



The mechanisms of North Atlantic CO₂ uptake in a large Earth System Model ensemble

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Abstract. The oceans currently take up around a quarter of the carbon dioxide (CO₂) emitted by human activity. While stored in the ocean, this CO₂ is not influencing Earth's radiation budget; the ocean CO₂ sink therefore plays an important role in mitigating global warming. CO₂ uptake by the oceans is heterogeneous, with the subpolar North Atlantic being the strongest CO₂ sink region. Observations over the last 2 decades have indicated that CO₂ uptake by the subpolar North Atlantic sink can vary rapidly. Given the importance of this sink and its apparent variability, it is critical that we understand the mechanisms behind its operation. Here we explore the combined natural and anthropogenic subpolar North Atlantic CO₂ uptake across a large ensemble of Earth System Model simulations, and find that models show a peak in sink strength around the middle of the century after which CO₂ uptake begins to decline. We identify different drivers of change on interannual and multidecadal timescales. Short-term variability appears to be driven by fluctuations in regional seawater temperature and alkalinity, whereas the longer-term evolution throughout the coming century is largely occurring through a counterintuitive response to rising atmospheric CO₂ concentrations. At high atmospheric CO₂ concentrations the contrasting Revelle factors between the low latitude water and the subpolar gyre, combined with the transport of surface waters from the low latitudes to the subpolar gyre, means that the subpolar CO₂ uptake capacity is largely satisfied from its southern bound-

ary rather than through air–sea CO₂ flux. Our findings indicate that: (i) we can explain the mechanisms of subpolar North Atlantic CO₂ uptake variability across a broad range of Earth System Models; (ii) a focus on understanding the mechanisms behind contemporary variability may not directly tell us about how the sink will change in the future; (iii) to identify long-term change in the North Atlantic CO₂ sink we should focus observational resources on monitoring lower latitude as well as the subpolar seawater CO₂; (iv) recent observations of a weakening subpolar North Atlantic CO₂ sink may suggest that the sink strength has peaked and is in long-term decline.

1 Introduction

Our limited understanding of how the CO₂ emission to atmospheric CO₂ (CO₂^{atm}) concentration ratio will evolve through time constitutes one of the largest components of uncertainty in future climate projections (Booth et al., 2012). To constrain how this airborne fraction of CO₂ might change, and thereby link physical climate understanding to the development of CO₂ emission policy, we need to understand the behaviour of the major terrestrial and marine CO₂ sources and sinks (Friedlingstein et al., 2006).

Earth System Models (ESMs) are the most advanced tools we have available to calculate the link between CO₂ emis-

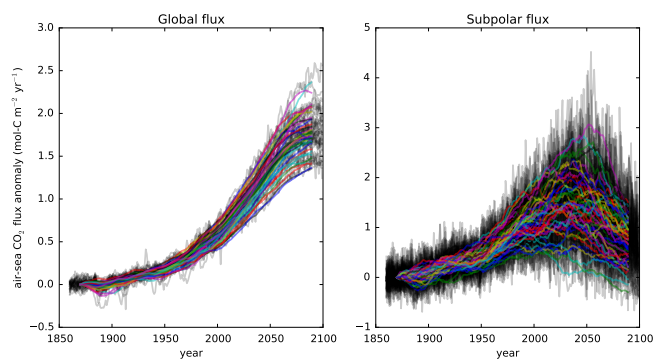


Figure 1. Left: globally averaged air–sea CO₂ flux, and right: North Atlantic subpolar region averaged air–sea CO₂ flux. Black lines represent annually averaged time series from all ESPPE members, and coloured lines represent those time series after application of a 20-year running mean.

sions and CO₂^{atm} concentrations. At a globally averaged scale, the current generation of Earth System Models, those developed and run for CMIP5 (Taylor et al., 2012), the 5th Climate Model Intercomparison Project, show good agreement on 21st century global ocean CO₂ uptake. With the exception of INM-CM4.0 (Volodin et al., 2010) the CMIP5 inter-model globally averaged ocean CO₂ uptake differences are smaller than the inter-scenario differences (Jones et al., 2013). At a regional level however, models do not agree. Furthermore, regional CO₂ uptake can behave very differently from that of the global mean (Fig. 1).

We need to understand the mechanisms behind differences in regional uptake to help us (i) validate models, and (ii) identify where and how to focus observations.

Whilst the carbon-cycle community is developing an increasingly comprehensive understanding of the mechanisms behind recent ocean CO₂ uptake variability in the North Atlantic (e.g. McKinley et al., 2004; Thomas et al., 2008; Ullman et al., 2009; Metzl et al., 2010; McKinley et al., 2011; Pèrez et al., 2013; Schuster and Watson, 2007), the Southern Ocean (e.g. Lenton and Matear, 2007; Le Quèrè et al., 2007; Lovenduski et al., 2013; Sallee et al., 2012; Ito et al., 2010; Lenton et al., 2009; Verdy et al., 2007), and potential broad-scale future ocean CO₂ uptake changes (e.g. Marinov et al., 2008; Murnane et al., 1999; Roy et al., 2011; Sarmiento and Le Quèrè, 1996), our understanding of the specific future mechanisms of change projected within comprehensive ESMs in these regions are much more limited (Sèfèrian et al., 2012; Russell et al., 2006; Halloran, 2012). Here we develop our understanding of the mechanisms controlling future subpolar North Atlantic CO₂ uptake within Earth System Models.

To understand why the North Atlantic CO₂ sink may be vulnerable to change, it is useful to review the factors that make the region such an intense CO₂ sink (Fig. 2; McKinley et al., 2011; Watson et al., 2009; Schuster et al., 2013).

