



Carbon-Cycle Feedbacks Operating in the Climate System

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Abstract

Climate change involves a direct response of the climate system to forcing which is amplified or damped by feedbacks operating in the climate system. Carbon-cycle feedbacks alter the land and ocean carbon inventories and so act to reduce or enhance the increase in atmospheric CO₂ from carbon emissions. The prevailing framework for carbon-cycle feedbacks connect changes in land and ocean carbon inventories with a linear sum of dependencies on atmospheric CO₂ and surface temperature. Carbon-cycle responses and feedbacks provide competing contributions: the dominant effect is that increasing atmospheric CO₂ acts to enhance the land and ocean carbon stores, so providing a negative response and feedback to the original increase in atmospheric CO₂, while rising surface temperature acts to reduce the land and ocean carbon stores, so providing a weaker positive feedback for atmospheric CO₂. The carbon response and feedback of the land and ocean system may be expressed in terms of a combined carbon response and feedback parameter, λ_{carbon} in units of $\text{W m}^{-2}\text{K}^{-1}$, and is linearly related to the physical climate feedback parameter, λ_{climate} , revealing how carbon and climate responses and feedbacks are inter-connected. The magnitude and uncertainties in the carbon-cycle response and feedback parameter are comparable with the magnitude and uncertainties in the climate feedback parameter from clouds. Further mechanistic insight needs to be gained into how the carbon-cycle feedbacks are controlled for the land and ocean, particularly to separate often competing effects from changes in atmospheric CO₂ and climate forcing.

Keywords Carbon cycle · Carbon emissions · Carbon feedbacks · Climate feedbacks

Introduction

Predicting our future climate is difficult due to feedbacks operating in the climate system [1], acting to amplify or damp the effect of climate forcing. To understand what a feedback is, consider the following every-day example: when a pan of water is heated on a stove, there is a direct warming response of the water from the heat supplied. At the same time, the pan is cooled by infrared radiation and this loss of heat is inhibited by the evaporation increasing the water vapour above the pan. Hence, the increase in the

water vapour is providing a positive feedback that reinforces the initial supply of heat and leads to greater warming [2]. There is a wide range of positive and negative responses and feedbacks operating in the climate system, such as involving radiative responses from changes in water vapour, clouds and surface albedo [3–7]; for a discussion of external forcing, climate response and feedbacks, see [8] and for reviews of linear models for climate feedbacks, see [9].

The present-day climate system is being forced by carbon emissions [10], increasing atmospheric CO₂ and surface temperature (Fig. 1a, blue and red arrows). The increase in the atmospheric carbon inventory is less than the cumulative carbon emission due to carbon being sequestered by the land and ocean systems. These changes in the land and ocean carbon inventories may be expressed in terms of a carbon-cycle framework with carbon-cycle parameters measuring the dependencies of the land and ocean carbon inventories on atmospheric CO₂ and surface temperature [11–14] (Fig. 1a, black arrows).

The carbon uptake by the land and ocean after an increase in atmospheric CO₂ may be viewed in terms of a combination of transient responses and feedbacks. The

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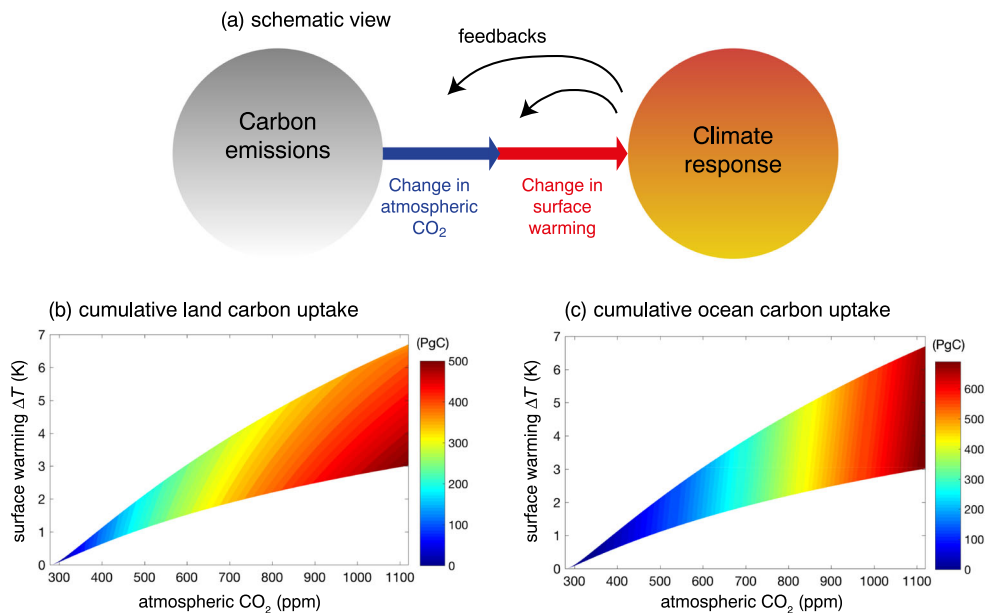


Fig. 1 (a) Schematic view of how carbon emissions drive a climate response by increasing atmospheric CO₂ and driving surface warming (blue and red arrows respectively). The climate response is modified by feedbacks (black arrows) enhancing the effects of the changes in the atmospheric CO₂ and surface temperature. An illustrative view of how cumulative carbon uptake (PgC, colours) for (b) the land and (c) the

ocean depends upon increases in atmospheric CO₂ (ppm) versus the increase in surface temperature, ΔT (K), which is generated from 100 simulations of an efficient WASP Earth system model [19] forced by an imposed 1% annual rise in atmospheric CO₂ under multiple warming responses with different cloud feedback parameters for 140 years until 4 × CO₂ is reached

transient response involves the ocean automatically taking up more carbon when there is higher atmospheric CO₂, analogous to how a warmer atmosphere drives a thermal transient response with the ocean taking up more heat. This transient response acts to oppose the original forcing and the associated air-sea exchange eventually declines in time as a new equilibrium state is approached, so it is viewed as a negative response with respect to the original forcing of the atmosphere. In addition, feedbacks act either to enhance or damp the original atmospheric perturbation and these feedbacks strengthen in time relative to the initial state. For example, there are the following carbon-cycle feedbacks acting on the land and in the ocean:

1. Photosynthesis by plants on land is expected to be stimulated by increasing atmospheric CO₂, so that the terrestrial uptake of carbon increases, which reduces the initial increase in atmospheric CO₂ and so provides a negative feedback to the original atmospheric perturbation;
2. The ocean carbon uptake is modified in strength by an ocean acidity feedback involving seawater becoming more acidic with greater atmospheric CO₂, so that the proportion of carbonate ions decreases and the proportion of dissolved CO₂ increases. This ocean acidity feedback leads to a smaller proportion of carbon

- emissions being subsequently taken up by the ocean and so acts as a positive feedback for atmospheric CO₂;
3. Increased surface warming may also enhance plant and soil respiration, so acting to increase atmospheric CO₂ and provides a positive feedback;
4. If there is surface warming of the ocean, the ventilation of the ocean decreases and so the ocean uptake of carbon decreases and instead atmospheric CO₂ increases, which enhances radiative forcing and provides a positive feedback that reinforces the initial perturbation.

