Carbon isotopes in the ocean model of the Community Earth System Model (CESM1)

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Abstract. Carbon isotopes in the ocean are frequently used as paleoclimate proxies and as present-day geochemical ocean tracers. In order to allow a more direct comparison of climate model results with this large and currently underutilized data set, we added a carbon isotope module to the ocean model of the Community Earth System Model (CESM), containing the cycling of the stable isotope 13C and the radioactive isotope 14C. We implemented the 14C tracer in two ways: in the “abiotic” case, the 14C tracer is only subject to air–sea gas exchange, physical transport, and radioactive decay, while in the “biotic” version, the 14C additionally follows the 13C tracer through all biogeochemical and ecological processes. Thus, the abiotic 14C tracer can be run without the ecosystem module, requiring significantly fewer computational resources. The carbon isotope module calculates the carbon isotopic fractionation during gas exchange, photosynthesis, and calcium carbonate formation, while any subsequent biological process such as remineralization as well as any external inputs are assumed to occur without fractionation. Given the uncertainty associated with the biological fractionation during photosynthesis, we implemented and tested three parameterizations of different complexity. Compared to present-day observations, the model is able to simulate the oceanic 14C bomb uptake and the 13C Suess effect reasonably well compared to observations and other model studies. At the same time, the carbon isotopes reveal biases in the physical model, for example, too sluggish ventilation of the deep Pacific Ocean.

1 Introduction

A large fraction of paleoclimatic reconstructions are based on isotopic measurements (e.g., Petit et al., 1999; McDermott, 2004; Curry and Oppo, 2005; Polka et al., 2013), yet there are many uncertainties associated with the interpretation of these records in terms of physical climate variables such as temperature, precipitation, and ocean circulation rates. More direct comparisons of paleo data with climate models would therefore be beneficial, both to test the interpretation of the isotopic proxy data and to allow for better comparisons of model simulations with proxy data. Furthermore, many isotope tracers are currently being measured in the ocean, and including them in ocean models can help us better understand the ocean circulation and diagnose model biases (e.g., Matsumoto et al., 2004). For all of these reasons, we have added a carbon isotope module to the ocean model of the Community Earth System Model (CESM) (Hurrell et al., 2013).

Carbon has two stable isotopes, 12C and 13C. More than 98.9% of carbon on earth is 12C, while 13C makes up most...
of the remaining 1%. The radioactive carbon isotope $^{14}$C, also called radiocarbon, is present only in trace amounts (approximately $1 \times 10^{-10}$ % of all carbon) and has a half-life of 5730 years (Godwin, 1962). Radiocarbon is a useful tracer to evaluate the ventilation of the deep ocean because it acts as a clock, measuring the time since water was last in contact with the atmosphere (e.g., Toggweiler et al., 1989; Orr, 2002; Meissner et al., 2003; Waugh et al., 2003; Key et al., 2004; Doney et al., 2004; Matsumoto et al., 2004; Meissner, 2007; Bardin et al., 2014). Because of the atmospheric nuclear weapons tests in the 1950s and 1960s and the well-known input function of radiocarbon during this time, radiocarbon is also useful to evaluate the recent penetration of anthropogenic carbon into the ocean (e.g., Graven et al., 2012). Furthermore, oceanic radiocarbon has been used to determine the mean gas exchange velocity used in ocean models (e.g., Wanninkhof, 1992; Sweeney et al., 2007; Naegler et al., 2006; Naegler, 2009). Oceanic $\delta^{13}$C has been used in paleoclimate studies as a tracer of the ocean circulation (e.g., Marchal et al., 1998; Curry and Oppo, 2005; Crucifix, 2005), to calculate the uptake of anthropogenic carbon dioxide (e.g., Keeling et al., 1980; Quay et al., 1992; Gruber et al., 1999; Sonnerup et al., 1999; Gruber and Keeling, 2001), and to diagnose biases in marine ecosystem models (e.g., Schmittner et al., 2013).

We added the carbon isotopes to the code so that they follow the cycling of total carbon through all ecosystem and physical/chemical processes. In this biotic formulation, a new $^{13}$C and $^{14}$C state variable was added to each carbon-bearing state variable, resulting in a total of 14 new state variables. For $^{14}$C, we also added the option of a simplified representation, where the isotope is only subject to the main chemical and physical processes during gas exchange and decay, but does not cycle through the ecosystem. This abiotic formulation of $^{14}$C was implemented based on the Ocean Carbon Model Intercomparison Project Phase 2 (OCMIP-2) protocol (Orr et al., 2000).

Abiotic radiocarbon had been added previously to the NCAR ocean model (in NCOM1.4, Orr, 2002, and POP1/CCSM3, Graven et al., 2012), and biotic $^{13}$C was implemented in the ecosystem model of the CCSM3 by X. Giraud and N. Gruber in 2009–2010. However, neither development was ever added to the trunk of the ocean model code of the CESM, so it was not maintained as the model evolved over the years and consequently none of these implementations still work in the current ocean model of the CESM. In contrast, the addition of a biotic radiocarbon tracer is completely new in this implementation in the CESM. In order to increase the chances of maintaining these developments as the model continues to evolve, the current implementation has been added to the code trunk of the current ocean model of CESM. By including carbon isotopes in the ocean model of the CESM1, the CESM1 joins the community of other comprehensive ocean general circulation models that include abiotic radiocarbon and/or biotic $^{13}$C in the ocean (e.g., Mo-

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3.1 Abiotic $^{14}$C

In this implementation, $\text{DI}_{14}^\text{C}$ is the model’s normalized concentration of total dissolved inorganic $^{14}$C, following the OCMIP2 protocol (Orr et al., 2000). $\text{DI}_{14}^\text{C}$ is used as normalized concentration in order to minimize the numerical error of carrying very small numbers. The normalization is done by dividing the real $\text{DI}_{14}^\text{C}$ by the standard ratio of $^{14}\text{C} / ^{12}\text{C} = 1.176 	imes 10^{-12}$ (Karlen et al., 1968). To obtain comparable $\text{DI}_{14}^\text{C}$ values as measured, we multiply the simulated $\text{DI}_{14}^\text{C}$ by this scaling factor of $1.176 \times 10^{-12}$. Since the abiotic radiocarbon is designed to be run without the ocean ecosystem active, we also carry an abiotic $\text{DI}_{12}^\text{C}$ tracer to calculate the isotope ratio $^{14}\text{R} = \text{DI}_{14}^\text{C} / \text{DI}_{12}^\text{C}$. For comparisons with observations, we calculate $\Delta^{14}\text{C}$ as a diagnostic variable:

$$\Delta^{14}\text{C} = (^{14}\text{R} - 1) \cdot 1000.$$

By construction, the abiotic $\text{DI}_{12}^\text{C}$ and $\text{DI}_{14}^\text{C}$ tracers only depend on the solubility of carbon in seawater and neglect all biological activity. The error in $\Delta^{14}\text{C}$ due to neglecting biology activity has been estimated to be on the order of 10% (Fiadiero, 1982).

Note that we do not multiply $^{14}\text{R}$ by $^{14}\text{R}_{\text{std}}$ in Eq. (1), as we are using a normalized $\text{DI}_{14}^\text{C}$ (following Orr et al., 2000). Given that this abiotic implementation does not account for the fractionation during gas exchange, we do not apply the correction for fractionation that is commonly applied to observational measurements of $^{14}\text{C} / ^{12}\text{C}$ ratios (as well as for the biotic $^{14}\text{C}$ implementation; see Eq. (27) in Sect. 3.2.4). The simulated abiotic $\Delta^{14}\text{C}$ is therefore directly comparable to observed data reported as $\Delta^{14}\text{C}$ (see Toggweiler et al., 1989, for more details).