Present-day high CO₂ uptake in the subpolar North Atlantic occurs because water that moves northwards as part of the Atlantic Meridional Overturning Circulation (AMOC) experiences steep thermal and chemical gradients and high biological activity (Rayner et al., 2003; Key et al., 2004; Carr et al., 2006). Biological activity exports carbon to depth in the form of sinking biological material, reducing surface carbon concentrations and increasing the air–sea CO₂ gradient. The cooling of water increases the solubility of CO₂ and speciates carbon into forms other than CO₂ (e.g. Zeebe and Wolf-Gladrow, 2001), further increasing the air–sea CO₂ gradient. Deep convection then removes water from contact with the atmosphere, potentially before it has had time to come into air–sea CO₂ equilibrium, maintaining a continuous strong air–sea CO₂ gradient – and therefore flux (Takahashi et al., 2009). A further complicating factor in the North Atlantic is that limited mixing between the subtropical and subpolar gyres allows the development of a strong biogeochemical gradient between waters with a high alkalinity to dissolved-carbon ratio (the warm and saline low-latitude waters), and waters with a low alkalinity to dissolved-carbon ratio (the cool and relatively fresh high-latitude waters; Key et al., 2004). This biogeochemical gradient results in a high CO₂ buffering capacity of low latitude water, permitting high anthropogenic CO₂ uptake, and a low buffering capacity at higher latitudes, limiting local future CO₂ uptake (Sabine et al., 2004). Combined with the advection of water from the subtropical to subpolar gyre, this latitudinal buffering gradient will likely impact the response of the sink to rising CO₂^{atm} (Völker et al., 2002).

Presently there is no agreement on the relative importance of the different factors described above in controlling past or future subpolar North Atlantic CO₂ uptake change. The hypothesised mechanisms for past decadal to multidecadal timescale changes in subpolar North Atlantic CO₂ uptake fall into four groups:

1. Biological drawdown. Evidence that CO₂ uptake variability may arise from the biological transport of carbon out of the surface ocean comes from the relative timing of observed surface ocean *p*CO₂ and chlorophyll change (Lefevre et al., 2004). The magnitude of this effect has however been questioned (Bennington et al., 2009).
2. Temperature. Both observational and model studies indicate that the temperature dependence of inorganic carbon speciation and CO₂ saturation is likely to have been an important player in air–sea CO₂ flux change on various timescales (Le Quèrè et al., 2000; Lefevre et al., 2004; McKinley et al., 2011; Omar and Olsen, 2006; Pèrez et al., 2013).
3. Vertical mixing. Changes in vertical mixing (through deep convection or stratification) has been proposed from both models and observations to be a dominant

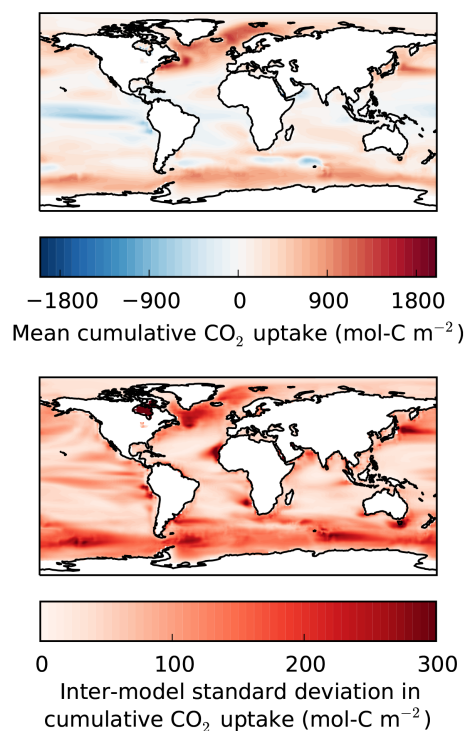


Figure 2. Cumulative sum of air–sea CO₂ flux between the years 1860 and 2100 (RCP8.5). (a) Mean and (b) inter-model standard deviation across ESPPE.

mechanism for changing the surface total Dissolved Inorganic Carbon (DIC) concentration and DIC-alkalinity ratio, and therefore changing the surface $p\text{CO}_2$ saturation (McKinley et al., 2004; Metzl et al., 2010; Schuster and Watson, 2007; Ullman et al., 2009), although this effect is likely to be damped by the associated changing vertical flux of nutrients and therefore biological CO₂ drawdown (McKinley et al., 2004).

4. Horizontal advection. Changes in surface ocean $p\text{CO}_2$ saturation driven by horizontal advection (rather than vertical transport) have been proposed from both modelling and observational studies (Omar and Olsen, 2006; Thomas et al., 2008). Debate however exists about the degree of long term DIC and alkalinity change, which brings in to question mechanisms implicating vertical and/or horizontal DIC and/or alkalinity transport (Corbière et al., 2007).

The diversity of proposed explanations for the observed subpolar North Atlantic CO₂ uptake variability could reflect different mechanisms dominating at different times and influencing uptake over different timescales. Many of the studies to-date have however examined approximately the same time periods. The range of proposed mechanisms therefore more likely reflects the difficulty of identifying causal drivers of change in a system, which despite huge effort, is still far

from completely observed. Similar problems apply to model-based studies. Proving causality in a model is straight forward when considering drivers external to the system (e.g. rising anthropogenic CO₂ emissions), because those drivers can be switched on and off, but when potentially important components of the mechanism are emergent properties of the model (e.g. the Meridional Overturning Circulation; MOC), these components can not simply be switched on and off, and even where they can be stopped (e.g. in the case of the AMOC by flooding the high-latitude North Atlantic/Arctic with freshwater), their role in the mechanism can not be isolated, because many other factors will change. To understand the mechanisms operating within ESMs, it can therefore often be useful to produce an even simpler model of the system (e.g. Good et al., 2011; Hooss et al., 2001; Meinshausen et al., 2011), one that emulates the complex model's behaviour, but also allows one to separately isolate the different components of the mechanisms. This is particularly valuable when attempting to understand common (or divergent) behaviours across a large suite of models.

