; Wadley et al., 2014) and on the global scale (Moore and Braucher, 2008). The strength of the sediment source is, however, not well constrained by measurements and both the total supply and the parameterization of iron from the sediments therefore varies widely between models (e.g. Aumont and Bopp, 2006; Moore and Braucher, 2008; Lancelot et al., 2009). A large variation in the strength of the sediment source of iron is possible because the magnitude of the removal of iron through particle scavenging is not well known either, and modelers thus have a large degree of freedom regarding the tuning of the strength of the iron sources parallel with the particle scavenging.

In the Southern Ocean, the iron availability affects the phytoplankton composition (e.g. Takeda, 1998; Hoffmann et al., 2006; Wang et al., 2014b), which has an impact on the mean intracellular ratio between silicate and carbon in diatoms (Baines et al.,
7.2 METHODS

7.2.1 Ocean model

For the current study we are using the Finite Element Sea-ice Ocean Model (FESOM) version 1.4 (Wang et al., 2014a) coupled to the biogeochemical model REcoM2 described below.

FESOM is a global sea-ice ocean model, which solves the hydrostatic equations under the commonly applied Boussinesq approximations using the finite element method. FESOM 1.4 is an update from the version used in previous work with REcoM2 (Schourup-Kristensen et al., 2014a; Schourup-Kristensen et al., 2014b). Important improvements include a change of the parameterization of the diapycnal mixing from the Pacanowski-Philander parameterization (PP, Pacanowski and Philander, 1981) to the k-profile parameterization (KPP, Large et al., 1994). This ensures that vertical turbulent mixing occurs to a larger degree in stable stratified water. Secondly, the Gent-McWilliams parameterization (Gent and Mcwilliams, 1990), which controls the eddy-induced horizontal advection, has been tuned to improve the slope of the isopycnals in the Southern Ocean, something that has a large impact on the mixed layer depth (MLD) in the area. And finally, the background vertical mixing coefficient now increases with depth, where it was previously constant. A full description of the current model version can be found in Wang et al. (2014a).

FESOM runs on a grid that is horizontally triangulated and unstructured, and vertically stratified. In the Southern Ocean south of 35°S, the grid used for the current study contains 16,000 surface nodes and a total of almost 477,000 nodes. The horizontal resolution ranges from 15 km along the coast of the Ross Sea to around 160 km on the southern edge of the subtropical gyres (Fig. 7.1). The bottom topography south of 60°S is based on the Rtopo-1 dataset (Timmermann et al., 2010) and in the rest of the Southern Ocean it is based on data from the General Bathymetric Chart of the Ocean (www.gebco.net). Both datasets have 1 minute resolution. The grid is similar to the one used in Schourup-Kristensen et al. (2014b).

7.2.2 Biogeochemical model

The Regulated Ecosystem Model (REcoM2) is a so-called Nutrient-Phytoplankton-Zooplankton-Detritus model (e.g. Hauck et al., 2013; Schourup-Kristensen et al., 2014b). The limiting nutrients include nitrogen, silicate and iron and the model additionally describes the cycling of carbon. It has two phytoplankton classes, nanophy-
Figure 7.1: The horizontally triangulated grid used in the study, overlaid the bottom topography.
toplankton and diatoms and one class of zooplankton. The detritus is contained in one class and has a sinking velocity that increases with depth following Kriest and Oschlies (2008). REcoM2 employs variable stoichiometry between carbon and the macronutrients following Geider et al. (1998).

The concentration of dissolved iron in the water is increased through remineralization of phytoplankton and detritus, as well as by zooplankton excretion and respiration. Iron is removed from the water by phytoplankton growth and scavenging. The intracellular concentration of iron is linearly correlated with the intracellular carbon concentration, with a Fe:C ratio of 0.005 μmol Fe (mmol C)$^{-1}$ following Sunda and Huntsman (1995), and uptake limitation is calculated using Michaelis-Menten kinetics.

The calculation of the scavenging rate is based on Parekh et al. (2005), where only a fraction of the dissolved iron is available for scavenging as the rest is bound to ligands. It has been altered to make scavenging linearly correlated with the concentration of detritus in the water column, with the effect that the scavenging rate becomes higher in areas of high biological production, and lower in low production areas. The scavenging rate ($\text{Scav}$) is calculated as follows:

$$\text{Scav} = \kappa_{Fe} \cdot \text{DetC} \cdot \text{DFe}' \left[ \mu\text{mol Fe m}^{-3} \text{day}^{-1} \right]$$

(7.1)

Here, $\text{DFe}'$ (μmol Fe m$^{-3}$) denotes the concentration of the dissolved iron that is not bound to ligands, $\text{DetC}$ (mmol C m$^{-3}$) is the detritus concentration and $\kappa_{Fe}$ (mmol C m$^{-3}$) day$^{-1}$) is the mass specific scavenging rate.

The removal of iron by scavenging is balanced by aeolian and sedimentary iron inputs in the model. The dust field applied in REcoM2 (Mahowald et al., 2003; Luo et al., 2003) is assumed to contain 3.5% of iron by weight, of which 1.5% dissolves immediately in the surface ocean upon deposition and the rest is lost from the system.

The input of dissolved iron from the sediments is calculated following Elrod et al. (2004), so the iron flux from the benthos to the bottom water ($\text{BenF}_{Fe}$) is proportional to the release of inorganic carbon from the sediments:

$$\text{BenF}_{Fe} = \frac{\text{BenthosC} \cdot q_{Fe:C}^{sed}}{\rho_{ben}^{C}} \left[ \mu\text{mol Fe m}^{-2} \text{day}^{-1} \right]$$

(7.2)

The vertically integrated carbon concentration in the benthos is denoted by BenthosC (mmol C m$^{-2}$), $\rho_{ben}^{C} = 200$ day is the remineralization time scale for benthos carbon and $q_{Fe:C}^{sed}$ (μmol Fe mmol C$^{-1}$) is the benthic iron to carbon release ratio. The Fe:C ratio in the sediments can become higher than in organic material in general as lithogenic iron in the sediments is mobilized under anoxic conditions. In the current study it is the mass specific scavenging rate ($\kappa_{Fe}$) and iron to carbon concentration in the benthos ($q_{Fe:C}^{sed}$) that are varied as described next.

### 7.2.3 Experiments

For the current study we have performed a total of nine model runs, using REcoM2-FESOM 1.4; one control run and eight sensitivity runs (Table 7.1). For all runs the ocean model was spun-up for 300 years under the CORE-I normal year forcing (Large and Yeager, 2009), after which the biogeochemical model was started and run for 50
Table 7.1: Experiments performed, including the values of the Fe:C-ratio in the sediment \( q^{Fe:C}_{sed} \) and mass specific scavenging rate \( \kappa_{Fe} \).

<table>
<thead>
<tr>
<th>MODEL RUN</th>
<th>SEDIMENT FE:C-RATIO [( \mu \text{mol Fe mmol C}^{-1} )]</th>
<th>MASS SPECIFIC SCAVENGING RATE [( \text{mmol C m}^{-3} \text{day}^{-1} )]</th>
</tr>
</thead>
<tbody>
<tr>
<td>CTRL</td>
<td>0.01</td>
<td>0.0312</td>
</tr>
<tr>
<td>SED1.1</td>
<td>0.10</td>
<td>0.312</td>
</tr>
<tr>
<td>SED1.2</td>
<td>0.10</td>
<td>0.156</td>
</tr>
<tr>
<td>SED1.3</td>
<td>0.10</td>
<td>0.078</td>
</tr>
<tr>
<td>SED1.4</td>
<td>0.10</td>
<td>0.039</td>
</tr>
<tr>
<td>SED2.1</td>
<td>0.68</td>
<td>1.061</td>
</tr>
<tr>
<td>SED2.2</td>
<td>0.68</td>
<td>0.531</td>
</tr>
<tr>
<td>SED2.3</td>
<td>0.68</td>
<td>0.266</td>
</tr>
<tr>
<td>SED2.4</td>
<td>0.68</td>
<td>0.133</td>
</tr>
</tbody>
</table>

years, likewise under \text{CORE-I} \ enforcing. The results presented here are the averages of the last 10 years of the runs.

To investigate the effect of the strength of the sediment source of iron on the biological production in the Southern Ocean south of 35°S, two sets of experiments were performed (Experiment 1 and 2), each with a constant benthic Fe:C release ratio \( q^{Fe:C}_{sed} \), Eq. 7.2). In experiment 1, the benthic Fe:C release ratio was ten times higher than in the CTRL run, and in experiment 2, it was set to 0.68 \( \mu \text{mol Fe mmol C}^{-1} \) as estimated by Elrod et al. (2004).

For the two experiments, the scavenging rate was increased as compared to the CTRL run to account for the additional iron input. For each experiment, four different scavenging rates were tested \( \kappa_{Fe} \), Eq. 7.1), leading to total of four sensitivity runs in each experiment. These runs were named SED1.1 to 1.4 and SED2.1 to 2.4 in experiment 1 and 2, respectively (Table 7.1).

7.3 RESULTS

7.3.1 Sensitivity study

Iron

In \text{REcoM2}, the magnitude of the released iron from the sediments depend on the prescribed benthic Fe:C release ratio as well as the carbon concentration within the sediments (Eq. 7.2). The release consequently differs between all of the model runs, with the total input to the Southern Ocean south of 35°S ranging from 0.22 × 10^9 \text{ mol Fe yr}^{-1} in the CTRL run to 15.8 × 10^9 \text{ mol Fe yr}^{-1} in SED2.3 (Table 7.2). The benthic Fe:C ratio has a large impact on the amount of iron released from the sediments, and the total iron input varies with almost a factor 10 between experiment 1 and 2, whereas the intra-experiment difference, brought on by differences in the scavenging rate, is small (Table 7.2). The total iron release nevertheless increases with decreased scavenging rate in the first three studies (1.1 to 1.3 and 2.1 to 2.3), but
Table 7.2: Total iron input to the Southern Ocean south of 35°S from the sediments in all runs, the mean iron flux into the water on the upper 500 m of the water column and the percentage of the external iron input contributed by the sediments. Input from dust: $0.28 \times 10^9$ mol Fe yr$^{-1}$.

