Dear Dr. Paul Halloran,

Thank you for serving as topical editor for our manuscript (GMD-2019-347) entitled 'Ocean biogeochemistry in the Norwegian Earth System Model version 2 (NorESM2)'.

We are also very grateful for two very thorough and constructive feedbacks from Dr. Segschneider and Dr. Palmieri. We believe that we have addressed all of the reviewers' comments in the latest revision.

As discussed earlier, we will submit a revised statement with regards to code availability in April, as we are currently still finalizing the licensing for the code release. Thank you for your understanding.

Please find attached:

- 1. Response letter with point-by-point response to Dr. Segschneider comments,
- 2. Response letter with point-by-point response to Dr. Palmieri comments, and
- 3. Marked-up revised manuscript.

Please don't hesitate to contact me with any questions.

Thank you again,

Jerry Tjiputra (on behalf of all authors)

Response letter to Reviewer1: Dr. Joachim Segscheneider

We thank Dr. Joachim Segschneider for his prompt and very thorough review of our manuscript. We have now addressed all of his comments (in **bold**) with point-by-point (i) responses (in *italics*) and/or (ii) description of changes made in the revision (in blue) detailed below. Changes in the revision can also be seen in the attached revised manuscript with highlighted changes (red and blue depicts older and revised version, respectively).

The paper is generally quite well written, with some room for improvement mainly in the 'Results' section, the figure captions, and the supplementary material. The supplementary material could be better justified and introduced, hardly any of the figures are discussed in the main text.

We agree with this assessment and have accordingly revised the 'Results' section, figure captions and supplementary material.

In the revised manuscript, we have made some improvements in the abovementioned components of the manuscript, add more references to the (revised) supplementary materials. Since many of the figures in the previous supplement materials, especially those showing results from NorESM2-LM version, closely resemble those of NorESM2-MM in the main manuscript, we have removed these figures.

Also, the reader's life could be made easier by a more specific referencing to the figures (like pointing to the specific panel, not only the Fig., and perhaps the particular feature like 'blue curve in Fig X.x'). In summary I recommend publication of the manuscript after a careful revision addressing the below comments. Due to the length of the paper, my comments are quite numerous, but by nature I would consider the requested changes as minor revisions.

We agree and have followed your advice accordingly. We've gone through the revised manuscript and inserted specific figure references where they fit.

Changes are applied throughout the manuscript, especially in the Results section.

Specific comments p5 ln 19 and 23 better: 'atmosphere-ocean coupling' instead of 'ocean coupling'

Done.

Both changed to 'atmosphere-ocean coupling'.

Section 2.3 p6 ln 6-ln10: I was a bit confused by this (DMS production) description, in particular there seems to be an error in the use of degradation/production? after reading through several papers, I now assume production should be changed do degradation in line 10 - or it should be 'detritus production' of opal/CaCO3. In Six et al. 2013 the k refers to sulphur to carbon ratios in cells of opal/CaCO3. In Six and Maier-Reimer 2006 the

gammas to degradation rates of opal/CaCO3; not clear to me what is meant in Eq. 2 please double-check and revise Section 2.3

Thank you for noticing these inconsistencies. We have updated and expanded Section 2.3, as follows:

- Changed the reference from Six and Maier-Reimer (2006) to Kloster et al. (2006), which is the updated parameterization currently used in our model.
- Revised equation (1) and provided an improved description for source and sink terms associated with detritus export production, bacterial consumption, and photolysis.
- Revised equation (2), through renaming the different terms.
- Added equations for loss terms due to bacterial consumption and photolysis are now added in the revision.

p6 Eq (1) capital P in Phy irritating - I suggest to use 'phyto' to avoid mixing up with physics; also, is there a reason why some terms carry the 'DMS', others not?

We follow your suggestion. Some terms initially carry 'DMS' as it is computed as a function of the DMS concentration in the model.

The term 'Phy' has been replaced with simply 'prod' (for production) and 'DMS's are removed in all terms in the revised eq. (1).

p9 ln16/18: perhaps the ref. to Fig. 1 could be moved to the end of the para, or even toward the end of the section, otherwise one is trying to connect the following statements to Fig. 1

We agree.

Moved the sentence referencing Fig. 1 to the end of the 3rd paragraph in the same section.

p10 ln25 I presume what is meant here is that preformed phosphate can be used to estimate the organic carbon pump, and preformed alkalinity to estimate the inorganic carbon pump? As the sentence stands now, either can be used to estimate both pumps. -> Preformed phosphate can be used to quantify the organic (), and preformed alkalinity to quantify the inorganic () carbon pump ().

Correct. We have revised the sentence to:

"Following Bernardello et al. (2014), PO_4^{pre} is used to quantify the organic (soft tissue, DIC^{soft}), whereas both PO_4^{pre} and ALK^{pre} are used to quantify the inorganic (carbonate, DIC^{carb}), biologically-mediated carbon pump (Eqs. 12-14)".

p10 ln26 (Eqs. 10-12; Bernardello et al.) is a bit misleading, perhaps (Eqs. 10-12, based on Bernardello et al.) is better suited (the corresponding eqs. are 1-4 in Bernardello e al. 2014)

We agree. The sentence has been revised (see previous response).

p11 eq.12 I do not quite follow how eq. 12 is derived (where does the +1 originate from?) +1 not in Bernardello et al. 2014 Eq. 4

In our updated code, we have assumed that during organic production and remineralization, both changes in nitrate and phosphate alter the proton concentration and therefore change the alkalinity (see also Section 3.1.3 of Paulmier et al., 2009). We have included the following statement in the revision:

Equation 14 is slightly different than that in Bernardello et al. (2014), $r_{N:P} + 1$ instead of $r_{N:P}$, because, in our updated code, we do not neglect the contribution of phosphate to alkalinity changes during biological production and remineralization. For instance, both nitrate and phosphate produced during organic remineralization increase the concentration of proton and therefore reduce the alkalinity (see also Section 3.1.3 of Paulmier et al., 2009).

p15 ln 18 'branched off into' is not a valid formulation. rephrase to sth like: the PI, control and historical exps are branched off from the spin-up. or 'the simulation is used as a starting point'....

We have rephrased the sentence as suggested to:

"The end of the model spin-up is used as a starting point for ..."

p15 ln 20 see above

Rephrased to:

"Simulation (3) is furthermore used as a starting point for"

p16 ln24 rephrase 'at subsurface 500 m'

Rephrased to:

"At 500 m depth, ..."

p16 ln25/26 'improvements ... in deviation' does not sound good. -> improvements in agreement, better agreement

Rephrased to:

"... shows better agreement with data throughout ..."

p18 ln 5 what is meant by: the simulated concentrations are mostly confined to the upper 1 km?

It was meant to describe that most of the high CFC-11 values are mostly simulated between the surface and 1 km depth. We have revised the sentence to:

"Similar to the observations, high values of CFC-11 are generally simulated in the upper 1 km in both the Atlantic and Pacific, with the exception of the North Atlantic ... "

p23 ln 31 ct Consistent with the lower oceanic (than atmospheric) pCO2 partial pressure (I guess partial pressure is meant here, since the oceanic growth rates are not shown in Fig. 20)

Yes, thank you. We have rephrased the sentence to:

"Consistent with the lower surface ocean (than the atmospheric) pCO_2 , the ocean carbon uptake continues to increase ..."

Fig. 2 since the panels are labelled a-d, better use 'a,c' in the caption instead of 'left', etc

Done.

Replaced left, right, top, and bottom with 'a,c', 'b,d', 'a,b', and 'c,d'.

Fig. 9 is never discussed in the text

In the revised section 5.6, towards the end, we have added the following (previously missing) discussion on Silicate (Fig. 9):

"The silicate spatial distribution in NorESM2 resembles closely the phosphate distribution (Figs. 9a-c and 7a-c), with similar biases across the vertical sections of Atlantic and Pacific (Figs. 9e-f and 7e-f). At high-latitudes, NorESM2 overestimates the surface silicate concentration (Fig. 9d), which could be attributed to the reduced opal export sinking speed relative to our earlier model version (30 instead of previously 60 m day⁻¹)."

Fig. 14 I am a bit sceptical about the term 'Southern Ocean' for latitudes between 20 deg S and 40 deg S. Why not just call it southern hemisphere mid latitudes.

We followed the reviewer's suggestion.

We have revised both figures 14 and 19 and renamed the region to "Mid-latitude southern hemisphere". The corresponding texts have been revised too.

The content description of the supplementary material is full of errors.

The supplementary material has been revised.

In addition, we have also reduced the number of supplemental figures (from 16 to 7; Figures S6 and S7 are new), since many of the figures (i.e., distributions of biogeochemical tracers, maps of

limiting nutrient, surface pCO₂ and CO₂ fluxes, DMS concentration, and mean primary production and transfer efficiency) from NorESM2-LM are qualitatively and visually very similar to those of NorESM2-MM (shown in the main manuscript).

Technical errors

general: Check all occurrences of 'to allow for' (which means to consider s.th. when planning for s.th., see e.g. dictionary.cambridge.org) and replace by 'allows' or 'enables us to' or similar change all occurrences of 'insight to' to 'insight into' check for sgl/pl and past/present mismatches

We have checked for and revised the above phrases as suggested.

in the following 'ct' stands for 'correct to' p1 ln 5 correct 'allow for'

Corrected to: "allow for".

p1 ln 7 riverine 'input'; p1 ln 7 'are recently' does not make sense -> have recently been ...

The sentence has been revised to: "have recently been...".

p2 ln 6 remove 'us' (who is 'us' here?) or better reformulate whole sentence

Rephrased to: "ESM simulations can therefore be used to estimate historical carbon budgets and future carbon emission pathways under specified scenarios."

p2 ln 8 remove 'us'

Rephrased to: " ... ESMs are state-of-the-art tools used to study ...".

p2 ln 13 ct ...hardware systems, higher resolution

Done. "(... new hardware systems, higher resolution, etc.)".

p2 ln 16 complex interplay between what and what?

Replaced "... complex interplay of internal climate variability." with "... complex interactions with the internal climate variability."

p2 ln 21 remove 'through'

Done.

p4 ln 3 check use of 'implications on' (correct: implications of sth. for sth.) should be 'consequences for', 'impact on', or similar

Replaced with 'consequences for'.

p4 ln 11 ct 'insight into the ocean's role'

Done. "... insights into the ocean's ...".

p4 ln 26 ct'in Section 4.'

Corrected.

p4 ln 28 ct'in Section 6.'

Corrected.

p4 ln 32 delete 'for'

Removed 'allows for a'.

p5 ln 2 ct we also now apply (or we also applied)

Revised to 'We also applied ...'.

p5 ln 29 ct 'closer to' (or better, 'which is within the range'....)

Used 'which is within the range'.

p6 ln 24 ct ... an early version of 'the' Global-NEWS model ...

Added 'the'.

p7 ln 15 ct and 'are' added to the nitrate pool (otherwise it is not clear if this is only assumed)

Removed 'assumed to be bio-available and'.

p7 ln 16 ct Particle export (without plural s)

Corrected.

p8 ln 1 ctinterior biogeochemistry 'using' the different (not 'in')

Replaced 'in' with 'using'.

p8 Eq. 4 dot between mu and max misplaced

Corrected.

p8 ln 14/15 change to: 1.25 moles dissolved oxygen and 1 mole of alkalinity...

Changed as suggested.

p9 ln 11 ct where the strength 'of' this

Corrected.

p 11 Eq. (12) dot after 0.5 misplaced

Corrected (now Eg. 14).

p11 Eq. (13) remove leading dot

Done (now Eq. 15).

p11 ln24 insight into

Corrected.

p11 ln29 correct 'allow for'

Rephrased to 'The inclusion of natural tracers provides substantial saving ...'.

p12 ln 5 remove 'to' before (ii)

Done.

p12 ln10 correct 'allow for'

Rephrased to 'In order to allow comparison between different ...'

p13 ln 27 ct ... different parameterizations, (add s)

Done.

p14 ln 16 ct , as follows (add s)

Done.

p14 ln 25 ct 'a quasi-equilibrium state' (remove pl. s)

Done.

p14 ln 30 ct ...carbon counterpart of 'the' respective ...

Done.

p15 ln 1 ct ...applied 'to' organic ... (not 'for')

Done.

p15 ln 3 -> atmospheric pCO2

Done.

p16 ln 8 correct 'the the'

Done.

p16 ln11 either 'the majority' or 'the major part' (but not 'majority part')

Used 'the majority of the mean ...".

p17 ln 29 ct 'In the subpolar North Atlantic'....

Corrected.

```
p17 ln 31 ct ....'a' deep MLD ....
```

Done.

p18 ln 5 delete 'depth' after upper 1 km

The sentence has been rephrased: "Similar to the observations, high values of CFC-11 are generally simulated in the upper 1 km in both the Atlantic and the Pacific, ..."

p18 ln 16 ct material being

Corrected.

p18 ln 17 ct ... 'at' intermediate depth

Corrected.

p18 ln 18 ct ... 100 and 1500 m depth).

Done.

p18 ln 20 ct ... concentrations of all nutrients ...

Replaced 'in' with 'of'.

p18 ln 27 ctNADW as the main watermass ...

Corrected.

p19 ln 3 ct ...(see also Fig 8i). - check if 8i was meant, 8e is Atlantic NorESM2

Yes, we meant Fig. 8i (not 8e). Thank you.

Corrected.

p19 ln 8/9 ct: ...low levels.... limit the phytoplankton growth.

Corrected.

p19 ln 21 correct 'Fig. 9' to 'Fig. 12'

Thank you.

Done.

p19 ln 29 ct ... and at high latitudes during summer months.

Done.

p20 ln 3 ct ... the spring blooms....

Done.

p20 ln 4 delete 'during the boreal spring months' (redundant in sentence)

Deleted.

p20 ln 9 coastal areas (not grids)

Corrected.

p20 ln 20/21 move ref to Weber after 'biogeochemistry' ctstill simulates too high transfer efficiencies...

Both corrected.

p20 ln 22 ct ... are comparable with observations.

Done.

p20 ln 26 correct 'Fig. 2c' to Fig. 2d; ctsimulates a lower ...

Both corrected.

p21 ln 8 ct ...at the lower end...

Done.

p21 ln 24 ct As at the surface...

Done. Replaced "As in the surface ..." with "As at the surface ...".

p22 ln 12 ct ...translates into stronger carbon sinks...

Done.

p22 ln 20 ct of DIC-rich deep watermasses...

Done.

p22 ln 21 ct ...strong bias ... is considerably reduced,

Corrected as suggested.

p22 ln 22 ct ... is approximately reversed compared to observations:

Corrected as suggested.

p22 In 26 ct Nevertheless, the

Done.

p23 ln 25 ct 0.7 oC, comparable to that from obs...

Replaced "... relatively comparable with that from the ..." with "comparable to that from ...".

p23 ln 26 ct the warming in esm-hist

Done.

p23 ln 30 ct (i.e., is lower than)...

Done.

p24 ln 7 ct For the 1980s and 1990s ...

Done.

p24 ln 9 correct allowing for

Done.

p24 ln 15 ct ...by the ocean for 1850-1994 and 1994-2011 is

Done.

p24 ln 27 correct allow for

Corrected to 'which enable us to'.

p24 ln 28 ct , and (iii) carbon isotopes that can be used e.g.

Corrected.

p25 In 8 ct the equatorial Pacific OMZ

Corrected.

p25 ln 9 ct ...Southern Ocean, and the equatorial and North Pacific.

Corrected.

p25 ln 14 replace allowing for by 'resulting in'

Done.

p25 ln 20 ct simulates a considerable bias

Corrected.

p25 ln 25 ct the CO2-fluxes' seasonal cycle...

Corrected.

p25 ln 26 attention to (not of)

Revised "attention of" to "attention to".

p25 ln 30 ct ... depth of the equatorial Pacific

Corrected.

p25 ln 33 ct penetrating too far north..

Corrected.

p26 ln 1 ct biogeochemical components in ESMs...

Corrected.

p26 ln 2 ct the mean climatological state. However, (check if the references have to be moved)

Checked and corrected.

p26 ln 3 ct a lot of effort

Corrected.

p26 ln 5 replace allowed for by e.g., 'provided'

Replaced 'allowed for additional constraints' with 'provided additional valuable constraints'

p26 ln 15 correct allow for

Replaced 'allow for efficient optimization of the current ecosystem parameterization' with 'enable us to perform optimization of the current ecosystem parameterization more efficiently'.

p26 ln 19 ct ..currently, we use fixed particulate organic carbon emissions ...

Corrected.

p26 ln 27 ct to investigate the sensitivity of

Corrected.

p26 ln 35 ct as the results of more complex model simulations

Corrected.

Table 1: ct ... ecosystem parameterisations that have been changed....

Corrected.

 Table 2: ct ... simulation periods over which their climatological values have been averaged.

 ct Average of the three remote sensing products...

Both phrases have been corrected.

Table 3: ct Annual mean biology-related metrics... (not only primary production is listed)

Agree, and corrected.

Fig. 1 ct Blue depicts components, processes, (there is only one blue)

Corrected.

Fig. 3 caption: ct 'Differences between ... are also Fig. 4, 7, 8, 9, 11, 13, 16, 17, 21, 22, 23

Corrected for all mentioned Figures.

Fig. 14 averaged over all months. also Fig. 19

Done.

Fig. 20 ct relative to the 1850-1879 period, (b)... Green depicts results from the simulation with NorESM2. The purple line in panel (c) represents (only one green, only one esm-hist with NorESM2-LM)

Corrected. Thank you.

Response letter to Reviewer2: Dr. Julien Palmieri

We thank Dr. Julien Palmieri for his prompt and constructive review of our manuscript. In particular, Dr. Palmieri raised important comments regarding the carbon isotopes evaluation in our manuscript. We agree with him that the stand-alone ocean simulation used to produce the carbon isotope results are not directly comparable with the coupled ESM configuration runs, which are the main focus of the paper. However, we believe and hope Dr. Palmieri concur, that the carbon isotope feature in NorESM2 is an important novel component and would still fit in the current manuscript. In the revision, we have emphasized the difference in the experimental setups. We have also shown (in new Supplementary figures) that despite the differences in model configurations and resolutions, both the stand-alone and coupled ESM simulations share very similar quasi-equilibrium interior biogeochemistry states (as seen in the interior distributions of phosphate and DIC concentrations). In addition to this, we have also addressed all of Dr. Palmieri's comments (in **bold**) in the revised manuscript. Point-by-point (i) responses (in *italics*) and/or (ii) description of changes made in the revision (in blue) are detailed below. Changes in the revision can also be seen in the attached revised manuscript with highlighted changes (red and blue depicts older and revised version, respectively).

The paper mainly describes the improvements made to the model since the previous version used for the CMIP5 exercise, comparing both to observation, and showing that the new version is mainly better the its predecessor. Some difficulties come from this presentation choice as sometime the authors consider that some description being made in the NorESM1 description papers, the reader is assumed to know them. In results, some basic description are almost missing. For example, the ocean grid description of which we only know it is tripolar.

We understand the reviewer concern and have added more basic description of the model, e.g., information on the ocean grid has been added (see also below). Indeed, the ocean carbon cycle model in NorESM2 has gone through several update iterations with respective documentations (e.g., Tjiputra et al., 2013: Schwinger et al., 2016). In order to avoid overlaps, we have opted to focus the current paper on key improvements in process representations and parameterizations. Improvements related to this comment can also be seen further below.

We have added the horizontal and vertical grid information in the last paragraph of Sect. 2.1.

Apart some little corrections of that kind, listed bellow, which should hopefully help to improve the paper, i have one concern about the isotope run. The isotopes are run on an ocean-only run, with a lower resolution grid than the coupled run, and a different atmosphere to force the run. I don't think this run should be used to evaluate the coupled run, unless the authors show that both model DIC steady states are comparable, or mention in the isotope results paragraph that these results are only informative, because from a different run. This needs - at least - to be reminded to the reader. Written as it is, the isotope results are in the middle of historical and esm-hist results. The reader can easily misunderstand and think they all are from the same simulation. Thank you for this thoughtful comment. Even though the isotope run was configured differently, we still think that the results of our isotope run represent an important advancement to the model and would still fit in the current manuscript. We agree with Dr. Palmieri that a clearer separation is needed.

We have moved the discussion on the isotope results toward the end and reminded again the reader that this is not from a coupled configuration run, in contrast to the other variables. Following your suggestion, we have also shown the interior DIC (and phosphate) concentrations under the preindustrial state (see also below) from the ocean-only simulation.

Overall, i have positive feelings about this paper. Most comments are minor like rephrasing or asking for missing details, what should translate in minor revision.

Thank you very much for all the positive and constructive feedbacks.

P2 11 : rephrasing : absorber of heat and the greenhouse gas CO2 to absorber of heat and of the greenhouse gas CO2 or absorber of heat and CO2 greenhouse gas ?? But the one in the paper sounds weird to my ears.

We have rephrase this to: "... absorber of heat and of the greenhouse gas CO2 ...".

P2 130 to P3 13 : Is there an equivalent paper to this one for NorESM1? It seems there is no reference description paper for the version 1 of the model. You list different papers of different development stages, but which one is the one that would describe the version of NorESM1 you use in this paper?

Yes, there is a model description paper for NorESM1 (Tjiputra et al., 2013).

we have clarified in the revision, i.e., towards the end of this paragraph, that the comparison presented here is relative to the version described in Tjiputra et al. (2013).

P3 13 : "which will contribute to CMIP6". You can probably write it at present tense.

We agree. Rephrased to "... which contributes to CMIP6."

P3 124-33 : This can not replace having 2 separate runs. Which nutrient does the biology see ? How important could that be within the scenario ? (PI carbonate system evolves with biology that feels high CO2 world, pH,...) probably not important for historical runs. that's a problem we faced within OMIP, and we could not have a definite answer (should run both to be sure...).

The reviewer is correct that the enhanced carbon sinks alter the carbonate chemistry in the model and, in turn, the dissolutions of $CaCO_3$ in our model. Aside from this, the biological formulation such as phytoplankton and nutrient dynamics are not affected by the high CO_2 (or other scenario runs). To address this, we have also added respective 'natural' components of alkalinity and $CaCO_3$ as described in subsection '3.2 Natural inorganic carbon tracers'. To

further clarify this and the 'non-effect' on biological production, we have added the following statements in subsection 3.2:

"This is because the anthropogenic CO_2 uptake alters the carbonate system and therefore the dissolution of $CaCO_3$. We also note that anthropogenic carbon does not influence the biological production (e.g., nutrient concentrations and phytoplankton growth rate) in our model."

P3 l34 to p4 l6 : 2 remarks : you say " The only external source [in NorESM1] is through atmospheric nitrogen fixation" you mean there is no dust (iron/Si/P) deposition in NorESM1? from this paragraph it sounds like you only add riverine nutrients, but you talk of dust somewhere else... could you clarify? Also, what is included in riverine nutrients? (actually you answer later on - maybe add a "see paragraph 2.4 for details")

You're right, we do have atmospheric deposition of dust (converted into dissolved iron).

We have corrected this error in the revised manuscript. In addition, in the last paragraph of the introduction, we have emphasized that "… more detailed descriptions of all improvements in biogeochemical processes are described in Section 2."

2.1 : you give several details in this paragraph but we miss some details of the ocean grid, like the basic ones: resolution, number of vertical levels,...

We have added the horizontal and vertical grid information in the last paragraph of Sect. 2.1.

p7 l4-7 : i don't understand what you explain there to get the riverine fluxes of DOC and POC, maybe you could try to make it clearer ?

We have rephrased the sentence as follows:

"In our model, the organic form of dissolved carbon (DOC) is connected to nutrients through the Redfield ratio (C:N:P=122:16:1), and therefore other forms of dissolved organic matters (e.g., DON, and DOP) are not explicitly simulated. Since Global-NEWS provides estimates of dissolved organic matter in carbon, nitrogen, and phosphate forms (DOC_{riv} , DON_{riv} , and DOP_{riv}) separately, only the minimum of the three riverine dissolved organic constituents is added to the DOC term in the model (i.e., $DOC=DOC+min(DOC_{riv}, r_{C:N}*DON_{riv}, r_{C:P}DOP_{riv})$). Any excess or remaining organic matter of the three constituents is then added to the corresponding inorganic pools (DIC, NO₃ or PO₄). The same concept also applies to riverine inputs of particulate organic carbon (POC) (see also Bernard et al., 2011)."

P9 paragraph 2.9 – So aerial dust deposition was already in NorESM1 - don't forget to add it P3.

This information has now been added to Page 4 (introduction).

P9 l11 - typo : strength *of* this

Corrected.

P10 112 : "during the spin-up" might be misleading. Maybe "for the spin-up" or "to start the spin-up" or just "initialized to zero" might be enough.

Rephrased to "at the beginning of the spin-up".

P12 114 : "... and CaCO3 formation, but the latter is neglected in our implementation". Why ? any idea of the impact ?

