

Global Biogeochemical Cycles®



RESEARCH ARTICLE

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Special section

Regional Carbon Cycle
 Assessment and Processes-2

Key Points:

- We synthesize air-sea fluxes of CO₂, nitrous oxide and methane in the global coastal ocean using observation-based products and ocean models
- The coastal ocean CO₂ sink is 60% larger in ocean models than in observation-based products due to systematic differences in seasonality
- Coastal nitrous oxide and methane emissions offset 30%–60% of the CO₂ coastal uptake in the net radiative balance

Supporting Information:

Supporting Information may be found in the online version of this article.

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



























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A Synthesis of Global Coastal Ocean Greenhouse Gas Fluxes

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Abstract The coastal ocean contributes to regulating atmospheric greenhouse gas concentrations by taking up carbon dioxide (CO₂) and releasing nitrous oxide (N₂O) and methane (CH₄). In this second phase of the Regional Carbon Cycle Assessment and Processes (RECCAP2), we quantify global coastal ocean fluxes of CO₂, N₂O and CH₄ using an ensemble of global gap-filled observation-based products and ocean biogeochemical models. The global coastal ocean is a net sink of CO₂ in both observational products and models, but the magnitude of the median net global coastal uptake is ~60% larger in models (−0.72 vs. −0.44 PgC year^{−1}, 1998–2018, coastal ocean extending to 300 km offshore or 1,000 m isobath with area of 77 million km²). We attribute most of this model-product difference to the seasonality in sea surface CO₂ partial pressure at mid- and high-latitudes, where models simulate stronger winter CO₂ uptake. The coastal ocean CO₂ sink has increased in the past decades but the available time-resolving observation-based products and models show large discrepancies in the magnitude of this increase. The global coastal ocean is a major source of N₂O (+0.70 PgCO₂-e year^{−1} in observational product and +0.54 PgCO₂-e year^{−1} in model median) and CH₄ (+0.21 PgCO₂-e year^{−1} in observational product), which offsets a substantial proportion of the coastal CO₂ uptake in the net radiative balance (30%–60% in CO₂-equivalents), highlighting the importance of considering the three greenhouse gases when examining the influence of the coastal ocean on climate.

Plain Language Summary The coastal ocean regulates greenhouse gases. It acts as a sink of carbon dioxide (CO₂) but also releases nitrous oxide (N₂O) and methane (CH₄) into the atmosphere. This synthesis contributes to the second phase of the Regional Carbon Cycle Assessment and Processes (RECCAP2) and provides a comprehensive view of the coastal air-sea fluxes of these three greenhouse gases at the global scale. We use a multi-faceted approach combining gap-filled observation-based products and ocean biogeochemical

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models. We show that the global coastal ocean is a net sink of CO₂ in both observational products and models, but the coastal uptake of CO₂ is ~60% larger in models than in observation-based products due to model-product differences in seasonality. The coastal CO₂ sink is strengthening but the magnitude of this strengthening is poorly constrained. We also find that the coastal emissions of N₂O and CH₄ counteract a substantial part of the effect of coastal CO₂ uptake in the atmospheric radiative balance (by 30%–60% in CO₂-equivalents), highlighting the need to consider these three gases together to understand the influence of the coastal ocean on climate.

1. Introduction

Coastal oceans play an important role in the global carbon cycle by serving as a hub of exchange between the land, tidal wetlands, estuaries, sediments, the atmosphere, and the open ocean (Bauer et al., 2013; Chen & Borges, 2009; Mackenzie et al., 1998; Ward et al., 2020). They contribute to the global oceanic uptake of anthropogenic carbon by absorbing carbon dioxide (CO₂) directly from the atmosphere and by burying, transforming, or outgassing the carbon delivered by terrestrial ecosystems to the coastal ocean (e.g., Regnier et al., 2022). A notable milestone in the efforts to quantify the CO₂ exchange between the atmosphere and coastal oceans was reached by Chen et al. (2013) during the first phase of the Regional Carbon Cycle Assessment and Processes (RECCAP), an international effort to establish the mean carbon balance and change over the period 1990–2009 for all subcontinents and ocean basins. These authors expanded on prior work at the scale of continental shelves (Borges et al., 2005; Cai et al., 2006; Chen & Borges, 2009; Laruelle et al., 2010) and examined the global atmospheric CO₂ uptake by coastal oceans using a compilation of surface ocean partial pressure of CO₂ (pCO₂) data available for 87 shelves. They concluded that most coastal ocean waters act as a sink for atmospheric CO₂, except for tropical coastal ocean systems that were identified as weak CO₂ sources and found the global coastal ocean CO₂ uptake to be 0.4 PgC year⁻¹ (for a surface area of coastal ocean of 30.3 million km²).