3.1.1 Surface fluxes

We follow the abiotic OCMIP-2 protocol (Orr et al., 2000) for most of the implementation of the abiotic radiocarbon surface fluxes, with the following notable differences.

- We use a coefficient $a$ of 0.31 cm h$^{-1}$ (Wanninkhof, 1992) instead of 0.337 cm h$^{-1}$ as used in OCMIP-2. This is higher than what most recent estimates suggest (e.g., Sweeney et al., 2007; Naegler et al., 2006; Naegler, 2009; Graven et al., 2012), but makes it consistent with the gas-transfer formulation used in other parts of the CESM.

- We use the daily mean of the squared 10 m wind speed (either from the prescribed CORE-II forcing or from the coupled atmospheric model) instead of the climatology of the squared monthly average of the instantaneous SSMI velocity and its instantaneous variance as used in OCMIP-2.

- We use the daily mean of the ice fraction and atmospheric pressure (either from the data models or the coupled sea ice and atmosphere models) instead of the monthly averaged climatology used in OCMIP-2.

- We use a constant reference value (1944 µmol m$^{-3}$) for the virtual fluxes of abiotic radiocarbon, rather than an annually updated average of the surface $\text{DI}_{14}^\text{C}$ as suggested in OCMIP-2. This is done to conserve total $^{14}\text{C}$ in the model (in the absence of radioactive decay).

To compute the partial pressure of CO$_2$ from the abiotic $\text{DI}_{12}^\text{C}$, we require an estimate of surface alkalinity. We follow again OCMIP-2, i.e., we estimate surface alkalinity (Alk) by scaling the ocean mean alkalinity, Alk$_{\text{ocean}} = 2310$ microeq kg$^{-1}$ with sea-surface salinity, SSS, i.e.,

$$\text{Alk} = \text{Alk}_{\text{ocean}} \cdot \rho_{\text{sw}} \cdot \text{SSS} / S_{\text{Ref}}$$

with $S_{\text{Ref}} = 34.7$ and $\rho_{\text{sw}} = 4.1/3.996$ g cm$^{-3}$ (these two are constants in the CESM). We alter this calculation in the Baltic Sea and the Black Sea to avoid unrealistic alkalinity values, following the procedure developed by K. Lindsay for creating initial conditions for the marine ecosystem model: in the Black Sea, the surface alkalinity is independent of SSS: alkalinity = 3300 · $\rho_{\text{sw}}$. In the Baltic Sea, we calculate alkalinity depending on the surface salinity, with alkalinity = 119 + 196 · SSS when SSS is equal to or below 7.3, and alkalinity = 1237 + 43 · SSS when the SSS is above 7.3. The computation of $p$CO$_2$ also requires an assumption about the surface ocean concentrations of silicic acid and phosphate, for which we use OCMIP-2’s global constants, i.e., 7.5 µmol kg$^{-1}$ for silicic acid, Si(OH)$_4$, and 0.5 µmol kg$^{-1}$ for phosphate, PO$_4$.

### Air–sea gas exchange

As in OCMIP-2, the air–sea gas exchange flux of $^{12}$C is calculated as

$$F = PV \cdot (C_{\text{sat}} - C_{\text{surf}})$$

with $PV$ being the CO$_2$ gas transfer velocity (called the piston velocity) in m s$^{-1}$, calculated as

$$PV = (1 - a_{\text{ice}}) \cdot a \cdot u_{10}^2 \cdot (660.0 / S_{\text{CO}_2})^{-1/2}.$$  

The coefficient $a$ is taken as 0.31 cm h$^{-1}$ as mentioned earlier, $a_{\text{ice}}$ is the fraction of the ocean covered by sea ice, $u_{10}$ is the squared 10 m wind speed from the coupler, and $S_{\text{CO}_2}$ is the Schmidt number of CO$_2$. $S_{\text{CO}_2}$ is calculated as in the ecosystem model, following Wanninkhof (1992):

$$S_{\text{CO}_2} = 2073.1 + S_{\text{ST}} \cdot (-125.62 + S_{\text{ST}} + 3.6276 + S_{\text{ST}} \cdot (-0.043219)).$$

$C_{\text{surf}}$ in the gas flux calculation above is the surface aqueous CO$_2$ concentration in mol m$^{-3}$ (also called CO$_2$, which is the aqueous CO$_2$ concentration in mol m$^{-3}$ in the ocean...
in general). $C_{\text{sat}}$ is the saturation concentration in mol m$^{-3}$, with $C_{\text{sat}} = \text{CO}_2 + \text{DCO}_2^+$ and $\text{DCO}_2^+$ being the difference in CO$_2$ concentration between the surface ocean and the atmosphere. SST is the sea surface temperature. CO$_2$ and DCO$_2^+$ in turn are calculated by the carbonate solver from the ecosystem model, based on SST, SSS, ALK, PO$_4$, Si(OH)$_4$, pH, atmospheric $p$CO$_2$, atmospheric pressure, and the abiotic DI$^{12}C$ and DI$^{14}C$ concentrations in the surface water.

As in OCMIP-2, we do not account for fractionation during gas exchange in this abiotic formulation, as the effect of isotopic fractionation is almost completely accounted for by the standard correction made when calculating $\Delta_{14}C$ from observations (see Toggweiler et al., 1989, for details). The gas flux of the normalized abiotic DI$^{14}C$ is calculated as

$$F^{14} = PV \cdot (C_{\text{sat}} \cdot R^{14}C_{\text{atm}} - C_{\text{surf}} \cdot R^{14}C_{\text{ocn}})$$

with

$$R^{14}C_{\text{atm}} = (1 + \Delta_{14}C_{\text{atm}}/1000)$$

and

$$R^{14}C_{\text{ocn}} = 1000 \cdot (D^{14}C/D^{12}C - 1).$$

The values of the atmospheric $p$CO$_2$ and $\Delta_{14}C_{\text{atm}}$ can be set to be constants or can be read in from a file. For atmospheric $p$CO$_2$, it can also be taken from the coupler, to ensure the use of a consistent atmospheric $p$CO$_2$ value across model components. Currently the code is set up to read in three files of $\Delta_{14}C_{\text{atm}}$ values, one each for the Northern Hemispheres, the equatorial region (20° N–20° S), and the Southern Hemisphere, in order to represent the spatial inhomogeneity of $\Delta_{14}C_{\text{atm}}$, for example, after the atmospheric nuclear bomb tests.

**Virtual fluxes**

The CESM ocean model is a volume-conserving model where water fluxes at the surface (from precipitation, evaporation, and river input) are added as virtual fluxes. These virtual fluxes represent the dilution or concentration effect from adding or removing freshwater. For the abiotic carbon isotope tracers, we have a virtual DI$^{12}C$ and DI$^{14}C$ flux. As for salinity and for DIC in the ecosystem model, we use a constant surface reference DI$^{12}C$ and DI$^{14}C$ for the calculation of virtual fluxes in order to conserve tracers. The reference values are 1944 µmol m$^{-3}$ for both DI$^{12}C$ and normalized DI$^{14}C$, the same as for DIC in the ecosystem model of CESM.