Our aim is to review how carbon-cycle responses and feedbacks are represented [11–14], providing guidance as to the underlying mechanisms and emphasize how the carbon responses and feedbacks may be combined together in terms of a carbon-cycle response and feedback parameter [13], directly analogous to how physical climate feedbacks are represented [7]. For carbon-cycle responses and feedbacks, we know that there are large inter-model differences in the magnitude of these feedbacks [13–15], but we do not understand how these responses and feedbacks evolve in time with changes in the state of the climate system and the carbon forcing being experienced. We need to solve this problem if we are to understand how our future climate is going to evolve and identify how much carbon may be emitted before reaching future warming targets.

Carbon-Cycle Feedbacks and Changes in Carbon Inventories

Cumulative carbon emissions, ΔC_{em} , drive an increase in the combined carbon inventories for the atmosphere, land and ocean [16],

$$\Delta C_{em} = \Delta C_{atmos} + \Delta C_{land} + \Delta C_{ocean}, \quad (1)$$

where ΔC_{atmos} , ΔC_{land} and ΔC_{ocean} represent the respective changes in the atmospheric, land and ocean inventories since the pre-industrial era and these carbon inventories changes are evaluated in PgC; the change in atmospheric carbon inventory, C_{atmos} , is dominated by the contribution from atmospheric CO_2 .

The land and ocean carbon inventories may be viewed as being dependent on the carbon state of the climate system [15], such that

$$C_{land} + C_{ocean} = F(\text{climate state variables}), \quad (2)$$

where F is a function defining the climate system and the climate-state variables may span a wide set of physical and biogeochemical variables. These climate-state variables would ideally be independent of each other within the climate function, but defining variables independent of each other is difficult to achieve due to the inter-connection of processes in the climate system.

The climate-state variables are usually taken to be the atmospheric carbon inventory, C_{atmos} , and global-mean surface air temperature, T , as proxies for climate change, such that

$$C_{land} + C_{ocean} = F(C_{atmos}, T). \quad (3)$$

For the terrestrial system, climate drivers are found to scale with the change in global-mean surface temperature [17], so supporting the choice of surface temperature in Eq. 3. For the ocean, the carbon uptake may initially depend on surface temperature, as solubility decreases in warmer waters and so inhibits carbon uptake from the atmosphere; however, the longer-term ocean carbon uptake is controlled more by the carbon storage in the ocean interior and might instead depend upon depth-mean ocean temperature [18].

Illustration of Land and Ocean Carbon Dependence on Atmospheric CO_2 and Temperature

The dependence of the cumulative carbon uptake of the land and ocean on atmospheric CO_2 and surface temperature is illustrated using 100 simulations of an efficient Earth system model [19] (Fig. 1b, c). The model is integrated with an imposed 1% annual rise in atmospheric CO_2 under multiple warming responses for 140 years until $4 \times CO_2$ is reached. This ensemble of simulations span parameter space by including variations in the cloud climate

feedback, altering the climate sensitivity from typically 2 to around 6 K per doubling of atmospheric CO_2 , although all simulations include the same representation of the terrestrial response.

Cumulative land and ocean carbon uptake increases with rising atmospheric CO_2 (Fig. 1b, c), but this carbon uptake is reduced with the rising surface temperature. While the cumulative carbon uptake alters in a consistent way with increased uptake with atmospheric CO_2 and decreased uptake with warming, there is a change in the curvature in this relationship in Fig. 1b, c. While the precise quantitative values for this dependence are likely to vary between climate models, the general dependencies of the carbon inventories depicted here on changes in atmospheric CO_2 and surface temperature are likely to be robust.

Linear Closure for Carbon-Cycle Feedbacks

Returning to the definition of the land and ocean carbon inventories in terms of the climate system (3), changes in the land and ocean carbon inventories may be expressed as

$$\Delta C_{land} + \Delta C_{ocean} = F(C_{atmos,0} + \Delta C_{atmos}, T_0 + \Delta T) - F(C_{atmos,0}, T_0), \quad (4)$$

where terms in the parentheses represent the independent variables and subscript 0 denotes the time of the pre-industrial.

The function defining the carbon state of the climate system, F , may be expanded as a Taylor series [15], such that

$$\begin{aligned} \Delta C_{land} + \Delta C_{ocean} = & \left. \frac{\partial F}{\partial C_{atmos}} \right|_0 \Delta C_{atmos} + \left. \frac{\partial F}{\partial T} \right|_0 \Delta T \\ & + \left. \frac{\partial^2 F}{\partial C_{atmos} \partial T} \right|_0 \Delta C_{atmos} \Delta T \\ & + \frac{1}{2} \left. \frac{\partial^2 F}{\partial C_{atmos}^2} \right|_0 \Delta C_{atmos}^2 + \frac{1}{2} \left. \frac{\partial^2 F}{\partial T^2} \right|_0 \\ & \times \Delta T^2 + R_3, \end{aligned} \quad (5)$$

where R_3 represents the effect of higher derivative terms and each differential term is formally evaluated with other variables kept constant at their pre-industrial value. Approximating by neglecting the second-order and higher derivative terms leads to a linear relationship between changes in the ocean and terrestrial carbon inventories and the changes in the climate state due to atmospheric carbon and global-mean temperature [11, 12],

$$\Delta C_{land} + \Delta C_{ocean} = \left. \frac{\partial F}{\partial C_{atmos}} \right|_0 \Delta C_{atmos} + \left. \frac{\partial F}{\partial T} \right|_0 \Delta T, \quad (6)$$

which may be written as

$$\Delta C_{land} + \Delta C_{ocean} = \beta \Delta C_{atmos} + \gamma \Delta T, \quad (7)$$

Table 1 Carbon-cycle concentration and climate feedback terms, β and γ , for the land and ocean from 5 CMIP5 Earth system models forced by an annual 1% increase in atmospheric CO₂. Diagnostics are from years 121 to 140 with decadal variability filtered using a moving-average filter

CMIP5 model	β ND	β_{land} ND	β_{ocean} ND	γ PgC K ⁻¹	γ_{land} PgC K ⁻¹	γ_{ocean} PgC K ⁻¹	Reference
CanESM2	0.80	0.47	0.33	−71.8	−61.0	−10.8	[21]
HadGEM2-ES	1.06	0.67	0.39	−70.9	−53.2	−17.7	[22]
IPSL-CM5A-LR	1.21	0.76	0.45	−73.8	−57.6	−16.2	[23]
MPI-ESM-LR	1.16	0.71	0.45	−100.6	−83.0	−17.6	[24]
Nor-ESM1-ME	0.55	0.15	0.40	−18.7	−1.8	−16.9	[25]
Mean	0.95	0.55	0.40	−67.1	−51.3	−15.8	
Std	0.28	0.25	0.05	29.8	30.0	2.9	
Std/mean	0.30	0.46	0.13	0.44	0.58	0.18	

where carbon-cycle parameters are defined by $\beta \equiv \left. \frac{\partial F}{\partial C_{\text{atmos}}} \right|_0$ and $\gamma \equiv \left. \frac{\partial F}{\partial T} \right|_0$, and $\beta \Delta C_{\text{atmos}}$ is referred to as the carbon-concentration feedback and $\gamma \Delta T$ as the carbon-climate feedback; β and γ measure the slope of the relationship relative to the pre-industrial (with the differentials being evaluated at pre-industrial time marked as 0), rather than by the local slope for a later time.

An increase in the land and ocean carbon inventories following an increase in atmospheric carbon is defined by a positive β , while an increase in these carbon inventories following an increase in surface temperature is defined by a positive γ . When viewed from an atmospheric perspective, positive values for β and γ correspond to a negative response and feedback as the atmospheric carbon inventory is decreased, while when viewed from a terrestrial or ocean perspective, these positive values for β and γ correspond to a positive response and feedback since the land and ocean carbon reserves are enhanced.