Here we explore the mechanisms controlling ocean CO₂ uptake across a large ensemble of HadCM3- (3rd Hadley Centre Climate Model) based ESMs in which parameters have been systematically varied to efficiently sample a wide range of model behaviours (Lambert et al., 2013). We refer to this ensemble as the Earth System Perturbed Parameter Ensemble, or the ESPPE. We make use of the Atlantic carbon-cycle box model presented by Völker et al., (2002) to emulate the more complex ESM and simplify this large suite of simulations. The value of simplifying our large suite of ESM simulations in this way is that:

1. By using a single box model that replicates the behaviour of a wide range of Earth System Model formulations using only a single set of parameters (i.e. not retuning the simple model to emulate each different version of the more comprehensive model), one can be confident that the box model contains (and therefore that one has identified) the key processes important to the change of interest within those Earth System Model formulations.
2. Within a box model one can isolate and quantify the importance of each of these drivers of change by separately holding the inputs representing that driver constant and re-running the ensemble, or filtering input data to remove and isolate the component of variability of interest. As discussed, this cannot be done in an Earth System Model where properties like overturning circulation emerge from the physics and are therefore impossible to prescribe.
3. Using a box model shown to replicate (without retuning) the behaviour of multiple Earth System Model formulations, one can undertake numerous idealised simulations, and by doing so develop a thorough understand-

ing of the mechanisms at play. To do this with a full ESM would be extremely time consuming and expensive.

2 Methods

We attempt to isolate the mechanisms controlling North Atlantic CO₂ uptake in a 27-member ESM ensemble based on a carbon cycle version of the 3rd Hadley centre Climate Model HadCM3C (an updated version of Cox et al. (2000), with increased horizontal resolution and improved aerosol representation (Lambert et al., 2013), and using the Hadley centre Ocean Carbon Cycle (HadOCC) sub-model; Palmer and Totterdell, 2001), in which the atmosphere and ocean physics, the atmospheric sulphur cycle and terrestrial biogeochemistry parameters have been systematically varied to optimally sample parameter space (Lambert et al., 2013). The HadCM3C perturbed parameter ensemble is referred to herein as ESPPE (Earth System Perturbed Parameter Ensemble). The original ESPPE ensemble contains 57 members, but data corruption meant that only 27 of these members could be used in the analysis presented here. The ESPPE ensemble follows the CMIP5 RCP8.5 pathway (Riahi et al., 2007), and has a fully interactive carbon cycle: CO₂ emissions are prescribed, and atmospheric CO₂ concentrations calculated.

The box model we use to simplify the behaviour of the ESPPE represents the major features of the Atlantic basin and Atlantic sector of the Southern Ocean, and is made up of six boxes, three surface and three deep. The surface boxes represent the top 300 m of the ocean south of 30° S, the top 150 m of the tropical ocean between 30° S and 48° N, and the upper 300 m of the subpolar region north of 48° N (Fig. 3). The three subsurface boxes represent the deep high-latitude ocean north of 48° N, the intermediate depth ocean between 150 and 1000 m in the tropical region (30° S–48° N), and the remaining deep Atlantic ocean. The volume fluxes between the six boxes, and the temperature, salinity and alkalinity of those boxes are prescribed, as is the atmospheric CO₂ concentration. The position and volume of the boxes, the mixing between the boxes, and the way advection is divided between boxes is based on observations and remains unchanged from that described in Völker et al. (2002). The model advects dissolved inorganic carbon (DIC) between boxes in quantities proportional to the prescribed overturning circulation strength, and mixes DIC between vertically adjacent boxes, as described in Völker et al. (2002). The box model does not include any representation of biological carbon fluxes, which were (and are commonly) considered to be of limited importance to anthropogenic carbon uptake (e.g. Völker et al., 2002; Pèrez et al., 2013). In each of the three surface boxes, the CO₂ concentration is calculated from the DIC, temperature, salinity and alkalinity. Any disequilibrium between partial pressures of CO₂ in the ocean and atmosphere then drives a flux which is rate limited by a prescribed pis-

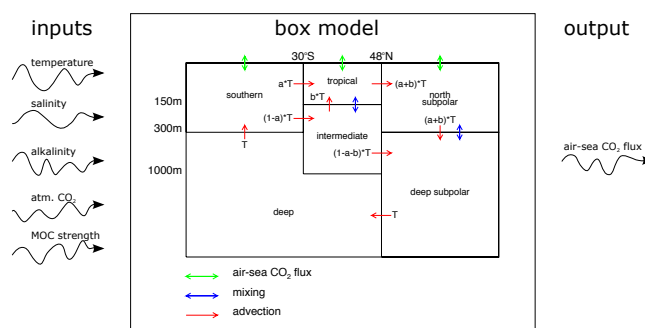


Figure 3. Schematic description of the box model.

ton velocity. The gas exchange is calculated by multiplying the piston velocity by the surface area of the box and the difference between the seawater CO₂ concentration and the seawater CO₂ value that would exist at equilibrium with atmospheric CO₂. The calculated air–sea CO₂ flux then modifies the concentration of DIC in each box. The formulation of the box model remains exactly as described in Völker et al. (2002) other than the tuning of the box model’s parameters (Table 1) to allow the box model to replicate results from the perturbed parameter ensemble. Note that by prescribing changes in alkalinity and allowing the DIC to adjust through air–sea flux, we are implicitly assuming that there is no significant freshwater-driven dilution/concentration of DIC and alkalinity.