This was addressed further in subsection 3.3.2:

"Isotope equilibrium fractionation during CaCO₃ formation increases $\delta^{I3}C$ of CaCO₃ and decreases seawater of $\delta^{I3}C$. Nevertheless, the fractionation effect during CaCO₃ formation is relatively small compared to the effects of air-sea gas exchange and photosynthesis and therefore is often omitted in modelling (Lynch-Stieglitz et al., 1995; Schmittner et al., 2013)."

We have further revised this statement to:

"Isotope equilibrium fractionation during CaCO₃ formation increases δ^{13} C of CaCO₃ and therefore depletes seawater of ¹³C (thereby lowering seawater δ^{13} C). Nevertheless, the fractionation effect during CaCO₃ formation is relatively small (i.e., in the order of -2 to +3‰, depending on species and environmental conditions; Grossman and Ku, 1986; Ziveri et al., 2003; Zeebe and Wolf-Gladrow, 2001) compared to the effects of air-sea gas exchange and photosynthesis. Therefore, fractionation during CaCO₃ formation is commonly omitted in modelling studies (e.g., Lynch-Stieglitz et al., 1995; Schmittner et al., 2013)."

P14 l24 to P15 l2 : The method to initialize the C isotopes is very clever, and must save a lot of computing time. But i wonder the impact of initializing them this way. Don't you introduce bias compare to a proper model integration? Do you note drifts in the isotopes concentration during the historical period?

Yes, the method will introduce a bias, but avoids a long-term drift towards the same bias. Under this configuration, the drifts for $DI^{12}C$ and $DI^{13}C$ at the end of spin-up are in the order of 0.005% per 1000 years. Secondly, we only performed a spin-up stand-alone ocean simulation where the carbon isotopes are activated (hence no 'historical' simulation). The results shown in the paper represent values at the end of the 5000 years spin-up which is a close representation of the preindustrial state.

We have clarified that only a preindustrial spin-up was performed at the end of revised Section 4 and beginning of Section 5.13.

P15 119-23 : might need a bit of rephrasing here. explanation about (3) are a bit confusing. i would suggest to call it emission-driven simulation as (1) and (2) are also "esm" runs. Or if you absolutely want to call it esm-... explain straight what it is like " (3) esm- experiments, the atmospheric CO2 is prognostically computed from ocean-atmosphere and land-

atmosphere CO2 -fluxes, as well as from prescribed anthropogenic emissions (for the historical period). (3) is composed of a esm-piControl-spinup simulation (100 model years) which is then branched off into (3a) an esm-piControl simulation (for 250 years), and (3b) a transient esm-hist (years 1850-2014) simulation." But calling this experiment "esm" is confusing for the reader.

The use of the 'esm-'naming in our emission-driven simulations follows the standard CMIP6 protocol (Eyring et al., 2016), so we would like to keep it.

As suggested by both reviewers, we have revised the text describing experiments (3), to make it clearer and easier to understand.

P15 130 : "The atmospheric forcing of the spin-up is the CORE normal year forcing (Large and Yeager, 2004), which represents a climatological mean year with a smooth transition between end and start of the year" Why did you use the CORE climatology ? that's a surprising choice, it removes all substantial link between your C-isotope/ocean- only run and the esm. Using CORE remove all potential isotope insights on the esm runs... there are other ways that could provide this insight possibility: you could extract atm fields from the end of the spin-up and use them as forcing field for this experiment, for example. Using CORE, the ocean dynamics will be different from the coupled, the spin-up steady state will have nothing to do with the esm's steady state. Already changing the ocean resolution affect the dynamic, but resolution plus forcing completely change the ocean. I don't think you can use the isotopes to evaluate the esm run with such a different set-up.

Yes, we completely agree with you that the stand-alone ocean simulation with CORE normal year forcing (in addition to the different spatial resolution) is not directly comparable with the simulation performed in a coupled ESM configuration. The main intention was simply to describe the carbon isotope implementation and furthermore we think that a first evaluation can still be done with a, computationally cheap, CORE forcing setup (e.g., the large-scale circulation pattern is not completely different).

We have emphasized this in the revised Section 5.1.3, and in addition, added two supplemental figures S6 and S7, which illustrate the similarity between the large-scale pattern of interior nutrient (phosphate) and DIC in the stand-alone ocean configuration and that in the coupled ESM setup.

P15 130 : Subsequent question concerning the isotope run. What happens after the spin-up? Is there anything special to run the historical period? Do you change the atm forcing field (i really wished you used atmospheric fields extracted from the coupled historical run)? or do you simply follow the historical atmospheric CO2 records for the last ~150years ?

As stated above, the simulation with carbon isotopes does not include any historical simulation (only preindustrial spin-up) and was meant to assess the performance of the new carbon isotopes under the preindustrial climate.

In the revision, we have clarified this (i.e., in Sections 4 and 5.13).

5.3 Salinity - P17 - no ref to Fig. 4

References to Fig. 4 have been added.

5.6 Nutrients - P18 - No ref to Fig. 7 and 9 The fig 2 is understandably well discussed as it is a very instructive figure, but all following figures are almost forgotten, and sometimes not introduced at all.

We have included references to Figs. 7-9 in the revised subsection '5.6 Nutrients', as suggested. In the revised manuscript, we have also included more direct references to specific Figures, in addition to Fig. 2.

P19 13 - if you refer to NorESM1, you should point to fig 8h and 8i

The previous sentence was referring to the Pacific Ocean, so a reference to Fig. 8h is not necessary.

We have, nevertheless, added a reference to NorESM2 (Fig. 8f) in the preceding sentence.

5.7 Dissolved oxygen - P19 - No ref to Fig. 11.

We have included references to Fig. 11 in the revision.

P19 l21 - Fig. 9 is the silicic acid; AOU is Fig. 12

Corrected.

5.12 - P22 : The Analysis done here are really great, but coming from a stand alone run forced with a different forcing field, these results do not inform about the main NorESM2 MM and LM runs i am afraid.

Although these simulations were done in a stand-alone ocean configuration, we think that the implementation and carbon isotopes capability in NorESM2 is novel and have decided to include them in this manuscript. While this simulation is not directly comparable with those coupled setups done with NorESM2-LM and NorESM2-MM, their interior biogeochemical mean states share very similar features.

To support this, we have added two new figures of phosphate and DIC from the stand-alone run (in the supplemental materials, Figs. S13 and S14) and referred to them in the main manuscript.

5.12 and 5.13 - double-check the figure order, you switched both section figures.

Checked and followed your suggestion.

We have switched subsections 5.12 and 5.13 such that the carbon isotopes discussions are now presented last. We have also reminded the readers that the carbon isotopes analysis is based on ocean-only, and not coupled ESM, configuration.

Fig.1 some features are missing. For instance the riverine inputs (with the different materials they contain). Also the silicic acid is missing.

Thank you for noticing this.

We have revised Fig. 1, adding arrows to indicate riverine inputs. Silicate has been added in the nutrient box. Figure caption has been revised accordingly.

Fig. 10 : The modeled Diatoms are never Si limited ? You say biological productivity, how is that done ? as both phyto are not necessarily limited by the same nutrient. Might be good to say 2 words in the figure caption about how you calculate that.

In our model, diatoms are not explicitly modeled. We only implement one prognostic bulk phytoplankton class. Silicate concentration in the model is used to determine the opal export, as stated in subsection 2.3. The bulk phytoplankton growth rate formulation is documented in Schwinger et al. (2016).

We have revised subsection 5.6 (last paragraph) to reflect this information:

"In NorESM, the bulk phytoplankton growth rate is limited by multiple nutrients (i.e., phosphate, nitrate, and dissolved iron), in addition to temperature and light (see also Section 2.3.2 of Schwinger et al. (2016). The model does not explicitly simulate diatom and calcifier classes."

As suggested, we have also revised Fig. 10 caption to inform how the limiting nutrient is determined.

Fig. 13 : I have difficulties to realize what's going on with contours superposed to colormaps and showing different things. It is OK-ish on vertical sections, but on global horizontal maps it is difficult to appreciate the differences, and spot what we should see. it is not straight forward. could you add a colored pictures for NorESM1 so it is not superposed as contour, for fig 13, 15, 18. I would ask to do that for all fig of that kind, but that would mean almost all pictures... i really find it difficult to interpret that way. it save space, but doesn't make it simpler for the reader, i think.

We have followed your suggestion.

We have revised Figs. 13, 15, and 18 and removed the contours in these figures. Respective captions have been revised to reflect these changes. For Fig. 15, DMS was not simulated in NorESM1-ME. For CO₂ fluxes, since the seasonal cycles of NorESM1-ME in selected regions are already shown in Fig. 19, we decided not to include them again in Fig. 18 to reduce the number of figures. A map of CO₂ fluxes from NorESM1-ME is also already available in Tjiputra et al. (2013).

Fig 14. and all alike : what is the difference between the solid red and dotted red lines ? i guess the dotted one is NorESM-LM but it is not written in the caption. you probably can remove the dotted line as there is no reference to it at all.

You're correct. Those are values for NorESM2-LM. We agree.

We have removed the red dotted-lines.

Ocean biogeochemistry in the Norwegian Earth System Model version 2 (NorESM2)

Jerry F. Tjiputra¹, Jörg Schwinger¹, Mats Bentsen¹, Anne L. Morée², Shuang Gao³, Ingo Bethke², Christoph Heinze², Nadine Goris¹, Alok Gupta¹, Yanchun He⁴, Dirk Olivié⁵, Øyvind Seland⁵, and Michael Schulz⁵

¹NORCE Norwegian Research Centre and Bjerknes Centre for Climate Research, Bergen, Norway
 ²Geophysical Institute, University of Bergen and Bjerknes Centre for Climate Research, Bergen, Norway
 ³Institute of Marine Research and Bjerknes Centre for Climate Research, Bergen, Norway
 ⁴Nansen Environmental and Remote Sensing Centre, Bergen, Norway
 ⁵Norwegian Meteorological Institute, Oslo, Norway

Correspondence: Jerry Tjiputra (jerry.tjiputra@norceresearch.no)

Abstract. The ocean carbon cycle is a key player in the climate system through its role in regulating atmospheric carbon dioxide concentration as well as other processes that alter the Earth's radiative balance. In the second version of the Norwegian Earth System Model (NorESM2), the oceanic carbon cycle component has gone through numerous updates that include, amongst others, improved process representations, increased interactions with the atmosphere, and additional new tracers. Oceanic

- 5 dimethyl sulfide (DMS) is now prognostically simulated and its fluxes are directly coupled with the atmospheric component, allowing for leading to a direct feedback to the climate. Atmospheric nitrogen deposition and additional external riverine inputs of other biogeochemical tracers through riverine are recently have recently been included in the model. The implementation of new tracers such as 'preformed' and 'natural' tracers enables a separation of physical from biogeochemical drivers as well as of internal from external forcings and hence a better diagnostic of the simulated biogeochemical variability. Carbon isotope
- 10 tracers have been implemented and will be relevant for studying long-term past climate changes. Here, we describe these new model implementations and present the evaluation of the model's performance in simulating the observed climatological states of water column biogeochemistry as well as in simulating the transient evolution over the historical period. Compared to its predecessor NorESM1, the new model's performance has improved considerably in many aspects. In the interior, the observed spatial patterns of nutrients, oxygen, and carbon chemistry are better reproduced, reducing the overall model biases. A new set
- 15 of ecosystem parameters and improved mixed layer dynamics improves the representation of upper ocean processes (biological production and air-sea CO₂ fluxes) at seasonal time scale. Transient warming and air-sea CO₂ fluxes over the historical period are also in good agreement with observation-based estimates. NorESM2 participates in the Coupled Model Intercomparison Project phase 6 (CMIP6) through DECK (Diagnostic, Evaluation and Characterization of Klima) and several endorsed MIP-simulations.

1 Introduction

Up to the early 1990s, climate models consisted only of physical atmospheric general circulation models (AGCMs) with prescribed ocean surface state variables or simplified ocean modules (swamp ocean, slab ocean). As the ocean is a huge reservoir and absorber of heat and of the greenhouse gas CO_2 (carbon dioxide), an expansion of climate models to include ocean

- 5 components and a representation of the carbon cycle became a necessity. Coupled atmosphere-ocean-land models (now referred to as Earth System Models ESMs) were eventually developed to account for further reservoirs and feedbacks including biogeochemical cycles (Bretherton, 1985; Cubasch et al., 2013; Heinze et al., 2019). The inclusion of the ocean carbon cycle is a first order requirement, because on timescales beyond a few decades, the ocean becomes the sole major sink for atmospheric excess CO₂ from anthropogenic emissions (Sarmiento and Gruber , 2006). This allows us ESM simulations can therefore be
- 10 used to estimate historical carbon budgets as well as future carbon emission pathways and future partitioning of carbon fluxes into Earth system reservoirs under specified scenarios.

Today, modern ESMs are state-of-the-art tools that allow us utilized to study the complex interactions and feedbacks between various components in the Earth system in a comprehensive way (Flato et al., 2013). They are applied regularly to simulate climate evolutions across time scales and to study transient climate change, its drivers, and its impact on the environment (e.g.,

15 Bopp et al., 2013; Gehlen et al., 2014). By prescribing plausible scenarios of future emissions and land use, these models provide projections for possible future climate states. ESMs are typically upgraded every few years with new and improved process representations as well as adaptations to technical advancements (e.g., new hardware system, higher resolutions resolution, etc.).

The ocean and its biogeochemistry play a crucial role in controlling the rate of anthropogenic climate change through a substantial negative feedback (Friedlingstein et al., 2006; Arora et al., 2013). However, the absorption rates of heat and carbon by the ocean are non-linear in space and time, due to the complex interplay of interactions with the internal climate variability (e.g., Tjiputra et al., 2012; Schwinger et al., 2014). Once passed the air-sea interface, the dissolved carbon dioxide reacts with seawater and is chemically transformed into different carbonate species. The ocean circulation and biological processes sequester parts of these into the deep ocean, where they stay isolated from the atmosphere, for decades to millennia (Sarmiento

- and Gruber , 2006). In addition to carbon dioxide, the ocean biogeochemistry also influences the Earth's climate through other seawater chemical constituents, such as through dimethyl sulfide emissions. This could result in a number of feedback mechanisms within the Earth system that can either amplify or dampen the rate of climate change (Rhein et al., 2013). ESM projections are therefore critical for narrowing uncertainties in the future global carbon budget, and consequently its climate feedback.
- 30 ESMs have been increasingly used to estimate the evolution of marine ecosystem stressors, such as ocean acidification and deoxygenation, under anthropogenic climate change (e.g., Gehlen et al., 2014; Tjiputra et al., 2018). ESM applications to support climate policy-making, such as quantifying future compatible anthropogenic carbon emissions under a set of predefined scenarios as well as addressing emerging questions associated with the United Nations sustainable development goals

have become more regular. With its growing relevance, a more realistic representation of the observed contemporary ocean biogeochemistry in Earth system models is urgently necessary to increase the fidelity of future projections.

In this manuscript, we present improvements in the ocean biogeochemical component of the Norwegian Earth System Model (NorESM). This component is based on the Hamburg Oceanic Carbon Cycle model (HAMOCC5.1), which was

- 5 originally developed by Maier-Reimer (1993) and has gone through several development iterations (e.g., Six and Maier-Reimer, 1996; Maier-Reimer et al., 2005). At the Bjerknes Centre for Climate Research in Norway, the model has been branched off, further developed (Assmann et al., 2010; Tjiputra et al., 2010, 2013; Schwinger et al., 2016), and coupled to the Bergen Layered (isopycnic coordinate) Ocean Model (BLOM-iHAMOCC). Within the CMIP5 experiments, we have identified several shortcomings in the previous model version of NorESM (NorESM1-ME, hereafter referred to as NorESM1).
- 10 Here, we discuss several developments that have been prioritized for the ocean biogeochemistry and are included, relative to the NorESM1 version (described in Tjiputra et al., 2013), in the current model version (NorESM2), which will contribute (NorESM2, Seland et al., 2020), which contributes to CMIP6.

NorESM1 does not include interactively coupled DMS emissions from the ocean, which is the largest natural source of atmospheric sulfur, and contributes to aerosol and cloud formation. Instead, the atmospheric chemistry module of NorESM1

- 15 prescribed mean fixed climatology oceanic fluxes of DMS, and there is no feedback associated with climate-induced change in marine biological production. However, a model study has suggested that the inclusion of interactive DMS in future climate projection could potentially result in spatially-varying changes in warming rate and a non-negligible feedback on the climate system (Schwinger et al., 2017). This has prompted us to include a fully-interactive DMS cycle (production, emissions, and radiative impact) into NorESM2.
- In the past years, the concept of emergent constraints was employed to identify key biogeochemical processes that lead to uncertanties in ESMs projections (e.g., Wenzel et al., 2014; Kwiatkowski et al., 2017). For instance, biases in the seasonal cycles of biological production and surface ocean pCO_2 have been identified as one of the factors contributing to the uncertainty in the projected carbon sinks and storage in the ESMs participating in CMIP5 (Kessler and Tjiputra, 2016; Goris et al., 2018). These studies motivate further improvement in the representation of biological processes, particularly focusing on the high
- 25 latitude regions, such as the Southern Ocean. The biological improvements in NorESM2 were mainly achieved through tuning of the poorly constrained parameters in the ecosystem module.

In NorESM1, as well as other ESMs participating in CMIP5, there is a large uncertainty in the simulated interior oxygen concentration when compared to observations. This is partly attributed to the short model spin-up and bias in the interior remineralization processes (Cabré et al., 2015; Séférian et al., 2016). To alleviate this, we have tested several parameterizations

30 of particulate organic carbon (POC) sinking schemes in our model (Schwinger et al., 2016). Currently, there are three different sinking schemes that can be selected in NorESM2. Based on our earlier assessments, we have used the scheme with sinking velocity that increases linearly with depth as the default parameterization in NorESM2.

Increasing atmospheric CO_2 and associated climate change will alter both the natural and anthropogenic components of ocean inorganic carbon chemistry. Disentangling future dynamical responses of these components separately is therefore needed to identify which processes (e.g., biological and physical, among others) regulate the simulated changes in ocean

35

carbon storage today and in future. In NorESM1, only a single dissolved inorganic carbon (i.e., total DIC) tracer was implemented, and there is no suitably accurate method to separate between the driving mechanisms of its changes. In NorESM2, we have introduced a new 'natural DIC' tracer, which simulates changes in the natural component of DIC (i.e., it only interacts with fixed preindustrial CO_2 concentration during the air-sea gas exchange). Other key diagnostic tracers such as preformed

5 phosphate, oxygen, and alkalinity have also been implemented, all of which allow us to better elucidate mechanisms driving the simulated ocean carbon chemistry changes (Bernardello et al., 2014). These tracers also allow to determine can also be used to infer interior ocean circulation-induced changes in biogeochemical tracers.

Nutrients are important constraints of ocean primary productivity and hence atmospheric carbon sinks. In NorESM1, the nutrient cycle was assumed to be an approximately closed system with a long-term loss due to sedimentation. The only external

- 10 source is through atmospheric nitrogen fixation and aerial dust deposition, which is converted into dissolved iron. Implementation of riverine input of nutrients in HAMOCC5.1 was initiated by Bernard et al. (2011) and a study by Gao et al. (in prep.) indicates that riverine nutrient inputs could improve the regional representation of marine biogeochemistry, with implications on consequences for future projections, i.e., they increase coastal carbon sinks and alleviate nutrient limitations due to warming-induced stratification. In NorESM2, we have accounted for riverine inputs of nutrients and other biogeochemical constituents.
- 15 Other processes representing the external nitrogen sources such as atmospheric nitrogen deposition and nitrogen fixation have also been implemented and updated, respectively.

Since ESMs are used to project future climate change, questions have arisen whether or not these models are able to simulate past climate change. To this end, NorESM1 has been applied to study the sensitivity of climate states to different boundary conditions in the past, going back from the last glacial to as far back as the mid-Pliocene (Zhang et al., 2012; Guo et al., 2019).

- 20 The model was also used to study the sensitivity of the ocean carbon cycle to different background climates, which could provide insights on the oceaninsight into the ocean's role in regulating past atmospheric CO₂ variability (Kessler et al., 2018; Luo et al., 2018). There has also been a growing interest to determine if such complex models are able to reproduce climate states directly inferred from proxy-records. These paleo proxies, e.g., carbon isotopes (δ^{13} C and Δ^{14} C), typically measured from foraminiferal samples collected from seafloor sediment, are natural archives that store watermass properties and are used
- 25 to infer large scale ocean circulation patterns as well as ventilation time scales during past climate changes (Peterson et al., 2014). In NorESM2, we have now implemented ¹³C and ¹⁴C tracers, which will be used as an additional ocean circulation and biogeochemistry constraint when simulating past climate variability.

Beyond the above mentioned biogeochemical processes, other improvements in tracer initialization, iron chemistry, air-sea gas exchange parameterization, code readability, and documentations of the codes have been made. Moreover, there have been

30 several updates within the physical ocean component of NorESM2 and a selection of these updates are discussed in this paper as they have direct impacts on the ocean biogeochemistry. In addition to describing the model improvements, we also evaluate the performance of the ocean biogeochemical component of NorESM2. Whenever possible, we compare and describe the performance of NorESM2 model with the first generation model, NorESM1.

The manuscript is organized as follows: advancements in the physics and <u>more detailed descriptions of all</u> improvements in 35 biogeochemical processes are described in Section 2. Implementations of new tracers are presented in Section 3. The different model simulations performed and used in the paper are described in <u>section_Section_4</u>. In Section 5, we present results from our coupled ESM simulations as well as that of the ocean-only (preindustrial) simulation to illustrate the performance of the simulated carbon isotope tracers. Summary and discussion are presented in <u>section_Section 6</u>.

2 Model changes and improvements

5 2.1 General configuration changes

In NorESM2, the ocean-ice model adopts a tripolar grid instead of the bipolar grid of NorESM1. The tripolar grid was chosen because it is more isotropic at northern high latitudes and therefore allows for a more efficient is more efficient with respect to time integration (longer time step) (Guo et al., 2019). Compared to NorESM1, NorESM2 has enabled a higher frequency ocean coupling to resolve the diurnal cycle in the flux- and state-exchange between ocean and atmosphere (i.e., hourly instead of daily

10 atmosphere-ocean coupling). We also now applied a full leap-frog time-stepping in both physical and ocean biogeochemical models to improve conservation of biogeochemical tracers (Schwinger et al., 2016).

There are several configuration options for NorESM2. For CMIP6, two configurations will mostly be used, differing only in the horizontal resolution of the atmospheric and land models with otherwise identical model components. They are named NorESM2-LM and NorESM2-MM and have atmosphere-land resolution of approximately 2° and 1°, respectively. In addition,

- 15 both versions have a different parameter tuning in the atmospheric component, necessary to achieve a top of the atmosphere radiative balance under preindustrial conditions. Both NorESM2-LM and NorESM2-MM share the same ocean physical ocean model (BLOM) and the same biogeochemical component (iHAMOCC) with a horizontal resolution of approximately 1°. The ocean model adopts 53 isopycnal layers with two non-isopycnic surface layers, which represent (1) the top 10 m depth and (2) the rest of the surface mixed layer. A more detailed description of the ocean physical component is provided in Bentsen et
- 20 <u>al. (in prep.).</u> The ocean biogeochemistry performs similarly in both model versions. Unless explicitly specified, our analysis refers to the results from NorESM2-MM (hereafter referred to as NorESM2).

2.2 Physical parameterizations

The simulated Atlantic Meridional Overturning Circulation (AMOC) strength in NorESM1 is on the strong side (30.8 Sv at 26.5°N; 1 Sv = 10⁶ m³ s⁻¹) compared to observation-based estimates and other CMIP5 models (Bentsen et al., 2013). The vigorous AMOC leads to a too warm and saline Atlantic Ocean at depth. Lack of upper ocean stratification at high latitudes is thought to contribute to this AMOC bias. Reformulating the eddy-induced transport (Gent and McWilliams, 1990) in order to make it more efficient in the upper ocean non-isopycnic regime of the model and adjustments to the parameterized restratification by mesoscale eddies (Fox-Kemper et al., 2008) lead to improved high latitude stratification and a more realistic MLD, particularly at high latitudes.

NorESM1 has a cold bias in the depth range 200–1000 m between 50°S and 50°N (Bentsen et al., 2013) that could be related to insufficient upper ocean vertical mixing. Further, the new hourly ocean atmosphere-ocean coupling and associated tuning of MLD parameterizations lead to a cold bias in the Pacific equatorial thermocline that also indicates too little vertical mixing.

- To improve this, two different approaches have been used. First, we allow surface turbulent kinetic energy (TKE) originating from wind stirring to penetrate below the mixed layer and be added to the TKE reservoir of the $k - \varepsilon$ turbulence model handling the shear instability mixing of the isopycnic interior. Secondly, with hourly ocean atmosphere-ocean coupling, high frequency winds are available to parameterize wind work on near-inertial motions. Inspired by the approach of Jochum et al. (2013), a fraction of this energy source is added to the mixed layer TKE reservoir, modifying the MLD estimation, while a fraction of the remaining energy is used to drive upper ocean mixing through assumed excitation and breaking of internal waves. Combining
- 10 these approaches, we were able to reduce the above-mentioned upper ocean biases significantly. The mid-depth warm bias seen in NorESM1 is also reduced. More details on the physical improvements are described in Bentsen et al. (in prep.). As a result of these improvements, NorESM2 simulates a more reasonable AMOC strength of approximately 21 Sv, which is closer and within the range of the observational estimates of 17.9 ± 3.3 Sv (Srokosz et al., 2012).