**3.1.2 Interior processes**

In the interior of the ocean, the only additional term to the transport of the tracers by the physical ocean model is the decay term for DI$^{14}C$, following the OCMIP-2 protocol:

$$\text{d}[\text{DI}^{14}C]/\text{d}t = L([\text{DI}^{14}C])$$

and

$$\text{d}[\text{DI}^{14}C]/\text{d}t = L([\text{DI}^{14}C]) - \lambda \cdot [\text{DI}^{14}C],$$

with $L$ being the 3-D transport operator and $\lambda$ being the radioactive decay constant for $^{14}C$ in s$^{-1}$, using a half-life of 5730 years (Godwin, 1962):

$$\lambda = \ln(2)/(5730 \times 31 556 926).$$

The radiocarbon age (relative to AD 1950 = 0 yr BP) is calculated from $\Delta_{14}C$ following

$$14C_{\text{age}} = -5730/\ln 2 \times \ln(1 + \Delta_{14}C/1000).$$

5730 years / ln 2 = 8267 years is the mean life of $^{14}C$, which differs from the often used mean life of 8033 years (e.g., Stuiver and Polach, 1977), which is based on the earlier Libby half-life of 5568 (Libby, 1955).

**3.2 Biotic $^{13}C$ and $^{14}C$**

In the biotic implementation of $^{13}C$ and $^{14}C$, we use the ocean ecosystem model (e.g., Moore et al., 2013) to compute the carbon pools as well as all other biological variables (like silicic acid and alkalinity). The ecosystem model currently has seven carbon pools: DIC, DOC (dissolved organic carbon), CaCO$_3$, diazotrophs, diatoms, small phytoplankton, and zooplankton. We carry passive tracers for each of these in the isotope-enabled version of the code. As $^{12}C$ makes up over 98% of the carbon earth and does not fractionate, we assume that the ecosystem carries $^{12}C$. This means that the isotope ratio $R$ can be calculated as the ratio of the new isotopic carbon pools to the ecosystem carbon pools. As for the abiomic radiocarbon, we use scaled variables for $^{13}C$ and $^{14}C$ in order to minimize the numerical error of carrying very small numbers (particularly for $^{14}C$). The scaling factor is the commonly used standard $\text{iso}^{13}C / ^{12}C$ for each isotope, i.e., 1.12372 × 10$^{-8}$ for $\text{iso} = ^{13}C$ (Craig, 1957) and 1.176 × 10$^{-12}$ for $\text{iso} = ^{14}C$ (Karlen et al., 1968). This means that we use $^{13}R_{\text{std}} = 1$ and $^{14}R_{\text{std}} = 1$ in the code, and that the model-simulated isotopic carbon pools are multiplied by the respective scaling factor to compare them with observations.

In the biotic formulation, we account for the fractionation of $^{13}C$ and $^{14}C$ during gas exchange and during biological processes. The fractionation ($\epsilon$) of $^{14}C$ is always twice that of $^{13}C$, as all relevant processes have a mass-dependent fractionation for carbon (Bigeleisen, 1952; Craig, 1954). The isotopic fractionation $\epsilon$ is related to the fractionation factor $\alpha$ through

$$\epsilon = (\alpha - 1) \cdot 1000.$$  

(13)

As diagnostic variables, we compute the $\delta^{13}C$ values by first computing the ratio $\text{iso} R = \text{DI}^{13}C / \text{DIC}$ and then using

$$\delta^{13}C = (\text{iso} R - 1) \cdot 1000.$$  

(14)
As for the abiotic $\Delta^{14}C$ calculation in Eq. (1), we do not multiply by $^{13}R_{\text{Std}}$ in the calculation of $\delta^{13}C$ because we are using normalized DI$^{13}C$.

### 3.2.1 Air–sea gas exchange of $^{13}C$

The air–sea flux of $^{13}C$ is calculated based on Zhang et al. (1995):

$$F^{13}=\text{PV} \cdot \alpha_{\text{aq}} \cdot \alpha_k \cdot (R^{13}C_{\text{atm}} \cdot C_{\text{sat}} - R^{13}C_{\text{DIC}} \cdot C_{\text{surf}} / \alpha_{\text{DIC}}). \quad (15)$$

Here, $C_{\text{sat}}$ and $C_{\text{surf}}$ are obtained from the ecosystem model. $\alpha_k = -0.99919$ is the constant kinetic fractionation factor from Zhang et al. (1995) (with $\epsilon = -0.81$ and $\alpha = \epsilon / 1000 + 1$). $\alpha_{\text{aq}}$ is the temperature (TEMP, in $^\circ$C) dependent isotopic fractionation factor during gas dissolution, based on the equation for $\epsilon_{\text{aq}}$ from Zhang et al. (1995):

$$\epsilon_{\text{aq}} = -0.0049 \cdot \text{TEMP} - 1.31. \quad (16)$$

The temperature and carbonate fraction ($f_{\text{CO}}$) dependent fractionation factor ($\alpha_{\text{DIC}}$) between total DIC and CO$_2$ is based on the empirical relationship for $\epsilon_{\text{DIC}}$ from Zhang et al. (1995):

$$\epsilon_{\text{DIC}} = 0.014 \cdot \text{TEMP} \cdot f_{\text{CO}} - 0.105 \cdot \text{TEMP} + 10.53. \quad (17)$$

$R^{13}C_{\text{atm}}$ is the $^{13}C$ to $^{12}C$ ratio in atmospheric CO$_2$, calculated using the atmospheric $\delta^{13}C_{\text{atm}}$ record and $R_{\text{atm}} = 1 + \delta^{13}C_{\text{atm}}/1000$ (scaled by $^{13}R_{\text{Std}}$). The values of $\delta^{13}C_{\text{atm}}$ can be set to be a constant or it can be read in from a file. Currently $\delta^{13}C_{\text{atm}}$ is assumed to be well mixed globally, so only one global value is read in. With small code modifications globally inhomogeneous $\delta^{13}C_{\text{atm}}$ values can easily be read in instead. $R^{13}C_{\text{DIC}}$ is the $^{13}C$ to $^{12}C$ ratio of dissolved inorganic carbon, calculated from the simulated biotic DIC and DI$^{13}C$.

### 3.2.2 Virtual fluxes of $^{13}C$

As stated in Sect. 3.1, we account for the dilution and concentration effect of surface freshwater fluxes in the model by adding a virtual flux, using a constant surface reference DI$^{13}C$ (and DI$^{14}C$) of 1944 µmol m$^{-3}$ for the calculation of virtual fluxes.

### 3.2.3 Biological fractionation of $^{13}C$

The isotopic carbon fixation by photosynthesis (photo$^{13}C$) is computed from the $^{12}C$ fixation during photosynthesis (photo$^C$, from the ecosystem model), using

$$\text{photo}^{13}C = \text{photo}^C \cdot R_p \quad (18)$$

with

$$R_p = 1000 \cdot R_{\text{CO}_2} / (\epsilon_p + 1000) \quad (19)$$

The strength of the biological fractionation of carbon during photosynthesis ($\epsilon_p$), as well as the key controlling parameters, are still being debated in the literature (e.g., Keller and Morel, 1999), and many of the existing $^{13}C$ implementations in models use different parameterizations. We therefore implemented three different parameterizations for $\epsilon_p$ to test the sensitivity of our results to the choice of biological fractionation.

The simplest model for $\epsilon_p$ by Rau et al. (1989) gives the same value for all types of autotrophs:

$$\epsilon_p = 1000 \cdot (\delta_{\text{CO}_2} - \delta_{\text{Cp}}) / (1000 + \delta_{\text{Cp}}). \quad (21)$$

This relationship is based on the empirical relationship found by Rau et al. (1989) between the isotopic composition of the autotroph (δ$\text{C}_p$) and CO$_2$:

$$\delta_{\text{C}_p} = -0.8 \cdot \text{CO}_2 - 12.6, \quad (22)$$

limiting $\delta_{\text{C}_p}$ to values between $-18$ and $-32\%$ (Rau et al., 1989).