To allow the box model to emulate the ESPPE, a single set of box model parameters was obtained by first running a 1000 member box model ensemble in which each of the box model parameters were varied within the ranges listed in Table 1. Parameter space was sampled using a latin hypercube. The fitness of each of the 1000 parameter sets was then judged by calculating the average coefficient of determination (R^2) across the 27 ESPPE members between the ESPPE subpolar North Atlantic air–sea flux, and the box model’s northern box air–sea flux. The ability of the box model to reproduce the ESM carbon flux is more dependent on the driving time series (CO₂^{atm}, temperature, salinity, alkalinity and overturning circulation strength) than it is dependent on the exact box model parameters. Indeed the ability of the box model is relatively insensitive to the box model parameters (Table 2 and Fig. S1 in the Supplement) suggesting that conclusions drawn on the drivers of the box model CO₂ flux are unlikely to be strongly dependent on the exact choice of box model parameters. The six parameter sets that gave the highest R^2 when compared with ESPPE output are presented in Table 2.

Variability on different timescales is separated using high and low-pass filtering. Filtering is achieved by applying a 5th order Butterworth fast Fourier transform filter. The mechanisms driving the modes of variability isolated using the high and low-pass filters are identified by manipulating the input time series (temperature, salinity, alkalinity, atmospheric

Table 1. Parameters used in box model.

Parameter Name	Parameter description	Parameter Range (for tuning)
T	overturning circulation strength (Sv)	n/a: as prescribed from ESM
a	fraction of overturning circulation strength	0–1
b	fraction of overturning circulation strength	1– a
mix_{eq}	vertical mixing (Sv)	0–20
$\text{mix}_{\text{north}}$	vertical mixing (Sv)	0–20
$\text{piston}_{\text{south}}$	southern box piston velocity (m hour ⁻¹)	0–0.4
$\text{piston}_{\text{eq}}$	equatorial box piston velocity (m hour ⁻¹)	0–0.4
$\text{piston}_{\text{north}}$	northern box piston velocity (m hour ⁻¹)	0–0.4

CO₂ and AMOC strength) used to force the box-model. These input time series are either filtered, held at a constant value, or left unchanged when supplied to the box-model. Initially only one input time series is manipulated at a time. In subsequent analysis, multiple input time series have been manipulated to examine their additive effect on the air–sea CO₂ flux.

To pick apart the contribution of different processes to the high and low frequency air–sea CO₂ flux simulated by the ESPPE, we sequentially control the inputs to the box model, isolating the role of that input in producing the overall change. Firstly, to understand the mechanism behind the high-frequency variability, we high-pass filter all of the inputs to the box model (temperature, salinity, alkalinity, atmospheric CO₂ concentrations and overturning circulation strength), adding to this the mean value from the original time series (since the high-pass filtering results in a time series varying around zero). This process removes any low-frequency variability. The high-pass filtered time series are used to drive the box model, and results compared to high-pass filtered results from the ESPPE (Fig. S2). The input variables for the North Atlantic are then sequentially held at their mean value (i.e. removing any variability) and the box model re-run (Fig. S2). Secondly, to understand the mechanisms driving the low-frequency variability the box model input time series are sequentially low-pass filtered (all other time series remain unchanged) and the box model run (Fig. S3), as described for the high-pass filter analysis.

3 Results and discussion

3.1 Box model validation

Using only a single set of parameters, the box model captures much of the variability in subpolar North Atlantic air–sea CO₂ flux simulated within and across the diverse ESPPE members (see the full data set in Fig. 4a and time series examples from that data set, Fig. S1). To test the predictive skill of the box model as an emulator for the ESPPE, we tuned the box model to emulate 13 randomly selected

ESPPE members, as described in the methods section, then ran the box model with inputs from the remaining ESPPE members, i.e. those ensemble members excluded from the tuning ensemble. Comparison of predicted and actual ESPPE subpolar North Atlantic air–sea CO₂ flux yields a coefficient of determination of 0.66 (Fig. S4). Comparison of the box model’s low latitude and southern box air–sea CO₂ flux with the ESPPE air–sea CO₂ flux shows that much of the variability outside of the subpolar region is also explained by the box model. This result holds independent of whether the box model is tuned to replicate the northern, low-latitude or southern box air–sea CO₂ flux (Fig. S5). The validation presented here gives us confidence that the box model represents the 1st order processes involved in the ESM simulation of North Atlantic CO₂ uptake, and provides us with a diagnostic tool to identify what drives CO₂ uptake variability in the ESPPE. Our findings imply that almost all of the ESPPE uncertainty is contained within the inputs to the box model rather than the parameters within the box model. The different processes of North Atlantic subpolar CO₂ uptake simulated by ESPPE ensemble members are therefore also captured within these box-model inputs.

3.2 Modes of variability

To explore the mechanisms behind the ESM’s variability we initially broke down the subpolar North Atlantic air–sea flux behaviour simulated within the Earth System Model ensemble by applying high and low pass filters to the data (Fig. 5). This allows us to identify discreet timescales of variability common across all ensemble members. We find that filtering the ESM results at < 5 years and > 30 years allows us to capture almost all of the ESM’s variability whilst cleanly separating the variability in to two components (Fig. 5). We will explore the mechanisms behind these two timescales of variability independently.

By splitting the ESPPE North Atlantic subpolar air–sea CO₂ flux into a high and low frequency component a number of things become clear. Firstly, the majority of the total signal can be described by these two separate components (Fig. 5). Secondly, we see that the high frequency compo-

Table 2. Box model parameter values.