2.3 Dimethyl sulfide

15 NorESM2 prognostically simulates DMS concentration according to the formulation of Six and Maier-Reimer (2006) Kloster et al. (2006) as follows:

$$\frac{\partial DMS}{\partial t} = \underline{ADV(DMS)S_{adv}} + S_{\underline{Phyprod}} - S_{bac}\underline{DMS} - S_{photo}\underline{DMS} - S_{gas} - S_{uv} - S_{flux},\tag{1}$$

where the first term on the right hand side denotes the physical transport, the second term denotes production due to phytoplankton degradation, DMS production, and the third, fourth, and fifth terms denote sinks due to bacterial activity, photolysis, and flux to the atmosphere. The production term depends on the is computed as a function of temperature, pH, and simulated detritus

20 to the atmosphere. The production term depends on the is computed as a function of temperature, pH, and simulated detritu export production (opal and CaCO₃, implicitly computed as a function of silicate concentration) productions: :

$$S_{\underline{Phyprod}} = \left(\gamma_{\underline{d}} \underline{Phy_{diao}} \cdot \underline{export_{opal}} + \gamma_{\underline{c}} \underline{Phy_{cal}} \cdot \underline{export_{CaCO_3}}\right) \cdot \left(\left(1 + \frac{1}{(T+10)^2}\right)\right) \cdot \left(1 + (pH - pH_{pi}) \cdot \alpha\right).$$
(2)

 $\gamma_d Phy_{dia}$ and $\gamma_c Phy_{cal}$ represent production rates of biogenic opal and within the mixed layer depth. The last term in Eq. 2 describes a pH-dependence of DMS production following (Six et al., 2013), which is parameterized based on the deviation of actual pH from the climatological monthly mean surface pH (pH_{pi}) calculated from the preindustrial control. In NorESM2, the atmospheric model receives the DMS emissions (S_{gas}) that are prognostically computed by the ocean biogeochemistry model. Due to uncertainties in the pH-control on DMS production, we have omitted the pH-dependency term by setting α to zero in the presented simulations. Parameters $\gamma_d \gamma_0$ and γ_c represent scaling factors for sulphur (S:Si and S:C) and are set to 0.025 and 0.125, respectively, such that DMS production is more sensitive to ealeifier-CaCO₃ production in the model. This assumption

30 is due to the fact that haptophyte-class phytoplankton generally has a higher DMSP to cell carbon ratio than diatoms (Keller et al., 1989). Since the diatom and calcifier portions of phytoplankton opal and CaCO₃ portions of export production are implicitly computed as a function of silicate concentration, e.g., diatom (calcifier) are opal (CaCO₃) production is high (low) when surface silicate is abundant and vice versa, the DMS production is also sensitive to the long-term trend in the silicate concentration.

The DMS sink due to bacteria consumption is estimated as a function of temperature [°C], DMS concentration, consumption rate ($\gamma_{bac}=0.0864 \text{ day}^{-1} \circ \text{C}^{-1}$), and a half-saturation constant ($k_{cb}=10 \text{ nmol } \text{L}^{-1}$), as follows.

5
$$S_{bac} = \gamma_{bac} \cdot (T+3.0) \cdot DMS \cdot \left(\frac{DMS}{k_{cb} + DMS}\right).$$
 (3)

The DMS photolysis term is defined as a function of photolysis rate ($\gamma_{uv}=0.0011 m^2 (W day)^{-1}$), incoming shortwave radiation attenuated by water and phytoplankton, I_z , and DMS concentration:

$$S_{uv} = \gamma_{uv} \cdot I_z \cdot DMS. \tag{4}$$

Lastly, in NorESM2, the atmospheric model receives the DMS emissions (S_{gas}) that are prognostically computed by the ocean 10 biogeochemistry model (see Section 2.8).

2.4 Riverine input

The influx of carbon and nutrients from over 6000 rivers to the coastal oceans has been implemented based on previous work of Bernard et al. (2011) with modifications. Bernard et al. (2011) have implemented the riverine fluxes based on the levels in year 1995, provided by an early version of the Global-NEWS model (Seitzinger et al., 2005). In addition to the riverine DIC,

- 15 dissolved organic carbon (DOC), inorganic nitrogen and phosphorus, particulate organic carbon (POC), particulate nitrogen and phosphorus, and dissolved silicate, we have also included riverine alkalinity (ALK) and iron (Fe). Except for DIC, ALK and Fe, all data are provided by the more recent Global-NEWS2 model (Mayorga et al., 2010), which is a hybrid of empirical, statistical and mechanistic model components that simulate steady-state annual riverine fluxes of carbon and nutrients. The DIC and ALK fluxes are taken from the work by Hartmann (2009). Riverine Fe-flux is calculated as a proportion of global
- 20 gross dissolved Fe input of 1.45 Tg yr⁻¹ (Chester, 1990), weighted by water runoff of each river. Only 1% of the riverine dissolved Fe is added to the oceanic dissolved Fe, since approximately 80 to 99% of the fluvial gross dissolved Fe is removed during estuarine mixing (e.g., Sholkovitz et al., 1981; Shiller and Boyle, 1991; Bruland and Lohan, 2014). Instead of releasing the riverine fluxes to the nearest ocean grid cell (Bernard et al., 2011), we have interpolated all fluxes at river mouths in the same way as the freshwater runoff (distributed as a function of river mouth distance with e-folding length scale of 1000 km and
- 25 cutoff of 300 km) to the ocean grid. The fluxes are specified to be constant over time at contemporary levels (year 2000). In the model, carbon and nutrients are taken up by phytoplankton according to a our model, the organic form of dissolved carbon (DOC) is connected to nutrients through the Redfield ratio (C:N:P=122:16:1). Therefore, the least abundant riverine organic matter (both dissolved and particulate species) is used to calculate the riverine fluxes of DOC and POC (detritus, particulate organic matter), and therefore other forms of dissolved organic matters (e.g., DON, and DOP) are not explicitly simulated.
- 30 Since Global-NEWS provides estimates of dissolved organic matter in carbon, nitrogen, and phosphate forms $(DOC_{riv}, DON_{riv}, and DOP_{riv}, respectively)$ separately, only the minimum of the three riverine dissolved organic constituents is added to the DOC

term in the model (i.e., $DOC = DOC + min(DOC_{riv}, r_{C;N} \cdot DON_{riv}, r_{C;P} \cdot DOP_{riv})$). Any excess riverine organic matter following the Redfield ratio conversion is merged into its inorganic form or remaining organic matter of the three constituents is then added to the corresponding inorganic pools (DIC, NO₃ or PO₄). The same concept also applies to riverine inputs of particulate organic carbon (POC) (see also Bernard et al., 2011).

5 2.5 Atmospheric nitrogen deposition

We apply anthropogenic nitrogen deposition fluxes that potentially affect the simulated ocean carbon uptake (through an increase of the available nitrate for biological production) to the ocean biogeochemical model. The monthly input fields, spanning the years 1850-2014, are simulated by chemistry transport models and provided by CMIP6 (Jones et al., 2016). Conservative remapping is used to interpolate the input data from a regular $2.5^{\circ}x1.9^{\circ}$ latitude-longitude grid to the tripolar

10 ocean model grid of NorESM2. Four species of wet or dry and oxidized or reduced nitrogen deposition rates are included in the input fields. All of them are assumed to be bio-available and added to the nitrate pool in the top-most ocean layer.

2.6 Particle exports export

In NorESM1, the export of particulate organic matter is parameterized with a constant sinking speed and a constant remineralization rate at depth, when sufficient oxygen is available. This simplistic formulation has been shown to have difficulty in

accurately representing the observed particle fluxes at depths (Kriest and Oschlies, 2008) and leads to biases in the interior distributions of biogeochemical tracers (e.g., as seen from remineralized phosphate concentrations). In NorESM2, we have developed two additional particle flux parameterizations that can be selected. They are referred to as WLIN and AGG schemes. The WLIN scheme is similar to the standard scheme but the sinking speed is a linear function of depth: w_{POC} = min(w_{min}+az, w_{max}). Here, we use w_{min} = 1 md⁻¹, w_{max} = 60 md⁻¹, and a = 60/2400 md⁻¹m⁻¹. When w_{min} and w_{max} are set to zero and infinity, respectively, this scheme is equivalent to the widely used Martin-curve formulation (Martin et al., 1987).

The AGG scheme implements a prognostic sinking speed, calculated according to a size distribution of sinking particles. The total concentration of particulate matter, formulated as a function of phytoplankton and detritus, forms sinking aggregates with explicitly computed size distribution that follows a power-law formula (Kriest and Evans, 1999; Kriest, 2002):

$$n(d) = Ad^{-\beta}, \quad l < d < \infty.$$
⁽⁵⁾

- 25 Here, *d* represents the particle diameter and *A* and β represent parameters of the size distribution. The minimum particle diameter *l* is set to 0.002 cm. The sinking speed of aggregates is computed according to the diameter of each aggregate. More details of the implementation of the AGG formulation are described in Schwinger et al. (2016). Based on the performance of the different schemes, we have set WLIN to be the default scheme and it is used in all experiments presented here. More details on the differences in the simulated interior biogeochemistry in using the different particle export schemes are available
- 30 in Schwinger et al. (2016).

2.7 Nitrogen fixation

Nitrogen fixation by cyanobacteria is computed implicitly and directly converted into nitrate concentration. In NorESM1, nitrogen fixation is implemented such that it can occur anytime and anywhere in the surface ocean as long as the nitrate concentration is lower than phosphate (multiplied by the stoichiometric nitrate to phosphate ratio $R_{N:P}$) following Maier-Reimer

5 et al. (2005). However, Breitbarth et al. (2007) provided some evidence that the large scale distribution of *Trichodesmium*, a type of cyanobacteria, is well constrained by seawater temperature between 20 and 30°C. Based on this, a temperature-dependent function f(T) according to Kriest and Oschlies (2015) has been added to the nitrogen fixation formulation in NorESM2. The rate of change of nitrate owing to nitrogen fixation is formulated as follows:

$$\frac{\partial NO_3}{\partial t} = f(T) \cdot \mu_{\underline{\cdot}} max(0, R_{N:P} \cdot PO_4 - NO_3), \tag{6}$$

10 with

25

$$f(T) = max\left(\left(0, \frac{-0.0042 \cdot T^2 + 0.2253 \cdot T - 2.7819}{0.2395}\right)\right),\tag{7}$$

where μ is the maximum N₂ fixation rate (0.005 day⁻¹). In addition, for each mole of N fixed to nitrate, 1.25 and 1 moles of dissolved oxygen and <u>1 mole of</u> alkalinity are consumed and lost in the surface layer. The corrected formula essentially limits the occurrence of N-fixation to warm low latitudes.

15 2.8 Air-sea gas exchange

In NorESM2, the air-sea gas exchange of CO₂, O₂, N₂, N₂O, DMS, CFC-11, CFC-12, and SF₆ is computed prognostically according to the updated formulation of Wanninkhof (2014). Here, the air-sea flux *F* of gas *X* is computed as a function of surface wind speed *U*, Schmidt number *Sc*, gas solubility K_o , and partial pressure difference of gas *X*, based on a bulk formulation:

20
$$F_X = 0.251 \cdot U^2 \cdot (Sc/660)^{-0.5} \cdot K_o \cdot (pX_{sw} - pX_{air}).$$
 (8)

The net fluxes are computed for the top layer of the ocean model (i.e., 10 m depth). The updated formulation includes a refitted temperature-dependent Schmidt number that can be applied for a temperature range of -2 to 40°C. The solubility of all gases is computed as a function of surface temperature and salinity following Weiss (1970) for O₂ and N₂, Weiss and Price (1980) for CO₂ and N₂O, Warner and Weiss (1985) for CFC-11 and CFC-12, and Bullister et al. (2002) for SF₆. The flux of DMS is only unidirectional (outgassing; $pDMS_{air} = 0$).

In the historical simulation, we use the annual atmospheric concentrations of CFC-11, CFC-12, and SF₆, divided into northern and southern hemispheric values, according to Bullister (2014). For the atmospheric CO_2 concentration, monthly globalmean values from the CMIP6 dataset are used. In the preindustrial simulation, a constant CO_2 concentration of 284.32 ppm is used.

2.9 Dissolved iron parameterization

Adjustments in the iron parameterization have been made to ensure that NorESM2 simulates the observed iron-limited primary production in the Southern Ocean, Equatorial and North Pacific. In the model, the consumption and release of dissolved iron associated with biological activities are determined using a fixed stoichiometry ratio ($R_{Fe:P} = 6.1e-4$ mol Fe mol P⁻¹, previously

5 3.7e-4 mol Fe mol P^{-1}). In the surface layer, a constant fraction (3.13e-6 kmol Fe kg⁻¹ day⁻¹, previously 6.27e-6 kmol Fe kg⁻¹ day⁻¹) of aerial dust deposition (Mahowald et al., 2005) is instantaneously converted into bio-available iron. The updated global input of bio-available iron from the atmosphere is 2.8 Gmol Fe yr⁻¹, which is within the range of values used by other models (Tagliabue et al., 2016). Finally, throughout the water column, complexation of iron by organic substances (ligands) is assumed:

10
$$\frac{\partial Fe_{cmpl}}{\partial t} = -\lambda_{Fe} \cdot max(0, Fe - Fe_{\underline{o}o}), \qquad (9)$$

where the strength if of this complexation λ_{Fe} is set to 0.05/365 (previously 0.005/365) day⁻¹ and the threshold Fe_{O} - Fe_{O} has been reduced from 0.6 to 0.5 nmol Fe m⁻³. The latter is motivated by the fact that the observed iron concentration in the deep Southern Ocean is lower than 0.6 nmol Fe m⁻³ (Boyd and Ellwood, 2010). The higher complexation rate allows for leads to a faster relaxation toward this value.

15 2.10 Ecosystem parameterization updates

The underlying marine ecosystem parameterization in NorESM2 remains the same as in NorESM1, but many of the parameters have been adjusted to reduce biases that are present in NorESM1. An updated schematic diagram of the marine ecosystem module in NorESM2 is presented in Fig. 1. The main deficiencies of the spatial annual mean primary productivity pattern in NorESM1 are a too strong primary production (PP) in the Southern Ocean, the Eastern Tropical Pacific, and to a lesser degree
the North Atlantic, contrasted by a very low PP in the subtropical gyres (most pronounced in the Pacific). The high productivity in the high latitudes is accompanied by an exaggerated annual cycle of PP showing a too strong spring bloom and a too fast decline of PP afterwards. Tuning of the ecosystem parameterization during the development of NorESM2 was focused on improving these regional shortcomings. The changes in ecosystem parameters listed in Table 1 mainly serve two purposes: (i) to increase top-down limitation by zooplankton grazing during phytoplankton peak bloom (but not before and after), and (ii)
to increase the fraction of nutrients that is routed through dissolved organic matter (DOM).

The first point is achieved through a reduction in zooplankton mortality as well as increases in the maximum grazing rate, assimilation efficiency, and half-saturation constant for zooplankton grazing. The latter point is mainly achieved by reducing the production of detritus by zooplankton (fecal pellets, $1 - \varepsilon_{zoo}$ of grazing) and increasing exudation and excretion rates. This allows, in combination with a decrease of the DOM remineralization constant, more nutrients to be laterally transported out of racions where nutrient transport i.e., the Transport Pacific

30 regions where nutrient trapping occurs, i.e., the Tropical Pacific.

Additionally, $CaCO_3$ and silicate to phosphate ratios were tuned to remove the high alkalinity bias of NorESM1 (see Schwinger et al., 2016, for details). Note that some adjustments of parameters given in Table 1 were also necessary because

of the new sinking scheme (increasing sinking speed with depth, see Section 2.6), and the different physical circulation fields between NorESM1 and NorESM2. An updated schematic diagram of the marine ecosystem module in NorESM2 is presented in Fig. 1.

In NorESM1, biological productivity was computed only for the top 100 m depth, which presumably represents the averaged euphotic layer. In an isopycnic model, there is no static interface separating this depth level. In cases where the bulk MLD extends below 100 m, we virtually split this at 100 m and the biological production is simulated only down to this interface. In NorESM2, we have omitted this virtual layer splitting and the biological production is allowed to occur below 100 m as long as it is within the bulk mixed layer depth and has sufficient light (attenuated with depth and chlorophyll concentration) for phytoplankton growth.

10 3 New tracers

3.1 Preformed tracers

Four new preformed tracers have been implemented in NorESM2, namely preformed dissolved oxygen (O_2^{pre}) , phosphate (PO_4^{pre}) , alkalinity (ALK^{pre}) , and dissolved inorganic carbon (DIC^{pre}) . They are initialized to zero during at the beginning of the spin-up. During the model integration, in the bulk mixed layer of the model (i.e., the top 2 levels), the preformed tracers are set to the respective total, non-preformed values at each time step. Below the mixed layer, the preformed tracers are advected

as passive tracers by the physical processes and hence are a measure of transport-induced (e.g., circulation, ventilation, etc.) changes. The preformed tracers are included for diagnostic purposes, such as identifying the sources of model-data misfits (Duteil et al., 2012). The preformed oxygen can be used to quantify the total and apparent oxygen utilizations (TOU and AOU) as well as O_2 disequilibrium (ΔO_2) in the interior ocean (Eqs. 10-11; Ito et al., 2004):

20
$$TOU = O_2^{pre} - O_2,$$
 (10)

and

25

30

15

$$\Delta O_2 = TOU - AOU = O_2^{pre} - O_2^{sat}. \tag{11}$$

Here, the saturated oxygen (O_2^{sat}) is determined as a function of temperature and salinity (Garcia and Gordon, 1992). Due to its closed coupling with ocean circulation, the preformed oxygen can be used to separate biologically- from physically-induced biases in the simulated interior biogeochemical tracers.

Both preformed phosphate and alkalinity can be Following Bernardello et al. (2014), PO_4^{pre} is used to quantify the organic (soft tissue, DIC^{soft}) and, whereas both PO_4^{pre} and ALK^{pre} are used to quantify the inorganic (carbonate, DIC^{carb}), biologicallymediated carbon pump (Eqs. 12-14; Bernardello et al., 2014) (Eqs. 12-14). We use a stoichiometry ratio of P: N: C = 1: 16: 122 in the model, such that DIC^{soft} represents remineralized particulate and dissolved organic materials and DIC^{carb} dissolution of particulate calcium carbonate in the water column:

$$DIC^{tot} = DIC^{pre} + DIC^{soft} + DIC^{carb},$$
(12)

$$DIC^{soft} = r_{C:P}(PO_4^{tot} - PO_4^{pre}), and$$

and

$$DIC^{carb} = 0.5 \underline{ALK^{tot} - ALK^{pre} + (r_{N:P} + 1) \cdot (PO_4^{tot} - PO_4^{pre}) \cdot [ALK^{tot} - ALK^{pre} + (r_{N:P} + 1) \cdot (PO_4^{tot} - PO_4^{pre})].$$
(14)

(13)

5 Equation 14 is slightly different from that of Bernardello et al. (2014), i.e., it uses the term $r_{N:P} + 1$ instead of $r_{N:P}$ because, in our updated code, we do not neglect the contribution of phosphate to alkalinity changes during biological production and remineralization. For instance, both nitrate and phosphate produced during organic remineralization increase the concentration of proton and therefore reduce the alkalinity (see also Section 3.1.3 of Paulmier et al., 2009).

The difference between total *DIC* and biologically altered DIC ($DIC^{soft}+DIC^{carb}$, or the biological carbon pump) therefore represents the physical carbon pump (DIC^{pre}), which comprises saturated and disequilibrium components (Eq. 15). Here, a tracer of saturated DIC (DIC^{sat}) has been implemented. It is computed at the surface layer as a function of atmospheric pCO_2 , total alkalinity, SSS, and SST. Below the first layer, DIC^{sat} is treated as a passive tracer advected by the circulation. Since the equilibration time-scale of the air-sea CO₂ fluxes is typically longer than the surface watermasses' residence time (e.g., in the deep water formation regions), a disequilibrium term (DIC^{diss}) is computed by subtracting the saturated from the preformed

15 DIC components. The DIC disequilibrium component is used to diagnose the importance of ventilation variability on the physical solubility pump, and to the overall DIC storage (Eggleston and Galbraith, 2018):

$$DIC^{pre} = DIC^{sat} + DIC^{diss}.$$
(15)

3.2 Natural inorganic carbon tracers

In order to comply with the Ocean Model Intercomparison Project (OMIP) of CMIP6 (Orr et al., 2017), we have implemented a natural tracer of DIC (DIC^{nat}), which is formulated in the same manner as DIC^{tot} , except that the air-sea gas exchange is computed with a fixed preindustrial atmospheric CO₂ concentration of 284.32 ppm. In a transient simulation with increasing atmospheric CO₂ concentrations, the difference between DIC^{tot} and DIC^{nat} represents anthropogenic DIC (DIC^{ant}):

$$DIC^{ant} = DIC^{tot} - DIC^{nat}.$$
(16)

The inclusion of a DIC^{nat} tracer also requires the implementation of respective 'natural' components for both the alkalinity and the particulate inorganic carbon (CaCO₃) tracers. This is because the anthropogenic CO₂ uptake alters the carbonate system and therefore the dissolution of CaCO₃. We note that anthropogenic carbon does not influence the biological production (e.g., nutrient concentrations and phytoplankton growth rate) in our model. Similarly, a 'natural' air-sea CO₂ flux has been added to the model output. These 'natural' tracers are only activated in simulations with atmospheric CO₂ that departs from the preindustrial value. Here, DIC^{nat} , $TALK^{nat}$, and $CaCO_3^{nat}$ are initialized in the same manner as DIC, TALK, and $CaCO_3$,

30 respectively. In transient historical and future scenario simulations, these tracers provide insights of insight into the natural carbon evolution under anthropogenic climate change.

In addition, DIC^{nat} also provides a more precise estimate of DIC^{ant} entering the ocean since the preindustrial period than simply taking the difference between DIC^{tot} at a specific time and its preindustrial mean value. Alternatively, these 'natural' tracers could be computed with a parallel transient simulation but with fixed preindustrial atmospheric CO₂ as the ocean carbon cycle boundary condition. The natural tracers hence allow for inclusion of natural tracers provides substantial saving

5 of computational time, especially for high-resolution simulations. We note that the DIC^{nat} tracer has not been implemented in the sediment module. Hence, in very long time-scale transient simulations where the sediment changes become substantial, the DIC^{nat} tracer may include some uncertainties.

3.3 Carbon isotopes

The ${}^{13}C$, and ${}^{14}C$ carbon isotope tracers have now been implemented in NorESM2. Naturally occurring stable carbon isotopes 10 (${}^{12}C$, ${}^{13}C$, and ${}^{14}C$) and their relative abundances provide valuable information about both past and present climate. Changes in the ${}^{13}C/{}^{12}C$ ratio are used, e.g., to (i) study glacial-interglacial atmospheric CO₂ changes from ice core air samples (Broecker and McGee, 2013; Bauska et al., 2016), to (ii) reconstruct bottom water oxygen concentrations (Hoogakker et al., 2015), (iii) reconstruct paleo deep water circulation (Toggweiler, 1999; Curry and Oppo, 2005; Crucifix, 2005), (iv) investigate oceanic anthropogenic CO₂ uptake involving the Suess effect (Gruber and Keeling, 2001; Quay et al., 2003; Holden et al., 2013), and

- 15 (v) evaluate model sensitivity and performance (Braconnot et al., 2012; Schmittner et al., 2013). The ${}^{14}C$: ${}^{12}C$ ratio is used as a circulation and age tracer (e.g., Skinner et al., 2017). Globally, the ${}^{12}C$: ${}^{13}C$: ${}^{14}C$ isotope ratio is about 99:1:10e⁻¹². In order to allow for a comparison between different carbon isotope studies, the ${}^{13}C$: ${}^{12}C$ ratio is standardized and expressed in permil as $\delta^{13}C$ (Eq. 24 for DIC) (Zeebe and Wolf-Gladrow, 2001). Our implementation of ${}^{13}C$ follows the OMIP guidelines (Orr et al., 2017). The ${}^{14}C$ is standardized as $\Delta^{14}C$ (see Sect. Section 3.3.3). Variations in $\delta^{13}C$ are due to isotopic fractionation processes
- 20 during air-sea gas exchange, photosynthesis and CaCO₃ formation, but the latter is neglected in our implementation (see also Section 3.3.2). For ¹⁴*C*, each fractionation factor is the quadratic of the respective ¹³*C* value (i.e., $\kappa^{14}C = \kappa^{13}C^2$). Lastly, ¹⁴*C* is radioactive and decays with a half-life of 5730 years to ¹⁴*N* following:

$$dX^{14}C/dt = -\lambda \cdot X^{14}C,\tag{17}$$

where

25
$$\lambda [day^{-1}] = ln(2)/(5730 \cdot 365).$$
 (18)

X represents any ${}^{14}C$ state variable. In the model, all ${}^{14}C$ tracers are normalized to prevent numerical errors from carrying values close to the precision of the model.