Diagnostics of Carbon Inventory Changes for CMIP5 Earth System Models

The carbon responses and feedbacks are now illustrated for 5 different representative coupled model intercomparison project phase 5 (CMIP5) Earth system models (Table 1); for a fuller range of 7 to 11 CMIP5 models, also see diagnostics by [14, 15]. All the climate models are forced by an idealised 1% annual increase in atmospheric CO₂ that drives enhanced radiative forcing and leads to surface warming (Fig. 2a, b). This addition of carbon increases each of the atmospheric, land and ocean carbon inventories (Fig. 2c, black, dashed and full lines) with a wider range in the land response than in the ocean response; the ocean uptake may be further interpreted by separating into different carbon pools (Fig. 2d).

Diagnostics of Carbon-Cycle Feedbacks for CMIP5 Earth System Models

The combined changes in land and ocean carbon inventories are now considered in terms of their separate relationships with atmospheric CO₂ and surface temperature. Given the application of a Taylor expansion, β should formally be evaluated with no climate change and surface temperature kept constant, while γ should be evaluated with atmospheric CO₂ kept constant. The carbon-cycle feedback parameters, β and γ , are estimated from the bulk land and ocean carbon changes diagnosed in coupled climate-carbon model experiments with different elements of the carbon-cycle or radiative forcing switched on or off [12–15]; there is also an alternative definition of carbon-cycle feedback parameters based upon the carbon flux to the atmosphere [20], rather than in terms of inventory changes. To diagnose the carbon-cycle feedbacks, three model versions have been traditionally used: a fully coupled, a radiatively coupled and a biogeochemically coupled version:

1. In the fully coupled simulation, the increase in atmospheric CO₂ affects the radiative forcing and biogeochemical processes acting in the land and the ocean.
2. In the radiatively coupled simulation, the increase in atmospheric CO₂ affects the radiative forcing and leads to associated climate change, while for the land and ocean biogeochemistry the pre-industrial constant atmospheric CO₂ is prescribed.
3. In the biogeochemically coupled simulation, the increase in the atmospheric CO₂ affects the land and ocean biogeochemical processes, but not the radiative forcing which retains its pre-industrial value.

Any combination of these three runs may be employed to estimate the carbon-cycle feedbacks. However, each

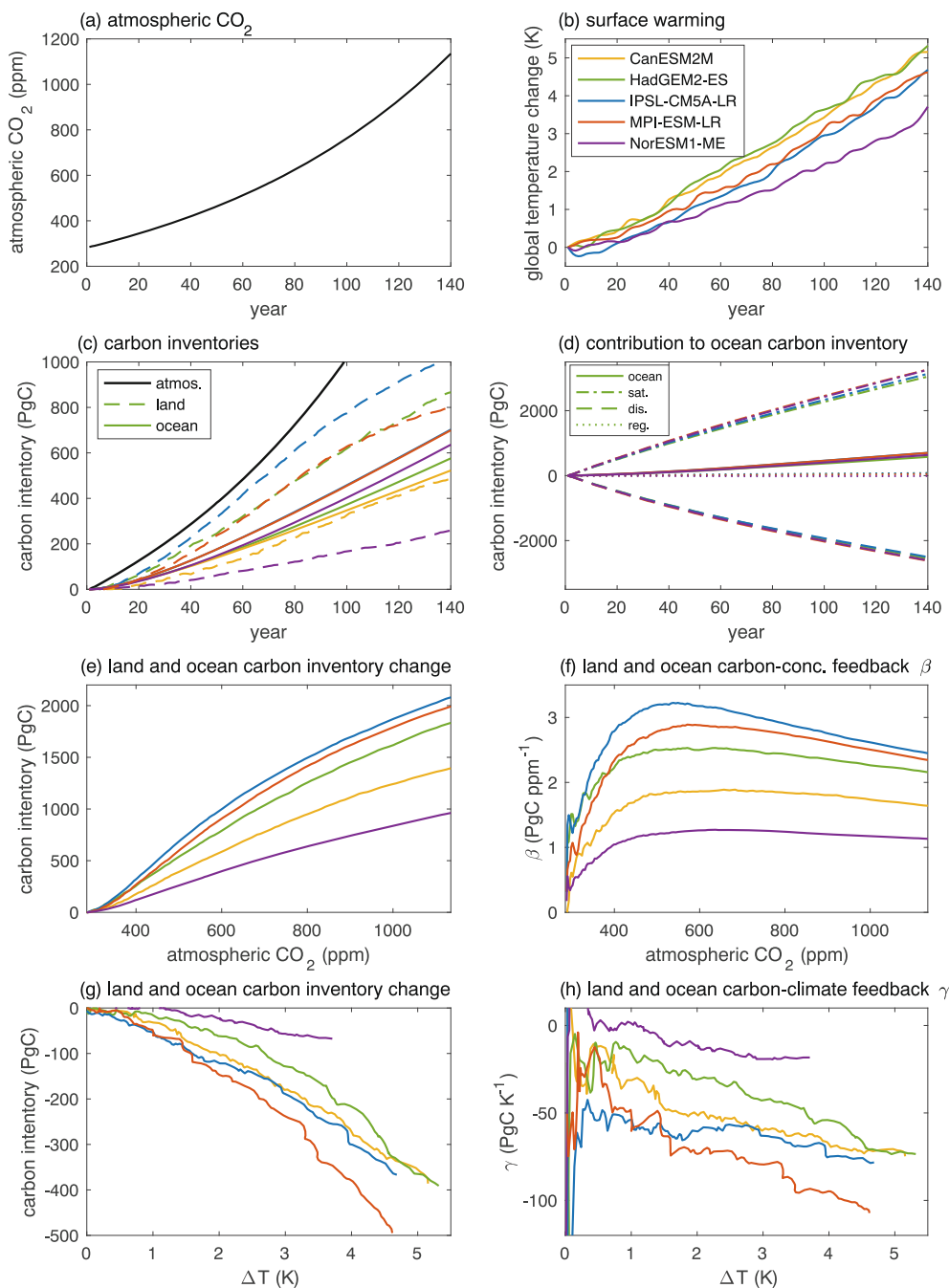


Fig. 2 Response of 5 CMIP5 Earth system models over 140 years (Table 1): (a) an imposed 1% annual rise in atmospheric CO₂ (ppm); (b) the global-mean change in surface air temperature (K); (c) increase in carbon inventories for the atmosphere (black line), the land (dashed lines) and the ocean (full coloured lines); and (d) the separate saturated, disequilibrium and regenerated contributions to the ocean carbon

inventory. The changes in the land and ocean carbon inventories (PgC) are presented in terms of separate changes due to changes in (e) atmospheric CO₂ (ppm) and (f) its carbon-concentration feedback term, $\beta^* = (C_{\text{atmos}}/C_{O_2})\beta$ in PgC ppm⁻¹ and in (g) surface temperature (K) and (h) its carbon-climate feedback term, γ in PgC K⁻¹

of the different combinations gives different results due to the nonlinearity of the system, such that the carbon fluxes diagnosed from the sum of the radiatively and the biogeochemically coupled simulations do not add up to that of the fully coupled simulation, and so that the

feedbacks from each of these combinations have different interpretations [15].

Here, we use the fully coupled simulation combined with the biogeochemically coupled simulations to estimate the feedbacks, such that the carbon-concentration feedback

parameter, β , is estimated directly from the biogeochemically coupled run and the carbon-climate feedback parameter, γ , is estimated from the difference between the fully coupled and the biogeochemically coupled runs and represents the effect of climate change under rising atmospheric CO_2 .