<table>
<thead>
<tr>
<th>RUN</th>
<th>IRON INPUT [10$^9$ mol Fe yr$^{-1}$]</th>
<th>MEAN FLUX [μmol Fe m$^{-2}$ day$^{-1}$]</th>
<th>BENTHIC INPUT [%]</th>
</tr>
</thead>
<tbody>
<tr>
<td>CTRL</td>
<td>0.22</td>
<td>0.12</td>
<td>44.0</td>
</tr>
<tr>
<td>SED1.1</td>
<td>2.00</td>
<td>1.33</td>
<td>87.7</td>
</tr>
<tr>
<td>SED1.2</td>
<td>2.10</td>
<td>1.30</td>
<td>88.2</td>
</tr>
<tr>
<td>SED1.3</td>
<td>2.14</td>
<td>1.20</td>
<td>88.4</td>
</tr>
<tr>
<td>SED1.4</td>
<td>2.00</td>
<td>0.96</td>
<td>87.7</td>
</tr>
<tr>
<td>SED2.1</td>
<td>13.8</td>
<td>9.60</td>
<td>98.0</td>
</tr>
<tr>
<td>SED2.2</td>
<td>14.9</td>
<td>9.70</td>
<td>98.2</td>
</tr>
<tr>
<td>SED2.3</td>
<td>15.8</td>
<td>9.42</td>
<td>98.3</td>
</tr>
<tr>
<td>SED2.4</td>
<td>15.1</td>
<td>7.83</td>
<td>98.2</td>
</tr>
</tbody>
</table>

then decreases again (Table 7.2). The increase happens as less scavenging means that more iron is available in the water for biological production, which leads to more carbon sinking into the sediments, thereby increasing the iron release (Eq. 7.2). This biological feedback mechanism will be discussed further below.

Turning to the vertical distribution of the mean benthic iron flux per unit area, it is largest in the sediments shallower than 500 m, with the maximum iron input occurring at a depth of 50 m in all runs (Fig. 7.2). The strength of the mean iron flux in the upper 500 m of the ocean differs greatly between experiment 1 and 2, concurring with the total iron input to the Southern Ocean (Table 7.2 and Fig. 7.2). The small intra-experimental dependence of the mean iron flux in the upper 500 m on the scavenging can be explained by the fact that the removal of iron through scavenging is less important in areas where iron is continuously supplied, such as the shelf area. The supply allows for biological production to continue in coastal regions despite of a high scavenging rate, and the organic carbon maintains the benthic iron release.

In contrast to the mean sediment flux, the mean iron profile in the Southern Ocean does not change substantially between the CTRL run, experiment 1 and experiment 2, but it changes slightly when the scavenging is changed (Fig. 7.3). On the shelves (> −1000 m, Fig. 7.3a and b), decreased scavenging leads to increased iron concentrations throughout the water column (Fig. 7.3a and b). The largest increases in the iron concentration on the shelves occur in the runs with the highest benthic Fe:C ratio, as more iron is delivered to the water in these runs (Fig. 7.3b). In the open water (< −1000 m, Fig. 7.3c and d), the same pattern occurs, with higher iron concentrations in the runs with less scavenging, though the effect is less clear in the deep water than in the shallow water. One thing that stands out is the fact that it is possible to virtually produce the open ocean iron profile of the CTRL run in both experiment 1 and 2 (Fig. 7.3c and d). Compared to the observations, all of the model runs have too low surface iron concentrations in the open ocean, whereas the deep iron concentrations
Figure 7.2: Vertical profiles of the mean flux from the sediment into the water calculated south of 35°S. a) Experiment 1 (Fe:C$_{sed}$ = 0.1 μmol Fe mmol C$^{-1}$), and CTRL run. b) Experiment 2 (Fe:C$_{sed}$ = 0.68 μmol Fe mmol C$^{-1}$), and CTRL run.

become too high, especially in SED2.1 to 2.4 where the sediment input is highest. For the shelves, the modeled iron profiles are closer to the observations than in the open ocean (Fig. 7.3).

Residence time of iron

The residence time of iron in the model ($T_{res}$) is calculated using the removal rate of iron, which depends on the scavenging of iron onto particles and on the sinking speed of iron-containing detritus. The residence time is calculated for the Southern Ocean south of 35°S for the upper 500 m and for the whole water column (Table 7.3).

\[
T_{res} = \frac{Fe_V}{Fe \text{ removal rate}_V} = \frac{Fe_V}{Scav_V + 0.005 \text{ μmol Fe mmol C}^{-1}\cdot Det_V \cdot w_{det} \cdot H^{-1}} \tag{7.3}
\]

Here the term $Fe_V$ denotes the water’s iron inventory (μmol Fe), $Scav_V$ denotes the total scavenging loss as described in Eq. (7.1) multiplied by the water volume (μmol Fe day$^{-1}$). $Det_V$ is the water’s inventory of detritus (mmol C), $w_{det}$ is the sinking speed of the detritus and $H$ is the depth of the water column considered.

The residence time in the model is clearly affected by the rate of scavenging, both in the upper 500 m of the water column and in the whole water column of the Southern Ocean. The longest upper ocean residence time of 18 years occurs in the CTRL run where the scavenging rate is the lowest, and the shortest residence time occurring in SED2.1 where the scavenging is highest (Table 7.3). The scavenging rate controls the residence time directly by removing iron from the water, but it also controls it indirectly as less scavenging leads to higher iron concentrations in the water, which in turn leads to more biological production and thereby also a higher export of iron through the sinking of detritus. In the CTRL run, the scavenging component and the sinking component of Eq. (7.3) are of equal magnitude, but as scavenging is increased in experiment 1 and 2, the scavenging becomes more important. In experiment 2 the scavenging is on average 50 times higher than the removal of iron through sinking of detritus.
Figure 7.3: Vertical profiles of the mean iron concentration in the Southern Ocean south of 35°S. a) Experiment 1, mean iron profiles on the shelves (>−1000 m). b) Experiment 2, mean iron profiles on the shelves (>−1000 m). c) Experiment 1, mean iron profiles in the open ocean (<−1000 m). d) Experiment 2, mean iron profiles in the open ocean (<−1000 m). All subplots additionally have the vertical profile from the CTRL run and from the observations from Tagliaabue et al. (2012), Table 1. Notice the double log-scale.

Table 7.3: Residence time of iron in the Southern Ocean south of 35° for the entire water column and waters shallower than 500 m.

<table>
<thead>
<tr>
<th>RUN</th>
<th>Upper 500 m</th>
<th>Whole Water Column</th>
</tr>
</thead>
<tbody>
<tr>
<td>CTRL</td>
<td>18.1 years</td>
<td>266.8 years</td>
</tr>
<tr>
<td>SED1.1</td>
<td>2.5 years</td>
<td>45.2 years</td>
</tr>
<tr>
<td>SED1.2</td>
<td>3.6 years</td>
<td>53.0 years</td>
</tr>
<tr>
<td>SED1.3</td>
<td>5.5 years</td>
<td>68.3 years</td>
</tr>
<tr>
<td>SED1.4</td>
<td>9.2 years</td>
<td>101.7 years</td>
</tr>
<tr>
<td>SED2.1</td>
<td>3.6 months</td>
<td>7.8 years</td>
</tr>
<tr>
<td>SED2.2</td>
<td>5 months</td>
<td>8.6 years</td>
</tr>
<tr>
<td>SED2.3</td>
<td>7.2 months</td>
<td>9.8 years</td>
</tr>
<tr>
<td>SED2.4</td>
<td>13 months</td>
<td>13.4 years</td>
</tr>
</tbody>
</table>
The residence time is consistently higher when calculated for the whole water column than when calculated for the upper 500 m of the water column. This is because the iron concentration is highest in the deep water (e.g. Fig. 7.3) and because the detritus concentration decreases with depth leading to less removal of iron through both sinking and scavenging.

**Biology**

The total NPP south of 35°S in the nine runs ranges from 4.9 Pg C yr\(^{-1}\) in run SED2.1 to 10.3 Pg C yr\(^{-1}\) in run SED1.4 (Fig. 7.4a). For both experiment 1 and 2, the total NPP increases with decreasing scavenging, as decreased scavenging means that more iron is available for biological production. The NPP from diatoms is especially dependent on the water’s iron concentration as diatoms have a higher requirement for iron than the smaller nanophytoplankton due to their higher half saturation constant. But when iron is available, diatoms also have a higher growth rate than the nanophytoplankton, and all in all this leads to larger differences in the fraction of NPP from diatoms within the experiments. The NPP from diatoms is largest in run SED1.4 and 2.4.

The intracellular Si:C ratio in the diatoms has been shown to increase under iron limitation (e.g. Takeda, 1998). This dependence is built into the REcoM2 code and when the Si:C ratio is averaged for the Southern Ocean south of 35°S, this ratio indicates in which runs the diatoms are more iron stressed. The intracellular ratio is allowed to vary between set limits of 0.004 to 0.8 mmol Si mmol C\(^{-1}\) in REcoM2, and in all runs, the mean Southern Ocean Si:C ratio lies between 0.5 and 0.8. The ratio does, however, decrease with decreased scavenging rate, as this means that more iron is available in the mixed layer (Fig. 7.3).

Another quantity that is related to iron limitation is export of biogenic silica (opal). A large fraction of the global export of opal, takes place in the Southern Ocean (e.g. Dunne et al., 2007). In REcoM2, the total export of opal is, not surprisingly, highly correlated with the diatom fraction of the NPP (Fig. 7.4), and thus increases with decreasing scavenging in the water. Of the sensitivity runs, only SED1.4 has a larger opal export than the CTRL run.

In our study it is clear that the mean state of the biology is dependent on the rate of scavenging and the resulting iron availability, but less on the magnitude of the sediment source. It is interesting to note that the CTRL is very similar to the run SED1.3 regarding the NPP and the mean Si:C ratio (Fig. 7.4a, b and c), whereas they differ regarding the export of opal (Fig. 7.4d). Regarding experiment 2, the NPP of the CTRL run lies between the result of SED2.3 and SED2.4.

### 7.3.2 Spatial distribution

For the spatial distribution we compare the CTRL run with SED1.3, as these runs produce a similar result regarding the total NPP, but not for the contribution of diatoms to NPP and the opal export. Regarding experiment 2, the NPP in the CTRL run lies between the results from SED2.3 and SED2.4 (Fig. 7.4) and we will to compare to run SED2.3.