Ranking	Parameter						
	piston _{south}	piston _{eq}	piston _{north}	mix _{eq}	mix _{north}	<i>a</i>	<i>b</i>
1st	0.0854	0.142	0.177	2.09	1.02	0.286	0.0103
2nd	0.138	0.211	0.168	12.7	19.2	5.37e-03	8.72e-02
3rd	0.321	0.129	2.82e-02	17.1	13.1	1.16e-02	0.727
4th	0.399	1.56e-03	0.130	8.65	6.76	0.423	2.42e-02
5th	0.199	0.104	8.56e-02	1.09	8.33	2.22e-03	8.60e-02
6th	0.0632	0.0136	0.159	13.1	10.9	0.608	0.288

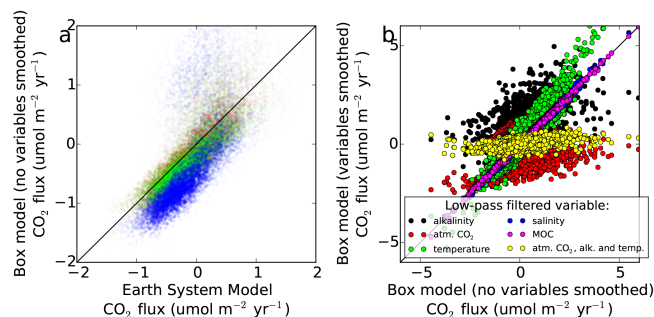


Figure 4. (a) ESPPE subpolar North Atlantic air–sea CO₂ flux anomaly plotted against box model estimates of that same flux using the top three box model parameter sets (Table 2) in red, blue and green respectively. (b) Results from box model driven with low-frequency variability in all input variables, plotted against: box model results when low-frequency alkalinity signal is removed (black), low-frequency atmospheric CO₂ signal removed (red), low-frequency temperature signal removed (green), low-frequency salinity signal removed (blue), low-frequency meridional overturning circulation (MOC) signal removed (purple), and low-frequency atmospheric CO₂ concentration, alkalinity and temperature signals all removed. The straight line represents the one-to-one line upon which results would fall if removal of the low-frequency variability in that variable did not influence CO₂ uptake.

ment occurs with little coherent structure across all ensemble members, and it also shows an increase in variability towards 2100 (Fig. 5). Thirdly, we see that the low-frequency signal tends to increase from its pre-industrial value through the 20th century, then in most cases peaks during the 21st century, then begins to decline (Fig. 5).

The peak and decline behaviour seen in the low-frequency air–sea CO₂ flux signal is unlike the globally averaged signal (Fig. 1), which under a CO₂ emission scenario like RCP8.5 (in which atmospheric concentrations are increasing throughout the 21st century) would be expected to (and indeed does – Fig. 1) continue increasing, but at a progressively reduced rate. The globally averaged response is consistent with our basic understanding of seawater carbon chemistry (Zeebe and Wolf-Gladrow, 2001; Revelle and Suess, 1957), and results from other ESMs (e.g. Friedlingstein et al., 2006). As long as the atmospheric CO₂ con-

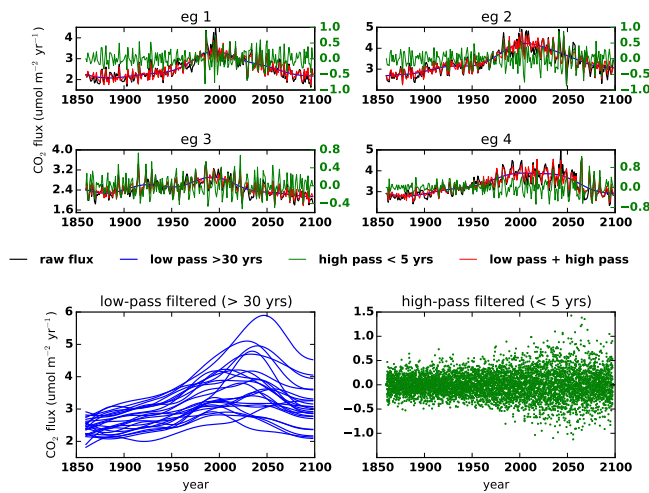


Figure 5. High and low pass filters are applied to the ESPPE subpolar North Atlantic air–sea CO₂ flux simulations to identify the separate time-scales of variability. Top panel: four random ensemble members' CO₂ flux is presented (black) alongside the low-pass (blue) and high-pass (green) processed fluxes. In red, the low and high pass filtered data are recombined to demonstrate that these timescales of variability together explain almost all of the original variability. Lower panel: the low-pass (blue, left) and high-pass (green, right) filtered results across all ensemble members are presented, demonstrating, in the case of the low-pass filters results, great diversity in model evolution.

centration is increasing, assuming no dramatic changes in ocean circulation or biology, there will always be an air-to-sea CO₂ concentration gradient, and therefore air-to-sea CO₂ flux. The decrease in this flux through time reflects the changing speciation of carbon in seawater in response to the increase in carbonic acid concentrations – which partitions carbon progressively in the direction of CO₂, elevating surface ocean CO₂ concentrations, and reducing the air–sea CO₂ concentration gradient (Zeebe and Wolf-Gladrow, 2001; Revelle and Suess, 1957).