The newly implemented marine carbon isotope code parallels the respective total DIC code, and in addition includes fractionation during photosynthesis and air-sea gas exchange (as well as decay for the ${}^{14}C$ tracers). In addition to the dissolved

30 i

inorganic tracers, the following ${}^{13}C$ and ${}^{14}C$ state variables have also been implemented: dissolved organic carbon, particulate organic carbon, calcium carbonate, phytoplankton, and zooplankton. Therefore, 12 new tracers are added (isotopic DOC, POC, CaCO₃, zooplankton, and phytoplankton). Due to the long equilibration time, the isotopic tracers are only activated in the computationally efficient configurations, such as the ocean carbon cycle stand-alone configuration (NorESM2-OC) or the low resolution version of the coupled model. Equilibrium times of the carbon isotopes are long due to the slow air-sea gas equilibration (Broecker and Peng, 1974) and long-term transient effects from the balance between sediment burial and weathering (Roth et al., 2014). Realistic equilibration times are therefore currently only obtained when bypassing the sediment module

- 5 of the model when running the carbon isotopes, as well as omitting carbon isotope input from rivers. When bypassing the sediment, mass balance is maintained by redistributing POC fluxes to the sediment over the entire overlying water column, and by dissolving inorganic carbon fluxes as well as opal fluxes at the bottom of the model immediately. Ongoing work will add the possibility for a fast sediment spin-up for use in future versions of the biogeochemical ocean model. Since $\Delta^{14}C$ is governed mainly by radioactive decay and circulation, we focus on $\delta^{13}C$ in our description of isotopic fractionation effects (Sect. 3.3.1 and 3.3.2). 10

3.3.1 Air-sea gas exchange fractionation

During air-sea gas exchange, the lighter ${}^{12}C$ isotope preferentially escapes to the atmosphere. This fractionation process increases $\delta^{13}C$ of surface ocean DIC, although the local net effect depends on the interplay between the local thermodynamic air-sea disequilibrium, the air-sea gas exchange rate, and the strength of the fractionation (Schmittner et al., 2013; Morée et al., 2018). The air-sea fractionation is a function of temperature $(T[^{\circ}C])$ and CO_3^{2-} fraction (fCO_3) , such that fractionation in-15 creases with decreasing temperatures, resulting in higher $\delta^{13}C$ in colder than in warmer surface water (Zhang et al., 1995). The fractionation during air-sea gas exchange, which varies between ~ 8 and 10.5 $\%_{00}$, is due to the combined effects of (1) the fractionation between CO₂ and the different carbon species, $\alpha_{CT_{e}}$, (2) kinetic fractionation, α_{k} , and (3) fractionation during gas dissolution, α_{aq_g} . The net air-sea ¹³CO₂ flux is formulated as follows:

20
$$F^{13}CO_2 = k_w \cdot \alpha_k \cdot \alpha_{aq_g} \cdot (p^{13}CO_2^{atm} - p^{13}CO_2^{sw} / \alpha_{CT_g}).$$
 (19)

Any fractionation α_i in Eq. 19 can be expressed as $\alpha_i = (\varepsilon_i/1000+1)^{\circ}/_{00}$, where

$$\varepsilon_k = -0.88,\tag{20}$$

$$\varepsilon_{aq_g} = 0.0049 \cdot T - 1.31, and \tag{21}$$

25 and

$$\varepsilon_{CT_g} = 0.0144 \cdot T \cdot fCO_3 - 0.107 \cdot T + 10.53. \tag{22}$$

In Eq. 19, k_w represents the gas transfer velocity for CO₂ according to Wanninkhof (2014) and T is in degrees C.

3.3.2 Biological fractionation

Phytoplankton prefers the lighter (¹²C) isotope during photosynthesis, thereby increasing δ^{13} C of DIC in the surface ocean and producing low- δ^{13} C POC. In the interior ocean, the low- δ^{13} C POC is released back into the water column during rem-30
ineralization/respiration, though without fractionation (Laws et al., 1995; Sonnerup and Quay, 2012). The 'biological isotope pump' thus creates a gradient of higher surface water δ^{13} C and lower deep water δ^{13} C. The average fractionation during photosynthesis is approximately 19 % (Lynch-Stieglitz et al., 1995; Tagliabue and Bopp, 2008). Even though many relationships for biogenic isotope fractionation have been proposed (e.g., Rau et al., 1996; Keller and Morel, 1999; Popp et al., 1998), the

5 modelled δ^{13} C distributions are not very sensitive to the different parameterization parameterizations, especially not in the surface ocean (Schmittner et al., 2013; Jahn et al., 2015). In addition, some relationships may be unsuitable for global scale modelling applications due to the dependency on unknown parameters (e.g., specific species, cell size, and cell membrane permeability).

A parameterization by Laws et al. (1997) is chosen in NorESM2, where the biological fractionation ε_{bio} depends on the ratio between phytoplankton growth rate μ [day⁻¹] at every model time step and the aqueous CO₂^{*} concentration [μ molkg⁻¹]:

$$\varepsilon_{bio} = \frac{6.03 + 5.5\mu/CO_2^*}{0.225 + \mu/CO_2^*}.$$
(23)

The growth rate μ is the brutto growth rate, uncorrected for losses such as mortality and exudation. ε_{bio} increases with increasing CO₂^{*} and decreasing μ . Fractionation values are kept within a realistic range of 5-26 % to correct for the influence of μ/CO_2^* extremes (similarly as done by Tagliabue and Bopp, 2008). This nonlinear parameterization of Laws et al. (1997) as described

- 15 in Eq. 23 is preferable over the linear formulation by Laws et al. (1995), because the linear formulation could result in unrealistic fractionation at high growth rates. Isotope equilibrium fractionation during CaCO₃ formation increases δ^{13} C of CaCO₃ and decreases therefore depletes seawater of ¹³C (thereby lowering seawater δ^{13} C). Nevertheless, the fractionation effect during CaCO₃ formation is relatively small (i.e., in the order of -2 to +3 %₀₀, depending on species and environmental conditions; Grossman and J to the effects of air-sea gas exchange and photosynthesisand therefore is often ., Therefore, fractionation during CaCO₃
- 20 formation is commonly omitted in modelling (Lynch-Stieglitz et al., 1995; Schmittner et al., 2013) studies (e.g., Lynch-Stieglitz et al., 1995; There are additional uncertainties related to temperature- and species-dependencies (Grossman and Ku, 1986; Zeebe and Wolf-Gladrow, 2001). Taking this into consideration, we have chosen not to implement fractionation during CaCO₃ formation in NorESM2.

3.3.3 Diagnostic and initialization

25 In order to evaluate the carbon isotopes against observations, we compute the diagnostic variables δ^{13} C and δ^{14} C according to their standard formulations, as follow: follows:

$$\delta^{13}C = \left[\frac{DI^{13}C/DI^{12}C}{(^{13}C/^{12}C)_{PDB}} - 1\right] \cdot 1000^{\circ}_{\circ\circ},\tag{24}$$

$$\delta^{14}C = \left[\frac{DI^{14}C/DIC}{NBS_{std}} - 1\right] \cdot 1000\%,\tag{25}$$

10

$$\Delta^{14}C = \delta^{14}C - 2 \cdot (\delta^{13}C + 25) \cdot (1 + \delta^{14}C/1000), \tag{26}$$

where $({}^{13}C/{}^{12}C)_{PDB}$ is the established Pee Dee Belemnite standard ratio of 0.0112372, and NBS_{std} is 1.170e⁻¹² (Orr et al., 2017). The $DI^{12}C$ is computed as the difference between total DIC and $DI^{13}C$.

In the model, the initialization of the carbon isotope tracers is done as follows: first, the model is spun up with preindustrial boundary conditions until the non-isotope carbon chemistry reaches a quasi-equilibrium states state. Next, we compute $DI^{13}C$

initial values by solving Eq. 24 for ¹³C using the simulated AOU and DIC together with the $\delta^{13}C$:AOU relationship reported 5 by Eide et al. (2017), i.e., $\delta^{13}C$ =-0.0075·AOU+1.72. The $DI^{14}C$ is initialized by first calculating $\delta^{14}C$ by solving Eq. 26 using pre-industrial $\delta^{13}C$ from Eide et al. (2017) (with the missing upper 200 m copied from the 200 m depth values) and the observational-based estimate of pre-industrial $\Delta^{14}C$ (Key et al., 2004). This $\delta^{14}C$ is then converted to absolute model $DI^{14}C$ using Eq. 25. The remaining isotope tracers are initialized as $C \cdot R \cdot \zeta$, with C as the total carbon counterpart of the respective isotope tracer, R as DI¹³C/DI¹²C or DI¹⁴C/DI¹²C ratio, and $\zeta = 0.98$, a typical value for biological fractionation (applied for 10

15

to organic carbon components only).

Here, the prognostic atmosphere atmospheric pCO_2 was initialized at 278 ppm from the start of the simulation. At initialization of the carbon isotopes, atmospheric $\delta^{13}C$ and $\Delta^{14}C$ are set to -6.5 % and 0 % respectively. Lastly, the results are presented as calibrated $\Delta^{14}C$ to account for long-term drift. That is, $DI^{14}C$ is multiplied with a factor F before standardization to $DI^{14}C$ corresponding to an atmospheric $\Delta^{14}C$ of zero permil (${}^{14}C_{atmzero}$).

$$F = \frac{{}^{14}C_{atmzero}}{{}^{14}C_{atm}},\tag{27}$$

where ${}^{14}C_{atm}$ in our simulation is found from the pre-calibrated $\Delta^{14}C_{atm}$ of approximately 36 $\%_{00}$ and ${}^{14}C_{atmzero}$ is likewise determined by solving Eqs. 25 and 26 for $\Delta^{14}C=0^{\circ}/_{00}$ in both cases using model $\delta^{13}C_{atm}$ (-7.5 $^{\circ}/_{00}$) and atmospheric pCO₂ (294 ppm). This leads to $F = \sim 2.5\%$.

Model simulations 20 4

Due to the long time scale of the large-scale ocean thermohaline circulation (flushing time is in the order of 1000-1500 years), a sufficiently long model integration in the order of at least 1000 years is required to achieve a quasi-equilibrium biogeochemical state in the interior ocean (Séférian et al., 2016). Prior to running any transient simulations, we have spun the model up for 1200 model years in a fully coupled configuration so that the ocean biogeochemistry reaches a quasi equilibrium state. In the spin-up

- simulation, all boundary conditions are fixed to constant preindustrial values, following the CMIP6 protocols (Eyring et al., 25 2016a). Due to the oceanic DMS fluxes, there is an internal feedback between the ocean biogeochemistry and the atmospheric radiative forcing during the spin-up. At the The end of the model spin-up, the simulation is branched off into is used as a starting point for (1) preindustrial control simulation (*piControl*, years 1850-2349), (2) transient historical simulation (years 1850-2014), and (3) esm-piControl-spinup simulation (100 model years). Simulation The end of simulation (3) is furthermore
- branched off into used as a starting point for (3a) an esm-piControl simulation (for 250 years), and (3b) a transient esm-hist 30 (years 1850-2014) simulation. In all The use of 'esm-experiments, esm-' (i.e., in experiments 3, 3a, and 3b) follows the CMIP6 naming convention (Eyring et al., 2016a), which indicates emission-driven simulation where the atmospheric CO₂ is prognos-

tically computed from ocean-atmosphere and land-atmosphere CO_2 -fluxes, as well as from prescribed anthropogenic emissions (for *esm-hist*). Here, the atmospheric CO_2 is transported by the atmospheric circulation model(see also Eyring et al., 2016a). In the *esm-piControl-spinup* the atmospheric CO_2 concentration is relatively stable with a small drift of -0.002 ppm yr⁻¹ and a long-term mean of 280.6 ± 0.4 ppm.

5 Except for the carbon isotopes (see next paragraph), all analyses shown in this paper were extracted from the two transient historical simulations (prescribed CO₂-*historical* and prognostic CO₂-*esm-hist*). Since both simulations produce nearly identical climatology states, we only present results from the prescribed CO₂-*historical* simulation.

For the newly implemented carbon isotopes, a considerably longer spin-up is required. Therefore, the carbon isotope tracers were not activated in our fully coupled simulations. Instead, simulations with carbon isotopes switched on have been performed

- 10 using the ocean carbon cycle stand-alone configuration of NorESM2. Additionally, a lower resolution ocean grid (using a similar tripolar grid as described above but with a nominally 2° resolution) has been used. This configuration avoids the high computational cost of running a long spin-up in a fully coupled configuration. The atmospheric forcing of the spin-up is the CORE normal year forcing (Large and Yeager, 2004), which represents a climatological mean year with a smooth transition between end and start of the year. Atmospheric CO₂, ¹³CO₂, and ¹⁴CO₂ concentrations are kept track of with a box-atmosphere
- 15 (i.e., assuming 100% instant mixing), which is updated each time-step according to the modelled air-sea CO₂ fluxes using a conversion factor of 2.13 ppm Pg \mathbb{C}^{-1}). These simplified prognostic atmospheric fields are simulated from the start of the spin-up (and the subsequent start of the carbon isotope simulation). As mentioned above, we have not implemented the carbon isotopes in the sediment compartment of the model. Therefore, the sediment module of HAMOCC was switched off for the carbon isotope simulations presented here. Otherwise, the setup of the ocean carbon cycle stand-alone configuration is as
- 20 described in Schwinger et al. (2016). The model spin-up was run for 5000 years, of which the first 1000 years were run without carbon isotopes. At year 1000 (4000), after the largest transient changes in biogeochemical tracers have flattened out, we initialised the the ^{13}C (^{14}C) tracers as described above.

Except for the carbon isotopes, all analyses shown in this paper were extracted from the two transient historical simulations (prescribed CO₂-*historical* and prognostic CO₂-*esm-hist*). Since both simulations produce nearly identical climatology states,

25 for the majority part of the mean stateevaluation, we only present results from the *historical* simulationAt the end of year 5000, the model reaches a quasi-equilibrium state, which approximately corresponds to the preindustrial state. Here, we evaluate the simulated carbon isotope simulations against the preindustrial estimates of the respective tracers.

5 Results

Here, we evaluate the model performance in simulating the mean climatological state of key biogeochemical tracers as well as
the evolution of air-sea CO₂ fluxes from the transient *historical* and *esm-hist* simulations. The carbon isotope results are from a stand-alone ocean simulation.

5.1 Statistical performance summary

We assess, relative to the observational estimates and the earlier version NorESM1, the simulated long-term annual mean of global hydrography and biogeochemical tracer distributions. Where relevant, the mean seasonal cycles are also evaluated for the surface values. The list of parameters, the averaging periods, and the respective observational data used to assess the

- 5 model performance are given in Table 2. We use spatial correlation and normalized-RMSE (root mean squared error) metrics to measure the model-data difference and determine whether or not the current model version has improved and performed better than the earlier version. The normalization was done by dividing the RMSE with the standard deviation of the respective observations.
- Figure 2b shows that, except for the surface layer, the normalized-RMSE of most of the NorESM2 variables' mean climatology is either noticeably improved or relatively similar to that of NorESM1. At subsurface 500 m depth, all variables but salinity have lower RMSE values. For many of the biogeochemical tracers (phosphate, nitrate, oxygen, and DIC), NorESM2 shows improvements in model-data deviation better agreement with data throughout the water column. Similarly, the respective spatial correlations with the observations are considerably improved for most variables, especially for nitrate, DIC, and alkalinity (Fig. 2a). For surface seasonality, NorESM2 performs fairly comparable to NorESM1, with improvements in all
 seasonal metrics (NRMSE and spatial pattern) for surface salinity and net primary production. NorESM2 simulates slightly
- larger NRMSE in its surface nutrients (phosphate, nitrate, and silicate) in both their annual and seasonal average values relative to NorESM1. This is attributed to the anomalously high concentration in the Arctic Ocean and parts of the Southern Ocean (just north of the Antarctic Circumpolar Current). More details on the performance of each specific variable are discussed in the following subsections.

20 5.2 Temperature

NorESM2 simulates a warm bias for both preindustrial conditions and the contemporary surface ocean (Fig. 3d), where a warm bias as high as 5° C is simulated in some regions of the Southern Ocean. Cold biases are simulated in parts of the north and equatorial Pacific, North Atlantic subpolar gyre, and the Arctic Ocean. The climatological mean temperatures are broadly similar between both NorESM versions with NorESM2 having less spatial correlations at 1 and 1.5 km depths (Fig. 2a,b). At

- 25 depths below 2 km, with the exception of North Pacific, NorESM2 simulates less biases (Fig. 3e,f,h,i). This improvement is related to the more accurately simulated Antarctic Bottom Water (AABW; see also Supplemental Fig. S1) in both the Pacific and Atlantic basins, as also depicted in Fig. 3. While NorESM1 simulates a too strong AMOC strength (roughly 31 Sv at 26°N), NorESM2 has a more reasonable strength of approximately 21 Sv. Nevertheless, the warm bias of 1-2°C in the North Atlantic Deep Water (NADW) remains when compared to observations. The cold bias seen previously in the intermediate watermass
- 30 in the Southern Ocean and North Pacific has been improved in the new version. In the Atlantic sector just south of 30°N at 1 km depth, there is a warm bias of up to 5°C (Fig. 3e), which could be related to the anomalously strong overflow watermass from the Mediterranean Sea. Nevertheless, the cold bias in the Atlantic and Pacific tropical/subtropical thermocline simulated in NorESM1 is now improved in NorESM2. Temperature values from NorESM2-LM are shown in Supplemental Fig. S2.

5.3 Salinity

The RMSE in surface salinity is reduced in NorESM2, mostly owing to improvements in the Arctic, where the previous model version simulates too saline water (Fig. 4d,g). Also at the surface, NorESM2 simulates anomalously too fresh and too saline biases in the Pacific and the Atlantic basins, respectively. In the subtropical south Pacific, the negative bias is as high as -2

- 5 psu, and is consistent with a too strong precipitation rate in this region simulated by the atmospheric component (not shown). Positive biases are simulated in the northern Indian Ocean, and along the Benguela current extension. In the interior, the salinity bias in the Pacific and the Atlantic is in the order of ± 0.11 and ± 0.4 psu, respectively (Fig. 4e, f). In the NADW, NorESM2 displays a positive bias, which is, however, smaller than that of NorESM1 (Fig. 4h; bias greater than 0.5 psu). Similar to temperature bias, NorESM2 produces too much saline water around 30°N at 1 km depth in the Atlantic (Fig. 4e), due to overflow
- 10 of high salinity Mediterranean watermasses. In the interior Pacific and Southern Ocean, both models' performances are generally comparable (Fig. 4e,f,h,i). Except for the surface salinity, simulation with lower atmospheric resolution (NorESM2-LM) depicts similar salinity pattern throughout the water column (Supplemental Fig. S3).

5.4 Mixed layer depth

Seasonally-averaged mixed layer depth (MLD) from NorESM2 and observations are shown in Fig. 5. To be consistent with

- 15 the observational estimates (de Boyer Montégut et al., 2004), we have computed our MLD using the $\sigma_T \sigma_T$ (density) criteria, which is the first depth where the change from surface $\sigma_T \sigma_T$ of 0.03 has occurred. In NorESM1, the MLD is generally too deep throughout most ocean regions. The new improvements in the MLD parameterization in NorESM2 reduce this bias considerably, especially in the low- and mid-latitude and the North Pacific regions. In the North Atlantic subpolar subpolar North Atlantic and parts of the Southern Ocean, the simulated MLD is still deeper than the observation-based estimates. In the
- 20 Weddell Sea, NorESM2 also persistently simulates <u>a</u> deep MLD throughout the year, a feature not seen in the observations. There is no significant difference between MLD simulated in NorESM2-MM and NorESM2-LM (Supplemental Fig. S4).

5.5 Ocean ventilation

To assess the simulated ventilation, we compare the passive chlorofluorocarbon (CFC-11) tracer distribution in the interior Atlantic and Pacific Oceans as simulated by NorESM2 with that from observations, as shown in Fig. 6. We compare the simulated CFC-11 from calendar year 2000 with climatological estimates of Key et al. (2004). Similar to the observations, the simulated high values of CFC-11 concentrations are generally simulated in the upper 1 km in both the Atlantic and the Pacificare mostly confined to the upper 1 km depth, with the exception of the North Atlantic, where up to 0.5 nmol CFC-11 m⁻³ penetrates down to 5 km depth. In the mid-latitude North Atlantic (i.e., 30°N), the model is unable to simulate the observed high values at around 1 km depth. This is likely due to the discrepancy in the watermass origin between the model

30 and observations. Here, the watermass in the model is too old, as also seen in the too high Apparent Oxygen Utilization and dissolved inorganic carbon (see below subsections on oxygen and DIC). The model simulates too strong deep water ventilation

in both the Atlantic and Pacific sectors of the Southern Ocean as well as in the high latitudes of the North Atlantic. <u>Similar</u> ventilation patterns are also produced by NorESM2-LM (Supplemental Fig. S5), with slightly less ventilated North Atlantic.

5.6 Nutrients

- In general, the climatological nutrient distributions (phosphate, nitrate, and silicate) in NorESM2 are either comparable or 5 improved compared to those of NorESM1, both in spatial correlation and normalized RMSE (Fig. 2a,b). At depth, these improvements are due to the new particulate organic carbon sinking scheme, where the sinking velocity increases linearly with depth. In NorESM1, the sinking speed is constant with depth, leading to more organic materials material being remineralized in the upper 1500 m1.5 km. The sinking scheme in NorESM1 produces anomalously high regenerated phosphate in at intermediate depth (approximately between 100 and 1500 m depthsdepth; Fig. 7h,i). A consequence of this is a too strong oxygen
- 10 minimum zone (OMZ; see also next subsection) and denitrification in the low latitudes, where the nitrate concentration becomes overly depleted. On the other hand, at depths greater than 1500 m, NorESM1 exhibits too low concentrations in of all nutrients relative to the observations (panels (h) and (i) of Figs. 7-9).

With the new sinking scheme, more organic material reaches the deep ocean and gets remineralized there. As a result, model data biases in phosphate and nitrate in the interior are noticeable improved (FigFigs. 2b, 7e, f, 8e, f). In the Pacific OMZ, a

- 15 reduced export production in NorESM2 also contributes to the reduced nutrient bias. In the interior North Atlantic, despite strong positive spatial correlation with the observations, NorESM2 shows a positive bias at depth below 3000 m and north of 30°S (Figs. 7e, 8e, 9e). Through analyzing the quasi-conservative 'PO' tracer (Broecker, 1974), we are able to attribute this to the bias in the simulated watermass (Supplemental Figure Fig. S1). In NorESM2, this region is dominated by Antarctic Bottom Water, whilst the observations indicate NADW as the main watermass. As with NorESM1, the current model still simulates too
- 20 low nutrient concentrations in the North Pacific intermediate watermass (Figs. 7f, 8f, 9f). This is likely associated with the too low surface primary production simulated in this region (see also Sect. Section 5.8), leading to limited organic matter available for remineralization at depth. We note that the circulation bias in this region could also contribute to the nutrient bias. In the Southern Ocean, too strong ventilation potentially leads to the negative nutrient biases in both Atlantic and Pacific sectors.
 Less-In NorESM2, less export production and deeper distribution of POC through the changed sinking scheme lead to a re-
- duced rate of denitrification in the Pacific oxygen minimum zone and a greatly improved nitrate distribution . In NorESM1-ME(Fig. 8f). In NorESM1, far too much nitrate was consumed for denitrification (see also Fig. 8ei). The silicate spatial distribution in NorESM2 closely resembles the phosphate distribution (Figs. 9a-c and 7a-c), with similar biases across the vertical sections of Atlantic and Pacific (Figs. 9e,f and 7e,f). At high-latitudes, NorESM2 overestimates the surface silicate concentration (Fig. 9d), which could be attributed to the reduced opal export sinking speed relative to our earlier model version (30 instead of previously 60 m.day⁻¹).