**Iron fields**

In the CTRL run, the mean iron concentration in the upper 100 m is highest in the belt between 40 and 60°S, with the highest concentrations occurring downstream
Figure 7.4: a) Total NPP south of 35°S for every run. b) The mean percentage of NPP by diatoms south of 35°S in all runs. c) Mean intracellular C:Si-ratio in diatoms south of 35°S in all runs. d) Total export of opal south of 35°S in all runs. The values written above the bars in subplot a) and c) denote the total NPP and opal export south of 50°S, respectively.
Figure 7.5: a) Mean surface iron concentration in the upper 100 m south of 50°S in the CTRL run, b) The difference in the mean iron concentration in the upper 100 m between SED1.3 and CTRL, c) The difference in the iron concentration in the upper 100 m between SED2.3 and CTRL.

Figure 7.6: a) Mean benthic input of iron in the Southern Ocean south of 50°S in model run a) CTRL, b) SED1.3, c) SED2.3. Notice the nonlinear scale.

of Patagonia, South Africa and Australia (Fig. 7.5a), where iron is supplied through dust and from the sediments (Fig. 7.6a). The CTRL run has a small input from the shelves around the Antarctica (Fig. 7.6a), but not enough to increase the iron concentration here (Fig. 7.5a), possibly because it is immediately removed from the surface by coastal downwelling or utilized for phytoplankton growth.

In SED1.3, the increased iron flux from the shelf of New Zealand increases the iron concentration in the Pacific sector of the Southern Ocean (Fig. 7.5b), the area with the lowest iron concentration in the CTRL run. The second place of elevated iron concentrations is around the Kerguelen Plateau, where the sediment flux also is relatively high (Fig. 7.5b and 7.6b). The open ocean part of the Southern Ocean does, however, also have areas with slightly lower iron concentrations in SED1.3 compared to the CTRL run due to the higher scavenging rate in SED1.3.

The much higher benthic input in SED2.3 along with the higher scavenging rate creates a strong off-shore gradient, with high concentrations close to the shore and very low in the open ocean (Fig. 7.5c).

Net primary production
The spatial distribution of the NPP is correlated with the iron concentration in the upper 100 m of the water column in the CTRL run, with higher rates between 40 and 60°S, especially in the Atlantic and Indian sectors of the Southern Ocean (Fig. 7.7a). When the sedimentary iron input is increased in SED1.3 and SED2.3, the total NPP is
increased downstream of the Patagonian shelf, in the Pacific sector of the Southern Ocean, downstream of New Zealand and around the Kerguelen plateau, where the iron concentrations were also elevated, but it is decreased in the rest of the Southern Ocean (Fig. 7.7b and c). The magnitude of the changes in both directions are largest in SED2.3 as both the near-shore iron input and the off-shore iron removal is the largest in this model run (Fig. 7.5).

Regarding the NPP from the diatoms in the CTRL run, it mainly takes place downstream of the Patagonian shelf, where iron is added from the dust and the sediments, but it also occurs in the Indian sector of the Southern Ocean south of 60°S, where iron must be supplied from below (Fig. 7.7d). When the sediment source is increased, the diatom growth is increased downstream of shelves, especially in the Pacific Ocean, and on the shelves of Patagonia, Kerguelen and along the Antarctic coast (Fig. 7.7e and f). In the open ocean south of 50°S, the production of diatoms is, however, decreased, especially in SED2.3 where the scavenging is strongest. The similarity in the patterns between the change in the total NPP and the NPP from the diatoms show that diatoms are to a large degree responsible for the changes in the NPP between the runs, especially south of 40°S.

**Silicon**

Due to a slow dissolution rate of biogenic silica, the export of opal is highly correlated with the diatom production in the CTRL run, with the largest export occurring downstream of New Zealand and the Patagonian shelf. A large export, however, also takes place south of 50°S where the external iron input is much smaller (Fig. 7.7a and 7.8a). When the benthic flux and scavenging rate of iron is increased in SED1.3 and
The diatom intracellular ratio between silicon and carbon affects the export of the two compounds. In the CTRL, low diatom Si:C ratios clearly occur in the same areas where the production from diatoms is high, but not necessarily where the iron concentrations are high. This indicates that the Si:C ratio is a proxy for good growth conditions for the diatoms, rather than for iron concentrations. The dependence of the Si:C ratio on the iron concentration is clear when we examine the changes between the CTRL run and SED1.3 and 2.3, as the ratio decreases south of 60°S, where the iron concentration also decreases and it increases downstream of the shelves where the iron concentration is increased (Fig. 7.8e and f).

7.4 Discussion

7.4.1 Magnitude of the sediment flux

Measurements of the flux of iron into the water on the Southern Ocean continental shelves cover a large range of values. In the lower end, a newer study found an average iron flux of 0.36 μmol m⁻² day⁻¹ for the Ross Sea (Marsay et al., 2014), whereas Blain et al. (2008) measured 136 μmol Fe m⁻² day⁻¹ on the Kerguelen plateau. These fluxes represent single sites, while our values are averaged over time and space (Table 7.2). While our CTRL run flux of 0.12 μmol m⁻² day⁻¹ is relatively close to the value found by Marsay et al. (2014), the mean value of the flux south of 35°S should most likely be somewhat higher than the value in the Ross Sea, and the value...
of $136 \, \mu\text{mol Fe m}^{-2} \text{day}^{-1}$, measured on the highly productive Kerguelen Plateau during the peak of iron release, is too high as a mean for the area south of $35^\circ\text{S}$. In between these is the mean value of $9.2 \, \mu\text{mol m}^{-2} \text{day}^{-1}$ found by de Jong et al. (2012) measured on the continental shelves in the Atlantic sector of the Southern Ocean, which is very similar to our mean values from experiment 2 (Table 7.2). Their value represents several sites and could indicate that the flux in our experiment 2 is realistic, thereby implying that the scavenging scheme should be changed in REcoM to deal with the increased iron input from such a large flux. A global mean flux of $4.3 \, \mu\text{mol m}^{-2} \text{day}^{-1}$ based on shelf fluxes was suggested by Elrod et al. (2004), and falls between our mean shelf fluxes in experiment 1 and 2 (Table 7.2). Since many of the observations Elrod et al. (2004) based their number on were located in high-productivity areas such as the Californian upwelling zone, it is probably too high as a mean for the Southern Ocean, where biological production generally is lower. This dataset thus points towards the flux in experiment 1 as the more realistic one.

The strength of the iron flux in experiment 1 (Table 7.2) also fits well with that used by some other OGCBMs; Moore et al. (2004) used a constant value of $2 \, \mu\text{mol m}^{-2} \text{day}^{-1}$ in areas shallower than $1100 \, \text{m}$, whereas Aumont and Bopp (2006) used a depth-dependent release rate of iron that was set to a maximum of $1 \, \mu\text{mol m}^{-2} \text{day}^{-1}$. The parameterization of the iron release used in the current study (Elrod et al., 2004), which leads to a spatially varying iron release, has also been used by Moore and Braucher (2008) and Wadley et al. (2014). In the CESM model (Moore and Braucher, 2008), it lead to an input rate ranging from $0$ to $16.5 \, \mu\text{mol Fe m}^{-2} \text{day}^{-1}$, very similar to the values in our SED2.3 model run (Fig. 7.6), and not unreasonable compared to the measurements mentioned above. But as is the case for the observations, the modeled fluxes vary significantly.

In the current study, the total input of iron to the Southern Ocean varied with more than an order of magnitude, from $0.22 \times 10^9 \, \text{mol Fe yr}^{-1}$ in the CTRL run to $15.8 \times 10^9 \, \text{mol Fe yr}^{-1}$ in SED2.3 (Table 7.2). Observation-based estimates of the total iron input do not exist, but it is possible to compare the iron input in REcoM to other OGCBMs. Tagliabue et al. (2009) had an input of $1.34 \times 10^9 \, \text{mol Fe yr}^{-1}$ in a model study covering the area south of $35^\circ\text{S}$. This is slightly lower than the input of $2.00 \times 2.14 \times 10^9 \, \text{mol Fe yr}^{-1}$ we see in our experiment 1 (Table 7.2). In another model study, Lancelot et al. (2009) had a total input of $0.38 \times 10^9 \, \text{mol Fe yr}^{-1}$ from the sediments south of $60^\circ\text{S}$. This is very similar to the input in our SED1.3 run, in which the input was $0.36 \times 10^9 \, \text{mol Fe yr}^{-1}$ in the same area.

### 7.4.2 Iron concentration and scavenging

The modeled mean iron profiles tend to have low surface iron concentrations and high deep iron concentrations compared to the mean profile of the observations, especially in the open ocean (Fig. 7.3). This shows that while most of the iron from the sediments is added from depths shallower than $500 \, \text{m}$ (Fig. 7.2), most of the scavenging, which is dependent on the detritus concentration, also removes iron in the upper part of the water column. In reality, the maximum mean values are generally close to $0.6 \, \mu\text{mol Fe m}^{-3}$, though it varies between the major ocean basins (e.g. Johnson et al., 1997; Moore and Braucher, 2008; Tagliabue et al., 2012). The modeled deep water concentrations close to $1 \, \mu\text{mol Fe m}^{-3}$ are thus likely overestimated. Johnson et al. (1997)
suggested that the scavenging rate of iron increases when the concentration reaches 0.6 μmol Fe m$^{-3}$, thereby explaining this apparent maximum. In REcoM2, binding of dissolved iron to ligands decreases scavenging when the iron concentration is lower than 1 μmol Fe m$^{-3}$. Decreasing the assumed ligand concentration would be a way to increase the scavenging rate in the model, something that could be necessary for realistic runs with an increased sediment source, especially when the strength is high as in experiment 2.

Regarding the surface iron concentrations, Klunder et al. (2011) observed a northward increase in the Atlantic sector of the Southern Ocean, ranging from 0.15 μmol Fe m$^{-3}$ in the Weddell Gyre to 0.25 μmol Fe m$^{-3}$ at 55°S. In the Indian sector of the Southern Ocean, Grand et al. (2015), on the contrary, found that the surface iron concentrations decreased northwards from Antarctica, with maximum coastal values of 1.7 μmol m$^{-3}$, and the concentration falling to 0.2 μmol m$^{-3}$ north of the Polar Front. In our runs, the iron concentration north of 60°S fit well with these results, and the surface iron concentration increases from south to north similar to the result from Klunder et al. (2011) (Fig. 7.5). Towards the south, the modeled iron concentrations are, however, relatively low, and while the coastal iron input is increased in SED1.3 and SED2.3 (Fig. 7.6b and c), the iron concentration around Antarctica has not increased towards the values from Grand et al. (2015).