The combined land and ocean carbon uptake increases with rising atmospheric CO_2 with the slope of the curve relative to the pre-industrial peaking around 500 ppm (Fig. 2e). The carbon-concentration feedback parameter, β , measures the slope of this curve, so that β increases to a maximum at around 500 ppm and then decreases for higher atmospheric CO_2 (Fig. 2f). This shape of response suggests that there is a competition of different mechanisms acting to enhance or diminish the rate of land and ocean uptake of carbon with increasing atmospheric CO_2 .

In contrast, the combined land and ocean carbon inventories decrease with the rising surface temperature (Fig. 2g) and, in most models, the rate of decrease becomes larger with more pronounced surface warming. The carbon-climate feedback parameter, γ , measures the slope of this curve relative to the pre-industrial and is negative with increasingly negative values with a higher surface temperature (Fig. 2h).

The dependence of the land and ocean carbon inventories on atmospheric CO_2 (Fig. 2e, f) implies that carbon-cycle parameters should only be compared for climate model integrations using the same forcing scenarios (as in Table 1), otherwise there will be automatically much larger differences due to the amount of carbon emitted to the atmosphere.

Separation of Land and Ocean Carbon-Cycle Feedbacks

The carbon response of the combined land and ocean system has so far been considered and naturally, it is of interest to separate their responses. However, the inter-connection of the land and ocean carbon systems makes this separation into individual land and ocean carbon responses more ambiguous than expected. For example, a carbon uptake by the land system may lead to an immediate decrease in atmospheric carbon, which will then cause an outgassing of carbon from the ocean to the atmosphere, acting to partly offset the original decrease in atmospheric carbon and instead decrease the ocean store of carbon. Hence, the coupled nature of the carbon system means that any changes in atmospheric carbon estimated solely from a change in a land or ocean carbon inventory should be viewed as an upper bound, as compensation by the other part of the carbon system is ignored.

Accepting this caveat, the carbon inventory changes are next separately considered in terms of global diagnostics for

the land and ocean in terms of their contributions to β and γ , such that $\beta = \beta_{\text{land}} + \beta_{\text{ocean}}$ and $\gamma = \gamma_{\text{land}} + \gamma_{\text{ocean}}$. The local regional responses may differ from these global-mean responses, particularly by having different signs for the climate response, γ .

Terrestrial Carbon-Cycle Feedbacks

The change in the land carbon inventory may be written in terms of the linear carbon-cycle framework (7) as

$$\Delta C_{\text{land}} = \beta_{\text{land}} \Delta C_{\text{atmos}} + \gamma_{\text{land}} \Delta T. \quad (8)$$

The carbon-concentration feedback, $\beta_{\text{land}} \Delta C_{\text{atmos}}$, represents how the land carbon inventory responds to higher atmospheric CO_2 . The primary contribution is through how photosynthesis is stimulated by elevated atmospheric CO_2 , so increasing the land carbon inventory and decreasing the atmospheric response to carbon emissions. There is positive β_{land} with carbon emissions with the highest values peaking for atmospheric CO_2 between 500 and 600 ppm (Table 1 for 5 different Earth system models) [12, 14].

The carbon-climate feedback, $\gamma_{\text{land}} \Delta T$, represents the effects of a warmer climate on the land carbon inventory. Surface warming leads to an increase in plant and soil respiration, so acting to decrease the land carbon inventory. Changes in precipitation and nutrient availability also alter plant photosynthesis and changes in the abundance and distribution of vegetation. The carbon-climate feedback parameter, γ_{land} , is negative and becomes increasingly negative with greater surface warming (Table 1) [12, 14].

To gain insight, the terrestrial carbon pool may be separated into carbon pools associated with vegetation and soil [26],

$$\Delta C_{\text{land}} = \Delta C_{\text{veg}} + \Delta C_{\text{soil}} \quad (9)$$

The total carbon stored in vegetation is increased by photosynthesis and reduced by plant respiration and litter fall, while the soil carbon is increased with litter fall and decreased with soil respiration [27]; further multiple components may be included, such as soft plant and woody components with the soft plant material decaying more rapidly than woody material. The land carbon feedback parameters, β_{land} and γ_{land} , may then be connected mechanistically to carbon cycled within these vegetation and soil pools. For example, Earth system model diagnostics reveal elevated atmospheric CO_2 stimulating plant productivity, as well as enhanced litter fall and heterotrophic respiration, which leads to a larger β_{land} [26].

In summary, changes in terrestrial carbon storage are controlled by a balance between an increase in carbon storage from photosynthesis being stimulated by increased atmospheric CO_2 and partly opposed by a decrease in carbon storage from plant and soil respiration being

enhanced by surface warming [27, 28]. These responses are also affected by changes in rainfall, nutrient availability and dynamic changes in vegetation [29, 30].

Ocean Carbon-Cycle Feedbacks

The change in the ocean carbon inventory may be again written in terms of a linear carbon-cycle framework (7) as

$$\Delta C_{\text{ocean}} = \beta_{\text{ocean}} \Delta C_{\text{atmos}} + \gamma_{\text{ocean}} \Delta T. \quad (10)$$

The carbon-concentration feedback, $\beta_{\text{ocean}} \Delta C_{\text{atmos}}$, represents how the ocean carbon inventory responds to higher atmospheric CO_2 . The primary contribution is how the ocean takes up more carbon in the form of dissolved inorganic carbon, DIC, with higher atmospheric CO_2 . There is a positive β_{ocean} with carbon emissions with the highest values peaking for atmospheric CO_2 around 500 ppm (Table 1) [12, 14, 15]. This bell-shaped response is due to the ocean increasing its ability to hold more carbon with higher atmospheric CO_2 until this increase is partly offset by the opposing effect of the ocean becoming more acidic, which inhibits the ability to hold more carbon.

The carbon-climate feedback, $\gamma_{\text{ocean}} \Delta T$, represents the effects of a warmer climate on the ocean carbon inventory. The possible effects of a warmer climate include (i) reduced ventilation of the ocean interior involving a thinner surface mixed layer that acts as the interface between the atmosphere and ocean, an increase in ocean stratification and a decrease in the ocean overturning circulation; (ii) a decrease in solubility of carbon with warmer waters; and (iii) a possible weakening in biological drawdown of carbon. The carbon-climate feedback parameter, γ_{ocean} , is negative (Table 1) and becomes increasingly negative with greater surface warming [12, 14, 15], probably due to how ocean ventilation is inhibited in a warmer climate [31]. However, this temperature-driven climate change in the carbon response is much larger for the land than the ocean on a centennial timescale (Table 1).