The difference in behaviour between the subpolar North Atlantic and the well-understood chemical response of the steady-state ocean (Revelle and Suess, 1957; as largely seen here in the global average: Fig. 1) indicates that CO₂ emis-

sion (and potentially associated climate change) forced physical, biological or chemical changes in the North Atlantic are modifying the capacity of this sink to take up atmospheric CO₂. Peak and decline North Atlantic CO₂ uptake has previously been identified in an idealised study by Völker et al. (2002), using the box-model applied in this study. Völker et al. (2002) demonstrate theoretically that the high latitude North Atlantic could take up less atmospheric CO₂ in the future than it did in the preindustrial, without invoking any change in ocean circulation or biology. The peak and decline demonstrated by Völker et al. (2002) occurred in response to proportionally more CO₂ being taken up under higher atmospheric CO₂ conditions in the low latitude Atlantic than in the subpolar North Atlantic – in response to the higher alkalinity (and therefore lower Revelle Factor (Revelle and Suess, 1957) and higher buffering of surface ocean *p*CO₂) in the low latitude waters, and that excess carbon being transported north into the subpolar gyre by the overturning circulation (explained further in Fig. 6 and the associated caption).

3.3 Drivers of multidecadal/centennial mode of variability

To assess the drivers of multidecadal/centennial variability, we first plot each annual-average value from the ESM simulations against the equivalent value generated within the box model (Fig. 4a). We then sequentially apply a low-pass filter to each input variable (and sets of input variables) to remove the low-frequency (> 30 year) variability from that/those input variable/variables, and using those input values run the box model. We then examine how the removal of low-frequency variability from the different input variables changes the output of the box model (Fig. 4b).

We find that the most important driver of the low-frequency (“peak and decline”) variability in the subpolar North Atlantic air–sea CO₂ flux comes from the progressive increase in atmospheric CO₂ concentrations (Fig. 4), which drives much of both the increase and decrease (Fig. S3) in CO₂ flux, as described under idealised conditions by Völker et al. (2002). Without a low-frequency signal in the atmospheric CO₂ concentrations fed into the box model however, a 21st century decline in air–sea CO₂ flux is still present (Fig. S5). This decline is driven by a slow reduction in subpolar alkalinity and to a lesser degree warming (Figs. 4 and S6). This finding confirms the applicability to our ESM ensemble of the idea proposed by Völker et al. (2002), and described in the preceding paragraph.

The similarity between the box model behaviour with no low-pass filtered inputs (i.e. optimally emulating the ESPPE), and with input salinity and AMOC low-pass filtered (Fig. 4), tells us that these two factors are not having an important impact on the low-frequency subpolar North Atlantic peak and decline air–sea flux time evolution (Fig. 4). The minimal impact of AMOC change on subpolar North Atlantic air–sea CO₂ flux likely reflects the fact that the AMOC

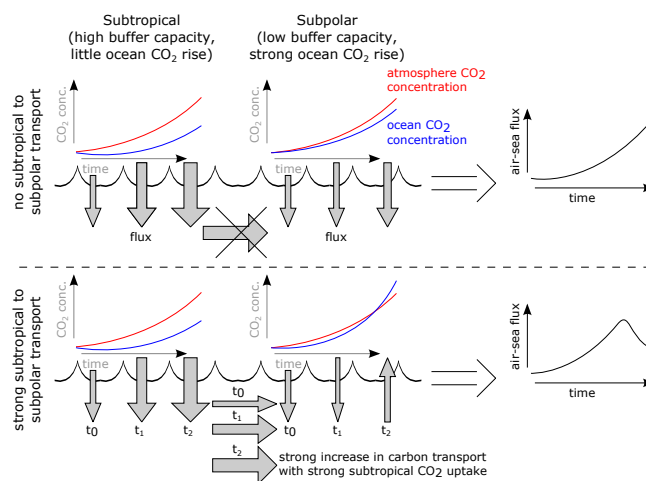


Figure 6. Diagrammatic explanation of the mechanism proposed in Völker et al. (2002) by which subpolar North Atlantic CO₂ concentration may peak then decline in response to continuously rising atmospheric CO₂ concentrations. The top half of the diagram explains what would happen if the low latitude and subpolar Atlantic were not connected by the circulation of the ocean (AMOC). Here, the higher alkalinity to dissolved-carbon ratio (the warm and saline low-latitude waters) of the subtropics means that these waters can strongly take up anthropogenic CO₂ without a big rise in surface ocean CO₂ concentrations. Similarly, the higher latitude subpolar waters (with low alkalinity to dissolved-carbon ratios) continuously take up CO₂, but the (relatively) small buffering capacity of these waters means that the surface ocean CO₂ concentration rises (relatively) quickly. A smaller air–sea CO₂ gradient is therefore maintained, and the air–sea CO₂ flux is (relatively) small. The bottom half of the diagram represents the situation in the real ocean, and the simulations considered in this study. Here the low latitude and subpolar Atlantic are linked by the near-surface limb of the Atlantic Meridional Overturning Circulation. In this situation, in response to rising atmospheric CO₂, the low latitude CO₂ uptake continues (in our idealised example) as in the top half of the diagram, but some of that extra carbon is being moved into the subpolar Atlantic, where the buffering capacity is lower, and the water does not have the capacity to hold as much extra carbon as CO₂. This could ultimately result in the subpolar Atlantic becoming a source for anthropogenic CO₂ rather than a sink, as it may not have the capacity to hold the extra CO₂ being passed to it from the south.

decline across the ESPPE is relatively modest (Fig. S6), and that only a fraction of the water moved by the AMOC has an opportunity to exchange CO₂ with the atmosphere. Removing the low-frequency signal from the temperature time series used by the box model has a minor effect (Fig. 4), causing the box model to over-predict the air–sea CO₂ flux at times of high flux, which translates in time series analysis to slightly underestimating the decline after peak air–sea CO₂ flux has been reached (Fig. S3). Similarly removing the low-frequency signal from the alkalinity time series input to the box model causes a slightly greater over-prediction of air–sea CO₂ flux values during the decline phase (Figs. 4 and S5).