In experiment 2, the combination of low surface mixed layer iron concentrations and high deep water iron concentration compared to the observations (Fig. 7.3) indicate that the value of 0.68 μmol Fe (mmol C$^{-1}$) for the benthic Fe:C ratio is too high for our model setup. In the open ocean that is not downstream from shelves, the iron concentration has been decreased in many areas in SED1.3 when compared to the CTRL run, and for SED2.3, the changes in the spatial distribution makes it clear that the scavenging needed to keep the sediment input in check is simply too large to be realistic as it removes too much dissolved iron in the open ocean.

One thing that might change the effect of the scavenging rate is the aeolian iron deposition. In the CTRL run, the sediment source make up 44% of the total external iron input, whereas it comprises approximately 88 and 98% in experiment 1 and 2, respectively. In the model study by Lancelot et al. (2009), the sediment source comprised 90% of the iron input south of 60°S, concurrent with SED1.3, whereas the sediment source comprised 72% of the input south of 35°S in the study by Tagliabue et al. (2009). For the global ocean, the sediment source made up 58% of the total iron supply in the study by Moore and Braucher (2008). Provided that these numbers are reasonable, and we do not know if they are, our CTRL run and experiment 1 have the more reasonable ratios between the sediment input and the dust input (Table 7.2).

### 7.4.3 Residence time of iron

The residence time of iron in the ocean is much lower than that of the macronutrients and also lower than the time scale of the thermohaline circulation (Johnson et al., 1997). The residence time can be calculated through the rate of iron removal or the rate of the input. As the flux of the iron input and associated scavenging varies widely between models, so does the residence time. At the higher end of the range lies the global mean residence time of 233 years found be Parekh et al. (2005). In this model, the only external iron source was aeolian, and the residence time in our CTRL run (267 years, Table 7.3) is relatively close to their result as this run only has
a small sedimentary iron input. Moore and Braucher (2008) conversely calculated a
global mean residence time of 12 years in a model with a relatively large input of
iron from external sources, and also a higher scavenging rate. Their surface water
residence time estimate ranged from 5 months in high-input areas to 1.4 years in
areas with less iron input. The sediment input in this model is, on the global scale,
approximately similar to ours in experiment 2, and the calculated residence time of
the current study fits relatively well with their estimates, both for the surface water
and for the whole ocean (Table 7.3). Iron removal or input rates are, however, difficult
to constrain and measurement-based estimates of the residence time therefore also
differ significantly.

In a study based on measurements, Bruland et al. (1994) estimated a residence time
of iron in the central North Pacific deep water ranging from 70 to 140 years. In the
North Atlantic deep water, the residence time has been found to be 270 years, while
it was estimated to be one to five months in the surface water (Bergquist and Boyle,
2006). In the Equatorial Atlantic the surface residence time is similarly short, ranging
from 280 days to 10.5 years have been found in the Equatorial Atlantic, which is un-
der influence from the Saharan dust plume (Croot et al., 2004b). These studies cover
the whole range of residence times calculated in the current study (Table 7.3). In this
study, the residence time is based on removal of iron, which is strongly controlled
by the scavenging. Decreasing the assumed concentration of ligands in the model
could move the high scavenging towards the coast, thereby reducing it in the open
ocean. This in turn would impact the residence time. But the processes are complex
and difficult to predict, and in order to utilize the residence times to optimize the
iron cycle in the model a more in depth study would be necessary.

7.4.4 Biological production

The higher values for NPP occur in the runs SED1.3 and 1.4 along with SED2.3 and 2.4,
in which the iron scavenging is relatively small and the surface iron concentrations
high. The fraction of the total NPP that is carried out by diatoms has previously been
shown to be dependent on the available iron (Wang et al., 2014b), and we also see
how the diatoms become more abundant in the runs where the iron concentration
is higher in the mixed layer (Fig. 7.4c). The increase in diatoms does, however, occur
downstream of islands and continental shelves, while diatoms decrease in abundance
in the open ocean (Fig. 7.7e and f). This means that it is only in the runs SED1.4, 2.3
and 2.4 that the diatom make up a larger fraction of the total NPP than in the CTRL
run (Fig. 7.4c). In a study based on satellite-based measurements, Bracher et al. (2009)
showed that diatoms dominate production in the Southern Ocean south of 30oS,
whereas other studies have shown less production by diatoms (e.g. Alvain et al., 2005;
Hirata et al., 2011). In a study focussing on the Southern Ocean, Soppa et al. (2014)
show that diatoms dominate in areas with high iron input, especially downstream
of the Patagonian shelf and in the Atlantic sector of the Southern Ocean in general,
but also south of Australia, and east of New Zealand and the Kerguelen Plateau.
In our study, the spatial distribution of the diatom production broadly follows the
same pattern (Fig. 7.7), but diatom productivity in the Atlantic sector of the Southern
Ocean is relatively low, most likely due to too little sedimentary input from the
shelves around the Antarctic Peninsula and South Georgia (Fig. 7.5 and 7.6). One
factor that plays a role in this, is the resolution of the model grid, which is increased
in coastal regions in FESOM (Fig. 7.1), but not enough to capture the narrow shelves adequately in both of these regions and especially around South Georgia (Fig. 7.1). We do consequently not capture the South Georgia bloom or the bloom downstream of the Antarctic Peninsula, which shows up in high-resolution runs (e.g. Wadley et al., 2014). If these blooms were captured, they would likely contain a large degree of diatoms as is the case in the Kerguelen bloom and downstream of the other major shelves (Fig. 7.7e and f).

In commonly used OGCBMs, the fraction of the biomass from diatoms in the Southern Ocean ranges from 20 to 100% (Vogt et al., 2013), and thereby differs widely, but they have in common that the diatoms are more abundant in areas with higher iron availability. Diatoms do not dominate production in any of our runs, with the highest percentage of NPP from diatoms occurring in SED1.4, where they comprise 28% of the total NPP. Our model results are thus at the lower end of what other models have found, especially SED1.1 and 1.2 (Fig. 7.4b), but close to the results by Alvain et al. (2005), as described above.

One thing that can help constrain the appropriate amount of diatom primary production in the Southern Ocean is by comparing the amount of vertically exported opal in our model to the amount found in previous studies. While our higher values of opal export fit well with the results in the lower end of the range from Dunne et al. (2007), the low opal export in the runs SED1.1 and 1.2 as well as SED2.1 to 2.3 show that too little iron is available in these runs to make them realistic (Fig. 7.4). As diatoms are dependent on a higher iron concentration than the nanophytoplankton to grow, the low open ocean surface iron concentrations in SED2.3 (Fig. 7.5c) along with the changes that occur in the spatial distribution of the diatom NPP (Fig. 7.7f) make it clear that although the NPP from diatoms has increased downstream of the major shelves in SED2.3, where the surface iron concentration has been increased, it has decreased in the open ocean due to the larger scavenging that occurs here, leading to a relatively small opal export.

Another factor affecting the silicon export is the intracellular Si:C ratio of the diatoms. In REcoMz, this number decreases with increasing iron availability, varying from 0.544 mol Si (mol C)\(^{-1}\) in SED1.4 to 0.76 mol Si (mol C)\(^{-1}\) in SED1.1 and 2.1 (Fig. 7.4). On a global scale, the Si:C ratio has been shown to vary with a factor of 4 to 5, with the mean value being 0.11 mol Si mol C\(^{-1}\) (e.g. Sarthou et al., 2005). Compared to the other major oceans, the Southern Ocean is, however, highly silificated, with measured values of the Si:C ratio ranging from 0.31 to 0.6 mol Si mol C\(^{-1}\), dependent on the iron availability (e.g. Brzezinski et al., 2003; Hoffmann et al., 2007; Baines et al., 2010). The mean numbers for the Si:C ratios in REcoMz are rather high compared to observed estimates, and consistent with a highly iron limited regime.

7.5 CONCLUSION

We have shown, that given different strengths of the sediment source of iron, combined with an appropriate rate of iron removal from the water through scavenging, it is possible to create virtually identical mean iron profiles for the Southern Ocean along with reasonable numbers for the total NPP. The phytoplankton composition and the spatial distribution of the NPP did, however, change with changed strengths
of the iron flux. When choosing a combination of the benthic iron influx and scavenging rate in a model it is therefore necessary to evaluate the model’s NPP from diatoms along with the amount of opal exported against previous studies, as these are highly dependent on the iron availability in the model. While our CTRL run produced reasonable values for the biogeochemistry, the sediment source of iron is relatively small in this run, and the NPP from diatoms is most likely also too small. Based on the knowledge we have about the benthic iron flux, as well as the above mentioned parameters, we found that our model run SED1.3 gave better results, with a mean shelf iron input of 1.2 μmol Fe m$^{-2}$ yr$^{-1}$, leading to a total benthic iron influx of 2.14 × 10$^{9}$ mol Fe yr$^{-1}$ to the Southern Ocean. The open ocean NPP is, however, still relatively small in this run, as is the fraction of the NPP from diatoms and the export of biogenic silicon. Further sensitivity runs with different parameterizations of the scavenging could improve this by decreasing the off-shore iron gradient.

REFERENCES


Part III

SYNTHESIS
Iron limitation of the Southern Ocean has been recognized as one of the main reasons for its High Nutrient Low Chlorophyll (HNLC) status since the early 1990’s (e.g. Martin, 1990; Martin et al., 1990). Since then, substantial effort has gone into assessing the role of iron in the Southern Ocean. In the current thesis, the role of the iron supply from the sediments and from the deeper reservoirs has been assessed. It has in particular been explored how the representation of the Southern Ocean iron supply in large scale Ocean General Circulation Biogeochemical Models (OGCBMs) affect the prediction of the net primary and export production (NPP and EP) in these models, and what these results teach us about the the iron supply in nature.

Publication I documents the results of the coupling between the biogeochemical model REzoM to the Finite Element Sea-ice Ocean Model (FESOM). The modeled global biogeochemical surface fields capture the large scale features, with low production in the subtropical gyres and higher around the Equator and in the temperate regions. The model results are comparable to those of other non-eddy resolving OGCBMs, with the spatial distribution of the chlorophyll \( a \) and NPP being especially good. In the Southern Ocean, the NPP and EP fits well with satellite-based estimates and the results of other OGCBMs. Production here is strongly iron limited and dominated by nanophytoplankton.