To gain insight into the changes in the ocean carbon inventory, the ocean dissolved inorganic carbon, DIC, may be separated into preformed and regenerated carbon pools,

$$\text{DIC} = \text{DIC}_{\text{pre}} + \text{DIC}_{\text{reg}}, \quad (11)$$

with the preformed carbon, DIC_{pre} , representing the carbon physically transferred from the surface ocean and the regenerated carbon, DIC_{reg} , representing the carbon that is regenerated from biological material. The preformed carbon may be further separated into saturated and disequilibrium pools, such that

$$\text{DIC} = \text{DIC}_{\text{sat}} + \text{DIC}_{\text{reg}} + \text{DIC}_{\text{dis}}, \quad (12)$$

where DIC_{sat} represents the amount of carbon the ocean holds if in equilibrium with atmospheric CO_2 at that time

and DIC_{dis} represents the extent that the ocean carbon departs from this equilibrium with the atmosphere [32–34]. These pools of dissolved inorganic carbon equate to ocean carbon inventories when integrating the DIC by mass over the global ocean, such that

$$\Delta C_{\text{ocean}} = \Delta C_{\text{ocean,sat}} + \Delta C_{\text{ocean,reg}} + \Delta C_{\text{ocean,dis}}. \quad (13)$$

For the 1% annual increase in atmospheric CO_2 experiments, the saturated ocean carbon inventory strongly increases in time following the rise in atmospheric CO_2 , although the rate of these increases slightly declines for higher atmospheric CO_2 (Fig. 2d, dash dot lines). There is only a very slight increase in the regenerated carbon inventory over this centennial timescale (Fig. 2d, dotted lines). Instead the disequilibrium carbon is large and negative (Fig. 2d, dashed lines), representing how the ocean carbon is not keeping pace with the saturated carbon expected if there was an atmosphere-ocean equilibrium for carbon. This disequilibrium carbon is affected with the rate of increase in atmospheric CO_2 and the rate of ocean ventilation, controlling the carbon transfer from the ocean surface mixed layer into the ocean thermocline and deep ocean. There is an overall ocean carbon uptake (Fig. 2d, full lines) so that the positive increase in the saturated response is not completely offset by the negative increase in the disequilibrium carbon.

The carbon-cycle feedback parameters, β_{ocean} and γ_{ocean} , may be diagnosed for the preformed and regenerated terms [18], and each of the saturated, regenerated and disequilibrium terms making up ocean DIC [31]. The advantage of this separation is that the saturated terms for β_{ocean} and γ_{ocean} may be defined from theory, although the disequilibrium terms, especially for γ_{ocean} still need to be diagnosed from model integrations.

The carbon-concentration feedback parameter, β_{ocean} , is positive and peaks in magnitude at between 400 and 500 ppm with variations in atmospheric CO_2 . The saturated contribution to β_{ocean} is strongly positive representing how the ocean holds more carbon with an increase in atmospheric CO_2 . This saturated contribution decreases in magnitude though with increasing atmospheric CO_2 due to the ocean taking up more carbon and acidifying and so decreasing its capacity to buffer changes in atmospheric CO_2 . The disequilibrium contribution to β_{ocean} is negative and represents the extent that the ocean has not kept pace with an atmosphere-ocean equilibrium during the period of increasing atmospheric CO_2 . The disequilibrium contribution to β_{ocean} decreases in magnitude with increasing atmospheric CO_2 .

The carbon-climate feedback parameter, γ_{ocean} , becomes generally more negative with greater surface warming, which involves three different contributions: (i) surface warming leads to the solubility of the ocean decreasing, making the saturated carbon pool smaller for the same atmospheric CO_2 , so that the saturated contribution to

γ_{ocean} is negative; (ii) surface warming decreases ocean ventilation and lengthens the residence time below the surface ocean, which increases the regenerated nutrient and carbon pool [15], so that the regenerated contribution to γ_{ocean} is positive; and (iii) the disequilibrium carbon response is sensitive to both changes in the saturated carbon and the ventilation, and the change in disequilibrium carbon provides a negative contribution to γ_{ocean} when evaluated from the difference between the fully coupled and the biogeochemically coupled model simulations. There is an overall negative contribution from the changes in solubility and disequilibrium dominating over the changes in the regenerated carbon [35].

In summary, the carbon storage in the ocean increases with elevated atmospheric CO_2 , controlled by ocean carbonate chemistry, although this additional ocean carbon uptake decreases in a warmer climate from a thinning of the mixed layer, weaker ventilation and overturning, and a decrease in solubility and possible changes in biology.

Summary for Carbon-Cycle Feedbacks for the Land and Ocean

Carbon-cycle feedbacks represent the extent that the land and ocean carbon inventories respond to the effects of carbon emissions, overall acting to decrease the expected rise in atmospheric CO_2 and so provide a negative response and feedback to the original perturbation in the atmosphere; as illustrated in Table 1 and in other diagnostics of Earth system models [12–14]. The dominant contribution is from the effect of the carbon-concentration feedback, $\beta \Delta C_{\text{atmos}} > 0$, acting to enhance the combined land and ocean carbon inventories. There is a smaller effect of the carbon-climate feedback, $\gamma \Delta T < 0$, with a warmer climate acting to reduce the rise in the combined land and ocean carbon inventories.

Limitations of the Carbon-Cycle Feedback Approach

The carbon-cycle feedback approach is very useful by providing an accessible measure of how the carbon system evolves in a climate model and may be used to define differences in the carbon cycling of these climate models [11–14]. However, at the same time, there are several limitations in the carbon-cycle framework that are important to acknowledge:

1. The original linearisation of the function defining the carbon state of the climate system, F in Eq. 6, ignores non-linear dependence in how carbon inventories alter with atmospheric carbon and surface temperature (represented by the curvature of the contours on Fig. 1b, c).

- This non-linear error is important in providing uncertainty in the estimate of the carbon feedback [13–15];
2. The land and ocean carbon systems may more directly be controlled by other variables. For the land, the dependent variables may be the water supply to plants and the nutrient availability to allow plants to photosynthesise more with elevated atmospheric CO_2 . This dependence on water and nutrient supply may not be captured by a general dependence on temperature. For the ocean, the dependent variable may be the extent of ocean ventilation, which is more directly related to the thickness of the surface mixed layer, the subduction rate defining the strength of exchange between the mixed layer and the ocean interior, and the strength of meridional overturning. The pattern of climate change is also potentially important in altering the regional mechanisms controlling the land and ocean carbon store;
 3. By themselves, the carbon-cycle feedback parameters, β and γ , are usually diagnosed from differences in climate-model experiments with selected components switched on and off [12–14] and so are difficult to diagnose from observations. While estimates of the carbon-cycle feedback parameters have been made using model simulations constrained by observations [36], it is important that the observations and model simulations cover the same period, rather than compare inferences made over a historical period with relatively small changes and over a climate projection with very large changes.

While noting these limitations, we next consider how the carbon-cycle feedback parameters, β and γ , representing the effects of a carbon response and feedback, may be converted into a carbon response and feedback parameter with the same units as the physical climate feedback parameter.

Connecting to a Carbon Response and Feedback Parameter

Our goal is now to connect the land and ocean carbon changes to a carbon response and feedback parameter following [13, 37], which is directly analogous to a physical climate feedback parameter. Both the carbon and climate feedback parameters when multiplied by a change in surface temperature provides a change in radiative forcing in W m^{-2} , which defines aspects of the climate response to carbon forcing. To make this connection as simple as possible, we choose to combine the effects of a carbon response and feedback together, even though in the physical system, the climate response involving planetary heat uptake and the climate feedback are usually separated from each other.

Derivation for the Carbon Response and Feedback Parameter

The radiative forcing expected from carbon emissions, $R(\Delta C_{em})$, may be defined in terms of the radiative forcing due to the increase in atmospheric carbon, $R(\Delta C_{atmos})$, plus the radiative forcing that would otherwise have occurred, $R_{CO_2}^{feedback}$, representing the effect of a combined carbon response and feedback due to changes in the land and ocean carbon inventories,

$$R(\Delta C_{em}) = R(\Delta C_{atmos}) + R_{CO_2}^{feedback}. \tag{14}$$

forcing from atmospheric land and ocean emissions forcing response and feedback

Here the carbon response and feedback, $R_{CO_2}^{feedback}$, are defined as positive for a land and ocean uptake of carbon, so acts to make the radiative forcing in the atmosphere, $R(\Delta C_{atmos})$, smaller in magnitude than the radiative forcing expected from carbon emissions, $R(\Delta C_{em})$, where Δ represents the change since the pre-industrial period; see Fig. 1 of [37] for this separation of radiative forcing and feedback. Next consider how $R(\Delta C_{atmos})$ and $R_{CO_2}^{feedback}$ on the right-hand side of (14) may be expressed.