In NorESM, the <u>bulk</u> phytoplankton growth rate is limited by multiple nutrients (i.e., phosphate, nitrate, and dissolved iron), in addition to temperature and light (see also Section 2.3.2 of Schwinger et al., 2016). The model does not explicitly simulate diatom and calcifier classes. The inclusion of iron is critical to simulate the observed HNLC (High Nutrients Low Chlorophyll) regions in the world ocean where year-long elevated concentrations of both nitrate and phosphate are not exhausted by phy-

previously of in day).

toplankton (de Baar et al., 1995; Martin and Fitzwater, 1988). The three major HNLC regions are the subarctic North Pacific, the eastern and central equatorial Pacific, and the Southern Ocean. Here, low levels of bio-available iron concentrations limits limit the phytoplankton growths. In NorESM1, only the subarctic North Pacific is simulated to be iron-limited, with the other two HNLC regions being mostly limited by nitrate, as shown in Fig. 10. Iron limitation is also incorrectly simulated in parts

5 of the subtropical North Pacific, South Pacific, and most parts of the western Pacific. Following improvements in the iron parameterization, the three HNLC regions are now shown to be iron-limited in NorESM2. Nevertheless, the iron limitation bias in the western equatorial Pacific remains.

5.7 Dissolved oxygen

At surface level, dissolved oxygen is mostly constrained by solubility and hence sea surface temperature. Both NorESM1 and

- 10 NorESM2 perform similarly for dissolved oxygen at the surface with subtle improvements in the seasonal spatial correlation coefficient of NorESM2 (Fig. 2a,c). As with nutrients, the climatological distributions of interior oxygen improved considerably when compared to NorESM1 (Fig. 11), which is one of the CMIP5 ESMs with too strong OMZ (Cabré et al., 2015). In NorESM2, the OMZ volume is reduced considerably in both the equatorial Atlantic and Pacific, and the absolute dissolved oxygen values are also much improved . The (Fig. 11e,f). Here the remaining bias is related to the nutrient trapping issue (Six
- 15 and Maier-Reimer, 1996), as can also be seen from too high apparent oxygen utilization (AOU; Fig. 912d). In the interior Pacific and Atlantic roughly at 2 km and deeper, NorESM2 simulates considerably higher oxygen concentration than the observations by as much as 50% (North Pacific deep water; Fig. 11f). In the North Pacific deep water, this is primarily due to the still too little organic matter flux reaching below the mesopelagic zone, leading to too low apparent oxygen utilization (Fig. 12d). The bias uncertainty in the dissolved organic carbon (not shown), which is currently not well constrained due to lack of
- 20 observational data, could also play a role in this. In addition to biological constraints, the remaining interior oxygen bias can be attributed to the too strong Antarctic Bottom Water (AABW) ventilation, which carries a colder (Fig. 3e,f) and higher oxygen saturated watermass into both the Atlantic and Pacific bottom waters (Fig. 11e,f).

5.8 Biological production

In NorESM1, the primary production is generally too strong in the equatorial Pacific and in the high latitude at high latitudes during summer months. In the subtropical oceans, the simulated production is too low relative to the observational estimates. With the newly tuned ecosystem parameters of NorESM2, the spatial productivity patterns are improved considerably with an average absolute bias in the open ocean of less than 5 mol C m⁻² yr⁻¹ (Fig. 13a). In NorESM1 these biases reach more than 10 mol C m⁻² yr⁻¹, especially in the Equatorial Pacific, the South Atlantic and the Southern Ocean. Seasonally, the spatial patterns are also improved when compared to the observations with a correlation of approximately r = 0.5 (Fig. 2Figs. 2c,

30 <u>13c</u>). The RMSE has also reduced considerably, particularly in the October-November-December months, where NorESM1 simulates a much too strong spring bloom in <u>parts of</u> the Southern Ocean which is overestimated by as much as a factor of eight (e.g., Figs. 13e, 14e and Nevison et al., 2015). Similarly, the spring bloom blooms in the North Atlantic and North Pacific are

overestimated in NorESM1 during the boreal spring months (Figs. 13e, 14a). The improved spring bloom characteristics at high latitudes, however, comes at the cost of a too low annual mean productivity in the North Atlantic and North Pacific.

Due to unresolved physical dynamics, the model is still unable to reproduce the observed high productivity in coastal regions. Globally, the contemporary total primary production in NorESM2 (NorESM1) is 35.3 (39.9) Pg C yr⁻¹. These estimates are

- lower than the observational (MODIS, excluding coastal gridsareas) estimates of 46.1 Pg C yr⁻¹ but within the broad range of CMIP5 models (Bopp et al., 2013). In our model, export production is estimated as the flux of particulate organic carbon exiting the base of the euphotic zone (here assumed to be 100 m depth), as a function of zooplankton and phytoplankton mortalities and zooplankton fecal materials. Table 3 shows that the export production in NorESM2 is 5.39 Pg C yr⁻¹, as compared to 7.90 Pg C yr⁻¹ simulated by NorESM1. Current estimates of the global export production remain highly uncertain, from 5 to >12
 Pg C yr⁻¹, with a more recent satellite-based approach that accounts for food-web processes revealing values of 6±1.2 Pg C
- yr^{-1} (Siegel et al., 2014).

Table 3 also summarizes other export production metrics from both model versions. As a consequence of the new particle sinking scheme, there is more POC exported into the deep ocean. This is especially reflected by the values of transfer efficiency $(T_{eff-1km})$, which is calculated as a ratio between POC fluxes at a depth of 1000 m and of 100 m. The $T_{eff-1km}$ in NorESM2 (0.24) is increased by a factor of four relative to that in NorESM1 (0.06). Compared to a recent estimate of transfer efficiency

- 15 (0.24) is increased by a factor of four relative to that in NorESM1 (0.06). Compared to a recent estimate of transfer efficiency reconstructed from observed interior biogeochemistry (Fig. 13b; Weber et al., 2016), NorESM2 still simulates strong values too high transfer efficiencies of >50% in the eastern Equatorial Pacific (Fig. 13 Weber et al., 2016) (Fig. 13d). On the other hand, the transfer efficiency in the Southern Ocean and northern high latitudes are comparable with observations. This represents a significant improvement when compared to NorESM1 (Fig. 13d,f), which simulates nearly uniform $T_{eff-1km}$ values of
- 20 less than 0.1 in nearly all ocean regions with the exception of eastern Equatorial Pacific and Equatorial Atlantic.

Figure 14 shows the seasonal cycle of biological production rates as simulated by NorESM1 and NorESM2 in averaged over different ocean regions together with observation-based estimates. The regional mean values in NorESM2 are generally lower than those in NorESM1, except in the tropies and the North Pacific North and tropical Pacific (Fig. 14b,c). Nevertheless, Fig. 2e-d shows that NorESM2 simulates a lower normalized RMSE than NorESM1 in all seasons. As stated above, this bias could be reduced even more if we excluded the continental shelf regions. Nevertheless, Fig. 14 shows that the seasonal phase and

amplitude of biological production rates in NorESM2 in the North Atlantic, North Pacific, and the Southern Ocean southern hemisphere regions are closer to the observations when compared to NorESM1. In the Southern Ocean southern hemisphere (south of 20°S), NorESM2 no longer produces the large summer bias seen in NorESM1.

5.9 DMS production and fluxes

25

30 In NorESM2, DMS production is tuned towards the climatological observations from Lana et al. (2011). Figure 15 shows the surface DMS concentration for each season together with the observation-based estimates. To first order, the DMS distribution follows the spatial pattern of the simulated primary production, with higher concentrations during spring/summer periods in high latitude regions. In the Equatorial Pacific and Atlantic upwelling regions, the model overestimates DMS concentrations.

Similarly, in the North Pacific, where the model underestimates productivity during the boreal spring bloom (see also Fig. 1314b), the DMS concentration is lower than observed (Fig. 15b,f).

The simulated DMS concentration in the North Atlantic is comparable with observations. In the Southern Ocean, the model produces the observed high concentrations during summer months (JFM) along the $4540-45^{\circ}S$ latitude band well Fig. 15e). In

5 winter periods, the DMS concentration approaches zero in the model owing to the too low productivity, whilst the observations still indicate values of approximately 1 μ mol S m⁻³. For the present day period (1971-2000), the simulated global DMS flux is 19.76±0.19 Tg S yr⁻¹, which is in at the lower end of observation-based estimates (17.6-34.4 Tg S yr⁻¹).

5.10 Dissolved inorganic carbon and alkalinity

Except for the polar Southern Ocean, the surface alkalinity concentrations in NorESM1 are approximately 100 µmol L⁻¹
higher than the observational estimate. And since the spin-up was forced with constant preindustrial atmospheric CO₂ concentrations, the surface DIC concentration adjusted (also anomalously high bias; Figs. 16g and 17g) to yield the 'correct' surface pCO₂. Consequently, biases in surface DIC and alkalinity compensate one another to yield the approximately 'correct' carbonate ion concentration (i.e., alkalinity minus DIC), seawater CO₂ buffer capacity, and air-sea CO₂ fluxes.

- In NorESM2, we alleviate the surface alkalinity bias by increasing the surface sink associated with calcium carbonate 15 production. This is done by increasing the CaCO₃ to phosphate uptake ratio ($R_{CaCO_3:P}$ in Table 1). As a result, the export of particulate CaCO₃ increases from 0.49 Pg C yr⁻¹ in NorESM1 to 0.66 Pg C yr⁻¹, with both estimates still within the range of the observation-based synthesis of 0.52±0.15 Pg C yr⁻¹ (Dunne et al., 2007). This modification increases the PIC:POC export ratio to 0.13 (0.06 in NorESM1, see also Table 3) and considerably improves the surface DIC and alkalinity concentrations in NorESM2 (Figs. 16and 17a,d and 17a,d). At the same time, the spatial correlations with observations are also improved as
- 20 well as the normalized RMSE (see Fig. 2a,b). We note that in few regions, e.g., the subtropical South Pacific, the model still underestimates the observed DIC and alkalinity concentrations. We note that the PIC:POC ratio is still within the large range of other studies (0.03 to 0.25; Koeve, 2002; Sarmiento et al., 2002)

As in at the surface, the spatial distributions and normalized bias of interior DIC and alkalinity in the NorESM2 improve considerably throughout the water column relative to NorESM1 (Fig. 2a,b). In both the interior Atlantic and Pacific basins,

- 25 the spatial distribution and magnitude of DIC biases closely resembles that seen in the nutrients (e.g., phosphate; Fig. Figs. <u>16e,f.</u>, 7e,f), when multiplied by the constant stoichiometry in the model ($R_{C:P}=122$). This suggests that the mechanisms driving the nutrient bias are also responsible for the bias in interior DIC. For instance, the positive anomaly in the equatorial Pacific between 1 and 3 km depths originates from too much biological remineralization in the model.
- The simulated interior alkalinity concentrations in NorESM2 between 1 and 3 km depths are anomalously high, which can 30 partially be attributed to too low interior remineralization in the Atlantic (see also AOU values Fig. 12c) or too high CaCO₃ export production (Table 3).

5.11 Surface pCO₂ and sea-air CO₂ fluxes

The NorESM2 pCO_2 spatial correlation relative to the observations is improved in almost all seasons when compared to NorESM1, while the NRSME NRMSE is reduced for JFM and AMJ months (Fig. 2c,d). Within these months, improvements are mostly seen in the Southern Ocean, where too strong surface mixing in NorESM1 leads to anomalously high pCO_2 . The

- 5 climatological mean of contemporary surface pCO₂ in NorESM2 compares well with the observational compilations, as seen in Fig. 18a,e. Similar improvements are also seen in the CO₂ fluxes, where NorESM2 simulates the broad spatial patterns relatively well (Fig. 18c,g). NorESM2 also produces moderately weaker annual mean carbon flux in the northern mid- to high-latitudes than NorESM1. Figure 18b,d-f,h shows the zonally-averaged monthly surface pCO₂ and CO₂ fluxes in NorESM2. Here, the model also agrees well with the observation-based estimates (contour-linesFig. 18b,d) in term of amplitude and
- 10 temporal variability, with distinct seasonal cycle poleward of between 20° latitudeand 45° south/north. In the extra-tropical oceans (between 30-6045-65°N and south of 60°S), the simulated winter (January-March and July-September in northern and southern hemispheres, respectively) pCO₂ is considerably lower than that seen in the observations. This feature also translates to a stronger carbon sinks In these regions, nevertheless, the carbon sink during these periods is comparable with observations (Fig. 18d,h).
- Figure 19 depicts the mean seasonal cycle of sea-air CO_2 fluxes from NorESM models and observations, averaged over different regions. In the North Atlantic, North Pacific, and the mid-latitude Southern Ocean southern hemisphere (between 20– 40° S), the seasonal phase and mean integrated annual CO_2 fluxes in NorESM2 compare better with the observations than those in NorESM1 (Fig. 19a). Nevertheless, NorESM2 simulates a stronger than observed carbon sink in the Labrador Sea and along the Kuroshio current extension during boreal winter months. We note that the Landschützer et al. (2014) data estimates higher
- 20 surface pCO₂ (hence a weaker carbon sink) in these regions than the observational estimates from Takahashi et al. (2009). Despite the warm SST bias in most of the tropics, NorESM2 simulates a moderately weaker than observed CO₂ outgassing throughout the year (red line as compared to blue line in Fig. 19c). A potential explanation for this is the weaker upwelling rates of DIC-rich deep watermass-watermasses in the model. In the Southern Ocean (south of 40°S), the strong bias in the seasonal amplitude in NorESM1 is now considerably improved considerably reduced (Fig. 19e,f), due to the improvements in
- 25 the biological production in this region. Nevertheless, here the seasonal phase in NorESM2 is approximately reversed <u>compared</u> to <u>observations</u>: the strongest sink is simulated in the austral winter months whereas observational estimates indicate summer months.

5.12 Distribution of $\delta^{13}C$ and $\Delta^{14}C$

Figure 21a-c shows that the spatial pattern of $\delta^{13}C$ is in fair agreement with the gridded preindustrial observational estimate 30 of Eide et al. (2017). Nevertheless, The ocean component of NorESM2 simulates weaker variability than observed, i.e., it is unable to reproduce both very low (negative) and high (positive) values. Therefore, the simulated $\delta^{13}C$ concentrations in the relatively younger watermasses, such as the North Atlantic Deep Water (NADW) and the Antarctic Intermediate Water (AAIW), have negative bias relative to the observations (Fig. 21c). On the contrary, positive biases are depicted in the older Antarctic Bottom Water (AABW) and in the deep North Pacific (Fig. 21e-f). The $\delta^{13}C$ can be decomposed into biological and residual (air-sea gas exchange and circulation) components (Broecker and Maier-Reimer, 1992; Eide et al., 2017) :-

$$\delta^{13}C = \delta^{13}C_{BIO} + \delta^{13}C_{AS}$$

where $\delta^{13}C_{BIO}$ can be estimated as a function of phosphate and DIC distribution and their global mean values:

5
$$\delta^{13}C_{BIO} = \frac{-19 \cdot r_{C:P}(PO_4 - PO_4)}{\overline{DIC}} + \overline{\underline{\delta^{13}C}}$$

Applying this decomposition, we can attribute the majority of the differences between model and observational estimates of $\delta^{13}C$ to biases in the biological component (Fig. 22). The $\delta^{13}C_{BIO}$ reveals that in the Southern Ocean (south of ~55°S), as well as in its exported deep waters, NorESM2-OC simulates too high $\delta^{13}C$ due to the too small regenerated signal in these waters. The negative $\delta^{13}C$ anomaly in the North Atlantic of about -0.5 % is a combination of a too low $\delta^{13}C_{BIO}$ signature in NADW

- 10 (Fig. 22e), combined with a too negative influence from air-sea gas exchange and circulation as compared to observational estimates (the residual term $\delta^{13}C_{AS}$). In the Pacific the positive anomaly from the Southern Ocean persists throughout most of the basin due to too weak remineralization, except for at intermediate depths where remineralization is high (see also AOU, Fig. 12).
- The simulated ∆¹⁴C distribution is in general agreement with the pre-industrial observational estimate by Key et al. (2004).
 Nevertheless, interior model-observation biases are in the range of ±50 %. Waters south of -60°S are positively biased as compared to observations (Fig. 23), indicating too young and well-ventilated waters in this region. This positive bias of 20-50 % is carried into the Atlantic and Pacific basins (Fig. 23e-f) and is equivalent to an underestimation of radiocarbon age by 200-400 years. North of the Equator, a negative bias of similar magnitude indicates too old water masses in the Atlantic basin below 3 km depth as well as throughout the water column in the northern half of the Pacific basin. The top 800m of the water
- 20 column has a negative bias of $100 \%_{00}$, which we attribute to the too negative atmospheric $p^{14}CO_2$ before calibration (-36 $\%_{00}$) and biases in air-sea equilibration and fractionation. Calibrated $\Delta^{14}C_{atm}$ is by definition $0 \%_{00}$, $\delta^{13}C_{atm}$ equilibrates at -7.5 $\%_{00}$, and atmospheric pCO₂ is 294 ppm at the end of the simulation.

5.12 Transient changes

In this subsection, we describe the global mean surface air temperature and oceanic carbon uptake over the historical period (1850–2014) as simulated by NorESM2. For this, we discuss the two transient simulations: (i) *historical* and (ii) *esm-hist*. Similar to SST (see Fig. 3), there is a warm-bias in the simulated surface air temperature. For the period of 1850-1859, the NorESM2 global mean surface air temperature is 14.5°C compared to 13.7°C from the HadCRUT4 dataset (Morice et al., 2012). Nevertheless, the simulated warming between 2005-2014 and 1850-1859 is 0.7°C, relatively comparable with that from the comparable to that from observations of 0.8°C (Fig. 20a). The warming in the *esm-hist* is lower, 0.6°C. Figure 20a also

³⁰ shows that NorESM2 (orange and green lines) simulates warming rates that are comparable to observations (blue line) between 1970 and 2010, but shows a stronger cooling between 1950 and 1970 (*historical* and *esm-hist*).

In both simulations, the global mean surface ocean pCO_2 follows the respective atmospheric CO_2 trend over the simulation periods. Towards the end of the historical period, the oceanic pCO_2 diverges from (solid green and orange lines) diverges from, i.e., lower than) is lower than, the atmospheric counterpart (dashed-green and -orange lines) by approximately 20 ppm (Fig. 20b). Consistent with the lower surface ocean (than the atmosphere) pCO_2 growth rates, the ocean carbon uptake continues to

- 5 increase over most of the transient period in both simulations (Fig. 20c). In the *historical* simulation, the carbon sink stabilizes at roughly 0.7 Pg C yr⁻¹ between 1910 and 1960. This flattening of the uptake strength is associated with the slow down in the atmospheric CO₂ growth rate. Figure 20d shows confirms that the atmospheric growth rate weakens during this time window (orange-bars orange bars in Fig. 20d). On the contrary, the carbon sink in *esm-hist* steadily increases during the same period, also consistent with the steady increase in the simulated prognostic atmospheric CO₂ (green-bars in Fig. 20d).
- From 1960 onward, the atmospheric CO₂ growth rates in both simulations increase almost linearly in time from 0.8 to more than 2.0 ppm yr⁻¹. Consequently, the ocean carbon uptake strengths increase as well, from approximately 1 Pg C yr⁻¹ in the 1960s up to 2.5 Pg C yr⁻¹ in 2014 (in both *historical* and *esm-hist*). Rapid strengthening in ocean carbon uptake is also evident in estimates from the Global Carbon Project (Fig. 20, Le Quéré et al., 2018) (GCP, blue line in Fig. 20; Le Quéré et al., 2018). For the decades of 1980s and 1990s, the simulated carbon uptake is also well within the IPCC (Intergovernmental Panel for
- 15 Climate Change, Denman et al., 2007) estimates (Table 4). In NorESM2, a new tracer *DIC^{nat}* has been implemented (see also Sect. Section 3.2), allowing for a more accurate estimate of and it enables us to more accurately estimate the anthropogenic carbon uptake and storage in the ocean. Using this tracer, we can also estimate the net flux of anthropogenic carbon into the ocean during the historical period, as depicted by the purple line in Fig. 20c). The strong resemblance between the orange and purple lines suggests that the long-term trend in oceanic carbon sinks is associated with the increasing atmospheric CO₂,
- ²⁰ rather than changes in climate states (as expected for the historical runs where the change in the climate state is relatively small). We note that, in the preindustrial control simulation, the ocean is a weak net carbon sink, of approximately 0.1 Pg C yr^{-1} . Table 4 also shows that the cumulative carbon uptake by the ocean for the 1850-1994 and 1994-2011 is well within the observation-based estimates. The cumulative carbon uptake in NorESM2 is lower than that of NorESM1 and compares better with observations.

25 **5.13** Distribution of $\delta^{13}C$ and $\Delta^{14}C$

As stated above, the carbon isotopes tracers are not included in the coupled ESM configuration and only simulated in a preindustrial stand-alone ocean configuration. We also note that although the stand-alone ocean simulation is not a direct representative of the coupled ESM simulation, its equilibrium ocean state is very comparable. For instance, the interior distribution of phosphate and DIC concentration in the stand-alone ocean simulation (Supplemental Figs. S6, S7) share very

³⁰

similar characteristics with that produced in the coupled configuration (Figs. 7a-c, 16a-c). At the end of preindustrial spin-up, Fig. 21a-c shows that the spatial pattern of $\delta^{13}C$ is in fair agreement with the gridded preindustrial observational estimate of Eide et al. (2017). Nevertheless, the ocean component of NorESM2 simulates weaker variability than observed, i.e., it is unable to reproduce both very low (negative) and high (positive) values. Therefore, the simulated $\delta^{13}C$ concentrations in the relatively younger watermasses, such as the North Atlantic Deep Water (NADW) and the Antarctic Intermediate Water (AAIW), have

negative bias relative to the observations (Fig. 21e). On the contrary, positive biases are depicted in the older Antarctic Bottom Water (AABW) and in the deep North Pacific (Fig. 21e,f). The $\delta^{13}C$ can be decomposed into biological and residual (air-sea gas exchange and circulation) components (Broecker and Maier-Reimer, 1992; Eide et al., 2017) :

$$\delta^{13}C = \delta^{13}C_{BIO} + \delta^{13}C_{AS},$$
(28)

5 where $\delta^{13}C_{BIO}$ can be estimated as a function of phosphate and DIC distribution and their global mean values:

$$\delta^{13}C_{BIO} = \frac{-19 \cdot r_{C:P}(PO_4 - \overline{PO_4})}{\overline{DIC}} + \overline{\delta^{13}C}.$$
(29)

Applying this decomposition, we can attribute the majority of the differences between model and observational estimates of $\delta^{13}C$ to biases in the biological component (Fig. 22). The $\delta^{13}C_{BIO}$ reveals that in the Southern Ocean (south of ~55°S), as well as in its exported deep waters, NorESM2-OC simulates too high $\delta^{13}C$ due to the too small regenerated signal in these waters.

- 10 The negative $\delta^{13}C$ anomaly in the North Atlantic of about -0.5 $\%_{00}$ is a combination of a too low $\delta^{13}C_{BIO}$ signature in NADW (Fig. 22e), combined with a too negative influence from air-sea gas exchange and circulation as compared to observational estimates (the residual term $\delta^{13}C_{AS}$). In the Pacific the positive anomaly from the Southern Ocean persists throughout most of the basin due to too weak remineralization, except for at intermediate depths where remineralization is high (see also AOU, Fig. 12).
- 15 The simulated $\Delta^{14}C$ distribution is in general agreement with the pre-industrial observational estimate by Key et al. (2004). Nevertheless, interior model-observation biases are in the range of $\pm 50 \%_{00}$. Waters south of 60° S are positively biased as compared to observations (Fig. 23), indicating too young and well-ventilated waters in this region. This positive bias of 20-50 $\%_{00}$ is carried into the Atlantic and Pacific basins (Fig. 23e-f) and is equivalent to an underestimation of radiocarbon age by 200-400 years. North of the Equator, a negative bias of similar magnitude indicates too old water masses in the Atlantic basin
- 20 below 3 km depth as well as throughout the water column in the northern half of the Pacific basin. The top 800m of the water column has a negative bias of 100 ‰, which we attribute to the too negative atmospheric $p^{14}CO_2$ before calibration (-36 ‰) and biases in air-sea equilibration and fractionation. Calibrated $\Delta^{14}C_{atm}$ is by definition 0 ‰, $\delta^{13}C_{atm}$ equilibrates at -7.5 ‰, and atmospheric pCO₂ is 294 ppm at the end of the simulation.

6 Summary and discussion

25 The ocean biogeochemical component of the Norwegian Earth System Model (NorESM) has been updated from version 1 (as used in CMIP5) to version 2 (NorESM2). These developments focus on alleviating known biases in mean states and seasonal cycles of key variables when compared to the observations. This paper describes new and improved processes, newly implemented diagnostic and carbon isotope tracers, and highlights the model improvements relative to the earlier model version.

On the biogeochemistry side, we have introduced a revision to the particulate organic carbon vertical sinking scheme, an improved tuning of ecosystem parameters, riverine inputs of nutrients and other biogeochemical constituents, an updated air-sea gas exchange parameterization, and atmospheric deposition of nitrogen. In NorESM2, the ocean biogeochemistry also simulates (i) fully interactive oceanic DMS fluxes coupled to the atmospheric model, (ii) newly implemented preformed tracers, which will allow for enable us to perform a more detailed diagnosis of physical and biogeochemical sources and sinks in future studies, and (iii) capability to simulate carbon isotope, carbon isotopes that can be used e.g., in extended paleo time-scale simulations.