Publication II and III, address the role of the iron input from the sediments and the vertical iron supply across the base of the mixed layer in the Southern Ocean. The ocean component of OGCBMs are known to have a large effect on the results for NPP and EP, especially in the Southern Ocean (e.g. Doney et al., 2004; Najjar et al., 2007). Publication II investigates how the vertical iron fluxes across the mixed layer, the NPP and the EP, differ between two model runs which only differ in the choice of Ocean General Circulation Model (OGCM) used; one was run with the Massachusetts Institute of Technology general circulation model (MITgcm) and one with FESOM. The study showed that the OGCM has a large impact on the vertical iron fluxes and the NPP. In MITgcm the total vertical flux is four times higher than in FESOM and the dominating pathway is entrainment, while it is diffusion in FESOM. The differences in the ocean circulation and mixing induce differences in the total NPP and the fraction of the NPP that is exported to the deep water. The study highlights the importance of assessing the the skill of the OGCM, and indicates that the vertical iron fluxes across the base of the mixed layer could play a large role regarding the differences in NPP and EP that has previously been established (e.g. Doney et al., 2004; Najjar et al., 2007).

The Southern Ocean is located far from major dust sources, and the sediment source of iron is thus thought to supply the bulk of new iron (e.g. Tagliabue et al., 2009a; Lancelot et al., 2009). The sediment source is, however, poorly constrained by measurements and thus varies widely between models, both in strength and in parameterization. Publication III presents a sensitivity study in which the sediment source of iron was varied significantly along with the scavenging rate. It shows that while the total NPP and the mean vertical iron profile can be recreated with very
different strengths of the sediment source of iron, it also changes the spatial distribution of the NPP. The amount of NPP from diatoms and the resulting vertical export of opal to the deep ocean is likewise sensitive to the changed inputs in the iron supply.

In the following, the relative importance of the different iron sources in the Southern Ocean will first be discussed, and it will be debated how these sources may change in the future ocean and what challenges modelers need to consider to take these changes into consideration. Secondly, advantages and typical applications of the new FESOM-REcoM are discussed, along with caveats and limitations. Finally, perspectives for future research areas and remaining open questions will be presented here.

### 8.1 Iron in the Southern Ocean

#### 8.1.1 Current iron sources

The model runs considered here are listed in Table (8.1), along with the factors that differ between them. The runs presented in publication I and II are identical, while the CTRL run in publication III use an updated version of FESOM. In the runs SED1.3 and SED2.3 from publication III, the sediment source was increased as compared to the other runs. Here, focus will mainly be on SED1.3 when the effect of the increased sediment source is discussed.

One strength of large scale OGCBMNs is that they make it possible to upscale various processes to larger areas, such as the Southern Ocean. In the following, focus will be on the iron supply to the surface water of the Southern Ocean. Comparing the input of iron from the sediments (Pub. III) and the vertical flux across the base of the mixed layer (Pub. II) is, however, not straightforward, as the fluxes refer to different volumes of water. The vertical iron fluxes represent supply into the surface mixed layer, while the sediment derived input is calculated into the upper 500 m of the water column (Fig. 8.1). This makes the analysis somewhat inconsistent, but still gives an impression of the relative importance of the different sources.

*Total iron input to the Southern Ocean*

In the model, the total input of iron to the surface mixed layer is approximately \(1.5 \times 10^8\) mol Fe yr\(^{-1}\), when counting the input from dust \((0.5 \times 10^8\) mol Fe yr\(^{-1}\), Pub. I, II and III), the vertical flux \((9 \times 10^8\) mol Fe yr\(^{-1}\), Pub. II) and the sedimentary input \((5 \times 10^8\) mol Fe yr\(^{-1}\), SED1.3, Pub. III) and while assuming that the iron added

<table>
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<th>RUN NAME</th>
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<tr>
<td>I</td>
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<td>III</td>
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<td>0.01</td>
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<td>III</td>
<td>Pub. III, SED1.3</td>
<td>0.1</td>
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<td>III</td>
<td>Pub. III, SED2.3</td>
<td>0.68</td>
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Figure 8.1: Conceptual plot of the iron fluxes in and out of the Southern Ocean surface water south of 50°S in FESOM-REcoM2. The input from the sediments and the scavenging is calculated for the upper 500 m of the water column. The uptake by phytoplankton is for the vertically integrated NPP. The dust input is identical in Pub. I, II and III. The numbers for the sediment source is from publication III, (CTRL and SED1.3), Table (7.2). The numbers for the vertical supply to the mixed layer is from publication II, Fig. 6.3, and counts entrainment, upwelling, horizontal advection and vertical eddy diffusion. Notice that dust is supplied to the surface ocean, the sediment flux takes place at the bottom at all depths, and the vertical fluxes are calculated across the base of the mixed layer. The plot gives a simplified view of the iron fluxes in the surface water of the Southern Ocean, and does not entail all fluxes represented in the model.
to the upper 500 m of the water column from the sediments is roughly equal to that added to the mixed layer (Fig. 8.1). For a total modeled NPP of 3.1 Pg C yr\(^{-1}\) in the Southern Ocean (Pub. I, II, III), the iron uptake is \(13 \times 10^8\) mol Fe yr\(^{-1}\). This number is very close to the \(13.5 \times 10^8\) mol Fe yr\(^{-1}\), which Boyd et al. (2012) argued biological production utilize in the Southern Ocean south of 50°S. The average iron input can thus on average sustain biological production in the Southern Ocean. Iron is, however, removed from the surface water through sinking of organic matter and through scavenging. The former accounts for a removal of \(3.5 \times 10^8\) mol Fe yr\(^{-1}\) (Fig. 8.1), and the remainder of the iron in organic matter is thus remineralized above that depth. As scavenging of iron onto particles in the water account for a removal of \(3.5\) to \(20 \times 10^8\) mol Fe yr\(^{-1}\) (CTRL and SED1.3, Pub. III), the remineralized iron also plays a role for production in the Southern Ocean.

The important role of the sediment source in supplying iron to the Southern Ocean is well established (e.g. Pollard et al., 2009; de Jong et al., 2012; Wadley et al., 2014) and confirmed in publication III. While the Southern Ocean NPP is very reasonable in publication I, II and the CTRL run of publication III, the surface iron concentrations were relatively low compared to measurements (e.g. Klunder et al., 2011; Grand et al., 2015) and the phytoplankton blooms downstream of islands such as South Georgia (e.g. Borrione and Schlitzer, 2013) and the Kerguelen Plateau (e.g. Blain et al., 2001) were missing, indicating that the sediment source of iron was too small. Increasing the strength of the sediment source tenfold from the CTRL run to the SED1.3 run, the total input to the upper 500 m of the water column of the Southern Ocean increased from \(0.5 \times 10^8\) mol Fe yr\(^{-1}\) to \(5 \times 10^8\) mol Fe yr\(^{-1}\) (Pub. III and Fig. 8.1). Both of these numbers are smaller than the sum of the vertical iron supply to the surface mixed layer of \(9 \times 10^8\) mol Fe yr\(^{-1}\) (Pub. II and Fig. 8.1). As explained above, the numbers for the two sources do not refer to exactly the same water volume (Fig. 8.1), and a fraction of the input from the sediment source is most likely released below the mixed layer and transported upwards by the physical transport mechanisms, thereby being counted twice. The difference in the iron input from the sediments and the vertical flux is nevertheless large enough to show that the vertical iron supply is significant.

**The sediment source of iron**

One problem that commonly arises when a sediment source of iron is added to a large scale OGCBM is that the coastal resolution is relatively coarse compared to the width of the shelves in some areas, for example around Antarctica and along the west coast of the American continent. Despite of an increased resolution around the coast (40 km), the grid used in this thesis is still relatively coarse, with the consequence that the sediment source of iron is underestimated in some coastal areas. Moore and Braucher (2008) solved this by scaling the iron input in coastal grid cells by the fraction of the cell that would be on the shelf in reality, while a reasonable solution in FESOM would be to increase the coastal resolution, at least in studies focussing or relying especially on the sedimentary input of iron.

It is well established that recurrent phytoplankton blooms appear downstream of South Georgia and the Kerguelen Plateau in the Southern Ocean (Fig. 8.2a and e). In publication III it was argued that the diatom abundance and the vertical export of biogenic opal ought to be used to evaluate the skill of OGCBMs, while another approach would be to assess whether the island blooms are adequately captured with a given strength of the sedimentary iron source. In the CTRL run of publication III both the
South Georgia and Kerguelen blooms are missing (Fig. 8.2b and f). While the modeled Kerguelen bloom increases with the increased strength of the sediment flux in run SED1.3 and 2.3 (Fig. 8.2f to h), the South Georgia bloom is not improved (Fig. 8.2b to d). North of the Kerguelen, the strength of the sediment source, combined with the size and depth of the shallow plateau fit together to produce a reasonable output, especially in SED1.3. For South Georgia, the missing bloom can be explained by the fact that the grid cells around South Georgia are all located at depths deeper than 1000 m (Fig. 7.1, Pub. III). At this depth only little iron is released (Fig. 7.2, Pub. III), and a bloom fueled by the iron released from the sediments does not evolve. This illustrates the importance of developing the coastal grid suitable for the aim of the study, both regarding depth and resolution.

**Vertical iron supply**

The sedimentary iron source (Pub. III) and the vertical iron supply (Pub. II) are not entirely separable, but the fact that the vertical flux dominates the supply even when compared to model run SED1.3, where the sediment source is relatively large, indicates that the vertical iron supply is on average the most important in FESOM. Especially the input from entrainment is interesting as it has previously been shown to dominate the supply from below (Tagliabue et al., 2014) and to sustain the summer bloom in some remote parts of the Southern Ocean (Carranza and Gille, 2015). The magnitude of the modeled entrainment does, however, depend heavily on the ocean circulation and mixing as is demonstrated by the fact that the entrainment driven input of iron, which at $20.2 \times 10^8$ mol Fe yr$^{-1}$, is eight times higher in MITgcm-REcoM than in FESOM-REcoMz (Pub. II). As the mean seasonal amplitude of the MLH is too small and the iron concentration at the base of the mixed layer low in FESOM as
compared to observations (Pub. II), the influx from entrainment is most likely underestimated in this model. In MITgcm, the mean seasonal amplitude is, on the other hand, too large, leading to the iron transport through entrainment probably being too high in this model. That the real entrainment should lie between the results from our study fits well with the observation-based result of $9.5 \times 10^8$ mol Fe yr$^{-1}$ suggested by Tagliabue et al. (2014).