Since the completion of RECCAP, the amount of available pCO₂ measurements in the coastal ocean has increased tremendously, reaching millions shortly after the RECCAP assessment was released (e.g., Surface Ocean CO₂ Atlas database SOCAT; Bakker et al., 2014) and ~19 million in the most recent publication (Bakker et al., 2022). In parallel, statistical gap-filling methods, initially developed for the open ocean, have been applied to these fast expanding data sets to resolve the spatio-temporal variability of the air-sea CO₂ flux in the coastal ocean (Chau et al., 2022; Landschützer et al., 2020; Laruelle et al., 2014; Roobaert et al., 2019). These global gap-filled observation-based coastal products led to a downward revision of the global coastal ocean CO₂ uptake to about half of the RECCAP value (0.15–0.20 PgC year⁻¹; Chau et al., 2022; Roobaert et al., 2019). This downward revision was corroborated by a recent synthesis of 214 regionally aggregated CO₂ flux estimates, leading to a net uptake of 0.25 PgC year⁻¹ (Dai et al., 2022), although these assessments covered slightly different periods and coastal areas (1985–2019 and ~22 million km² in Chau et al. (2022); 1998–2015 and 28 million km² in Roobaert et al. (2019); 1998–present and ~30 million km² in Dai et al. (2022)).

While coastal ocean waters are a sink of CO₂, they are also the main oceanic source of two other important greenhouse gases: nitrous oxide (N₂O) and methane (CH₄) (e.g., Saunio et al., 2020; Wan et al., 2022; Weber et al., 2019; Yang et al., 2020). RECCAP did not consider N₂O and CH₄, but recent studies have compiled oceanic N₂O and CH₄ measurements (Kock & Bange, 2015) and applied statistical gap-filling techniques similar to those employed for CO₂ to assess the global ocean air-sea N₂O and CH₄ fluxes (Weber et al., 2019; Yang et al., 2020). These studies have greatly improved the quantification of N₂O and CH₄ air-sea fluxes at the global scale, but coastal ocean N₂O and CH₄ emissions remain highly uncertain and the extent to which these emissions offset the present-day coastal CO₂ uptake is unknown.

Coastal air-sea fluxes of CO₂, N₂O and CH₄ have strong spatial and seasonal variability. Regional-scale observational and modeling studies have greatly improved the quantification of the mean and temporal variability of air-sea fluxes of greenhouse gases in individual regions across the globe (e.g., Anderson et al., 2009; Arévalo-Martínez et al., 2015; Fennel et al., 2019; Gomez et al., 2020; Gülzow et al., 2013; Hauri et al., 2021; Louchard et al., 2021; Mayer et al., 2018; Pipko et al., 2017; Turi et al., 2014). However, the limited spatial coverage of these studies largely inhibits a global-scale perspective. Global gap-filled observational products and global ocean biogeochemical models now have horizontal resolutions of ~25–50 km to estimate coastal CO₂ (Bourgeois et al., 2016; Lacroix et al., 2020, 2021; Roobaert et al., 2022) and N₂O (Berthet et al., 2023; Ganesan

et al., 2020; Stell et al., 2022) fluxes, recently complemented these regional-scale studies in global-scale studies (e.g., Friedlingstein et al., 2022).