5 On the physical side, the simulated Atlantic Meridional Overturning Circulation strength is considerably improved. An hourly coupling (previously daily) across the ocean-atmosphere-ice interfaces has been implemented. Intensive fully coupled testing prompted further attempts to optimize mixed layer physics parameters, which slighly reduce the bias in MLD when compared to our earlier model version.

In connection to our contribution to the CMIP6 simulations, two model configurations have been prepared, NorESM2-LM and NorESM2-MM, where the latter adopts a higher atmospheric resolution. In both versions, the ocean components are identical. Following the CMIP6 protocols (Eyring et al., 2016a), we have performed preindustrial control (*piControl*, years 1850-2349) and transient climate (*historical*, years 1850-2014) simulations. Here, we analyzed the simulations performed using both model versions.

As a result of the above developments, the overall performance of NorESM2 in simulating the climatological state of in-15 terior ocean biogeochemistry has improved considerably. Some of the key improvements include (i) better representation of equatorial Pacific oxygen minimum zonethe equatorial Pacific OMZ, (ii) improved interior nutrient distributions, and (iii) representation of iron limitation in the Southern Ocean, the equatorial and North Pacific.

With respect to transient ocean carbon sinks, previous studies have indicated the importance of the background biogeochemical state, such as alkalinity, dissolved inorganic carbon, and buffer capacity, in order to adequately simulate the long-term

- 20 trends (Hauck and Völker, 2015; Fassbender et al., 2018; Lebehot et al., 2019). NorESM1 exhibits a considerable bias for the background biogeochemical state, especially in surface alkalinity concentrations. These biases have been substantially reduced in NorESM2, allowing for resulting in a more adequate representation of the background buffer factor and improved model fidelity in simulating the temporal evolution of the ocean carbon sink.
- The seasonality of upper ocean biogeochemistry has been identified as critical for quantifying a model's uncertainty in its long-term projections of oceanic carbon uptake (Fassbender et al., 2018). Further, with ongoing anthropogenic CO₂-emissions, the seasonal variability of regional sea-air CO₂-fluxes is expected to amplify. This amplification, which is linear for the thermal component and nonlinear for the biophysical component of oceanic pCO₂ (Fassbender et al., 2018), can also impact the longterm evolution of sea-air CO₂-fluxes. In this aspect, NorESM1 simulates a considerable bias in the seasonal cycle of biological production and air-sea CO₂ fluxes, particularly in the Southern Ocean. The updated set of ecosystem parameters in NorESM2
- 30 is able to mitigate these biases to some extent. In the North Atlantic, North Pacific, and most parts of the Southern Ocean, the seasonal cycle (both in phase and amplitude) of primary production and CO₂-fluxes in NorESM2 reflects observational estimates and their driving mechanisms more accurately (Landschützer et al., 2018). We note however that there is a mismatch between the CO₂-fluxes' seasonal cycle in observations and NorESM2 in the Southern Ocean between 40°N and 60 °N. A closer attention of to the associated physical (e.g., solubility) and biogeochemical (e.g., biological productivity) drivers in this

region is needed when considering future scenarios in the future. However, general seasonal characteristics are improved and we are confident that NorESM2 will provide a better estimate of future changes.

Despite numerous model improvements, several biases remain in the updated model. A long-standing issue of nutrient trapping in the intermediate depth of the equatorial Pacific persists. Here, the simulated phosphate concentration is still higher than observed, but the bias is less severe than that of NorESM1. Consistently, dissolved inorganic carbon is still anomalously high and oxygen too low in the same region. The simulated Southern Ocean still has a too strong ventilation, leading to biases in the newly formed bottom watermass. Other physical biases such as the Antarctic Bottom Water penetrating too northward also contribute far north also contributes to the simulated model-data differences in the interior North Atlantic.

Currently, evaluations of ocean biogeochemical component components in ESMs have been largely focused on assessing the model ability in simulating the mean climatology state, howeverclimatological state. However, further options exist (Eyring et al., 2016b; Heinze et al., 2019). The ocean biogeochemical observing community has put a lot of efforts effort in sustaining key monitoring networks, such as the Surface CO₂ Atlas (SOCAT; Bakker et al., 2014) and the Global Ocean Data Analysis Project (GLODAP; Olsen et al., 2016). This critical observational network has allowed for additional provided additional valuable constraints for ESMs, such as by confronting the model ability to simulate seasonality of surface ocean pCO₂ as well

as long-term trends in surface ocean pCO₂ (Tjiputra et al., 2014), ocean acidification (Lauvset et al., 2016), and deoxygenation (Tjiputra et al., 2018). As ESMs simulate emerging climate change signals in the coming decades (Henson et al., 2017), sustaining these observations into the future is becoming more important than ever.

In order to tackle the challenges associated with climate change, ocean modules in Earth system models could be further upgraded to include interactive methane (CH₄), nitrous oxide (N₂O), and biogenic volatile organic compounds (BVOCs)

20 cycles for enabling model projections in respective emission driven frameworks. Moreover, NorESM2 still has a relatively simple marine biology that would benefit from more key functional groups as well as their sensitivity to climatic stressors. In preparation for the anticipated increasing spatial resolution in the physical components of the model, different data assimilation methods (combined with state-parameter estimation on the basis of observations) are currently being explored to allow for efficient enable us to perform optimization of the current ecosystem parameterization more efficiently (Tjiputra et al., 2007;

Further NorESM-internal improvements may consider online estimated or progonostic organic carbon emissions from the ocean biogeochemistry to the atmosphere. Currently, we use a fixed particulate organic carbon emissions from the ocean, linked to an upper-ocean chlorophyll-a climatology. The spatial distribution and temporal characteristics of DMS emissions may have an impact on the strength of the feedback involving DMS and aerosol-cloud interactions. Dust and iron or phosphate contained

- 30 therein as well as nitrogen deposition will be computed in the future version of the atmospheric NorESM module and may provide more realistic nutrient input and may thus lead to feedbacks in a coupled ESM. The biogeochemical exchange processes are also dependent on sea ice cover and require better knowledge of air-sea-interaction processes in polar and subpolar regions. Mineralisation Mineralization processes in land regions may change under climate change conditions, leading to important changes in nutrient fluxes into the ocean biogeochemical systems. Biogenic Volatile Organic Compounds (BVOC) and
- 35 halocarbons emitted from the ocean have been suspected to influence atmospheric chemistry, e.g., in the stratosphere. It will

²⁵ Gharamti et al., 2017).

certainly be the ambition of further developments to investigate sensitive the sensitivity of these processes to climate change to increase complexity where needed.

Beyond their current applications, future ESMs should be developed further to provide societal relevant information that is directly relevant for policy decisions on climate mitigation and adaptation measures. This includes potential climate hazards

- 5 to societies (such as marine heatwaves, sudden pH drops, and spreading low oxygen zones in the oceans) as well as the risks emerging from the combination of hazards, vulnerabilities, and exposures (Oppenheimer et al., 2014). Future inclusion of socio-economic dynamics in ESMs and their feedback to climate through societal transformative action remains a big challenge (Giupponi et al., 2013) while a new ocean related socio-economic scenario framework (OSPs) has been developed mirroring the established Socioeconomic Pathways or SSPs (Maury et al., 2017). On the other hand, refining process complexity and
- 10 increasing resolution may not be the only alternative for further model development, as the results of a-more complex model simulations become more difficult to interpret (Bony et al., 2011). Therefore, simplified models and ESMs of intermediate complexity including the respective simplified ocean biogeochemistry modules, will need to be developed and used in parallel (for instance Steinacher et al., 2013).

Data availability. All observational data used in this paper are publicly available: World Ocean Atlas (https://www.nodc.noaa.gov/OC5/
 woa13/), Global Ocean Data Analysis Project version 2 (https://www.glodap.info/), surface mixed layer depth (http://www.ifremer.fr/cerweb/
 deboyer/mld/home.php), surface pCO₂ and air-sea CO₂ fluxes (https://www.nodc.noaa.gov/ocads/oceans/SPCO2_1982_2011_ETH_SOM_
 FFN.html), ocean primary production (https://www.science.oregonstate.edu/ocean.productivity/), Dimethyl Sulphide (https://www.bodc.ac.
 uk/solas_integration/implementation_products/group1/dms/), Δ¹⁴C and CFC-11 (https://www.nodc.noaa.gov/ocads/oceans/glodap/), and δ¹³C (https://www.bcdc.no/data-products.html).

20 *Code and data availability.* The codes of Bergen Layered Ocean Model and isopycnic-based Hamburg Oceanic Carbon Cycle (BLOMiHAMOCC) can be downloaded from https//github.com/NorESMhub/BLOM. The NorESM CMIP5 and CMIP6 model outputs can be accessed through the Earth System Grid Federation (ESGF) decentralized database (https://esgf-node.llnl.gov).

Table 1. Parameter values of the ecosystem parameterisation parameterizations that have been changed between NorESM1, NorESM-OC1.2, and NorESM2. NorESM-OC1.2 was an intermediate model version (Schwinger et al., 2016) and is included here for completeness and traceability. Naming of parameters follows Ilyina et al. (2013).

Parameter	Symbol	NorESM1	NorESM-OC1.2	NorESM2	Unit
Half-saturation constant PO ₄ uptake	K _{PO4}	2×10^{-7}	4×10^{-8}	4×10^{-8}	$kmol P m^{-3}$
Half-saturation constant grazing	Kzoo	4×10^{-8}	4×10^{-8}	$8 imes 10^{-8}$	$\mathrm{kmol}\mathrm{P}\mathrm{m}^{-3}$
Half-saturation constant silicate uptake	K _{Si}	$1.5 imes10^{-6}$	$1 imes 10^{-6}$	$5 imes 10^{-6}$	${\rm kmol}{\rm Si}{\rm m}^{-3}$
Phytoplankton mortality rate	$\lambda_{ m phy}$	0.008	0.008	0.004	d^{-1}
Zooplankton mortality rate	$\lambda_{ m zoo}$	$5 imes 10^{-6}$	$3 imes 10^{-6}$	$3 imes 10^{-6}$	$(\text{kmol}Pm^{-3}d)^{-1}$
Phytoplankton exudation rate	$eta_{ m phy}$	0.03	0.03	0.04	d^{-1}
Zooplankton excretion rate	β_{zoo}	0.03	0.06	0.06	d^{-1}
Maximum grazing rate	$\mu_{ m zoo}$	1	1	1.2	_
Zooplankton assimilation efficiency	$\omega_{\rm zoo}$	0.5	0.6	0.7	_
Silicate to phosphorus uptake ratio	R _{Si:P}	25	30	33	$mol Si mol P^{-1}$
CaCO ₃ to phosphorus uptake ratio	$R_{CaCO_3:P}$	35	40	45	$mol CaCO_3 mol P^{-1}$
Fraction of grazing ingested	\mathcal{E}_{zoo}	0.8	0.8	0.85	_
Detritus remineralization rate	λ_{det}	0.03	0.025	0.025	d^{-1}
DOC remineralization rate	λ_{DOC}	0.025	0.025	0.004	d^{-1}

Model variables	Periods	Observations and references
Temperature	1971–2000	World Ocean Atlas 2013 (Locarnini et al., 2013)
Salinity	1971–2000	World Ocean Atlas 2013 (Zweng et al., 2013)
Mixed layer depth	1971-2000	de Boyer Montégut et al. (2004)
Oxygen	1971-2000	World Ocean Atlas 2013 (Garcia et al., 2013a)
Phosphate	1971-2000	World Ocean Atlas 2013 (Garcia et al., 2013b)
Nitrate	1971-2000	World Ocean Atlas 2013 (Garcia et al., 2013b)
Silicate	1971-2000	World Ocean Atlas 2013 (Garcia et al., 2013b)
Dissolved inorganic carbon	1997–2007	Global Ocean Data Analysis Project (GLODAPv2; Lauvset et al., 2016)
Alkalinity	1997–2007	Global Ocean Data Analysis Project (GLODAPv2; Lauvset et al., 2016)
Surface pCO ₂	1982–2011	Landschützer et al. (2015)
Sea-air CO ₂ fluxes	1982–2011	Landschützer et al. (2015)
Net primary production	2003-2012	Averaged of the three remote sensing products (VGPM, Eppley-VGPM, and
		and CbPM) from Moderate Resolution Imaging Spectroradiometer
		(Behrenfeld and Falkowski, 1997; Westberry et al., 2008)
Dimethyl sulfate	1971-2000	Lana et al. (2011)
CFC-11	2000	Key et al. (2004)
δ^{13} C	Last 10 yrs of piControl	Eide et al. (2017)
$\Delta^{14}C$	Last 10 yrs of piControl	Key et al. (2004)

Table 2. List of the simulated variables and the respective historical simulation periods at over which their climatological values are have been averaged over. The last column indicates the observational reference used to validate the model output.

Table 3. Annual mean primary production biology-related metrics simulated by NorESM1-ME, NorESM2-LM, and NorESM2-MM.

	NorESM1-ME	NorESM2-LM	NorESM2-MM	Units
Primary production	39.86	33.27	35.29	$Pg C yr^{-1}$
POC export	7.90	4.94	5.39	$Pg C yr^{-1}$
CaCO ₃ export	0.49	0.66	0.74	$Pg C yr^{-1}$
Opal export	105.93	76.89	81.38	Tmol Si yr $^{-1}$
<i>f</i> -ratio	0.20	0.15	0.15	-
CaCO ₃ to POC ratio	0.06	0.13	0.14	-
$T_{eff-1km}$	0.06	0.24	0.24	-

Table 4. Annual and cumulated ocean carbon uptakes simulated by NorESM1, NorESM2, and from observation-based estimates.

	NorESM1-ME	NorESM2-LM	NorESM2-MM	Observations
Annual uptake 1980s [Pg C yr ⁻¹]	2.03	1.80	1.85	1.8±0.8 (Denman et al., 2007)
Annual uptake 1990s [Pg C yr ⁻¹]	2.24	2.04	2.05	2.2±0.4 (Denman et al., 2007)
Cumulated uptake (1850-1994) [Pg C]	127.91	111.02	107.82	111±21* (Gruber et al., 2019)
Cumulated uptake (1994-2007) [Pg C]	33.68	30.57	31.04	29±5 (Gruber et al., 2019)

*1800-1994



Figure 1. Schematic diagram of the ecosystem module in the ocean biogeochemical component of the NorESM2 model. The diagram is an updated version of a similar diagram in Six and Maier-Reimer (2006). Blue <u>colors depict componentdepicts components</u>, processes, and parameters that have been modified in NorESM2 relative to the previous model version NorESM1-ME. <u>Dashed-arrows indicate air-sea</u> interactions, while wavy arrows depict riverine inputs.



Figure 2. Summary of (lefta,c) spatial correlation coefficients and (rightb,d) normalized-RMSE between the observations and models (NorESM1-ME, NorESM2-LM and NorESM2-MM). In all (topa,b) 3D fields, the metric values are computed over a 2D horizontal global domain ranging from the surface to 3km depth for 3D fields, while for (bottomc,d) seasonal metrics, only surface values are evaluated. Each square contains three metric scores, i.e., for NorESM1-ME (top), NorESM2-LM (bottom-left), and NorESM2-MM (bottom-right). Black dots indicate which model performes the best.



Figure 3. Climatological temperature values at the (a) surface and across the vertical sections of (b) Atlantic and (c) Pacific in NorESM2-MM (color-shadings) and from observations (contour lines). Difference Differences between NorESM2-MM (or NorESM1-ME) and observations are shown in panels (d)-(i).



Figure 4. Climatological salinity values at the (a) surface and across the vertical sections of (b) Atlantic and (c) Pacific in NorESM2-MM (color-shadings) and from observations (contour lines). Difference Differences between NorESM2-MM (or NorESM1-ME) and observations are shown in panels (d)-(i).



Figure 5. Climatological seasonal mixed layer depth as (left-column) simulated in NorESM2-MM and (right-column) estimated from observations.



Figure 6. Concentration of CFC-11 across the vertical sections of (a) Atlantic and (b) Pacific from NorESM2-MM (color-shadings) and observations (contour lines). Differences between NorESM2-MM and observations are shown in panels (c)-(d).



Figure 7. Climatological phosphate values at the (a) surface and across the vertical sections of (b) Atlantic and (c) Pacific in NorESM2-MM (color-shadings) and from observations (contour lines). Difference Differences between NorESM2-MM (or NorESM1-ME) and observations are shown in panels (d)-(i).



Figure 8. Climatological nitrate values at the (a) surface and across the vertical sections of (b) Atlantic and (c) Pacific in NorESM2-MM (color-shadings) and from observations (contour lines). Difference Differences between NorESM2-MM (or NorESM1-ME) and observations are shown in panels (d)-(i).



Figure 9. Climatological silicate values at the (a) surface and across the vertical sections of (b) Atlantic and (c) Pacific in NorESM2-MM (color-shadings) and from observations (contour lines). Difference Differences between NorESM2-MM (or NorESM1-ME) and observations are shown in panels (d)-(i).



Figure 10. Maps of limiting nutrients for biological productivity as simulated in (a) NorESM1 and (b) NorESM2. The limiting nutrient is determined as the minimum between PO_4 , $R_{P:N} \cdot NO_3$, and $R_{P:Fe} \cdot Fe$.



Figure 11. Climatological dissolved oxygen values at the (a) surface and across the vertical sections of (b) Atlantic and (c) Pacific in NorESM2-MM (color-shadings) and from observations (contour lines). Difference Differences between NorESM2-MM (or NorESM1-ME) and observations are shown in panels (d)-(i).



Figure 12. Climatology of apparent oxygen utilization (AOU) values across the vertical sections of (a) Atlantic and (b) Pacific in NorESM2 (color-shadings) and from observations (contour lines). Differences between NorESM2-MM (or NorESM1-ME) and observations are shown in panels (c)-(f).



Figure 13. Annual mean (a) vertically-integrated contemporary primary production and (b) transfer efficiency ($T_{eff-1km}$) as simulated in NorESM2 (color-shadings) and from observations (contour-lines). Respective difference between NorESM2-MM (NorESM1-ME) and observations are shown in color-shadings (contour-lines) in panels (c)-(d).



Figure 14. Mean seasonal cycle of vertically-integrated contemporary primary production in the (a) North Atlantic (north of 20° N), (b) North Pacific (north of 20° N), (c) tropics (20° S- 20° N), (d) mid-latitude Southern Ocean southern hemisphere (20° S- 40° S), (e) Southern Ocean (40° S- 60° S), and (f) polar Southern Ocean (south of 60° S). Shown are values from NorESM1-ME (black), NorESM2-MM (red) and observational estimates (blue). Numbers within each panel depict the regional mean values averaged for over all months.



Figure 15. Seasonally-averaged surface concentration of DMS from observations (a-d) and as simulated in NorESM2-MM (color-shadings) and from observations (contour-linesc-h).



Figure 16. Climatological dissolved inorganic carbon (DIC) values at the (a) surface and across the vertical sections of (b) Atlantic and (c) Pacific in NorESM2-MM (color-shadings) and from observations (contour lines). Difference-Differences between NorESM2-MM (or NorESM1-ME) and observations are shown in panels (d)-(i).



Figure 17. Climatological alkalinity values at the (a) surface and across the vertical sections of (b) Atlantic and (c) Pacific in NorESM2-MM (color-shadings) and from observations (contour lines). Difference Differences between NorESM2-MM (or NorESM1-ME) and observations are shown in panels (d)-(i).


Figure 18. Surface Observed surface climatology maps of (a) partial pressure of CO₂ and (c) sea-air CO₂ fluxes together with Hovmöller plots of zonally-averaged seasonal cycle of (b) surface pCO₂ and (d) sea-air CO₂fluxes as simulated in NorESM2-MM (color-shadings) and . <u>Corresponding values</u> from observations NorESM2-MM simulation are shown in panels (contour-linese-g). 51



Figure 19. Mean seasonal cycle of vertically-integrated contemporary sea-to-air CO_2 fluxes in the (a) North Atlantic (north of 20°N), (b) North Pacific (north of 20°N), (c) tropics (20°S-20°N), (d) mid-latitude Southern Ocean southern hemisphere (20°S-40°S), (e) Southern Ocean (40°S-60°S), and (f) polar Southern Ocean (south of 60°S). Shown are values from NorESM1-ME (black), NorESM2-MM (red) and observational estimates (blue). Numbers within each panel depict the regional mean values averaged for over all months.



Figure 20. Time-series of global mean annual (a) surface air temperature anomalies (relative to the 1850-1879 periods) period, (b) surface ocean and atmospheric pCO_2 , (c) ocean CO_2 uptake, and (d) atmospheric CO_2 concentration. Shown are values from NorESM2-MM (grey) preindustrial control and (orange) historical simulations. Green colors depict values depicts results from the emissions prescribed historical simulations simulation with NorESM2-LM. Purple The purple line in panel (c) represent represents the anthropogenic carbon uptake estimated by substracting the natural carbon uptake from the total uptake (see also Sect. Section 3.2). Data for temperature and ocean carbon sinks are estimates from HadCRUT4 and the Global Carbon Project (Morice et al., 2012; Le Quéré et al., 2018), respectively, in blue colors.



Figure 21. Climatological δ^{13} C values at (a) 500 m depth and across the vertical sections of (b) Atlantic and (c) Pacific from NorESM2-OC (color-shadings) and observations (contour lines). Difference Differences between NorESM2-OC and observations are shown in color-shadings in panels (d)-(f).



Figure 22. Climatological fields of the biological component of δ^{13} C (i.e., δ^{13} C_{*BIO*}) at (a) 500 m depth and across the vertical sections of (b) Atlantic and (c) Pacific from NorESM2-OC (color-shadings) and observations (contour lines). Difference Differences between NorESM2-OC and observations are shown in color-shadings in panels (d)-(f).



Figure 23. Climatology Δ^{14} C values at (a) 500 m depth and across the vertical sections of (b) Atlantic and (c) Pacific from NorESM2 (color-shadings) and observations (contour lines). Difference Differences between NorESM2 and observations are shown in color-shadings in panels (d)-(f).

Appendix A

5

Author contributions. JT, JS, AM, MB, CH, IB, and SG contributed to developing and implementing the iHAMOCC model code changes. JT, JS, MB, AM, AG, DO, and ØS designed, tested, and performed the experiments described in the manuscript. JT, JS, AM, and YH analyzed the model simulations. All authors contributed to the writing of the manuscript.

Competing interests. The authors declare that they have no conflict of interest.

Acknowledgements. The authors thank Joachim Segschneider and Julien Palmieri for their valuable and constructive feedbacks on the manuscript. We acknowledge support from the Research Council of Norway funded projects EVA (229771), INES (270061), and Key-Clim (295046), Horizon 2020 European Union's Framework Programme for Research and Innovation project CRESCENDO (Coordinated

- 10 Research in Earth Systems and Climate: Experiments, Knowledge, Dissemination and Outreach, no. 641816), and funding from the Bjerknes Centre for Climate Research. JT also acknowledges project Biodiversa (295340) and COLUMBIA (275268). High performance computing and storage resources were provided by the Norwegian infrastructure for computational science (through projects nn2345k, nn2980k, nn1002k, nn9560k, ns2345k, ns2980k, ns1002k, and ns9560k). We also acknowledge Johan Liakka for providing technical assistance and Jöran Maerz for the Teff – 1km data. *Disclaimer.*This article reflects only the authors' view – the funding agencies as well as their
- 15 executive agencies are not responsible for any use that may be made of the information that the article contains.