Due to the complexity of the processes controlling marine NPP, it is impossible to exactly pin-point which processes that are responsible for the differences in the two model runs presented in publication II. Since this study was carried out, REcoM2 has been coupled to an updated version of FESOM (Pub. III and App. A), which can help to give a better understanding of the effect of the different modeled processes. In the updated version, the mean seasonal cycle of FESOM’s mixed layer depth is deeper and closer to both the observations and the MLD in MITgcm than before (Fig. 8.3). The entrainment of iron is also positive across the Southern Ocean, as is the case in MITgcm-REcoM2 (Fig. 6.8). But despite of these changes, all of the vertical iron fluxes are virtually identical in FESOM 1.3 (Pub. II, Fig. 6.3) and 1.4 (Fig. 8.3), and are thus still low compared to MITgcm (Pub. II). This shows, that even though the mixed layer indeed plays a role, especially for the entrainment and for the access to the deeper iron inventories, the iron supply across the base of the mixed layer is to a higher degree dependent on the supply of iron to the base of the mixed layer. The strength of the upwelling in the Southern Ocean, thus largely controls the iron fluxes into the surface mixed layer by delivering iron to the permanent thermocline, below which the MLD does not reach. The upwelling of iron to the base of the mixed layer is controlled by the iron concentration in the deep water and by the vertical velocity of the water. As only the circulation component differs between the two model runs in publication II, it is tempting to conclude that the vertical velocity, and thereby the different strengths of the meridional overturning circulation are responsible for the differences. But since the deep iron concentration is dependent on a number of processes, such as the sinking and remineralization of organic matter, which again are affected by the ocean circulation in multiple ways, this view is too simplified, though it is likely that the strength of the overturning circulation is a large part of the explanation for the differences in the vertical iron fluxes between the models.

**Other iron sources**

As the deep water iron concentrations are important for sustaining the vertical iron fluxes, the deeper iron sources could play an important role for the biological production in the Southern Ocean. These sources include the deeper sediment input, input from hydrothermal vents and the deep inflow of iron rich water to the deep Southern Ocean in the North Atlantic and Upper Circumpolar Deep Water. In models, it is thus both the implementation of the external sources into the code as well as a correct representation of the meridional overturning circulation and the mixed layer depth that is crucial for the amount of iron available for transport into the mixed layer from below. In the current study, the input from hydrothermal vents has not been considered, as they are most likely relatively unimportant on the timescale of the model runs presented here. On longer time scales they will, however, have noticeable effects, especially in the Pacific sector of the Southern Ocean where iron limitation is strongest (Tagliabue et al., 2010).
8.1 Iron in the Southern Ocean

8.1.2 Iron supply in a changing climate

Martin (1990) was the first to suggest that aeolian iron deposition to the Southern Ocean plays a crucial role in the global climate. The argument behind his "iron hypothesis" is that the large unused pool of macronutrients in the mixed layer of the Southern Ocean can be utilized for biological production when the aeolian deposition of iron is increased. This increased production leads to an increased drawdown of CO$_2$ from the atmosphere to the ocean, which reduces the atmospheric greenhouse effect and hence lowers the atmospheric temperature. Later studies have proven the correlation between aeolian dust deposition and changes in atmospheric CO$_2$ (e.g. Martínez-Garcia et al., 2011; Martínez-García et al., 2014) but it has also been argued that while dust indeed plays a role in the glacial/interglacial cycles, the changes in dust input are not large enough to be the controlling factor (e.g. Watson et al., 2000).

As the climate is currently changing more rapidly than ever before, the iron hypothesis becomes interesting again, as an increased temperature is likely to change the iron supply to the Southern Ocean in several ways; increased temperature will change the stratification and thereby the MLD (Rhein et al., 2013), changes in biological production may change the amount of iron released from the sediments (Elrod et al., 2004), changed circulation patterns will impact the transport of iron released by the sediments (Misumi et al., 2014) and a higher temperature will change the supply of iron from dust (Maher et al., 2010). These changes will affect biological production in the Southern Ocean and thereby also the feedbacks regarding the global carbon cycle and the climate (Boyd and Ellwood, 2010).

Ocean circulation and mixing
One consequence of a warmer ocean is that the mixed layer becomes shallower and the stratification stronger on a global scale, leading to decreased vertical mixing of
nutrients to the surface water along with increased light in the surface mixed layer (Rhein et al., 2013). The decreased nutrient supply leads to decreased NPP when averaged globally (e.g. Steinacher et al., 2010; Bopp et al., 2013). In the Southern Ocean, model studies, however, show that shallower mixed layers alleviate light limitation while nutrient supply is sustained by the large seasonal amplitude in the MLD, leading to a net increase in the NPP towards the end of the century (e.g. Bopp et al., 2001; Steinacher et al., 2010; Bopp et al., 2013; Misumi et al., 2014). The biogeochemical responses to 21st century forcing do, however, vary between models, especially in the Southern Ocean (Steinacher et al., 2010; Bopp et al., 2013).

One reason for the relatively deep mixed layers in the Southern Ocean are the subpolar westerlies, which have their maximum between the subtropical front and the polar front. The westerlies are expected to increase and move polewards over the coming century due to an increase in the index of the so-called Southern Annular Mode (SAM) (Sen Gupta et al., 2009; Zheng et al., 2013). This phenomenon acts against the strengthening of the stratification that warmer temperatures bring, and thus adds to the complexity of the future ocean circulation and mixing.

As shown in publication II, the ocean circulation and mixing differs greatly between OGCMs, even under similar prescribed forcing, with large effects on the modeled NPP. The response of OGCMs to the counteracting effects of the warming signal and the SAM-signal is thus bound to differ (e.g. Hauck et al., 2015), especially when forcing additionally differs between earth system models. Knowing the biases of the modeled ocean circulation and mixing under current forcing is thus critical for understanding it’s response to future changes, though a large uncertainty is further introduced by the atmospheric component of the models.

Sedimentary supply
The strength of the iron release from the sediments depends on the redox state within the sediments, and thereby the amount of organic matter deposited on the sea bed (e.g. Elrod et al., 2004). As the rate of NPP is expected to increase over the next century in the Southern Ocean (e.g. Steinacher et al., 2010; Bopp et al., 2013), the supply of organic matter to the sediments will likely also increase, thereby increasing the release of iron from the sediments. Whether the additional iron supply from the sediments will have an effect on the biological production in the open ocean depends on the strength and location of the horizontal advection responsible for transport of the iron. As the subtropical gyres are currently expanding (Polovina et al., 2008), they will enter the Southern Ocean over the coming century, thereby transporting iron-rich water further towards the south (Misumi et al., 2014). The southwards shift of the subtropical gyres also intensifies the coastal currents, something that could increase the effect of the sedimentary supply of iron as they increase the off-shore transport from the shelves. In large scale OGCMs the coastal currents are, however, not necessarily adequately captured. In these models, resolution thus impedes the representation of the sediment source two-fold; by not resolving the shelves from which the major part of the sediment derived iron is released, and also by not representing the strong surface currents necessary for advection adequately. Here, FESOM has the advantage of the unstructured grid that makes it possible to resolve the shelves in future studies regarding the sediment derived iron input to the ocean.

Other iron sources
Dependent on the changes in land use and climate, the dust supply is likely to
increase in the future (e.g. Maher et al., 2010). Large parts of the Southern Ocean are, however, located far from land sources, and the input of iron from the sediments is still likely to dominate the input in spite of enhanced aeolian iron inputs (Tagliabue et al., 2009a), especially if the sediment source also increases.

As sea-ice forms, it incorporates and up-concentrates iron from the underlying sea water, while iron supplied by dust is also stored in the ice (van der Merwe et al., 2011). The sea-ice mainly acts to change the location and timing of the iron-input to the water, and its primary effect is to change the phytoplankton composition (e.g. Wang et al., 2014b). Recently, the calving of ice bergs has been shown to increase in the Southern Ocean (Mouginot et al., 2014). Icebergs transport iron to relatively remote areas of the Southern Ocean, releasing iron as they melt (Raiswell et al., 2008; Lin et al., 2011). While the iron input from ice is most likely small compared to other iron sources (Table 1.1) it is likely to directly impact biological production locally, and can potentially be investigated with FESOM-REcoM2 as will be discussed in section (8.4.1).

Irradiance and temperature
Future changes in radiation and temperature do not just change the iron sources directly, but also the conversion of Fe(III) to the more bioavailable Fe(II) (Fig. 1.5). Using an OGCBM, Tagliabue et al. (2009b) showed that including the irradiance in the modeled iron cycle increases the amount of bioavailable iron by up to 53 % in the Southern Ocean, while including the effect of temperature on the iron speciation decreased the amount of available iron slightly. Commonly used OGCBMs do take the changed availability of light for phytoplankton growth into account (e.g. Steinacher et al., 2010), but the direct effect on the iron cycle is so far only considered in a single model (Tagliabue et al., 2009b; Tagliabue and Völker, 2011). This direct effect of irradiance on the bioavailable iron should thus be tested in a larger amount of models and in runs covering the next century in order to gain a better understanding of the coming changes in the Southern Ocean. As studies already suggest that the iron concentration in the Southern Ocean surface water will increase in the coming century (e.g. Misumi et al., 2014), this effect could either be enhanced or decreased as changes in the MLD also affects the average irradiation in the surface mixed layer and thereby the iron availability.

8.2 Global iron supply

As FESOM is a global model, the different iron inputs must result in an adequate representation of the biogeochemical fields for the global ocean. During the initial coupling of REcoM2 to FESOM, iron was not released to the water from the sediments. Test runs did, however, indicate that large parts of the Arctic Ocean, which in reality receives iron from the broad shelves among the Siberian coastline (e.g. Klunder et al., 2012; Hioki et al., 2014), became iron limited in the absence of the sediment source. A relatively weak sediment source was consequently added. In the Pacific Ocean, the iron limitation is, however, strong, both in publication I and when FESOM was updated (App. A, Fig. A.8). One explanation for this is the very narrow coastal shelf along the west coast of South and North America, which are not captured in the current grid, thus diminishing the sediment source in the region. Another explanation is that scavenging removes iron from deep water, thereby reducing the Fe:N
ratio in the upwelled water. Based on measurements of iron’s isotopic composition, it has recently been argued that the sediment input from Papua New Guinea is transported across the Pacific Ocean and sustains biological production in the eastern Pacific (Radic et al., 2011). Increasing the sediment source tenfold in publication III does increase the surface iron concentrations in the western Pacific Ocean as iron is brought eastwards from the coast of Japan and New Zealand, again showing that an increased strength of the sediment source improves the skill of REcoM. Whether the sedimentary iron input from Papua New Guinea plays a role in supplying iron to the surface water of the modeled Pacific Ocean is, however, an open question.