Firstly, in an empirical global radiative balance [38, 39], the radiative forcing from atmospheric CO₂, $R(\Delta C_{atmos})$, drives a climate feedback and climate response, which are expressed in terms of a physical feedback term, $\lambda_{climate}\Delta T$, plus a planetary heat uptake, N ,

$$R(\Delta C_{atmos}) = \lambda_{climate}\Delta T + N, \tag{15}$$

atmospheric forcing climate feedback heat uptake

where R is expressed as a global-mean radiative forcing per unit surface area in $W\ m^{-2}$ and is a function of the change in the atmospheric carbon inventory ΔC_{atmos} , N is the global-mean heat uptake per unit area in $W\ m^{-2}$, $\lambda_{climate}$ is the physical climate feedback parameter in $W\ m^{-2}K^{-1}$ and ΔT is the change in the global-mean surface air temperature since the pre-industrial era in K. The climate feedback parameter includes the effect of the Planck feedback of enhanced longwave radiation from a warmer surface together with the radiative feedbacks from changes in water vapour, surface albedo and cloud (Fig. 3, grey and black circles from [7]). By defining each of the radiative terms in Eq. 15 as positive, the Planck radiative response of enhanced longwave radiation from a warmer surface is represented by a positive $\lambda_{climate}$, which provides a negative feedback acting to partly offset the effect of the radiative forcing and decrease the magnitude of surface warming.

Secondly, drawing upon the analogy with the physical climate system, we define the radiative response and feedback from the changes in the land and ocean carbon

inventories, $R_{CO_2}^{feedback}$, in terms of the product of a carbon response and feedback parameter, λ_{carbon} , and the change in global-mean surface temperature, ΔT ,

$$R_{CO_2}^{feedback} \equiv \lambda_{carbon}\Delta T, \tag{16}$$

where λ_{carbon} is in $W\ m^{-2}K^{-1}$ and ΔT in K.

Hence, combining (14) to (16), the global-mean radiative balance may be expressed in terms of the radiative forcing expected from carbon emissions, $R(\Delta C_{em})$, as

$$R(\Delta C_{em}) = \lambda_{carbon}\Delta T + \lambda_{climate}\Delta T + N. \tag{17}$$

forcing from carbon response climate heat uptake emissions and feedback feedback

The radiative forcing expected from emissions, $R(\Delta C_{em})$, in Eq. 17 may then be viewed as being offset by a carbon response and feedback, $\lambda_{carbon}\Delta T$, representing the land and ocean removal of carbon from the atmosphere, plus a climate feedback, $\lambda_{climate}\Delta T$, radiating heat to space, and plus a heat uptake, N , representing a planetary gain in heat [37].

In order to derive an expression for λ_{carbon} in Eq. 17, we now apply a series of approximations following [13]:

1. The radiative forcing from atmospheric CO₂ is usually represented by a saturating logarithmic dependence [40], $R(\Delta C_{atmos}(t)) = a \ln(C_{atmos}(t)/C_{atmos}(t_0))$, which may be linearly approximated by

$$R(\Delta C_{atmos}) \simeq \phi\Delta C_{atmos}, \tag{18}$$

where a is a radiative forcing coefficient for atmospheric CO₂ and $\phi \simeq a/C_{atmos}(t_1)$ is a linearised estimate of the slope of the radiative forcing with respect to the atmospheric carbon inventory (with $C_{atmos}(t_1)$ the atmospheric carbon inventory at the time t_1 when the radiative forcing is estimated); ϕ has a mean and standard deviation of $0.004 \pm 0.0006\ W\ m^{-2}PgC^{-1}$ when evaluated from years 121 to 140 in our diagnostics of 5 Earth system models (Table 2).

2. The radiative forcing, $R_{CO_2}^{feedback}$, from the carbon response and feedback of the land and ocean system is assumed to depend upon the combined land and ocean carbon inventory changes, $\Delta C_{land} + \Delta C_{ocean}$, which is linearly connected to the atmospheric carbon and global-mean temperature (7) with a functional relationship given by

$$R_{CO_2}^{feedback} = R_{CO_2}^{feedback}(\beta\Delta C_{atmos} + \gamma\Delta T), \tag{19}$$

and then approximated as in Eq. 18 by

$$R_{CO_2}^{feedback} \simeq \phi(\beta\Delta C_{atmos} + \gamma\Delta T). \tag{20}$$

3. Returning to the empirical global radiative balance (15), the planetary heat uptake is approximated by a diffusive

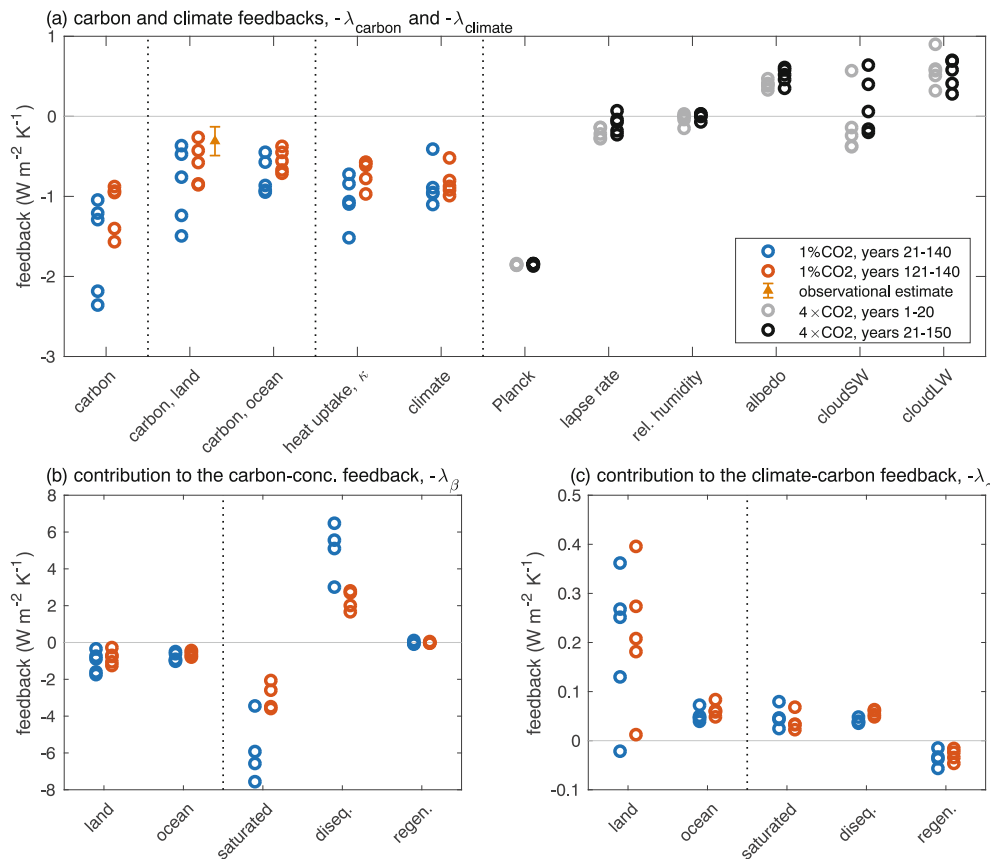


Fig. 3 (a) Carbon response and feedback parameter, $-\lambda_{\text{carbon}}$, for the land and ocean, the efficiency of heat uptake, $-\kappa$ and the climate feedback parameter, $-\lambda_{\text{climate}}$, all in $W m^{-2} K^{-1}$; their values are plotted multiplied by a negative sign, so that an overall positive value indicates surface warming and a positive feedback for surface temperature. Diagnostics follow [13] and are from 5 CMIP5 Earth system models (Table 1) with a 1% annual increase in atmospheric CO_2 for years 21 to 140 (blue circles) and years 121 to 140 (orange circles). An estimate of the land carbon feedback parameter from observational analyses (orange triangle for the mean and bounds for two standard deviations) is included following [37], which accounts for an ocean adjustment. In