Laws et al. (1995) assumed that CO$_2$ enters the cell by diffusion and that the fractionation depends on the rate of photosynthesis, and therefore parameterized $\epsilon_p$ as a function of CO$_2^*$ and the specific photosynthesis rate of each phytoplankton group ($\mu$, in s$^{-1}$, calculated by the ecosystem model):

$$^{13}\epsilon_p = (\mu / \text{CO}_2^* \cdot 86400 - 0.371) / (-0.015). \quad (23)$$

Keller and Morel (1999) argued that only considering diffusive CO$_2$ transport into cells and assuming a linear relationship between $\epsilon_p$ and CO$_2^*$ concentration and the specific growth rate ($\mu$) does not agree with laboratory and field data, citing work by Sikes et al. (1980), Tortell et al. (1997), and Laws et al. (1997). Keller and Morel (1999) therefore proposed to use phytoplankton type-specific (constant) cell parameters (see Table 1) to compute the fractionation during photosynthesis:

$$^{13}\epsilon_p = \epsilon_{\text{diff}} + (C_{\text{up}} / (C_{\text{up}} + 1 / \text{var})) \cdot \delta_{\text{d13C}} + \theta \cdot (\epsilon_{\text{fix}} - \epsilon_{\text{diff}}) \quad (24)$$

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Small phytoplankton</th>
<th>Diatom</th>
<th>Diazotroph</th>
</tr>
</thead>
<tbody>
<tr>
<td>$Q_c$ (mol C cell$^{-1}$)</td>
<td>$69.2 \times 10^{-14}$</td>
<td>$63.3 \times 10^{-14}$</td>
<td>$3 \times 10^{-14}$</td>
</tr>
<tr>
<td>cell$_{avp}$ (m$^2$)</td>
<td>$1.8 \times 10^{-5}$</td>
<td>$3.3 \times 10^{-5}$</td>
<td>$3.0 \times 10^{-8}$</td>
</tr>
<tr>
<td>cell$_{surf}$ (m$^2$)</td>
<td>$87.6 \times 10^{-12}$</td>
<td>$100.6 \times 10^{-12}$</td>
<td>$5.8 \times 10^{-12}$</td>
</tr>
<tr>
<td>$C_{\text{up}}$</td>
<td>2.2</td>
<td>2.3</td>
<td>7.5</td>
</tr>
<tr>
<td>$\epsilon_{\text{fix}}$</td>
<td>25.3</td>
<td>26.6</td>
<td>30.0</td>
</tr>
</tbody>
</table>

Table 1. Parameters used in the parameterization of $\epsilon_p$ for the implementation following Keller and Morel (1999). The values for small phytoplankton are based on *E. huxleyi*, the value for diatoms are based on *P. tricornutum*, and the values for diazotrophs are based on *Synechococcus* sp. (Keller and Morel, 1999; Popp et al., 1998).
where
\[ \theta = (1 + (C_{up} - 1) \cdot \text{var})/(1 + C_{up} \cdot \text{var}) \] (25)
and
\[ \text{var} = \mu / \text{CO}_2 \cdot 1000 \cdot Q_c / (\text{cellpermea} \cdot \text{cellsurf}) \] (26)
with Qc being the cell carbon content, cellpermea being the cell wall permeability to \( \text{CO}_2 \) (aq), cellsurf being the surface areas of cells, \( C_{up} \) being the ratio of active carbon uptake to carbon fixation, \( \epsilon_{fix} \) being a constant phytoplankton type-dependent fractionation effect of carbon fixation, \( \epsilon_{diff} = 0.7 \) representing the fractionation by diffusion (O’Leary, 1984), and \( \delta_{13C} = -9.0 \) being the difference between the isotopic compositions of the external \( \text{CO}_2 \) and the organic matter pools (Goericke et al., 1994).

While the fractionation during calcium carbonate formation is much smaller than the fractionation during photosynthesis (Turner, 1982), we include a small constant fractionation of 2% for calcium carbonate formation, based on work by Ziveri et al. (2003) that found a range of 3% to −2% for different species. Other implementations of 13C in ocean models have used values of 1% (e.g., Sonnerup et al., 1999; Tagliabue and Bopp, 2008) or have assumed no isotopic fractionation for calcification (e.g., Marchal et al., 1998; Schmittner et al., 2013). However, as shown by Schmittner et al. (2013), the effect of the calcium carbonate pump on \( \delta_{13C} \) is small, so the choice of the value for the fractionation during calcium carbonate formation is not expected to have a big impact on the results in the current ecosystem model with one species of calcium carbonate.

### 3.2.4 Biotic \(^{14}\text{C}\)

The \(^{14}\text{C}\) air–sea flux is calculated in the same way as shown in Eq. (15) for \(^{13}\text{C}\), but with the fractionation for \(^{14}\text{C}\) being twice as large as for \(^{13}\text{C}\) (\( \epsilon_{14} = 2 \cdot \epsilon_{13} \)). Zeebe and Wolf-Gladrow (2001) and with \( R^{14}\text{C}_{\text{atm}} \) and \( R^{14}\text{C}_{\text{DIC}} \) instead of \( R^{13}\text{C}_{\text{atm}} \) and \( R^{13}\text{C}_{\text{DIC}} \). The biological fractionation is also the same as for \(^{13}\text{C}\), except that \( \epsilon_{14} = 2 \cdot \epsilon_{13} \) everywhere in Sect. 3.2.3. The surface reference value for \( D^{14}\text{C} \) for the virtual flux calculation is 1944 \( \mu\text{mol} \cdot \text{m}^{-3} \), the same as for \( D^{13}\text{C} \) (and \( D^{12}\text{C} \)).

In contrast to \(^{13}\text{C}\), \(^{14}\text{C}\) decays in all carbon pools, following the decay equation (see Eq. (11) in Sect. 3.1.2).

To compare the model-simulated \( \delta^{14}\text{C} \) values that we save as diagnostics (see Eq. 14) with published observations of \( \Delta^{14}\text{C} \), we apply the same fractionation correction to it that is used for observations to convert \( \delta^{14}\text{C} \) to \( \Delta^{14}\text{C} \):
\[ \Delta^{14}\text{C} = \delta^{14}\text{C} - 2(\delta^{13}\text{C} + 25)(1 + \delta^{14}\text{C} / 1000). \] (27)

In the following we always show \( \Delta^{14}\text{C} \).

As for the abiotic \(^{14}\text{C}\) implementation, the value of \( \Delta^{14}\text{C}_{\text{atm}} \) can be set to be a constant or it can be read in from three files (one for the Northern Hemisphere, one for the equatorial region, and one for the Southern Hemisphere).