8.3 TWO DIFFERENT OCEAN MODELS

Coupling a biogeochemical model to a new ocean model requires a significant amount of effort and time, both regarding the actual work of the coupling, but even more so for the process of tuning the coupled model to obtain reasonable results when compared to observations or satellite-based estimates of the state variables. The current thesis presents the results of such an effort, with the coupling of the biogeochemical model REcoM to the Finite Element Sea-ice Ocean Model (FESOM). The question is, however: When is it reasonable to choose FESOM-REcoM for a study, considering that REcoM already runs coupled to the MITgcm in a successful set-up?

FESOM’s main strength is the capability to highly resolve certain areas with seamless transition to areas of lower resolution. This feature has not been fully utilized in the current study, but should be further investigated in the future. As a higher resolution is likely to improve the skill of the ocean model, it is probable that it will also affect the biogeochemical results (e.g. Doney et al., 2004; Najjar et al., 2007; Sinha et al., 2010). The effect of resolution was investigated by McKiver et al. (2014) who showed that a higher resolution especially improved the distribution of surface chlorophyll $a$ due to a better description of the vertical velocity and MLD in the OGCM.

While it is in theory easy to change the mesh in FESOM, it does in reality require a lot of work to first create a mesh and then tune it for the ocean. Further, when running the coupled FESOM-REcoM model it may be necessary to tune the biogeochemical model again. It is therefore relevant to look for tested configurations of FESOM before embarking on a new study with FESOM-REcoM. An additional advantage of using a tested configuration is that knowledge about the ocean state exists, and as shown in publication II, the ocean circulation and mixing is key for successful biogeochemical modeling. For FESOM, several tuned configurations with specialized grids exist. One example is a mesh in which the resolution is scaled towards eddy activity, which would for instance make it possible to examine the role of mesoscale eddy transport of nutrients into the mixed layer. Other possibilities include using a high resolution Arctic Ocean or a mesh where the equatorial region is highly resolved. Another aspect to keep in mind when using a grid with local high resolution is its increased computational demand. This is due to the larger amount of nodes that such a mesh has, and to the smaller time step that is necessary in the whole model domain. Higher resolution thus increases the runtime of the model.

The complexity of FESOM’s grid is an advantage for resolving complicated topography, but it demands more of the people that process the output in terms of mathematical and programming abilities. The complexity of the grid thus also increases the risk of simple mathematical errors during processing.
The MITgcm-setup used with REcoM2 is in contrast to FESOM relatively simple to work with as calculations are performed on a regular grid. The MITgcm also comes with extensive experience and support in forms of a very large user group with associated output in forms of papers and knowledge of the model’s behavior under different circumstances. The drawback of this model is that REcoM2 runs on a single grid of relatively coarse resolution that is not easily changed.

When choosing which model to use for a certain research question it thus makes sense to look at the question at hand; does it require a high resolution in a certain area, does a grid exist for this area for FESOM and is the length of the planned model run reasonable compared to the required runtime of FESOM? Otherwise, it is most likely a good idea to choose to work with MITgcm-REcoM2.

8.4 Future directions for fesom-recom2

8.4.1 Iron sources

In publication III, the model’s sensitivity to the strength of the sediment source was investigated, but to learn more about its effect, a mesh with higher resolution on the continental shelves would be necessary. Higher coastal resolution would make it possible to examine the role of the sediment source in generating the increased NPP downstream of islands as well as the distance travelled by the iron from the coast.

Resolving the area of the ACC would make it possible to look into the frontal role for the iron distribution, and whether frontal activity increases the iron supply by vertical mixing as argued by Sokolov and Rintoul (2007). If resolution permits, the role of the horizontal eddy transport could also be investigated.

Another possibility is to add iron to the surface water from melting icebergs (e.g. Raiswell et al., 2008; Lancelot et al., 2009). An iceberg transport model has been coupled to FESOM (Rackow, 2011; Wesche et al., 2013) and ongoing work adds a realistic amount of calved icebergs to the Southern Ocean. As the melt rate is calculated in the iceberg model, iron could be added to this melt water in realistic concentrations and thereby be added to the water. As the iron supply from icebergs is small relative to the other source (E.g. Table 1.1), it most likely has a local effect, and such a study would be more reasonable if performed using a relatively high spatial resolution, as is possible in FESOM.

8.4.2 The Arctic Ocean

The current global version of FESOM (Wang et al., 2014a) has been used for a number of studies regarding the Arctic Ocean (e.g. Wang et al., 2012; Wekerle et al., 2013). Due to the rapidly decreasing sea-ice extent in the Arctic Ocean, with associated changes in mixed layer depth and light availability, NPP is also likely to change drastically in the area. While contemporary model runs show that the NPP predicted by the models agrees relatively well with the satellite-based observations, the controlling limitation factors of light and nutrient concentrations differ widely between models (Popova et al., 2012). The modeled biogeochemistry of the Arctic Ocean also is thus also sensitive to the performance of the ocean model. As a consequence, the models respond differently to the retreat of sea-ice over the coming century, with
the Arctic NPP increasing in some models while decreasing in others (Vancoppenolle et al., 2013).

As grids with a high resolution in the Arctic Ocean and associated tuned ocean states exist for FESOM, it would be logical to investigate how FESOM-REcoM2 describes the biogeochemistry in the Arctic. An important factor in the future Arctic Ocean is the representation of sea-ice and associated biogeochemistry (e.g. Vancoppenolle et al., 2013). FESOM does a good job in modeling the seasonal and interannual extent of Arctic sea-ice (Wekerle et al., 2013), but the biogeochemistry within the sea-ice is not incorporated into the REcoM2 code, and the light penetration through the sea-ice is crudely parameterized. The impact of these two factors would be interesting to further investigate with FESOM-REcoM2.

8.4.3 Climate runs

The model runs in the present thesis have all been performed with the prescribed CORE-I and II forcing (Large and Yeager, 2004; Large and Yeager, 2009), which is a good choice here as it reduces the complexity of the model set-up and it makes hindcast experiments and thereby skill assessment against existing observational data possible.

Currently, it is not possible to perform runs covering future climate changes with FESOM-REcoM2, as the interannually varying CORE-II forcing data set ends in 2008 (Large and Yeager, 2009). For climate runs over climate relevant time scales, different forcing must thus be considered. One option is to use the output from an atmospheric model and use it as prescribed forcing. This method was used by Hauck et al. (2015) to force the MITgcm-REcoM2 to the end of the 21st century. Another option is to use an interactive atmospheric model component. For FESOM this has become a realistic possibility after its coupling to the atmospheric model ECHAM6 (Sidorenko et al., 2015; Rackow et al., 2014). In a coupled model, ocean and atmosphere are fully interactive and inconsistencies from prescribed forcing are thus avoided. The free interaction, however, adds to the degrees of freedom, with the consequence that larger errors between the modeled and observed fields may occur as the model develops its own equilibrium state. Running FESOM-REcoM2 with an interactive atmosphere is interesting in relation to the current work, as it could shed light on the effect of future changes in the MLD on the vertical iron supply to the Southern Ocean, and its impact on EP and NPP. For studies regarding the Arctic Ocean, it would further be interesting to see how the high resolution of FESOM-REcoM2 would change the model’s reaction to future climate change. But, as highlighted in publication II of the current thesis, such studies will, not give a definitive answer to the future changes in net primary production, as biogeochemical changes are highly dependent on the ocean model.

When running future scenarios with FESOM-REcoM2, it is necessary to first evaluate the carbon cycle in the model. When evaluating the skill of FESOM-REcoM2 in publication I, the carbon cycle was not considered as it was not yet in equilibrium due to the relatively short runtime. A logical next step is thus to assess the skill of the model with respect to Dissolved Inorganic Carbon (DIC), Total Alkalinity (TA), pH and calcium carbonate export as done by Yool et al. (2013). This would open for the opportunity to investigate the atmosphere-ocean flux of CO2 and the effect of ocean acidification on the phytoplankton communities.


In this thesis, the results of the coupling between the biogeochemical model REcoM2 and the Finite Element Sea-ice Ocean Model (FESOM) were first documented. Secondly, the new model was used to gain a better understanding of the sources providing iron to the surface mixed layer of the Southern Ocean. The work in particular focused on the input of iron from the sediment and from the deeper reservoirs through vertical transport across the base of the mixed layer, and how these sources, and their model parameterization and strength, affect the net primary production in the Southern Ocean.

The coupled FESOM-REcoM2 model captures the main spatial surface patterns of the global biogeochemical fields, with relatively low biological production in the subtropical gyres and higher in the temperate regions. The results are comparable to those of other non-eddy resolving Ocean General Circulation Biogeochemical Models (OGCBMs), with the spatial distribution of the chlorophyll $a$ and net primary production (NPP) being in the better end of previous results.

The thesis shows that the representation of the ocean circulation and mixing in the Ocean General Circulation Model (OGCM) has a large impact on the vertical supply of iron to the surface mixed layer. The vertical iron supply in turn affects the representation of net primary and export production as well as the phytoplankton composition in the models, though other factors from the OGCM, such as irradiance in the water, also plays a role. As coupled Ocean General Circulation Biogeochemical Models (OGCBMs) are routinely used to investigate the carbon cycle in the Southern Ocean, the study highlight the importance of evaluating the skill of the OGCM in order to understand the biogeochemical results.

The input of iron to the water from the sediments is thought to play an important role in the Southern Ocean, but measurements provide a large range of flux rates. The strength of the sediment source thus varies widely between OGCBMs. In FESOM-REcoM2 it is possible to vary the strength of the sediment source considerably, and still obtain reasonable and similar results for the surface iron concentration in the Southern Ocean. The strength of the sediment source does, however, impact the phytoplankton composition and its spatial distribution in the model, especially downstream of shelf regions, as well as the magnitude of the opal export. One way to constrain the rate of iron release from the sediments is thus to evaluate the model’s results regarding diatom production and opal export to previous estimates of these factors.

In conclusion, OGCBMs are useful tools to gain an understanding of the marine biogeochemical cycles. Their results are, however, dependent on the representation of the physical and biogeochemical processes in the models and these must therefore be critically addressed. In this thesis this has been done with a focus on the representation of iron limitation in the Southern Ocean.
Part IV

APPENDIX
A.1 INTRODUCTION

The Regulated Ecosystem Model with two phytoplankton classes (REcoM2) was presented in a set-up with the Finite Element Sea-ice Ocean Model (FESOM) in publication I. For publication III FESOM was updated to the newest version (v. 1.4, Wang et al., 2014).