References

- Arora Vivek K., Boer, G. J., Friedlingstein, P., Eby, M., Jones, C., Christian, J., Bonan, G., Bopp, L., Brovkin, V., Cadule, P., Hajima, T., Ilyina, T., Lindsay, K., Tjiputra, J., and Wu., T.: Carbon Concentration and Carbon–Climate Feedbacks in CMIP5 Earth System Models, J. Climate, 26, 5289–5314, 2013.
- 5 Assmann, K. M., Bentsen, M., Segschneider, J., and Heinze, C.: An isopycnic ocean carbon cycle model, Geosci. Model Dev., 3, 143–167, https://doi.org/10.5194/gmd-3-143-2010, 2010.
 - Bakker, D. C. E., Pfeil, B., Smith, K., Hankin, S., Olsen, A., Alin, S. R., Cosca, C., Harasawa, S., Kozyr, A., Nojiri, Y., O'Brien, K. M., Schuster, U., Telszewski, M., Tilbrook, B., Wada, C., Akl, J., Barbero, L., Bates, N. R., Boutin, J., Bozec, Y., Cai, W.-J., Castle, R. D., Chavez, F. P., Chen, L., Chierici, M., Currie, K., de Baar, H. J. W., Evans, W., Feely, R. A., Fransson, A., Gao, Z., Hales, B., Hardman-
- Mountford, N. J., Hoppema, M., Huang, W.-J., Hunt, C. W., Huss, B., Ichikawa, T., Johannessen, T., Jones, E. M., Jones, S. D., Jutterström, S., Kitidis, V., Körtzinger, A., Landschützer, P., Lauvset, S. K., Lefévre, N., Manke, A. B., Mathis, J. T., Merlivat, L., Metzl, N., Murata, A., Newberger, T., Omar, A. M., Ono, T., Park, G.-H., Paterson, K., Pierrot, D., Ríos, A. F., Sabine, C. L., Saito, S., Salisbury, J., Sarma, V. V. S. S., Schlitzer, R., Sieger, R., Skjelvan, I., Steinhoff, T., Sullivan, K. F., Sun, H., Sutton, A. J., Suzuki, T., Sweeney, C., Takahashi, T., Tjiputra, J., Tsurushima, N., van Heuven, S. M. A. C., Vandemark, D., Vlahos, P., Wallace, D. W. R., Wanninkhof, R., and Watson, A. J.:
- 15 An update to the Surface Ocean CO2 Atlas (SOCAT version 2), Earth Syst. Sci. Data, 6, 69–90, https://doi.org/10.5194/essd-6-69-2014, 2014.
 - Bauska, T. K., Baggenstos, D., Brook, E. J., Mix, A. C., Marcott, S. A., Petrenko, V. V., Schaefer, H., Severinghaus, J. P., and Lee, J. E.: Carbon isotopes characterize rapid changes in atmospheric carbon dioxide during the last deglaciation, P Natl Acad Sci USA, 113, 3465–3470, https://doi.org/10.1073/pnas.1513868113, 2016.
- 20 Behrenfeld, M. and Falkowski, P.: Photosynthetic rates derived from satellite-based chlorophyll concentration, Limnol. Oceanogr., 42, 1–20, 1997.
 - Bentsen, M., Bethke, I., Debernard, J. B., Iversen, T., Kirkevåg, A., Seland, Ø., Drange, H., Roelandt, C., Seierstad, I. A., Hoose, C., and Kristjánsson, J. E.: The Norwegian Earth System Model, NorESM1-M - Part 1: Description and basic evaluation of the physical climate. Geoscientific Model Development, 6(3), 687–720, https://doi.org/10.5194/gmd-6-687-2013, 2013.
- 25 Bernard, C. Y., Dürr, H. H., Heinze, C., Segschneider, J., and Maier-Reimer, E.: Contribution of riverine nutrients to the silicon biogeochemistry of the global ocean – a model study, Biogeosciences, 8, 551–564, 2011.
 - Bernardello, R., Marinov, I., Palter, J. B., Sarmiento, J. L., Galbraith, E. D., and Slater, R. D.: Response of the ocean natural carbon storage to projected twenty-first-century climate change, J. Climate, 27(5), 2033–2053, https://doi.org/10.1175/JCLI-D-13-00343.1, 2014.

Bony, S., Stevens, B., Held, I., Mitchell, J., Dufresne, J. L., Emanuel, K., Friedlingstein, P., Griffies, S. and Senior, S.: Carbon Dioxide and Climate: Perspectives on a Scientific Assessment, Position paper prepared for the WCRP Open Science Conference, Denver, USA, Oct

- 30 Climate: Perspectives on a Scientific Assessment, Position paper prepared for the WCRP Open Science Conference, Denver, USA, Oct 24-28, 21pp., 2011.
 - Bopp, L., Resplandy, L., Orr, J. C., Doney, S. C., Dunne, J. P., Gehlen, M., Halloran, P., Heinze, C., Ilyina, T., Séférian, R., Tjiputra, J., and Vichi, M.: Multiple stressors of ocean ecosystems in the 21st century: projections with CMIP5 models, Biogeosciences, 10, 6225–6245, https://doi.org/10.5194/bg-10-6225-2013, 2013.
- 35 Boyd, P. W. and Ellwood, M. J.: The biogeochemical cycle of iron in the ocean, Nature Geoscience, 3, 675–682, 2010.
 Braconnot, P., Harrison, S. P., Kageyama, M., Bartlein, P. J., Masson-Delmotte, V., Abe-Ouchi, A., Otto-Bliesner, B., and Zhao., Y.: Evaluation of climate models using paleoclimatic data, Nature Climate Change, 2, 417–424, https://doi.org/10.1038/nclimate1456, 2012.

- Breitbarth, E., Oschlies, A., and LaRoche, J.: Physiological constraints on the global distribution of Trichodesmium effect of temperature on diazotrophy, Biogeosciences, 4, 53–61, https://doi.org/10.5194/bg-4-53-2007, 2007.
- Bretherton, F. P.: Earth System Science and Remote-Sensing, P. IEEE, 73, 1118-1133, https://doi.org/10.1109/Proc.1985.13242, 1985.
- 5 Broecker W, S., and Peng, T. H.: Gas exchange rates between air and sea, Tellus, 26, 21–35, https://doi.org/10.1111/j.2153-3490.1974.tb01948.x, 1974.
 - Broecker, W. S.: "NO" a conservative water-mass tracer, Earth Planet. Sci. Lett., 23(1), 100-107, https://doi.org/10.1016/0012-821X(74)90036-3, 1974.
 - Broecker, W. S., and Maier-Reimer E.: The influence of air and sea exchange on the carbon isotope distribution in the sea, Global Bio-

10 geochem. Cycles, 6(3), 315–320, https://doi.org/10.1029/92GB01672, 1992.

- Broecker, W. S., and McGee, D.: The 13C record for atmospheric CO2: What is it trying to tell us?, Earth and Planetary Science Letters, 368, 175–182, https://doi.org/10.1016/j.epsl.2013.02.029, 2013.
 - Bruland, K.W., Middag, R., and Lohan, M.C.: 8.2 Controls of Trace Metals in Seawater. In: H.D. Holland and K.K. Turekian (Editors), Treatise on Geochemistry (Second Edition). Elsevier, Oxford, pp. 19–51, 2014.
- 15 Bullister, J. L., Wisegarver, D. P., and Menzia, F. A.: The solubility of sulfur hexafluoride in water and seawater, Deep-Sea Res. Pt. I, 49, 175–187, 2002.

Bullister, J.: Updated (2014) Atmospheric CFC-11, CFC-12, CFC-113, CCl4 and SF6 Histories, available at: http://cdiac.ornl.gov/ftp/oceans/ CFC_ATM_Hist/CFC_ATM_Hist_2014, last access: 5 December 2014.

Cabré, A., Marinov, I., Bernardello, R., and Bianchi, D.: Oxygen minimum zones in the tropical Pacific across CMIP5 models: mean state

- differences and climate change trends, Biogeosciences, 12, 5429–5454, https://doi.org/10.5194/bg-12-5429-2015, 2015.
 Chester, R.: Marine Geochemistry, 1st ed., 702p, Springer, Netherlands, 1990.
 - Crucifix, M.: Distribution of carbon isotopes in the glacial ocean: A model study, Paleoceanography, 20, PA4020, https://doi.org/10.1029/2005PA001131, 2005.

Cubasch, U., Wuebbles, D., Chen, D., Facchini, M. C., Frame, D., Mahowald, N., and Winther, J.-G.: Introduction. In: Climate Change 2013:

- 25 The Physical Science Basis. Contribution of Working Group I to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change [Stocker, T.F., Qin, D., Plattner, G.-K., Tignor, M., Allen, S. K., Boschung, J., Nauels, A., Xia, Y., Bex, V. and Midgley, P. M. (eds.)]. Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA, 2013.
 - Curry, W. B., and Oppo, D. W.: Glacial water mass geometry and the distribution of δ 13C of Σ CO2 in the western Atlantic Ocean, Paleoceanography, 20, PA1017, doi:10.1029/2004PA001021, 2005.
- 30 de Baar, H. J. W., de Jong, J. T. M., Bakker, D. C. E., Löscher, B. M., Veth, C., Bathmann, U., and Smetacek, V.: Importance of iron for phytoplankton spring blooms and CO₂ drawdown in the Southern Ocean, Nature, 373, 412–415, 1995.
 - de Boyer Montégut, C., Madec, G., Fischer, A. S., Lazar, A., and Iudicone, D.: Mixed layer depth over the global ocean: an examination of profile data and a profile-based climatology, J. Geophys. Res., 109, 12003, doi:10.1029/2004JC002378, 2004.

Denman, K. L., Brasseur, G., Chidthaisong, A., Ciais, P., Cox, P. M., Dickinson, R. E., Hauglustaine, D., Heinze, C., Holland, E., Jacob, D.,

35 Lohmann, U., Ramachandran, S., da Silva Dias, P. L., Wofsy, S. C., and Zhang, X.: Couplings Between Changes in the Climate System and Biogeochemistry, in: Climate Change 2007: The Physical Science Basis, Contribution of Working Group I to the Fourth Assessment Report of the Intergovernmental Panel on Climate Change, edited by Solomon, S., Qin, D., Manning, M., Chen, Z., Marquis, M., Averyt, K. B., Tignor, M., and Miller, H. L., Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA, 2007.

- Dunne, J. P., Sarmiento, J. L., and Gnanadesikan, A.: A synthesis of global particle export from the surface ocean and cycling through the ocean interior and on the seafloor, Global Biogeochem. Cycles, 21, GB4006, https://doi.org/10.1029/2006GB002907, 2007.
- Duteil, O., Koeve, W., Oschlies, A., Aumont, O., Bianchi, D., Bopp, L., Galbraith, E., Matear, R., Moore, J. K., Sarmiento, J. L., and Segschneider, J.: Preformed and regenerated phosphate in ocean general circulation models: can right total concentrations be wrong?,
- 5 Biogeosciences, 9, 1797–1807, https://doi.org/10.5194/bg-9-1797-2012, 2012.
- Eide, M., Olsen, A., Ninnemann, U. S., and Johannessen, T.: A global ocean climatology of preindustrial and modern ocean δ13C, Global Biogeochem. Cycles, 31, 515–534, https://doi.org/10.1002/2016GB005473, 2017.
- Eggleston, S. and Galbraith, E. D.: The devil's in the disequilibrium: sensitivity of ocean carbon storage to climate state and iron fertilization in a general circulation model, Biogeosciences, 15, 3761-3777, https://doi.org/10.5194/bg-15-3761-2018, 2018.
- 10 Eyring, V., Bony, S., Meehl, G. A., Senior, C. A., Stevens, B., Stouffer, R. J., and Taylor, K. E.: Overview of the Coupled Model Intercomparison Project Phase 6 (CMIP6) experimental design and organization, Geosci. Model Dev., 9, 1937–1958, https://doi.org/10.5194/gmd-9-1937-2016, 2016a.
 - Eyring, V., Gleckler, P. J., Heinze, C., Stouffer, R. J., Taylor, K. E., Balaji, V., Guilyardi, E., Joussaume, S., Kindermann, S., Lawrence, B. N., Meehl, G. A., Righi, M., and Williams, D. N.: Towards improved and more routine Earth system model evaluation in CMIP, Earth
- 15 Syst. Dynam., 7, 813–830, https://doi.org/10.5194/esd-7-813-2016, 2016b.

30

- Fassbender, A. J., Rodgers, K. B., Palevsky, H. I., and Sabine, C. L.: Seasonal asymmetry in the evolution of surface ocean pCO₂ and pH thermodynamic drivers and the influence on sea-air CO₂ flux, Global Biogeochem. Cycles, 32, 1476–1497, https://doi.org/10.1029/2017GB005855, 2018.
- Flato, G., Marotzke, J., Abiodun, B., Braconnot, P., Chou, S. C., Collins, W., Cox, P., Driouech, F., Emori, S., Eyring, V., Forest, C., Gleckler,
- 20 P., Guilyardi, E., Jakob, C., Kattsov, V., Reason, C. and Rummukainen, M.: Evaluation of climate models, in Climate Change 2013: The Physical Science Basis, Contribution of Working Group I to the IPCC AR5, pp. 741–882, Cambridge Univ. Press, Cambridge, U. K., and New York, 2013.

Friedlingstein, P., Cox, P., Betts, R., Bopp, L., et al.: Climate-carbon cycle feedback analysis: Results from the C⁴MIP model intercomparison,
 J. Climate, 19, 3337–3353, 2006.

Garcia, H. E. and Gordon, L. I.: Oxygen solubility in seawater: Better fitting equations, Limnol. Oceanogr., 37, 6, 1307–1312, 1992.

Garcia H. E., Boyer, T. P., Locarnini, R. A., Mishonov, A. V., Antonov, J. I., Baranova, O. K., Zweng, M. M., Reagan, J. R., and Johnson, D. R.: World Ocean Atlas 2013, Volume 3: Dissolve Oxygen, Apparent Oxygen Utilization, and Oxygen Saturation. S. Levitus, Ed., A. Mishonov technical editor, NOAA Atlas NESDIS 75, 29 pp., 2013a.

- Garcia H. E., Boyer, T. P., Locarnini, R. A., Mishonov, A. V., Antonov, J. I., Baranova, O. K., Zweng, M. M., Reagan, J. R., and Johnson, D. R.: World Ocean Atlas 2013, Volume 4: Dissolved Inorganic Nutrients (phosphate, nitrate, silicate). S. Levitus, Ed., A. Mishonov technical editor, NOAA Atlas NESDIS 76., 2013b.
 - Gehlen, M., Séférian, R., Jones, D. O. B., Roy, T., Roth, R., Barry, J., Bopp, L., Doney, S. C., Dunne, J. P., Heinze, C., Joos, F., Orr, J.
- 35 C., Resplandy, L., Segschneider, J., and Tjiputra, J.: Projected pH reductions by 2100 might put deep North Atlantic biodiversity at risk, Biogeosciences, 11, 6955–6967, https://doi.org/10.5194/bg-11-6955-2014, 2014.

Gent, P. R. and McWilliams, J. C.: Isopycnal mixing in ocean circulation models, J. Phys. Oceanogr., 20, 150–155, 1990.

Fox-Kemper, B., Ferrari, R., and Hallberg, R.: Parameterization of Mixed Layer Eddies. Part I: Theory and Diagnosis. J. Phys. Oceanogr., 38, 1145–1165., 2008.

- Gharamti, M. E., Tjiputra, J., Bethke, I., Samuelsen, A., Skjelvan, I., Bentsen, M., and Bertino, L.: Ensemble data assimilation for ocean biogeochemical state and parameter estimation at different sites, Ocean MOdelling, 112, 65–89, http://dx.doi.org/10.1016/j.ocemod.2017.02.006, 2017.
- Giupponi, C., Borsuk, M. E., de Vries, B. J. M., and Hasselmann, K.: Innovative approaches to integrated global change modelling, Envi ronmental Modelling and Software, 44, 1–9, http://dx.doi.org/10.1016/j.envsoft.2013.01.013, 2013.
- Goris, N., Tjiputra, J. F., Olsen, A., Schwinger, J., Lauvset, S. K., and Jeansson, E.: Constraining Projection-Based Estimates of the Future North Atlantic Carbon Uptake. J. Climate, 31, 3959–3978, https://doi.org/10.1175/JCLI-D-17-0564.1, 2018.
- Grossman, E. L., and Ku, T.-L.: Oxygen and carbon isotope fractionation in biogenic aragonite: temperature effects, Chemical Geology: Isotope Geoscience section, 59, 59–74, 1986.
- 10 Gruber, N. and Keeling, C. D.: An improved estimate of the isotopic air-sea disequilibrium of CO2: Implications for the oceanic uptake of anthropogenic CO2, Geophysical Research Letters, 28, 555–558, https://doi.org/10.1029/2000GL011853, 2001.
 - Gruber, N., Clement, D., Carter, B. R., Feely, R. A., van Heuven, S., Hoppema, M., Ishii, M., Key, R. M., Kozyr, A., Lauvset, S. K., Lo Monaco, C., Mathis, J. T., Murata, A., Olsen, A., Perez, F. F., Sabine, C. L., Tanhua, T., and Wanninkhof, R.: The oceanic sink for anthropogenic CO₂ from 1994 to 2007, Science, 363, 1193–1199, https://doi.org/10.1126/science.aau5153, 2019.
- 15 Guo, C., Bentsen, M., Bethke, I., Ilicak, M., Tjiputra, J., Toniazzo, T., Schwinger, J., and Otterå, O. H.: Description and evaluation of NorESM1-F: a fast version of the Norwegian Earth System Model (NorESM), Geosci. Model Dev., 12, 343–362, https://doi.org/10.5194/gmd-12-343-2019, 2019.
 - Hartmann, J.: Bicarbonate-fluxes and CO2-consumption by chemical weathering on the Japanese Archipelago Application of a multilithological model framework, Chemical Geology, 265, 237–271, 2009.
- 20 Hauck, J. and C. Völker: Rising atmospheric CO₂ leads to large impact of biology on Southern Ocean CO₂ uptake via changes of the Revelle factor, Geophys. Res. Lett., 42, 1459–1464,10.1002/2015GL063070, 2015.
 - Henson, S. A., Beaulieu, C., Ilyina, T., John, J. G., Long, M., Séférian, R., Tjiputra, J., and Sarmiento, J. L.: Rapid emergence of climate change in environmental drivers of marine ecosystem, Nature Communications, 8, 14682, https://doi.org/10.1038/ncomms14682, 2017.
 - Heinze, C., Eyring, V., Friedlingstein, P., Jones, C., Balkanski, Y., Collins, W., Fichefet, T., Gao, S., Hall, A., Ivanova, D., Knorr, W., Knutti,
- 25 R., Löw, A., Ponater, M., Schultz, M. G., Schulz, M., Siebesma, P., Teixeira, J., Tselioudis, G., and Vancoppenolle, M., ESD Reviews: Climate feedbacks in the Earth system and prospects for their evaluation, Earth Syst. Dynam., 10, 379–452, https://doi.org/10.5194/esd-10-379-2019, 2019.
 - Holden, P. B., Edwards, N. R., Müller, S. A., Oliver, K. I. C., Death, R. M., and Ridgwell, A.: Controls on the spatial distribution of oceanic δ13C_{DIC}, Biogeosciences, 10, 1815–1833, https://doi.org/10.5194/bg-10-1815-2013, 2013.
- 30 Hoogakker, B. A. A., Elderfield, H., Schmiedl, G., McCave, I. N., and Rickaby, R. E. M.: Glacial-interglacial changes in bottom-water oxygen content on the Portuguese margin, Nature Geosci., 8, 40-43, https://doi.org/10.1038/ngeo2317, 2015.
 - Ilyina, T., Six, K. D., Segschneider, J., Maier-Reimer, E., Li, H., and Núñez Riboni, I.: Global ocean biogeochemistry model HAMOCC: model architecture and performance as component of the MPI-Earth System Model in different CMIP5 experimental realizations, J. Adv. Model. Earth Syst., 5, 287–315, https://doi.org/10.1029/2012MS000178, 2013.
- 35 Ito, T., Follows, M. J., and Boyle, E. A.: Is AOU a good measure of respiration in the oceans?, Geophys. Res. Lett., 31, L17305, https://doi.org/10.1029/2004GL020900, 2004.

- Jahn, A., Lindsay, K., Giraud, X., Gruber, N., Otto-Bliesner, B. L., Liu, Z., and Brady, E. C.: Carbon isotopes in the ocean model of the Community Earth System Model (CESM1), Geoscientific Model Development, 8, 2419-2434, https://doi.org/10.5194/gmd-8-2419-2015, 2015.
- Jochum, M., Briegleb, B. P., Danabasoglu, G., Large, W. G., Norton, N. J., Jayne, S. R., Alford, M. H., and Bryan, F. O.: The Impact of
 Oceanic Near-Inertial Waves on Climate. J. Climate, 26, 2833–2844, doi:10.1175/JCLI-D-12-00181.1. 2013.
- Jones, C. D., Arora, V., Friedlingstein, P., Bopp, L., Brovkin, V., Dunne, J., Graven, H., Hoffman, F., Ilyina, T., John, J. G., Jung, M., Kawamiya, M., Koven, C., Pongratz, J., Raddatz, T., Randerson, J. T., and Zaehle, S.: C4MIP – The Coupled Climate–Carbon Cycle Model Intercomparison Project: experimental protocol for CMIP6, Geosci. Model Dev., 9, 2853–2880, https://doi.org/10.5194/gmd-9-2853-2016, 2016.
- 10 Keller, M., Bellows, W., and Guillard, R.: Dimethyl sulfide production in marine phytoplankton, in: Biogenic sulfur in the environment, edited by: Saltzman, E. and Cooper, W., ACS-Symposium series, New Orleans, Louisiana, American Chemical Society, 167–181, 1989.
 - Keller, K., and Morel, F. M. M.: A model of carbon isotopic fractionation and active carbon uptake in phytoplankton, Marine Ecology Progress Series, 182, 295–298, https://doi.org/10.3354/meps182295, 1999.

Kessler, A. and Tjiputra, J.: The Southern Ocean as a constraint to reduce uncertainty in future ocean carbon sinks, Earth Syst. Dynam., 7,

15 295–312, https://doi.org/10.5194/esd-7-295-2016, 2016.

- Kessler, A., Galaasen, E. V., Ninnemann, U. S., and Tjiputra, J.: Ocean carbon inventory under warmer climate conditions the case of the Last Interglacial, Clim. Past, 14, 1961–1976, https://doi.org/10.5194/cp-14-1961-2018, 2018.
 - Key, R. M., Kozyr, A., Sabine, C. L., Lee, K., Wanninkhof, R., Bullister, J. L., Feely, R. A., Millero, F. J., Mordy, C., and Peng, T. H.: A global ocean carbon climatology: Results from Global Data Analysis Project (GLODAP), Global Biogeochem. Cycles, 18, GB4031,
- 20 https://doi.org/10.1029/2004GB002247, 2004.
 - Kloster, S., Feichter, J., Maier-Reimer, E., Six, K. D., Stier, P., and Wetzel, P.: DMS cycle in the marine ocean-atmosphere system a global model study, Biogeosciences, 3, 29–51, https://doi.org/10.5194/bg-3-29-2006, 2006.
 - Koeve, W.: Upper ocean carbon fluxes in the Atlantic Ocean: The importance of the POC:PIC ratio, Global Biogeochem. Cycles, 1(4), 1056, https://doi.org/10.1029/2001GB001836, 2002.
- 25 Kriest, I.: Different parameterizations of marine snow in a 1-D model and their influence on representation of marine snow, nitrogen budget and sedimentation, Deep-Sea Res. Pt. I, 49, 2133–2162, 2002.
 - Kriest, I. and Evans, G.: Representing phytoplankton aggregates in biogeochemical models, Deep-Sea Res. Pt. I, 46, 1841–1859, 1999.
 - Kriest, I. and Oschlies, A.: On the treatment of particulate organic matter sinking in large-scale models of marine biogeochemical cycles, Biogeosciences, 5, 55–72, https://doi.org/10.5194/bg-5-55-2008, 2008.
- 30 Kriest, I. and Oschlies, A.: MOPS-1.0: towards a model for the regulation of the global oceanic nitrogen budget by marine biogeochemical processes, Geosci. Model Dev., 8, 2929-2957, https://doi.org/10.5194/gmd-8-2929–2015, 2015.
 - Kwiatkowski, L., Bopp, L., Aumont, O., Ciais, P., Cox, P. M., Laufkötter, C., Li, Y., and Séférian, R.: Emergent constraints on projections of declining primary production in the tropical oceans, Nature Climate Change, 7, 355–359, https://doi.org/10.1038/NCLIMATE3265, 2017.
 Lana, A., Bell, T. G., Simó, R., Vallina, S. M., Ballabrera-Poy, J., Kettle, A. J., Dachs, J., Bopp, L., Saltzman, E. S., Stefels, J., Johnson,
- J. E., and Liss, P. S.: An updated climatology of surface dimethlysulfide concentrations and emission fluxes in the global ocean, Global Biogeochem. Cycles, 25, GB1004, https://doi.org/10.1029/2010GB003850, 2011.
 - Landschützer, P., Gruber, N., Bakker, D., and Schuster, U.: Recent variability of the global ocean carbon sink, Global Biogeochem. Cycles, 28, 927–949, https://doi.org/10.1002/2014GB004853, 2014.

- Landschützer, P., Gruber, N., and Bakker, D. C. E.: A 30 years observation-based global monthly gridded sea surface pCO2 product from 1982 through 2011 (NCEI Accession 0160558). Version 2.2. NOAA National Centers for Environmental Information. Dataset. https://doi: 10.3334/cdiac/otg.spco2_1982_2011_eth_somffn, 2015.
- Landschützer, P., Gruber, N., Bakker, D. C. E., Stemmler, I., and Six, K. D.: Strengthening seasonal marine CO₂ variations due to increasing atmospheric CO₂, Nature Climate Change, 8, 146–150, https://doi.org/10.1038/s41558-017-0057-x, 2018.
- Large, W. and Yeager, S.: Diurnal to Decadal Global Forcing for Ocean and Sea-Ice Models: The Data Sets and Flux Climatologies, Tech. Note NCAR/TN-460+STR, National Center of Atmospheric Research, Boulder, Colorado, USA, 2004.
- Lauvset, S. K., Key, R. M., Olsen, A., van Heuven, S., Velo, A., Lin, X., Schirnick, C., Kozyr, A., Tanhua, T., Hoppema, M., Jutterström, S., Steinfeldt, R., Jeansson, E., Ishii, M., Perez, F. F., Suzuki, T., and Watelet, S.: A new global interior ocean mapped climatology: the 1°x
 1°GLODAP version 2, Earth Syst. Sci. Data, 8, 325–340, https://doi.org/10.5194/essd-8-325-2016, 2016.
- Laws, E. A., Popp, B. N., Bidigare, R. R., Kennicutt, M. C., and Macko, S. A.: Dependence of phytoplankton carbon isotopic composition on growth rate and [CO₂]_{*aq*}: Theoretical considerations and experimental results, Geochimica et Cosmochimica Acta, 59, 1131–1138, http://dx.doi.org/10.1016/0016-7037(95)00030-4, 1995.