![Figure 1. Schematic of the passive tracer modules with the new ecosystem driver and carbon isotope modules. Existing modules are shown in blue, new modules are shown in red, and edited modules are shown in blue with a red box. Dashed lines indicate future developments. This schematic shows how the ecosystem driver acts as an interface between the ecosystem-related modules and the passive tracers module that drives all tracer modules as well as how ecosys_share is used to share variables computed by the ecosystem model and used by other modules beside the ecosystem model.](image-url)

### 3.3 Ecosystem driver

We added an ecosystem driver (ecosys_driver) to the ocean model of the CESM in order to make it easier to expand the model to carry additional passive tracers that require variables from the ecosystem model, without adding these additional tracers to the ecosystem model itself. The ecosystem driver is structured similarly to the passive_tracers subroutine that calls all passive tracer modules, but it handles only the passive tracers that use the ecosystem model (see Fig. 1). It is called from the passive_tracers subroutine, and determines how many ecosystem-related passive tracers the model carries based on the namelist options set at buildtime. It then calls all subroutines in the ecosystem model and the related tracer modules, after being called by passive tracers with the corresponding tracer indices. Variables computed in the ecosystem model but used by other modules are shared via the new ecosys_share module. Only the ecosystem model changes the value of the variables in ecosys_share at this point. Other modules currently only read them from there, but do not modify them. With this infrastructure in place, additional tracers can be easily added without changing the ecosystem model too much. The only changes to the ecosystem model should be the copying of ecosystem variables to ecosys_share if they need to be shared with a new module as well as potentially the addition of new definitions and calculations of derived ecosystem variables that are needed.
but that are not currently computed in the ecosystem model (or not present in the required format, i.e., defined as local 2-D variables instead of a global 3-D variable). Nitrogen isotopes in the ocean model have already been added using this infrastructure (S. Yang, personal communication, 2014).

4 Results

4.1 Simulations and spin-up

We have performed several simulations with the new carbon-isotope-enabled model. As described in Sect. 2, we used the ocean-only version of the CESM1.0.5, at a nominal 3° horizontal resolution, forced by CORE-II climatological forcing (Large and Yeager, 2009). To spin up the carbon isotopes, we performed spin-up simulations that lasted several thousands of years. Radiocarbon takes a long time (5000–15 000 years, according to Orr et al., 2000) to equilibrate, due to the long timescale of deep ocean ventilation.

The abiotic radiocarbon has been spun up for 10 000 years using an atmospheric CO$_2$ concentration of 284.7 ppm and a $\Delta^{13}$C value of 0‰. The abiotic D$^{14}$C and D$^{13}$C were started from the standard ecosystem initial conditions, scaled to yield a global initial state of 0‰ $\Delta^{14}$C (following Orr et al., 2000), in order to simplify early interpretation and code verification. After 10 000 simulated years, the models satisfy the OCMIP2 surface CO$_2$ flux criterion of less than 0.01 Pg C year$^{-1}$. In terms of the drift in $\Delta^{14}$C, 91% of the ocean volume is spun up to the OCMIP2 criterion of a drift of less than 0.001 ‰ year$^{-1}$ (compared to the required 98% for OCMIP2). Compared to the fully spun-up solution (obtained using a new online Newton–Krylov method, manuscript in preparation by K. Lindsay, NCAR), differences are seen in the deep ocean only.

For the biotic carbon isotopes, we spun up the carbon isotopes for 6010 years, starting from the initial conditions of the ecosystem model, scaled to give a $\delta^{13}$C of 0‰ and a $\Delta^{14}$C of −100‰. The atmospheric CO$_2$ concentration was set to 284.7 ppm, the atmospheric $\Delta^{14}$C was set to 0‰, and the atmospheric $\delta^{13}$C was set to −6.379‰. In order to study the different biological fractionation parameterizations, two additional spin-up simulations were branched from the first spin-up simulation at year 2560 and run to year 6010. After 6010 years, the surface CO$_2$ flux is well below the OCMIP2 criterion of less than 0.01 Pg C year$^{-1}$, and over 99.99% of the ocean volume shows a drift of less than 0.001‰ year$^{-1}$ in $\delta^{13}$C. However, only 26% of the ocean satisfies the OCMIP2 criterion of a drift of less than 0.001‰ year$^{-1}$ for $\Delta^{14}$C. Another 4000 years or more are likely required to get the biotic $\Delta^{14}$C fully spun-up according to the OCMIP2 criterion. However, if we weaken the OCMIP2 criterion by an order of magnitude to less than 0.01‰ year$^{-1}$, 99.98% of the ocean satisfies this new criterion for $\Delta^{14}$C. Due to the long time required to run the ocean model with the ecosystem and the biotic carbon isotopes (the 6010 years took over 7 months of constant running on a supercomputer), we are currently not able to run the biotic radiocarbon to full equilibrium. In order to reach equilibrium in the future, a fast spin-up technique for the ecosystem model is currently in development by Keith Lindsay and will be applied to the biotic carbon isotopes when it is ready. We believe that for the purpose of this paper, which documents the implementation of the carbon isotopes in the model, the current spin-up is sufficient, in particular because future science applications will use the fully coupled 1° CESM, rather than the 3° ocean-only model version used here for model development. For these future science applications, the biotic radiocarbon should be spun up further in order to be fully trustworthy, and the carbon isotope simulation should be carefully validated in this model configuration.

After the long spin-up, we then performed experiments from 1765 to 2008, with the initial conditions from the end of the spin-up simulations in year 6010 for the biotic carbon isotopes and in year 10 000 for the abiotic radiocarbon. The atmospheric CO$_2$, $\Delta^{14}$C, and $\delta^{13}$C were prescribed based on the OCMIP-2 files (Orr et al., 2000) up to 1989, and H. Graven’s formulation of the global average for 1990–2008 (personal communication, 2012). The atmospheric state was the same repeating climatological CORE-II forcing as used for the spin-up, so changes related to warming or changes in the wind forcing over the 20th century are not included. We continued with the climatological CORE-II forcing rather than use the interannually varying CORE-II forcing for 1948–2007 in order to avoid shocks to the ocean when switching the forcing and when the forcing jumps from 2007 back to 1948 every 60 years. This jump in the forcing impacts the simulation for 10 years or more (as described in Danabasoglu et al., 2014), and would overlap with the start of the introduction of bomb radiocarbon into the atmosphere.

We also continued the spin-up simulations for 243 years, so that we could remove the influence of any continuing drift on the radiocarbon results shown in Sect. 4.2.2. To investigate the influence of the net CO$_2$ uptake on the simulation results in the second part of the 20th century, we also performed sensitivity experiments where the atmospheric CO$_2$ was fixed at 1949 conditions, while $\Delta^{14}$C$_{atm}$ and $\delta^{13}$C$_{atm}$ changed as usual.

4.2 $^{14}$C results

4.2.1 Simulated distributions of $\Delta^{14}$C

The radiocarbon simulation shows good agreement with the gridded GLODAP data for the 1990s (Key et al., 2004), reflecting the main features of the $\Delta^{14}$C distribution: (i) at the surface (see Fig. 2) the model shows the observed M shape of $\Delta^{14}$C distribution, with the highest values in the relatively stable subtropical waters, intermediate values in the equatorial upwelling zone, and low values in the polar regions,
The largest bias is located in the deep Pacific, where the residence time is short and sea ice limits the uptake of atmospheric $^{14}$C, with the overall lowest values in the Southern Ocean, where the upwelling of old, low $^{14}$C waters further dilutes the surface waters. (ii) In the zonal mean (see Fig. 3), newly formed deepwater with high $^{14}$C values can clearly be separated from old water masses with low $^{14}$C values. (iii) In the depth profiles (see Fig. 4), it is obvious that the $^{14}$C in the deep water decreases from the Atlantic Ocean over the Indian Ocean to the Pacific Ocean, which has the lowest $^{14}$C values (i.e., oldest water). Consistently, the abiotic $^{14}$C values are higher than the biotic $^{14}$C values, but both show the same general features also shown in GLODAP (Key et al., 2004) and in the cruise data compiled by Schmittner et al. (2013) because their distribution is set mainly by the physical ocean simulation. The difference between the abiotic and biotic simulations due to biological effects has been estimated to be on the order of 10% (Fiadiero, 1982), but since the biotic radiocarbon simulation is less spun-up than the abiotic simulation at this point, a detailed investigating of the impact of the biological effects is premature and will be the topic of a future study when we can spin up both radiocarbon implementations using a fast-spin up technique.