addition, the components of the physical climate feedback, $-\lambda_{\text{climate}}$, from Planck, lapse rate, relative humidity, surface albedo, shortwave and longwave radiation changes from clouds are included for the same climate models from [7] for a $4 \times CO_2$ experiments from years 1 to 20 (grey circles) and years 21 to 150 (black circles). The carbon feedback parameter, λ_{carbon} , is separated into contributions depending on (b) the carbon-concentration feedback, λ_{β} , and (c) the carbon-climate feedback, λ_{γ} , connected with changes in the land and ocean carbon inventories, and the saturated, disequilibrium and regenerated ocean carbon inventories

closure, $N = \kappa \Delta T$,

$$R(\Delta C_{\text{atmos}}) \simeq (\lambda_{\text{Climate}} + \kappa) \Delta T. \quad (21)$$

so providing a link between the change in the atmospheric carbon inventory and surface temperature change using Eq. 18, such that

$$\phi \Delta C_{\text{atmos}} \simeq (\lambda_{\text{Climate}} + \kappa) \Delta T, \quad (22)$$

where κ is the ocean heat uptake efficiency in a closure for ocean heat uptake; κ has a mean and standard deviation of $0.70 \pm 0.17 W m^{-2} K^{-1}$ when evaluated from years 121 to 140 in our diagnostics of 5 Earth system models (Table 2).

Combining the definition of the carbon response and feedback parameter λ_{carbon} in Eqs. 16 with Eqs. 20 and 22,

then provides

$$R_{CO_2}^{\text{feedback}} \equiv \lambda_{\text{carbon}} \Delta T \simeq (\beta(\lambda_{\text{Climate}} + \kappa) + \phi \gamma) \Delta T, \quad (23)$$

so that λ_{carbon} is defined by

$$\lambda_{\text{carbon}} \simeq \underbrace{\beta(\lambda_{\text{climate}} + \kappa)}_{\text{carbon-concentration response and feedback}} + \underbrace{\phi \gamma}_{\text{carbon-climate response and feedback}}. \quad (24)$$

The carbon response and feedback parameter, λ_{carbon} , then is made up of two contributions, each including combined carbon and climate effects:

1. The carbon-concentration cycle parameter, β , measuring a dependence on atmospheric carbon, multiplied by the sum of the physical climate feedback, λ_{climate} , and transient heat uptake via κ , and

Table 2 Carbon-cycle feedback parameters, λ_{carbon} , separated into land and ocean contributions, and then further into β and γ contributions from 5 CMIP5 Earth system models

Model	λ_{carbon} $\text{Wm}^{-2}\text{K}^{-1}$	λ_{land} $\text{Wm}^{-2}\text{K}^{-1}$	λ_{ocean} $\text{Wm}^{-2}\text{K}^{-1}$	$\lambda_{\beta,\text{land}}$ $\text{Wm}^{-2}\text{K}^{-1}$	$\lambda_{\beta,\text{ocean}}$ $\text{Wm}^{-2}\text{K}^{-1}$	$\lambda_{\gamma,\text{land}}$ $\text{Wm}^{-2}\text{K}^{-1}$	$\lambda_{\gamma,\text{ocean}}$ $\text{Wm}^{-2}\text{K}^{-1}$	λ_{climate} $\text{Wm}^{-2}\text{K}^{-1}$	κ $\text{Wm}^{-2}\text{K}^{-1}$	ϕ $\text{Wm}^{-2}\text{PgC}^{-1}$
CanESM2	0.88	0.43	0.45	0.70	0.50	−0.27	−0.049	0.92	0.57	0.0045
HadGEM2-ES	0.96	0.58	0.38	0.76	0.44	−0.18	−0.060	0.52	0.61	0.0034
IPSL-CM5A-LR	1.40	0.84	0.56	1.05	0.62	−0.21	−0.059	0.80	0.58	0.0036
MPI-ESM-LR	1.58	0.86	0.72	1.25	0.80	−0.40	−0.084	0.99	0.78	0.0048
Nor-ESM1-ME	0.94	0.27	0.67	0.28	0.73	−0.01	−0.062	0.88	0.97	0.0036
Mean	1.16	0.60	0.56	0.81	0.62	−0.21	−0.063	0.82	0.70	0.0040
Std	0.32	0.26	0.14	0.37	0.15	0.14	0.013	0.18	0.17	0.0006
Std/mean	0.28	0.43	0.25	0.46	0.24	0.67	0.21	0.22	0.24	0.15

These carbon-feedback parameters are compared with climate feedback parameter, λ_{climate} , ocean heat uptake efficiency, κ , and derivative of radiative forcing from CO_2 on the atmospheric carbon inventory, ϕ . All of these parameters are evaluated from years 121 to 140 for the CMIP5 Earth system models forced by a 1% atmospheric CO_2 annual increase

- The carbon-climate cycle parameter, γ , measuring a dependence on climate change, multiplied by a dependence of the radiative forcing on atmospheric carbon, ϕ .

The carbon and climate responses and feedbacks are directly connected to each other through the carbon response and feedback parameter, λ_{carbon} , being proportional to the sum of the physical climate feedback parameter, λ_{climate} , and the ocean heat uptake efficiency, κ , plus the effect of the carbon-climate parameter, γ , in Eq. 24. A stronger physical climate feedback, λ_{climate} , such as with the Planck response dominating, acts to decrease the magnitude of surface warming, which in turn enhances the ability of the combined land and ocean uptake of carbon and provides a stronger carbon response offsetting the effect of carbon emissions. In a similar manner, a greater ocean heat uptake efficiency, κ , leads to a reduction in the magnitude of surface warming, so similarly increases the combined land and ocean uptake of carbon.

Carbon Response and Feedback Parameter Diagnosed from Earth System Models

The carbon response and feedback parameter, λ_{carbon} , is diagnosed using Eq. 24 from the combination of β , γ , λ_{climate} , κ and ϕ for each of the 5 Earth system models: λ_{carbon} ranges from 0.88 to 1.58 $\text{W m}^{-2}\text{K}^{-1}$ for the 5 models with a mean and standard deviation of $1.16 \pm 0.32 \text{ W m}^{-2}\text{K}^{-1}$ diagnosed from years 121 to 140 (Table 2); here, ϕ is estimated from the local slope of the radiative forcing versus atmospheric carbon inventory for years 121 to 140 using $R(\Delta C_{\text{atmos}}(t)) = a \ln(C_{\text{atmos}}(t)/C_{\text{atmos}}(t_0))$, with a from [41].

A positive λ_{carbon} represents the effect of an increased land and ocean carbon inventory, so acting to decrease the atmospheric carbon inventory, reduce the magnitude of the additional radiative forcing from atmospheric CO_2 and so decreases surface warming. Hence, a positive λ_{carbon} is acting as a negative feedback for surface temperature, so these diagnostics are plotted with a negative sign in Fig. 3 to have the same convention as physical feedbacks.