Laws, E. A., Bidigare, R., R., and Popp, B. N.: Effects of growth rate and CO2 concentration on carbon isotopic fractionation by the marine diatom Phaeodactylum tricornutum, Limnol. Oceanogr., 42, 1552–1560, 1997.

- Lebehot, A. D., Halloran, P. R., Watson, A. J., McNeall, D., Ford, D. A., Landschützer, P., Lauvset, S. K., and Schuster, U.: Reconciling Observation and Model Trends in North Atlantic Surface CO₂, Global Biogeochem. Cycles, 33 (10), 1204–1222, https://doi.org/10.1029/2019GB006186, 2019.
 - Le Quéré, C., Andrew, R. M., Friedlingstein, P., Sitch, S., Hauck, J., Pongratz, J., Pickers, P. A., Korsbakken, J. I., Peters, G. P., Canadell,
- J. G., Arneth, A., Arora, V. K., Barbero, L., Bastos, A., Bopp, L., Chevallier, F., Chini, L. P., Ciais, P., Doney, S. C., Gkritzalis, T., Goll, D. S., Harris, I., Haverd, V., Hoffman, F. M., Hoppema, M., Houghton, R. A., Hurtt, G., Ilyina, T., Jain, A. K., Johannessen, T., Jones, C. D., Kato, E., Keeling, R. F., Goldewijk, K. K., Landschützer, P., Lefèvre, N., Lienert, S., Liu, Z., Lombardozzi, D., Metzl, N., Munro, D. R., Nabel, J. E. M. S., Nakaoka, S., Neill, C., Olsen, A., Ono, T., Patra, P., Peregon, A., Peters, W., Peylin, P., Pfeil, B., Pierrot, D., Poulter, B., Rehder, G., Resplandy, L., Robertson, E., Rocher, M., Rödenbeck, C., Schuster, U., Schwinger, J., Séférian, R., Skjelvan, I.,
- 25 Steinhoff, T., Sutton, A., Tans, P. P., Tian, H., Tilbrook, B., Tubiello, F. N., van der Laan-Luijkx, I. T., van der Werf, G. R., Viovy, N., Walker, A. P., Wiltshire, A. J., Wright, R., Zaehle, S., and Zheng, B.: Global Carbon Budget 2018, Earth Syst. Sci. Data, 10, 2141–2194, https://doi.org/10.5194/essd-10-2141-2018, 2018.
 - Locarnini, R. A., Mishonov, A. F., Antonov, J. I., Boyer, T. P., Garcia, H. E., Baranova, O. K., Zweng, M. M., Paver, C. R., Reagan, J. R., Johnson, D. R., Hamilton, M., and Seidov, D.: World Ocean Atlas 2013, Volume 1: Temperature. S. Levitus, Ed., A. Mishonov technical
- 30 editor, NOAA Atlas NESDIS 73, 2013.

15

- Luo, Y., Tjiputra, J. F., Guo, C., Zhang, Z., and Lippold, J.: Atlantic deep water circulation during the last interglacial, Sci. Rep.-UK, 8, 4401, https://doi.org/10.1038/s41598-018-22534-z, 2018.
- Lynch-Stieglitz, J., Stocker, T. F., Broecker, W. S., and Fairbanks, R. G.: The influence of air-sea exchange on the isotopic composition of oceanic carbon: Observations and modeling, Global Biogeochem. Cycles, 9, 653–665, https://doi.org/10.1029/95GB02574, 1995.
- 35 Mahowald, N., Baker, A., Bergametti, G., Brooks, N., Duce, R., Jickells, T., Kubilay, N., Prospero, J., and Tegen, I.: Atmospheric global dust cycle and iron inputs to the ocean, Global Biogeochem. Cycles, 19, 4025, https://doi.org/10.1029/2004GB002402, 2005.
 - Maier-Reimer, E.: Geochemical cycles in an ocean general circulation model. Preindustrial tracer distribution, Global Biogeochem. Cycles, 7, 645–677, 1993.

- Maier-Reimer, E., Kriest, I., Segschneider, J., and Wetzel, P.: The HAMburg Ocean Carbon Cycle Model HAMOCC5.1 Technical Description Release 1.1, Berichte zur Erdsystemforschung 14, ISSN 1614-1199, Max Planck Institute for Meteorology, Hamburg, Germany, 50pp, 2005.
- Martin, J., Knauer, G., Karl, D., and Broenkow, W.: VERTEX: Carbon cycling in the Northeast Pacific, Deep-Sea Res., 34, 267–285, 1987.
- Martin, J. H. and Fitzwater, S. E.: Iron deficiency limits phytoplankton growth in the northeast Pacific subarctic, Nature, 331, 341–343, 1988.
 Mayorga, E., Seitzinger, S. P., Harrison, J. A., Dumont, E., Beusen, A. H. W., Bouwman, A. F., Fekete, B. M., Kroeze, C., and Van Drecht, G.: Global Nutrient Export from WaterSheds 2 (NEWS 2): Model development and implementation, Environmental Modelling and Software, 25, 837–853, 2010.
 - Maury, O., Campling, L., Arrizabalaga, H., Aumont, O., Bopp, L., Merino, G., Squires, D., Cheung, W., Goujon, M., Guivarch, C., Lefort, S.,
- 10 Marsac, F., Monteagudo, P., Murtugudde, R., Osterblom, H., Pulvenis, J. F., Ye, Y., and van Ruijven, B. J.: From shared socio-economic pathways (SSPs) to oceanic system pathways (OSPs): Building policy-relevant scenarios for global oceanic ecosystems and fisheries, Global Environ Chang, 45, 203–216, https://10.1016/j.gloenvcha.2017.06.007, 2017.
 - Morée, A. L., Schwinger, J., and Heinze, C.: Southern Ocean controls of the vertical marine δ13C gradient a modelling study, Biogeosciences, 15, 7205–7223, https://doi.org/10.5194/bg-15-7205-2018, 2018.
- 15 Morice, C. P., Kennedy, J. J., Rayner, N. A. and Jones, P. D.: Quantifying uncertainties in global and regional temperature change using an ensemble of observational estimates: the HadCRUT4 dataset. Journal of Geophysical Research, 117, D08101, https://doi.org/10.1029/2011JD017187, 2012.
- Nevison, C. D., Manizza, M., Keeling, R. F., Kahru, M., Bopp, L., Dunne, J., Tiputra, J., Ilyina, T., and Mitchell, B. G.: Evaluating the ocean biogeochemical components of Earth system models using atmospheric potential oxygen and ocean color data, Biogeosciences, 12, 193–208, https://doi.org/10.5194/bg-12-193-2015, 2015.
- Paulmier, A., Kriest, I., and Oschlies, A.: Stoichiometries of remineralisation and denitrification in global biogeochemical ocean models. Biogeosciences, 6, 923–935, https://doi.org/10.5194/bg-6-923-2009, 2009.
 - Olsen, A., Key, R. M., van Heuven, S., Lauvset, S. K., Velo, A., Lin, X., Schirnick, C., Kozyr, A., Tanhua, T., Hoppema, M., Jutterström, S., Steinfeldt, R., Jeansson, E., Ishii, M., Pérez, F. F., and Suzuki, T.: The Global Ocean Data Analysis Project version 2 (GLODAPv2) – an
- 25 internally consistent data product for the world ocean, Earth Syst. Sci. Data, 8, 297–323, https://doi.org/10.5194/essd-8-297-2016, 2016. Oppenheimer, M., Campos, M., Warren, R., Birkmann, J., Luber, G, O'Neill, B., and Takahashi, K.: Emergent risks and key vulnerabilities. In: Climate Change 2014: Impacts, Adaptation, and Vulnerability. Part A: Global and Sectoral Aspects. Contribution of Working Group II to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change [Field, C. B., Barros, V. R., Dokken, D. J., Mach, K. J., Mastrandrea, M. D., Bilir, T. E., Chatterjee, M., Ebi, K. L., Estrada, Y. O., Genova, R. C., Girma, B., Kissel, E. S., Levy, A. N.,
- 30 MacCracken, S., Mastrandrea, P. R., and White L. L. (eds.)]. Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA, pp. 1039–1099, 2014.
 - Orr, J. C., Najjar, R. G., Aumont, O., Bopp, L., Bullister, J. L., Danabasoglu, G., Doney, S. C., Dunne, J. P., Dutay, J.-C., Graven, H., Griffies, S. M., John, J. G., Joos, F., Levin, I., Lindsay, K., Matear, R. J., McKinley, G. A., Mouchet, A., Oschlies, A., Romanou, A., Schlitzer, R., Tagliabue, A., Tanhua, T., and Yool, A.: Biogeochemical protocols and diagnostics for the CMIP6 Ocean Model Intercomparison Project
- 35 (OMIP), Geosci. Model Dev., 10, 2169-2199, https://doi.org/10.5194/gmd-10-2169-2017, 2017.
- Peterson, C. D., Lisiecki, L. E., and Stern, J. V.: Deglacial whole-ocean δ¹³C change estimated from 480 benthic foraminiferal records, Paleoceanography, 29, 549–563, https://doi.org/10.1002/2013PA002552, 2014.

- Popp, B. N., Laws, E. A., Bidigare, R. R., Dore, J. E., Hanson, K. L., and Wakeham, S. G.: Effect of phytoplankton cell geometry on carbon isotopic fractionation, Geochimica Et Cosmochimica Acta, 62, 69–77, https://doi.org/10.1016/S0016-7037(97)00333-5, 1998
- Quay, P., Sonnerup, R., Westby, T., Stutsman, J., and McNichol, A.: Changes in the 13C/12C of dissolved inorganic carbon in the ocean as a tracer of anthropogenic CO2 uptake, Global Biogeochem. Cycles, 17, 1004, https://doi.org/10.1029/2001GB001817, 2003.
- 5 Rau, G. H., Riebesell, U., and Wolf-Gladrow, D.: A Model Of Photosynthetic 13C Fractionation By Marine Phytoplankton Based On Diffusive Molecular CO2 Uptake, Marine Ecology Progress Series, 133, 275–285, 1996.
 - Rhein, M., Rintoul, S. R., Aoki, S., Campos, E., Chambers, D., Feely, R. A., Gulev, S., Johnson, G. C., Josey, S. AS. A., Kostianoy, A., Mauritzen, C., Roemmich, D., Talley, L. D., and Wang, F.: Observations: Ocean. In: Climate Change 2013: The Physical Science Basis.
 Contribution of Working Group I to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change [Stocker, T.F., et al.
- 10 (eds.)]. Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA, 2013.
- Roth, R., Ritz, S. P., and Joos, F.: Burial-nutrient feedbacks amplify the sensitivity of atmospheric carbon dioxide to changes in organic matter remineralisation, Earth Syst Dynam, 5, 321–343, https://doi.org/10.5194/esd-5-321-2014, 2014.
 - Sarmiento, J. L., Dunne, J., Gnanadesikan, A., Key, R. M., Matsumoto, K., and Slater, R.: A new estimate of the CaCO3 to organic carbon export ratio, Global Biogeochem. Cycles, 16(4), 1107, https://doi.org/10.1029/2002GB001919, 2002.
- 15 Sarmiento, J. L. and Gruber, N., Ocean Biogeochemical Dynamics, Princeton University Press, Princeton, 2006. Schmittner, A., Gruber, N., Mix, A. C., Key, R. M., Tagliabue, A., and Westberry, T. K.: Biology and air-sea gas exchange controls on the distribution of carbon isotope ratios (δ13C) in the ocean, Biogeosciences, 10, 5793–5816, https://doi.org/10.5194/bg-10-5793-2013, 2013.
 - Schwinger, J., Tjiputra, J. F., Heinze, C., Bopp, L., Christian, J. R., Gehlen, M., Ilyina, T., Jones, C. D., Salas-Mélia, D., Segschneider, J.,
- 20 Séférian, R., and Totterdell, I.: Nonlinearity of ocean carbon cycle feedbacks in CMIP5 Earth System Models, J. Climate, 27, 3869–3888, https://doi.org/10.1175/JCLI-D-13-00452.1, 2014
 - Schwinger, J., Goris, N., Tjiputra, J. F., Kriest, I., Bentsen, M., Bethke, I., Ilicak, M., Assmann, K. M., and Heinze, C.: Evaluation of NorESM-OC (versions 1 and 1.2), the ocean carbon-cycle stand-alone configuration of the Norwegian Earth System Model (NorESM1), Geosci. Model Dev., 9, 2589-2622, https://doi.org/10.5194/gmd-9-2589-2016, 2016.
- 25 Schwinger, J., Tjiputra, J., Goris, N., Six, K. D., Kirkevåg, A., Seland, Ø., Heinze, C., and Ilyina, T.: Amplification of global warming through pH dependence of DMS production simulated with a fully coupled Earth system model, Biogeosciences, 14, 3633–3648, https://doi.org/10.5194/bg-14-3633-2017, 2017.
 - Séférian, R., Gehlen, M., Bopp, L., Resplandy, L., Orr, J. C., Marti, O., Dunne, J. P., Christian, J. R., Doney, S. C., Ilyina, T., Lindsay, K., Halloran, P. R., Heinze, C., Segschneider, J., Tjiputra, J., Aumont, O., and Romanou, A.: Inconsistent strategies to spin-up models in CMIP5:
- 30 implications for ocean biogeochemical model performance assessment, Geosci. Model Dev., 9, 1827–1851, https://doi.org/10.5194/gmd-9-1827-2016, 2016.
 - Seitzinger, S. P., Harrison, J. A., Dumont, E., Beusen, A. H. W., and Bouwman, A. F.: Sources and delivery of carbon, nitrogen, and phosphorus to the coastal zone: An overview of Global Nutrient Export from Watersheds (NEWS) models and their application, Global Biogeochem. Cycles, 19, GB4S01, doi:10.1029/2005GB002606, 2005.
- 35 Seland, Ø., Bentsen, M., Seland Graff, L., Olivié, D., Toniazzo, T., Gjermundsen, A., Debernard, J. B., Gupta, A. K., He, Y., Kirkevåg, A., Schwinger, J., Tjiputra, J., Schancke Aas, K., Bethke, I., Fan, Y., Griesfeller, J., Grini, A., Guo, C., Ilicak, M., Hafsahl Karset, I. H., Landgren, O., Liakka, J., Onsum Moseid, K., Nummelin, A., Spensberger, C., Tang, H., Zhang, Z., Heinze, C., Iverson, T., and Schulz.

M.: The Norwegian Earth System Model, NorESM2 – Evaluation of theCMIP6 DECK and historical simulations, Geosci. Model Dev. Discuss., https://doi.org/10.5194/gmd-2019-378, in review, 2020.

- Sholkovitz, E. R., and Copland, D.: The coagulation, solubility and adsorption properties of Fe, Mn, Cu, Ni, Cd, Co and humic acids in a river water, Geochim Cosmochim Ac, 45, 181–189, 1981.
- 5 Shiller, A. M., and Boyle, E. A.: Trace elements in the Mississippi River Delta outflow region: Behavior at high discharge, Geochim Cosmochim Ac, 55, 3241–3251, 1991.
 - Siegel, D. A., Buesseler, K. O., Doney, S: C., Salley, S. F., Behrenfeld, M. J., and Boyd, P. W.: Global assessment of ocean carbon export by combining satellite observations and food-web models, Global Biogeochem. Cycles, 28,181–196, https://doi.org/10.1002/2013GB004743, 2014.
- 10 Six, K. D. and Maier-Reimer, E.: Effects of plankton dynamics on seasonal carbon fluxes in an ocean general circulation model, Global Biogeochem. Cycles, 10, 559–583, 1996.
 - Six, K. D. and Maier-Reimer, E.: What controls the oceanic dimethylsulfide (DMS) cycle? A modeling approach, Global Biogeochem. Cycles, 20, GB4011, https://doi.org/10.1029/2005GB002674, 2006.

Six, K. D., Kloster, S., Ilyina, T., Archer, S. D., Zhang, K., and Maier-Reimer, E.: Global warming amplified by reduced sulphur fluxes as a

- 15 result of ocean acidification, Nature Climate Change, 3, 975–978, https://doi.org/10.1038/NCLIMATE1981, 2013.
 Skinner, L. C., Primeau, F., Freeman, E., de la Fuente, M., Goodwin, P. A., Gottschalk, J., Huang, E., McCave, I. N., Noble, T. L., and Scrivner, A. E.: Radiocarbon constraints on the glacial ocean circulation and its impact on atmospheric CO2, Nature Communications, 8, 16010, https://doi.org/10.1038/ncomms16010, 2017.
- Sonnerup, R. E., and Quay, P. D.: 13C constraints on ocean carbon cycle models, Global Biogeochem. Cycles, 26, GB2014, https://doi.org/10.1029/2010GB003980, 2012.
- Srokosz, M., Baringer, M., Bryden, H., Cunningham, S., Delworth, T., Lozier, S., Marotzke, J., and Sutton, R.: Past, present, and future changes in the Atlantic meridional overturning circulation. Bulletin of the American Meteorological Society, 93(11), 1663–1676. https://doi.org/10.1175/BAMS-D-11-00151.1, 2012.
- Steinacher, M., Joos, F., and Stocker, T. F.: Allowable carbon emissions lowered by multiple climate targets, Nature, 499, 197-201,
- 25 https://doi.org/10.1038/nature12269, 2013.
 - Tagliabue, A., and Bopp, L.: Towards understanding global variability in ocean carbon-13, Global Biogeochem. Cycles, 22, GB1025, https://doi.org/10.1029/2007GB003037, 2008.
 - Tagliabue, A., Aumont, O., DeAth, R., Dunne, J. P., Dutkiewicz, S., Galbraith, E., Misumi, K., Moore, J. K., Ridgwell, A., Sherman, E., Stock, C., Vichi, M., Völker, C., and Yool, A.: How well do global ocean biogeochemistry models simulate dissolved iron distributions?,
- Global Biogeochem. Cycles, 30, 149–174, https://doi.org/10.1029/2015GB005289, 2016.
 Takahashi, T., Sutherland, S., Wanninkhof, W., Sweeney, C., Feely, R. A., Chipman, D. W., Hales, B., Friederich, G., Chavez, F., Sabine, C., Watson, A., Bakker, D. C. E., Schuster, U., Metzl, N., Yoshikawa-Inoue, H., Ishii, M., Midorikawa, T., Nojiri, Y., Körtzinger, A., Steinhoff, T., Hoppema, M., Olafsson, J., Arnarson, T. S., Tilbrook, B., Johannessen, T., Olsen, A., Bellerby, R., Wong, C. S., Delille, B., Bates, N. R., and de Baar, H. J. W.: Climatological mean and decadal change in surface ocean pCO₂, and net sea-air CO₂ flux over the
- 35 global oceans, Deep-Sea Res. II, 56, 554–577, https://doi.org/10.1016/j.dsr2.2008.12.009, 2009.
 - Tjiputra, J. F., Polzin, D., and Winguth, A. M. E.: Assimilation of seasonal chlorophyll and nutrient data into an adjoint three-dimensional ocean carbon cycle model: Sensitivity analysis and ecosystem parameter optimization, Global Biogeochem. Cycles, 21, GB1001, https://doi.org/10.1029/2006GB002745, 2007.

- Tjiputra, J. F., Assmann, K., Bentsen, M., Bethke, I., Otterå, O. H., Sturm, C., and Heinze, C.: Bergen Earth system model (BCM-C): model description and regional climate-carbon cycle feedbacks assessment, Geosci. Model Dev., 3, 123–141, doi:10.5194/gmd-3-123-2010, 2010.
- Tjiputra, J. F., Olsen, A., Assmann, K., Pfeil, B., and Heinze, C.: A model study of the seasonal and long-term North Atlantic surface pCO2
 variability, Biogeosciences, 9, 907–923, https://doi.org/10.5194/bg-9-907-2012, 2012.
- Tjiputra, J. F., Roelandt, C., Bentsen, M., Lawrence, D. M., Lorentzen, T., Schwinger, J., Seland, Ø., and Heinze, C.: Evaluation of the carbon cycle components in the Norwegian Earth System Model (NorESM), Geosci. Model Dev., 6, 301–325, https://doi.org/10.5194/gmd-6-301-2013, 2013.

Tjiputra, J. F., Olsen, A., Bopp, L., Lenton, A., Pfeil, B., Roy, T., Segschneider, J., Totterdell, I., and Heinze, C.: Long-term surface pCO₂ trends from observations and models, Tellus B, 66, https://doi.org/10.3402/tellusb.v66.23083, 2014.

Tjiputra, J. F., Goris, N., Lauvset, S. K., Heinze, C., Olsen, A.,Schwinger, J., and Steinfeldt, R.: Mechanisms and Early Detections of Multidecadal Oxygen Changes in the Interior Subpolar North Atlantic, Geophys. Res. Lett., 45, 4218–4229, https://doi.org/10.1029/2018GL077096, 2018.

Toggweiler, J. R.: Variation of atmospheric CO₂ by ventilation of the ocean's deepest water, Paleoceanography, 14, 571-588,

15

35

https://doi.org/10.1029/1999PA900033, 1999.

10

Wanninkhof, R.: Relationship between wind speed and gas exchange over the ocean revisited, Limnol. Oceanogr.: Methods, 12, 351–362, https://doi.org/10.4319/lom.2014.12.351, 2014.

- Warner, M. and Weiss, R.: Solubilities of chlorofluorocarbons 11 and 12 in water and seawater, Deep-Sea Res. Pt. I, 32, 1485–1497, 1985. Weber, T., Cram, J. A., Leung, S. W., DeVries, T., and Deutsch, C.: Deep ocean nutrients imply large latitudinal variation in particle transfer
- efficiency, Proceedings of the National Academy of Sciences, 113, 13, 8606–8611, https://doi.org/10.1073/pnas.1604414113, 2016
 Weiss, R.: Solubility of nitrogen, oxygen and argon in water and seawater, Deep-Sea Res., 17, 721–735, doi:10.1016/0011-7471(70)90037-9, 1970.

Weiss, R. and Price, B.: Nitrous oxide solubility in water and seawater, Mar. Chem., 8, 374–359, 1980.

Wenzel, S., Cox, P. M., Eyring, V., and Friedlingstein, P.: Emergent constraints on climate-carbon cycle feedbacks in the CMIP5 Earth system
 models, J. Geophys. Res. Biogeosci., 119, 794–807, https://doi.org/10.1002/2013JG002591, 2014.

- Westberry, T., Behrenfeld, M. J., Siegel, D. A., and Boss, E.: Carbon-based primary productivity modeling with vertically resolved photoacclimation, Global Biogeochem. Cycles, 22, GB2024, http://dx.doi.org/10.1029/2007GB003078, 2008.
 - Zeebe, R., and Wolf-Gladrow, D.: CO2 in Seawater: Equilibrium, Kinetics, Isotopes, Elsevier Oceanography Series, edited by: Halpern, D., Elsevier Science B.V., Amsterdam, The Netherlands, 346 pp., 2001.
- 30 Zhang, J., Quay, P. D., and Wilbur, D. O.: Carbon isotope fractionation during gas-water exchange and dissolution of CO2, Geochimica et Cosmochimica Acta, 59, 107–114, http://dx.doi.org/10.1016/0016-7037(95)91550-D, 1995.
 - Zhang, Z. S., Nisancioglu, K., Bentsen, M., Tjiputra, J., Bethke, I., Yan, Q., Risebrobakken, B., Andersson, C., and Jansen, E.: Pre-industrial and mid-Pliocene simulations with NorESM-L, Geosci. Model Dev., 5, 523–533, https://doi.org/10.5194/gmd-5-523-2012, 2012.

Ziveri, P., Stoll, H., Probert, I., Klaas, C., Geisen, M., Ganssen, G., and Young, J: Stable isotope 'vital effects' in coccolith calcite, Earth and Planetary Science Letters, 210, 137–149, https://doi.org/10.1016/S0012-821X(03)00101-8, 2003.

Zweng, M. M, Reagan, J. R., Antonov, J. I., Locarnini, R. A., Mishonov, A. V., Boyer, T. P., Garcia, H. E., Baranova, O. K., Paver, C. R., Johnson, D. R., Seidov, D., and Biddle, M.: World Ocean Atlas 2013, Volume 2: Salinity. S. Levitus, Ed., A. Mishonov, technical editr, NOAA Atlas NESDIS 74, 2013.