Figure 2. Surface values of total $^{14}$C from the 1990s (including bomb $^{14}$C) from (a) cruise data compiled by Schmittner et al. (2013), (b) the gridded GLODAP data (Key et al., 2004), (c) simulated biotic $^{14}$C, and (d) simulated abiotic $^{14}$C.

Above 1000 m, the depth structure of the simulated $^{14}$C agrees reasonably well with observations, with the best agreement with the GLODAP $^{14}$C in the upper 250 m of the Indian Ocean (see Fig. 4). The largest biases are found at depths below 1000 m (see Fig. 4), with the model showing $^{14}$C values that are too negative (i.e., water that is too old). The largest bias is located in the deep Pacific, where the $^{14}$C is up to 100% too negative (see Figs. 3 and 4). In terms of radiocarbon age, the maximum bias in the deep Pacific is 1000 years compared to GLODAP, revealing that the deep Pacific Ocean in the model is not ventilated as much as it should be. This is a well-known bias in the CESM, which was also present in the ocean model of a previous version of the CESM, the CCSM3 (Graven et al., 2012), as well as in the nominal 1° resolution version of the current CESM1 ocean model (Bardin et al., 2014), and is related to too weak Antarctic Bottom Water formation in the CESM (Danabasoglu et al., 2011) and too shallow mixed layers in the Southern Ocean (Moore et al., 2013). Currently radiocarbon is being used to test improvements to the ventilation in the Southern Ocean in the ocean model in the CESM, in order to improve this bias in future versions of the CESM (K. Lindsay, personal communication, 2014).

4.2.2 $^{14}$C bomb inventory

The excess oceanic radiocarbon inventory is frequently used to investigate the ocean uptake of anthropogenic carbon (e.g., Key et al., 2004; Graven et al., 2012) and to determine the mean gas exchange velocity used in ocean models (e.g., Wanninkhof, 1992; Sweeney et al., 2007; Naegler et al., 2006; Naegler, 2009). To establish how well the newly developed radiocarbon tracer compares to observations, we here compare the simulated excess radiocarbon inventory with observational estimates. The excess radiocarbon in the ocean includes change in the oceanic radiocarbon from the atmospheric nuclear bomb tests of the 1950s and 1960s, as well as from the Suess effect and changes in net CO$_2$ uptake, compared to the reference period of the 1940s, following Naegler (2009). In 1975, the excess radiocarbon inventory in the abiotic and biotic simulation is $297 \times 10^{26}$ atoms $^{14}$C and $295 \times 10^{26}$ atoms $^{14}$C, respectively. This lies within the range of observational estimates of the excess radiocarbon in 1975, which range from $225 \times 10^{26}$ atoms $^{14}$C to $314 \pm 35 \times 10^{26}$ atoms $^{14}$C (see Table 2). It has been shown that the earlier estimates from Broecker et al. (1985, 1995) were high by about 25% (e.g., Hesshaimer et al., 1994; Peacock, 2004; Sweeney et al., 2007), which suggests that the simulated values are probably on the high end of the observational range. One reason for this could be the choice of the coefficient $\alpha = 0.31 \text{ cm h}^{-1}$ in Eq. (3), which has been shown...
Figure 3. Zonal averages of total $\Delta^{14}$C for the Atlantic, Pacific, and Indian oceans for the 1990s, from cruise data compiled by Schmittner et al. (2013) (top row), the gridded GLODAP data (Key et al., 2004) (second row), the $\Delta^{14}$C from the biotic model (third row), and the abiotic model (bottom row). Note that due to the sparse observational data (see Fig. 2a for the coverage at the surface), the zonal average from the cruise data in the top row is more of a zonal composite than a zonal average.

Figure 4. Depth profiles of $\Delta^{14}$C for (a) the global ocean, (b) the Atlantic Ocean, (c) the Pacific Ocean, and (d) the Indian Ocean. The simulated biotic (green) and abiotic (blue) $\Delta^{14}$C is compared to the global gridded GLODAP $\Delta^{14}$C (black) data set (Key et al., 2004). In addition, dashed lines show the cruise data compiled by Schmittner et al. (2013) (gray) and the model-simulated data subsampled at the same locations as these data (green and blue dashed lines).

to be high (e.g., Sweeney et al., 2007; Naegler, 2009). Graven et al. (2012) showed that in the ocean model of the CCSM3, the simulated excess radiocarbon inventory was lower when a coefficient $a = 0.23 \text{ cm h}^{-1}$ rather than $a = 0.31 \text{ cm h}^{-1}$ was used in Eq. (3). However, since $a = 0.31 \text{ cm h}^{-1}$ is the parameter used in the CESM in general to compute air–sea gas fluxes, we did not change it here. For 1995, the excess radiocarbon inventories in the abiotic and biotic simulation are $389 \times 10^{26}$ atoms $^{14}$C and $390 \times 10^{26}$ atoms $^{14}$C, respectively, which is close to but slightly higher than the observational estimates of $313–383 \times 10^{26}$ atoms $^{14}$C (see Table 2).

The natural radiocarbon inventory, before anthropogenic disturbances from the Suess effect and from increased oceanic net CO$_2$ uptake, has been estimated to be $19 \, 000 \pm$
1200 × 10^{26} \text{atoms of } ^{14}\text{C} \text{ (Naegler, 2009). In the simulation the inventory is just outside the error bar for the biotic model (17 763–17 770 × 10^{26} \text{atoms of } ^{14}\text{C}, depending on the biological fractionation used), and slightly lower for the abiotic model (16 190 × 10^{26} \text{atoms of } ^{14}\text{C}). The natural radiocarbon inventories are calculated for years 6185–6194 of the control simulations, which corresponds to the same total run time as years 1940–1949 in the 1765–2008 experiments, which were started from the control in year 6010. However, the biotic model estimate of the natural radiocarbon inventory might still not be the final value, as the biotic radiocarbon is still spinning up. In terms of the anthropogenic radiocarbon inventories presented next, this biases should not play any large role, however, as we remove any remaining drift. Specifically, to calculate the early anthropogenic radiocarbon inventory present in the 1940s, we take the difference between the natural radiocarbon inventory in simulation years 6185–6194 (with constant atmospheric CO_2, Δ^{14}\text{C}, and δ^{14}\text{C}) and the inventory in the 1940s (with changing atmospheric CO_2, Δ^{14}\text{C}, and δ^{14}\text{C} since 1765). By taking this difference between years of equal total run time, we remove the impact of any remaining drift in Δ^{14}\text{C}. We find an anthropogenic radiocarbon inventory of 20 × 10^{26} \text{atoms of } ^{14}\text{C} for the abiotic model and 5 × 10^{26} \text{atoms of } ^{14}\text{C} for the biotic model (independent of the biological fractionation used). Both of these anthropogenic radiocarbon inventories for the 1940s are within the error bar of the estimate of 4 ± 20 × 10^{26} of $^{14}\text{C}$ from Naegler (2009), with the biotic model giving a very good match.