There are comparable contributions with the carbon feedback parameter from the land and ocean: λ_{carbon} for the land ranges from 0.27 to 0.86 $\text{W m}^{-2}\text{K}^{-1}$, while λ_{carbon} for the ocean ranges from 0.38 to 0.72 $\text{W m}^{-2}\text{K}^{-1}$ (Table 2). The land and ocean λ_{carbon} are made up of positive contributions for $\lambda_{\beta} \equiv \beta(\lambda_{\text{climate}} + \kappa)$ and smaller negative contribution from $\lambda_{\gamma} \equiv \phi\gamma$ in Eq. 24, although the latter ocean term is much smaller in magnitude (Fig. 3b,c; Table 2).

The inter-model spread in the estimates of the carbon feedback parameter reduces in time with a greater spread when diagnosed from years 21 to 140, than over the last two decades of the integration (Fig. 3, blue and orange circles).

Observational Estimate of the Land Carbon Feedback Parameter

The radiative feedback from the land may again be defined in terms of a carbon feedback parameter for the land multiplied by the change in surface temperature,

$$R_{\text{CO}_2,\text{land}}^{\text{feedback}} = \lambda_{\text{carbon,land}} \Delta T. \quad (25)$$

Following [37], the carbon feedback parameter for the land may be connected to the change in the land carbon inventory, ΔC_{land} ,

$$\lambda_{\text{carbon,land}} = \frac{a}{I_B} \frac{\Delta C_{\text{land}}}{\Delta T}, \quad (26)$$

where a is a radiative forcing coefficient for atmospheric CO_2 and I_B is the sum of the atmospheric carbon inventory and the ocean buffered carbon inventory [42, 43]. This relationship takes into account how a change in the land carbon inventory, ΔC_{land} , leads to changes in both the atmospheric and ocean carbon inventories [37].

Using constraints from the Global Carbon Budget [10], the land carbon feedback parameter, $\lambda_{\text{carbon,land}}$, is estimated to have a mean and standard deviation of $0.33 \pm 0.09 \text{ W m}^{-2}\text{K}^{-1}$ in the present day [37] (Fig. 3a, orange triangle). This estimate of $\lambda_{\text{carbon,land}}$ from Eq. 26 is smaller in magnitude than most of those model diagnostics from Eq. 24 due to different choices in how changes in the carbon inventory affect changes in the partitioning of carbon between the atmosphere and ocean (Fig. 3a, blue and orange circles).

Comparison With Climate Feedback Processes

The carbon feedback parameter, λ_{carbon} , is directly comparable with the climate feedback parameter, λ_{climate} , and the ocean heat uptake efficiency, κ , all having the same units of $\text{W m}^{-2}\text{K}^{-1}$ (Fig. 3a, blue and orange circles): λ_{carbon} is $1.16 \pm 0.32 \text{ W m}^{-2}\text{K}^{-1}$, while λ_{climate} is $0.82 \pm 0.18 \text{ W m}^{-2}\text{K}^{-1}$ and κ is $0.70 \pm 0.17 \text{ W m}^{-2}\text{K}^{-1}$ for the mean and standard deviations of the 5 climate models (Table 2).

The different components for λ_{carbon} for the carbon-concentration feedback for the land and ocean and the carbon-climate feedback for the land are comparable to the different contributions to λ_{climate} for the years 121 to 140 (Fig. 3a, grey and black circles).

The Planck response provides enhanced longwave radiation from a warmer surface feedback, defined here as a positive λ , and so provides a cooling with a negative feedback. This physical feedback is augmented by an overall negative feedback from water vapour through changes in the lapse rate and relative humidity, which is partly opposed by a positive feedback from changes in surface albedo and clouds [7] (Fig. 3a, grey and black circles). The cloud effect includes partly opposing longwave and shortwave effects: a warming effect from a thicker tropospheric clouds providing less longwave heat loss to space, a warming effect from less reflected solar radiation from decreasing low clouds in the tropics and midlatitudes, and a cooling effect from more reflected solar radiation from more low cloud at high latitudes [44]. Hence, the importance and complexity of the carbon-cycle feedbacks, involving partly opposing physical and biogeochemical responses, is comparable to the importance and complexity of the cloud feedbacks in the climate system.

Conclusions

The climate system is being systematically perturbed by carbon emissions [1, 10], driving rising atmospheric CO_2 and surface warming and ocean heat uptake. While there is clearly a warming of the climate system, the amount of carbon that may be emitted before exceeding warming targets is uncertain [45, 46]. Part of this uncertainty is due to how much of the emitted carbon is sequestered by the land and ocean systems, which may be viewed in terms of a carbon-cycle response and feedback providing an overall negative feedback to carbon emissions in the climate system.

The carbon-cycle framework provides a methodology to evaluate the carbon-cycle feedbacks, separated into effects due to rising atmospheric CO_2 and surface temperature [12–14]. This methodology is very useful by providing an accessible measure of how the carbon cycle operates in a complex climate model. The relative importance of the land and ocean is identified in terms of the carbon-cycle feedback, although there are inherent approximations through linearising their carbon response relative to the pre-industrial [15]. The usual practice of forcing climate models with prescribed atmospheric CO_2 automatically acts to combine the effect of carbon emissions and carbon-cycle feedbacks and possibly under-estimate their combined effect. Instead, there is a greater spread in climate projections when climate models are forced by emissions and there is an interaction with carbon-cycle feedbacks [47].

The carbon-cycle feedback framework by design identifies how the land and ocean carbon inventories depend upon atmospheric CO_2 and surface temperature, so ignores the complicating effects of non- CO_2 radiative forcing from other greenhouse gases and aerosols. Extending the framework to include the effect of other non- CO_2 greenhouse gases, such as methane, and their dependence on surface temperature and atmospheric mixing ratios is possible: there would need to be model integrations with different forcing scenarios and additional differences in coupled model integrations (with components switched on and off) to identify the changes in the surface temperature and atmospheric non- CO_2 greenhouse gas due to the non- CO_2 radiative forcing and the cycling of the non- CO_2 greenhouse gases.

While there are underlying approximations, extending the carbon-cycle feedback framework to evaluate a carbon response and feedback parameter [13, 37] allows direct comparison with other physical feedback processes contributing to a climate feedback parameter. The carbon and climate responses and feedbacks are directly connected to each other, such as with the carbon response and feedback parameter being proportional to the physical climate feedback parameter. The overall carbon feedback parameter is

comparable in magnitude to the physical climate feedback parameter and the uncertainty in particular components, particularly the land carbon feedback, are comparable to the uncertainty in the cloud feedback parameter.

Future work should focus on gaining mechanistic insight into those parts of the carbon-cycle feedback that are most significant, which may be achieved by considering further sub-components for the land and ocean carbon system or by considering the regional response of the carbon system. On centennial timescales, the carbon-cycle feedbacks are mainly due to the dependence of the land and ocean carbon inventories on atmospheric CO₂ and the dependence of the land carbon inventory to surface warming. However, on longer multi-centennial timescales after carbon emissions cease, the carbon-cycle response is likely to be much more dominated by the ocean rather than the land [48]. Greater mechanistic insight may be achieved by focussing on how physical and biogeochemical processes affect the carbon response for the land and ocean, such as by identifying the carbon responses for the soil and vegetation, and the carbon responses for the ocean saturated, regenerated and disequilibrium pools.

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Compliance with Ethical Standards

Conflict of Interests On behalf of all authors, the corresponding author states that there is no conflict of interest.

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