The box modelling results obtained using low-pass filtered ESM output are consistent with the mechanism proposed by Völker et al. (2002), i.e. because low Revelle factor low-latitude water maintains strong CO₂ uptake throughout the coming century while the CO₂ uptake in subpolar North Atlantic begins to decline as a result of its high Revelle factor, the increasing supply of anthropogenic carbon to the subpolar North Atlantic carbon from the south reduces subpolar North Atlantic CO₂ uptake. The pathway by which this additional anthropogenic carbon enters the subpolar North Atlantic within the box model appears to vary depending on the parameter set employed within the box model (Table 2 and Table S1). The 1st, 3rd, 4th and 6th ranking parameter sets (Table 2) describe the basic scenario of increased surface ocean anthropogenic carbon advection from low to high northern latitudes. The 2nd ranking parameter set describes interior transport of anthropogenic carbon between surface low and high northern latitude boxes through strong vertical mixing in/out of these boxes. Finally, the 5th ranking parameter set emulates the ESM air–sea flux despite a low subpolar North Atlantic piston velocity, because of a strong air–sea gradient maintained by vertical mixing. This strong gradient also appears to make subpolar North Atlantic air–sea CO₂ flux sensitive to the changes in surface anthropogenic carbon flux from its southern boundary, despite the magnitude of that flux being small.

3.4 Drivers of annual/inter-annual mode of variability

Considering the high-frequency variability simulated within the ESPPE (Fig. 7 and S2), we compare box model simulations run with all input time series high-pass filtered, with high-pass filtered ESPPE subpolar North Atlantic air–sea CO₂ flux data. We then sequentially (and then together) hold the input time series constant at their average values (Fig. 7), and re-run the box model to isolate the contribution of variability in each of the input time series to the ESPPE results. We find that the box model captures the temporal variability but tends to underestimate the magnitude of variability (Fig. 7a). Holding temperature and alkalinity (yellow dots) constant we find near-complete breakdown of the box model's ability to capture the ESM's CO₂ flux variability (Fig. 7b). Independently holding temperature and alkalinity constant we find that these factors separately account for much of the correlation between the box model and ESPPE high-frequency variability. Holding salinity, meridional overturning circulation strength and atmospheric CO₂ concentrations constant (in turn) we find little impact on the correlation between the box model and the ESPPE results (Fig. 7b). It is therefore clear that the high-frequency variability simulated by the ESM within the ESPPE is almost completely driven by variability in temperature and alkalinity, and is largely insensitive to the model's variability in salinity, AMOC and atmospheric CO₂ on these timescales.

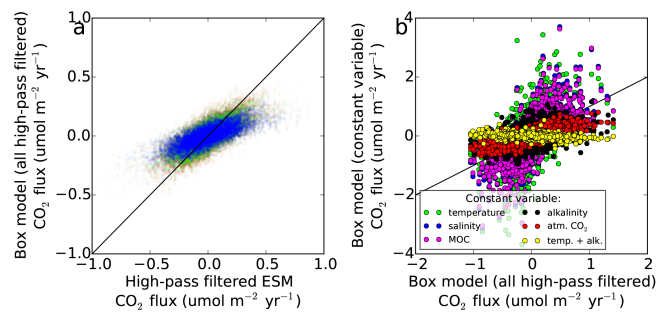


Figure 7. (a) High-pass filtered ESPPE subpolar Atlantic air–sea CO₂ flux plotted against box model estimates of that same flux using the top three box model parameter sets (Table 2) in red, green and blue respectively, but forced with high-pass filtered input time series. (b) All box model inputs high-pass filtered plotted against all box model inputs high-pass filtered but one variable held constant. The constant variable in each case is named within the legend. All results are presented as anomalies from the mean.

3.5 Evidence for these mechanisms occurring in the Earth System Model ensemble

By emulating the ESPPE using the box model we have simplified the system to a level at which we can explore the mechanisms at play in detail. Using the box model we have identified what appears to be the dominant mechanisms controlling high-latitude Atlantic CO₂ uptake on short (< 5 years) and long (> 30 years) timescales. We finally ask whether these mechanisms are consistent with evidence derived purely from the Earth System Model simulations.

Earth System Model *p*CO₂ is calculated interactively from DIC and alkalinity concentrations, temperature and salinity. It is possible to repeat this calculation offline (e.g. Halloran, 2012), and by doing so assess the relative importance of these different variables in determining the model's *p*CO₂, and by inference air–sea CO₂ flux, on different timescales. We perform this analysis on the low and high-pass filtered time series to ask whether the same variables are controlling air–sea CO₂ flux in the ESM ensemble members as have been identified in the box model analysis.

Our box model analysis suggests that the low-frequency behaviour of the box model is primarily driven by changing atmospheric CO₂ (and therefore changing DIC), with secondary controls from alkalinity and temperature and no significant salinity control (Fig. 4). We find this to be consistent with the behaviour of the ESM ensemble members. Recalculating *p*CO₂ whilst holding DIC constant causes a large deviation from the ESM ensemble's interactively calculated *p*CO₂ (i.e. a large deviation from the one-to-one line in Fig. 8d), recalculating *p*CO₂ with alkalinity or SST held constant results in small deviations from the ESM's interactively calculated *p*CO₂ (Fig. 8a and b), and recalculating *p*CO₂ whilst holding salinity constant results in a very small deviation from the ESM's interactively calculated *p*CO₂ (Fig. 8c).