Here, the the results of the CTRL run of publication III are compared to observational data in a manner similar to what was done in publication I, starting with the physical fields, then the nutrients and ending with the biological production. Focus is on the mean surface fields, calculated based on the last ten years of 50-year simulations as described in publication III.

The results are first presented in Taylor diagrams (Taylor, 2001), in which the model results are compared to observations. A number of plots additionally illustrate the bias between modeled and observed surface fields. For the mixed layer depth (MLD) and net primary production (NPP) we additionally show the seasonal cycle in the individual ocean basins illustrated in Fig. A.1. The data we compare the model results to is listed in Table 1 (Pub. I) along with its temporal extend and resolution.

A.2 RESULTS

A.2.1 Physics: Mixed Layer Depth

The MLD was calculated using the density threshold criteria of 0.03 kg m$^{-3}$ as suggested by de Boyer Montegut et al. (2004). The correlation between the observed and the modeled mean MLD equals 0.73 (Fig. A.2a), and has thus been improved from the correlation of 0.68 found by (Schourup-Kristensen et al., 2014).

Overall, the spatial distribution of the MLD$_{max}$ is well captured, with the shallower MLD in the tropics and deeper in the temperate areas (Fig. A.3). In the Southern Ocean, the deep mixed layers in the Indian and Pacific sectors are captured, though their meridional extend is somewhat off. Close to the Antarctic, the MLD$_{max}$ is too shallow as was also the case in the previous model version (Fig. A.3).

Taking the spatial distribution of each month into account, the correlation between the observed and the modeled mean MLD is 0.58 (Fig. A.2b). The seasonal cycle of the individual ocean basins illustrated in Fig. (A.1) all have significant correlations with the observations (Fig. A.4). Additionally, all ocean basins, except for the North Indian and Equatorial Pacific Basins, have correlations that are higher than 0.9. In the Southern Ocean, the depth of the MLD over the year has been improved compared to the previous version of FESOM, as the bias between the modeled and the observed MLD has become smaller. Overall, FESOM does a good job in reproducing the seasonal cycle of the MLD.
Figure A.1: Division of the global surface ocean into individual basins used to show the seasonal cycle of the MLD and NPP.

Figure A.2: Taylor diagrams (Taylor, 2001) showing correlation, normalized standard deviation and the normalized root mean square error between values of the model results and observations (Table 1, Schourup-Kristensen et al., 2014), weighted by area. a) Spatial distribution. b) Spatial-seasonal distribution.

All values are surface values, except the mixed layer depth and the vertically integrated NPP. All fields have been interpolated to a 1° by 1° grid, using linear interpolation.
Figure A.3: Mean MLD\textsubscript{max} for a) observations (de Boyer Montegut et al., 2004), b) model.

Figure A.4: Mean MLD over the year in the basins depicted in Fig. A.1 for the model results and observations de Boyer Montegut et al. (2004). The correlation coefficient between model results and observations is written in each subplot, and the statistically significant correlations (p values < 0.05) are marked with *.
For dissolved inorganic nitrogen (DIN) and silicon (DSi), the correlations between the spatial distribution of the observations and the model results are 0.88 and 0.86, respectively. These relatively high correlations occur as the main distribution with high concentrations of DIN and DSi in the Southern Ocean and the northern seas, along with low concentrations in the subtropical gyres are captured in the model (Fig. A.5 and A.6). For DIN, the surface concentrations are, however, too high in the tropical Pacific Ocean, as was also the case when REcoM2 was coupled to FESOM 1.3.

Due to scarcity of iron measurements, the water’s concentration of dissolved iron can not be directly compared to a climatology of observations. The spatial distribution of surface iron nevertheless appears reasonable, with the highest concentrations in the Equatorial Atlantic and Indian Oceans, which are under influence of the dust plume from the Sahara. The lowest concentrations occur in the Pacific Ocean, whereas the Southern Ocean has intermediate surface iron concentrations (Fig. A.7). When comparing to mean values for various ocean basins it is apparent that the modeled iron field is closest to the observations in the Atlantic and Indian Oceans, while the mean concentrations are too low, especially in the Pacific, but also in the Southern Ocean (Table A.1).

The relatively low iron concentrations in the Pacific and Southern Ocean lead to iron being the most limiting nutrient in these areas, both for nanophytoplankton and diatoms (Fig. A.8). Nanophytoplankton is on average limited by iron in 57% of the global ocean, while the number for diatoms is 54%. These are relatively high numbers, showing that the iron limitation is relatively strong in the model, especially
Figure A.7: Mean surface concentration of dissolved iron. Plotted to the scale of Schneider et al. (2008)

Table A.1: Modeled mean surface iron concentrations (0–100 m) in the different ocean basins shown in Fig. A.1. Observed values are from Moore and Braucher (2008), except those marked with *, which are from Tagliafu et al. (2012), Table 2. The latter is the mean of the values given for the Antarctic and Subantarctic regions.

<table>
<thead>
<tr>
<th>BASIN</th>
<th>LATITUDINAL EXTEND</th>
<th>MODEL</th>
<th>OBS.</th>
</tr>
</thead>
<tbody>
<tr>
<td>North Atlantic</td>
<td>45 – 70° N</td>
<td>0.42</td>
<td>0.68</td>
</tr>
<tr>
<td>North-central Atlantic</td>
<td>10 – 45° N</td>
<td>1.25</td>
<td>0.68</td>
</tr>
<tr>
<td>South-central Atlantic</td>
<td>45 – 10° S</td>
<td>0.86</td>
<td>0.44</td>
</tr>
<tr>
<td>North Indian</td>
<td>45 – 10° S</td>
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<td>1.21</td>
</tr>
<tr>
<td>North Pacific</td>
<td>45 – 70° N</td>
<td>0.14</td>
<td>0.31</td>
</tr>
<tr>
<td>Equatorial Pacific</td>
<td>10° S – 10° N</td>
<td>0.02</td>
<td>0.84</td>
</tr>
<tr>
<td>South-central Pacific</td>
<td>45 – 10° S</td>
<td>0.02</td>
<td>0.31</td>
</tr>
<tr>
<td>Atlantic Southern Ocean</td>
<td>90 – 45° S</td>
<td>0.13</td>
<td>0.39*</td>
</tr>
<tr>
<td>Indian Southern Ocean</td>
<td>90 – 45° S</td>
<td>0.07</td>
<td>0.33*</td>
</tr>
<tr>
<td>Pacific Southern Ocean</td>
<td>90 – 45° S</td>
<td>0.07</td>
<td>0.15*</td>
</tr>
</tbody>
</table>
in the Pacific Ocean. For diatoms, silicon is dominating the nutrient limitation in the remainder of the global ocean.

### A.2.3 Biological production

The global NPP is 32.4 Pg C yr$^{-1}$ and the export production (EP) 5.9 Pg C yr$^{-1}$, very similar to the numbers in the previous version of FESOM-REcoM2. The global export of biogenic silicon (opal) sums up to 67 Tmol Si yr$^{-1}$.

For the spatial distribution of chlorophyll a and NPP, the correlation between the model results and the satellite-based estimates are 0.73 and 0.75, respectively. For both fields, the spatial distribution, with low concentrations in the subtropical gyres and higher between 40 and 70°S and N, is captured (Fig. A.9 and A.10). But especially the surface chlorophyll concentrations are too low in the gyres, and both fields have too low concentrations in the Pacific Ocean due to the strong iron limitation here (Fig. A.8).

The seasonal cycle of NPP is plotted in Fig. (A.11) for the ocean basins shown in Fig. (A.1). The seasonal cycle of the modeled NPP has significant correlations with the satellite based estimates in eight of the fourteen basins, and the best seasonal cycles are found in the subtropical gyres. North and south of 45°N and S, respectively, the spring bloom starts earlier in the model than in the satellite-based estimates, and in the Pacific Ocean it is again clear that the modeled NPP is relatively low.

For export production (EP), the correlations are 0.28 and 0.46 when compared to the fields from Schlitzer (2004) and Laws et al. (2000), respectively. Export production is, however, difficult to measure, and the fields we compare to are thus uncertain. Looking at the spatial distribution, the modeled EP (Fig. A.12) is highly correlated with the NPP (Fig. A.10), with large export rates occurring in the upwelling zones around the Equator, in the Southern Ocean and in the north Atlantic. As indicated by the correlation coefficients, this is closest to the EP-field from Laws et al. (2000),

**Figure A.8:** Annual mean limiting nutrients calculated following Schneider et al. (2008). a) Nanophytoplankton (Fe = 56.8%, DIN = 41.3% of total area. b) Diatoms (Fe = 53.9%, DSi = 40.6% and DIN = 5.5% of total area.)
A.3 conclusion

Figure A.9: Spatial distribution of the mean surface concentration of chlorophyll $a$ plotted to the scale of Schneider et al. (2008). a) Satellite-based estimate (www.globcolour.info). b) Modeled mean.

Figure A.10: Spatial distribution of the vertically integrated net primary production plotted to the scale of Schneider et al. (2008). a) Satellite-based estimate (SEAWIFS and Behrenfeld and Falkowski (1997)). b) Modeled mean.

while the field from Schlitzer (2004) has less export in the North Atlantic and more in the coastal zones.

A.3 CONCLUSION

The biogeochemical model REcoM2 has been coupled to the newest version of the Finite Element Sea-ice Ocean Model (FESOM, v. 1.4). The results of the new model are very similar to those of REcoM2 coupled to FESOM,1.3. The largest differences occur in the Southern Ocean, where the mixed layer has improved from the previous version, leading to more production in the Pacific sector of the Southern Ocean, and generally to higher levels of chlorophyll $a$ in the surface water.
Figure A.11: Mean NPP over the year in the basins depicted in Fig. A.1 plotted for the model mean and satellite-based observation (SeaWIFS and Behrenfeld and Falkowski (1997)). The correlation coefficient between model results and observations is written in each subplot, and the statistically significant correlations (p values < 0.05) are marked with *.

Figure A.12: Spatial distribution of the export production. a) Satellite-based estimate (Laws et al., 2000). b) Inverse model based estimate (Schlitzer, 2004) c + d) Modeled EP.
REFERENCES