Using sensitivity experiments from 1950 to 2008 with atmospheric CO_2 held constant at 1949 levels but normally increasing atmospheric Δ^{14}\text{C}, we can calculate the impact of increased ocean uptake of anthropogenic CO_2 on the excess radiocarbon inventory: in 1975, the excess oceanic radiocarbon inventory relative to the 1940s due to atmospheric Δ^{14}\text{C} changes alone (from the atmospheric bomb tests and the Suess effect) is $282 \times 10^{26} \text{atoms of } ^{14}\text{C}$ for the abiotic model and $280 \times 10^{26} \text{atoms of } ^{14}\text{C}$ for the biotic model, while for 1995 the numbers are $353 \times 10^{26} \text{atoms of } ^{14}\text{C}$ and $354 \times 10^{26} \text{atoms of } ^{14}\text{C}$, respectively. This means that the increase in net CO_2 uptake contributed $15 \times 10^{26} \text{atoms of } ^{14}\text{C}$ in 1975 and $36 \times 10^{26} \text{atoms of } ^{14}\text{C}$ in 1995 compared to the 1940s (for both the abiotic and biotic models), which is 5 and 9% of the total radiocarbon excess in these years, respectively. These changes are in excellent agreement with calculations from Naegler (2009), which showed an excess radiocarbon inventory in 1995 of $346 \pm 98 \times 10^{26} \text{atoms of } ^{14}\text{C}$ due to atmospheric Δ^{14}\text{C} changes, and $27 \pm 9 \times 10^{26} \text{atoms of } ^{14}\text{C}$ due to net CO_2 uptake. The percentage contribution of the net CO_2 uptake to the total radiocarbon excess was given as 3% in 1975 and 8% in 1995 in Naegler (2009), which again compares very well with our model simulation.

Table 2. Excess oceanic radiocarbon inventory, measured in $10^{26}$ atoms of $^{14}\text{C}$, from various sources for 1975 (GEOSECS) and 1995 (WOCE). Corrections by Naegler et al. (2006) are for neglected ocean regions; corrections by Naegler (2009) are for neglected contributions from increasing DIC. The values from this study are listed at the bottom, for the abiotic and biotic implementation. The biotic excess radiocarbon inventories are the same for all biological fractionation choices tested.

<table>
<thead>
<tr>
<th>Publication</th>
<th>1975 (GEOSECS)</th>
<th>1995 (WOCE)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Broecker et al. (1980)</td>
<td>314 ± 35</td>
<td></td>
</tr>
<tr>
<td>Broecker et al. (1985)</td>
<td>289</td>
<td></td>
</tr>
<tr>
<td>Lassey et al. (1990)</td>
<td>303</td>
<td></td>
</tr>
<tr>
<td>Hesshaimer et al. (1994)</td>
<td>225</td>
<td></td>
</tr>
<tr>
<td>Broecker and Peng (1994)</td>
<td>300</td>
<td></td>
</tr>
<tr>
<td>Broecker et al. (1995)</td>
<td>305 ± 30</td>
<td></td>
</tr>
<tr>
<td>Peacock (2004) multitracer correlation</td>
<td>241 ± 60</td>
<td>335 ± 15</td>
</tr>
<tr>
<td>Corrected by Naegler et al. (2006)</td>
<td>245 ± 60</td>
<td>340 ± 15</td>
</tr>
<tr>
<td>Corrected by Naegler (2009)</td>
<td>252 ± 60</td>
<td>367 ± 15</td>
</tr>
<tr>
<td>Peacock (2004) silicate approach</td>
<td>262 ± 26</td>
<td></td>
</tr>
<tr>
<td>Corrected by Naegler et al. (2006)</td>
<td>264 ± 26</td>
<td></td>
</tr>
<tr>
<td>Key et al. (2004)</td>
<td></td>
<td>313 ± 47</td>
</tr>
<tr>
<td>Corrected by Naegler et al. (2006)</td>
<td></td>
<td>355 ± 50</td>
</tr>
<tr>
<td>Corrected by Naegler (2009)</td>
<td></td>
<td>383 ± 50</td>
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<tr>
<td>Naegler and Levin (2006)</td>
<td>258 ± 13</td>
<td>367 ± 17</td>
</tr>
<tr>
<td>Sweeney et al. (2007)</td>
<td>225</td>
<td>343 ± 40</td>
</tr>
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<td>370 ± 40</td>
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<tr>
<td>Naegler (2009)</td>
<td></td>
<td>373 ± 98</td>
</tr>
<tr>
<td>This study, abiotic $^{14}\text{C}$</td>
<td>297</td>
<td>389</td>
</tr>
<tr>
<td>This study, biotic $^{14}\text{C}$</td>
<td>295</td>
<td>390</td>
</tr>
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</table>
4.3 $^{13}$C results

4.3.1 Simulated $\delta^{13}$C and the impact of different biological fractionation parameterizations

In the literature, models of biological fractionation are still under debate (e.g., Keller and Morel, 1999). We therefore tested three different parameterizations of biological fractionation, to investigate the impact on the simulated $\delta^{13}$C (as described in Sects. 3.2.3 and 4.1). As shown in Fig. 5a, the simulated globally averaged $\epsilon_p$ depth profiles differ when these different parameterizations are used, with $\epsilon_p$ values ranging from 15 to 30. By design, $\epsilon_p$ is the same for diatoms, diazotrophs, and small phytoplankton when using Rau et al. (1989), while $\epsilon_p$ shows large variations between species for the method of Keller and Morel (1999), due to the dependence on species-specific cell parameters (see Table 1). The method of Laws et al. (1995) leads to small differences between species in the surface ocean only. Below 200 m, only the $\epsilon_p$ following Rau et al. (1989) still changes with depth (see Fig. 5a), due to the sole dependence of $\epsilon_p$ on CO$_2^*$ and the export of organic carbon and carbonates to depth.

The impact of the different biological fractionation choices on $\delta^{13}$C$_{DIC}$ is noticeable (see Fig. 5b), with the globally averaged $\delta^{13}$C$_{DIC}$ based on $\epsilon_p$ from Rau et al. (1989) being larger below 150 m compared to the $\delta^{13}$C$_{DIC}$ from Laws et al. (1995) and Keller and Morel (1999), but slightly smaller at the surface. Despite the more complex formulation of $\epsilon_p$ in Keller and Morel (1999) compared to Laws et al. (1995) and the significantly different $\epsilon_p$ profiles, the resulting $\delta^{13}$C$_{DIC}$ from both methods is very similar and only differs slightly at depth (most notably between 150 and 2000 m). To compare the simulated $\delta^{13}$C$_{DIC}$ to the cruise data of $\delta^{13}$C$_{DIC}$ compiled by Schmittner et al. (2013), we re-gridded the model output to subsample the model at the same points as covered by the cruise data. The resulting globally averaged depth profiles are shown in Fig. 5c, and are remarkably similar to the full globally averaged model results in Fig. 5b. Both show the expected increase in $\delta^{13}$C$_{DIC}$ directly below the surface, due to the preferential uptake of the light isotope during photosynthesis, followed by the expected decrease in $\delta^{13}$C$_{DIC}$ with depth due to the remineralization of the isotopically light organic material back into the water column. The model-simulated global depth profile of $\delta^{13}$C$_{DIC}$ lies within the error range of $\pm0.2\%e$ around the cruise $\delta^{13}$C$_{DIC}$ data between the surface and 150 m and below 1000 m, but shows smaller $\delta^{13}$C$_{DIC}$ values than observed between 150 and 1000 m.

For individual basins, the model bias compared to the cruise data is smallest in the Atlantic, with the $\delta^{13}$C$_{DIC}$ based on the biological fractionation from Rau et al. (1989) almost entirely within the uncertainty range of the data (see Fig. 5d). All three basins contribute to the bias seen between 150 and 2000 m in the global average, with the Indian Ocean contributing the most to this bias in the upper ocean and the Pacific Ocean contributing the most at intermediate depths (see Fig. 5e–f). In general, the model simulated $\delta^{13}$C$_{DIC}$ tends to be smaller than the observed $\delta^{13}$C$_{DIC}$. While the difference between the full global average in the model and the subset global average based on the cruise locations is small, the difference between the total basin average (shown as dashed lines in Fig. 5d–f) and the subset basin averages (shown as solid lines) is larger for the individual basins.