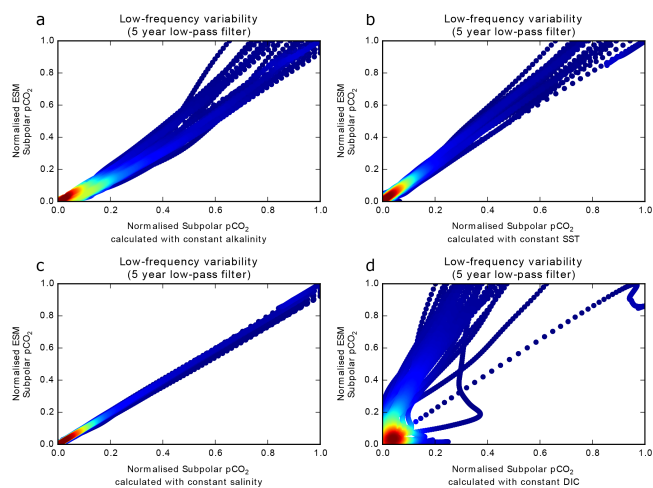


Figure 8. Low-pass filtered ESM Subpolar Gyre (SPG) $p\text{CO}_2$ plotted against low-pass filtered ESM SPG $p\text{CO}_2$ calculated with alkalinity only (a), SST only (b), salinity only (c) and DIC only (d) held constant through time. Points represent annually averaged values. Colours from blue to red represent an increasing density of points.

Note however that this analysis is simply indicative of what is occurring, because we cannot separate out the different contributors to DIC change.

Our box model analysis indicates that the high-frequency behaviour of the box model is primarily driven by changing SST and alkalinity (Fig. 7). We find this to be consistent with the behaviour seen in the ESM ensemble. The relationship with $p\text{CO}_2$ calculated interactively in the ESM, and that calculated offline using ESM temperature, salinity, DIC and alkalinity still holds if salinity is held constant (Fig. 9c), is less strong if DIC is held constant (Fig. 9d), and is weak where alkalinity or SST are held constant (Fig. 9a or b).

3.6 Limitations to the application of the box model

Whilst we find the box modelling approach extremely valuable in simplifying the behaviour of the ESM ensemble, and through doing so facilitating a mechanistic understanding of the ESM behaviour, this approach has limitations. As discussed in Sect. 3.3, the box model appears to emulate the ESM ensemble well under a range of different parameter sets, despite the pathway of DIC supply to the high-latitude North Atlantic not being consistent under all parameter sets. It would therefore appear that whilst the box model is useful in identifying the generic mechanism operating across the ESM ensemble, it does not allow us to isolate the detail of that mechanism. This limitation is likely to reflect the fact that in representing the ocean using a small number of boxes, one has to make overly simplified divisions between dynamic water-masses and oceanic features.

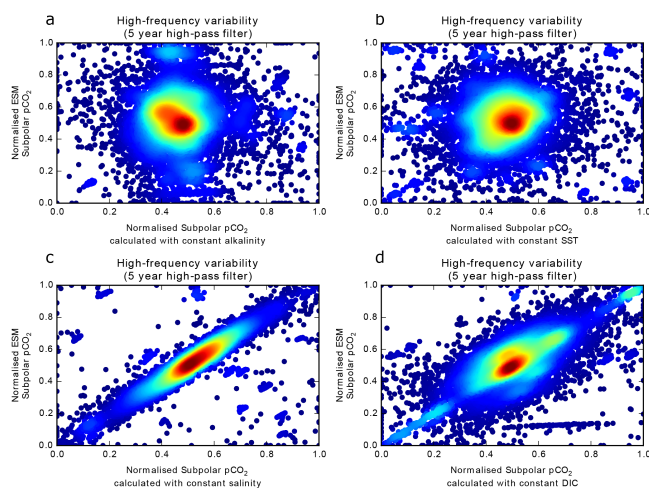


Figure 9. High-pass filtered ESM Subpolar Gyre (SPG) $p\text{CO}_2$ plotted against high-pass filtered ESM SPG $p\text{CO}_2$ calculated with alkalinity only (a), SST only (b), salinity only (c) and DIC only (d) held constant through time. Points represent annually averaged values. Colours from blue to red represent an increasing density of points.

4 Conclusions

We find that different mechanisms are controlling the inter-annual and centennial subpolar North Atlantic CO₂ variability in our large ensemble of perturbed parameter ESM simulations. The interannual variability appears to be controlled by rapid changes in the local seawater temperature and alkalinity fields, whereas the centennial variability is largely controlled by the anthropogenically driven increase in atmospheric CO₂ concentrations interacting with the background chemical gradient (high to low alkalinity), and DIC transport, in the North Atlantic. Our findings suggest that while it is important to understand the mechanisms behind recent inter-annual variability in the subpolar North Atlantic CO₂ flux, that understanding might not directly inform us about how the sink is likely to change in the future.

CO₂ uptake change can be driven by the basic chemical response of seawater to rising atmospheric CO₂, change in the ocean's physical circulation or state, or change in biological activity. We have the greatest confidence in predicting future change based on the former and least confidence in change based on the latter. This is because the chemistry is well understood and largely independent of the climate system response, whilst the physical change is subject to uncertainty in the climate system dynamics and the biological change adds structural and parameter uncertainty to the already uncertain physical response. The fact that the 21st century subpolar North Atlantic CO₂ uptake change appears to be largely controlled by the basic chemical response of seawater to rising atmospheric CO₂ concentrations therefore implies that similar behaviour can be expected in the real world. This raises the question, if the real-world North Atlantic CO₂ sink is to

follow a peak and decline trajectory, where on this trajectory do we presently sit? Perhaps the suggestion that the strength of the subpolar North Atlantic CO₂ sink has been decreasing (e.g. McKinley et al., 2011; Schuster and Watson, 2007) indicates that the real-world system is already in long-term decline.

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