At the surface, the simulated $\delta^{13}$C$_{DIC}$ values show a systematic bias in that they are generally larger than the observational data suggest, but the same general spatial pattern is visible (see Fig. 6). While both gas-exchange and biological processes are important for the surface ocean $\delta^{13}$C$_{DIC}$ pattern (Schmittner et al., 2013), the details of the biological fractionation parameterizations appear to have a very small impact at the surface, as shown in the almost identical surface distributions from the model (see Fig. 6c–e). The zonal means of $\delta^{13}$C$_{DIC}$ from the different biological fractionation parameterizations on the other hand do show some small differences (see Fig. 7), with the biological fractionation from Rau et al. (1989) leading to the largest $\delta^{13}$C$_{DIC}$ values in all three ocean basins, while the fractionation based on Keller and Morel (1999) shows the lowest $\delta^{13}$C$_{DIC}$ values. Overall, all three parameterizations lead to the expected pattern of high values of $\delta^{13}$C$_{DIC}$ in water that has recently been in contact with the surface (e.g., North Atlantic Deep Water) and low $\delta^{13}$C$_{DIC}$ values in water that has been out of contact with the atmosphere for a long period of time and has accumulated a large amount of remineralized (isotopically light) organic matter (e.g., in the deep Pacific).

We choose the biological formulation from Laws et al. (1995) as the default biological fractionation in our model, as it considers the growth rate of different species, but the differences in the simulated $\delta^{13}$C$_{DIC}$ compared to the more complex formulation from Keller and Morel (1999) are small. The other parameterizations of biological fractionation remain an option in the model that can be chosen at build time.

4.3.2 Oceanic surface $^{13}$C Suess effect

The surface oceanic Suess effect, which is the decrease in the surface ocean $\delta^{13}$C due to the penetration of carbon originating from the burning of fossil fuels, has been calculated from observational data as well as from other models that include $^{13}$C, and it is often used to derive the oceanic anthropogenic carbon uptake (e.g., McNeil et al., 2001; Tagliabue and Bopp, 2008). In our model simulation, the surface $\delta^{13}$C change between 1975 and 1995 is $-0.164$ to $-0.167\%e$ decade$^{-1}$ (the range is for the different biological fractionations used). This compares well with other estimates of $-0.171\%e$ decade$^{-1}$ (Bacastow et al., 1996), $-0.18\%e$ decade$^{-1}$ (Gruber et al., 1999), $-0.15\%e$ decade$^{-1}$ (Sonnerup et al., 1999), and $-0.174\%e$ decade$^{-1}$ (Tagliabue and Bopp, 2008). As already shown by Quay et al. (1992) and Gruber et al. (1999), the surface ocean Suess effect is not uniform (see Fig. 8), and the model simulation of the spatial
Figure 5. (a) Depth profiles over the top 500 m (where $\epsilon_p$ is important because of primary production) of the globally averaged values of $\epsilon_p$ produced by the three tested parameterizations for biological fractionation for diatoms (solid line), diazotrophs (short dashes), and small phytoplankton (large dashes). The simulated globally averaged depth profile (0–6000 m) of $\delta^{13}$C$_{\text{DIC}}$ in the 1990s is shown in (b), and the global average depth profile of the subset model $\delta^{13}$C$_{\text{DIC}}$ for the same grid points as in the cruise data compiled by Schmittner et al. (2013) is shown in (c). Basin average depth profiles are shown in (d–f), with dashed lines showing the full basin average from the model and solid lines showing the subset averages for the same points as the cruise data compiled by Schmittner et al. (2013). The uncertainty for the cruise data is shown as gray shading in (c), and is $\pm 0.2$‰ (Schmittner et al., 2013). Note that the irregular $y$ axis in (b–f) emphasizes the upper ocean.

Figure 6. Surface values of $\delta^{13}$C for the 1990s from (a) cruise data compiled by Schmittner et al. (2013), (b) $5^\circ$ extrapolated gridded data from Gruber and Keeling (1999) and Gruber and Keeling (2001), and (c–e) the biotic model, using the biological fractionation from (c) Rau et al. (1989), (d) Laws et al. (1995), and (e) Keller and Morel (1999).

Suess effect agrees well with the model results of Tagliabue and Bopp (2008): the largest changes (i.e., most negative values in Fig. 8) occur in regions with little deep ventilation and therefore longer residence times of water at the surface (e.g., the subtropical gyres), while the smallest changes (i.e., least negative or zero in Fig. 8) occur in regions of reduced air–sea gas exchange (e.g., under sea ice), in regions with active deep convection (and therefore short residence times at the surface, e.g., around the Antarctic), as well as in regions with upwelling (which dilutes the surface $\delta^{13}$C, for example, off the western coast of South America).

Compared to the pre-industrial ocean, the total surface ocean $^{13}$C Suess effect is $-0.064$ to $-0.065$‰ decade$^{-1}$ for 1860–2000 (depending on the different biological fractionations used), compared to $-0.07$‰ decade$^{-1}$ found by Tagliabue and Bopp (2008). The fact that the simulated oceanic
Figure 7. Zonal ocean basin composites from the cruise data compiled by Schmittner et al. (2013) (top row) compared to 1990s zonal basin averages from the model simulation using the biological fractionation from Rau et al. (1989) (second row), Laws et al. (1995) (third row), and Keller and Morel (1999) (bottom row).

Figure 8. Surface ocean Suess effect (the change in $\delta^{13}C$) between 1970 and 1990, in \%$\text{e}$ decade$^{-1}$.

$^{13}$C Suess effect calculated over different periods agrees reasonably well with other available estimates suggests that our model is able to simulate the change in the oceanic $\delta^{13}C$ inventory correctly, despite some mean biases in the distribution of $\delta^{13}C$ described and shown in Sect. 4.3.1.

5 Summary

We have developed carbon isotope tracers in the ocean model of the CESM, including a biotic and an abiotic radiocarbon tracer and a biotic $^{13}$C tracer. The details of the implementation are described here in order to serve as a reference for future users of these new model features and/or for model developers planning to modify the code or add carbon isotopes to other ocean models. In particular, we tested three different formulations for the fractionation during phytoplankton growth that have been discussed in the literature, and show that the effect on the simulated $\delta^{13}C$ in the ocean is relatively minor. A comparison of the simulation results from the coarse nominal 3$^\circ$ resolution ocean model forced with climatological CORE-II atmospheric forcing and with present-day data for $\Delta^{14}C$ and $\delta^{13}C$ shows that the simulated carbon isotopes can represent the large-scale features of the observed distributions as well as the anthropogenic changes due to nuclear bomb tests and the burning of fossil fuels. The carbon isotopes also reflect some known model biases, for example too sluggish ventilation of the deep Pacific Ocean. Once
a fast spin-up technique for the biotic carbon isotopes has been implemented, we are planning to further validate the carbon-isotope simulation in the fully coupled CESM framework at 1° resolution. Ultimately, we plan to use the carbon isotopes for both present-day and paleo simulations in the fully coupled framework of the CESM at the standard nominal 1° resolution in the ocean, in order to investigate details of changes in the ocean circulation over the 20th century, the last Millennium, and at the Last Glacial Maximum.

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