As a result of observational and modeling advances since RECCAP, a global view of the coastal ocean's spatial and seasonal patterns in air-sea greenhouse gas fluxes has started to emerge, at least for CO₂ fluxes. Polar and subpolar coastal oceans, such as the northwest North Atlantic along the Canadian and US coast (Cahill et al., 2016; Fennel & Wilkin, 2009; Gustafsson et al., 2019; Lachkar & Gruber, 2013; Laruelle et al., 2015; Previdi et al., 2009; Signorini et al., 2013; Thomas et al., 2004), the European shelves (Cossarini et al., 2015; Gustafsson et al., 2019; Neumann et al., 2022; Thomas et al., 2004) and Arctic and Antarctic shelf (Arrigo et al., 2008; Ouyang et al., 2022; Pipko et al., 2017, 2021) generally are strong sinks of CO₂ characterized by large seasonal variations, and likely account for about 90% of the annual global coastal CO₂ uptake (while representing ~45% of the global coastal surface area, see Dai et al., 2022; Laruelle et al., 2014; Roobaert et al., 2019). There are exceptions to subpolar and polar shelves where outgassing has been identified, such as the Scotian Shelf (Rutherford et al., 2021; Rutherford & Fennel, 2022) or the Laptev Sea in the Arctic (Anderson et al., 2009). Coastal upwelling regions, such as the nearshore California Current, are sources of CO₂ to the atmosphere with a marked seasonality that follows the upwelling dynamics (Dai et al., 2013; Damien et al., 2023; Fiechter et al., 2014; Lachkar & Gruber, 2013; Turi et al., 2014). Tropical systems, such as the Gulf of Mexico (Laurent et al., 2017; Xue et al., 2016) and the South China Sea (Wan et al., 2022), are mostly identified as weak CO₂ sources with weak seasonal variability (Dai et al., 2022; Laruelle et al., 2014, 2015; Roobaert et al., 2019). Our knowledge of N₂O and CH₄ variability in the global coastal ocean is more limited, but gap-filled products and global models suggest that N₂O and CH₄ annual emissions strongly vary between coastal regions (e.g., Ganesan et al., 2020; Stell et al., 2022; Weber et al., 2019; Yang et al., 2020). These products and models offer a remarkable opportunity to establish a greenhouse gas budget for the global coastal ocean and improve our understanding of its spatial and seasonal variability.

Rising atmospheric CO₂ levels influence coastal CO₂ uptake on multi-decadal time-scales. Prior syntheses at the global scale including RECCAP (Bauer et al., 2013; Cai et al., 2006; Chen et al., 2013; Laruelle et al., 2010; Regnier et al., 2013) and at the regional scale (Fennel & Testa, 2019; Legge et al., 2020; Liu et al., 2018) clearly support the view that the coastal ocean is currently a sink of atmospheric CO₂, but the extent to which it has changed on longer time-scales remains controversial (see Dai et al., 2022, for a review). Mackenzie et al. (2005) from a modeling perspective and later Cai (2011) from observations first hypothesized that the potential of the coastal ocean to act as a sink for CO₂ might be increasing with time. This view is increasingly supported by time series analyses that suggest that trends in sea surface pCO₂ are overall weaker than the atmospheric pCO₂ trend in most coastal regions. This finding further implies an intensified CO₂ uptake or decreased outgassing, although potential trends in winds and sea ice may also play a role (Bauer et al., 2013; Dai et al., 2022; Laruelle et al., 2018; Wang et al., 2017). However, exceptions have been identified in regions where coastal ocean pCO₂ increases at a similar rate (i.e., near-zero changes in the flux) or even at higher rates (i.e., reduced CO₂ uptake or intensified outgassing) than atmospheric pCO₂ (e.g., California Current, South and Mid Atlantic Bight, Baltic Sea; Dai et al., 2022; Laruelle et al., 2018; Reimer et al., 2017; Schneider & Müller, 2018). The quantification of coastal CO₂ flux trends from observations is, however, still strongly restricted by the limited spatial coverage and/or the relatively short duration of time series.

Global ocean biogeochemical models offer an attractive means of assessing long-term trends in air-sea CO₂ flux densities in the coastal ocean and how they differ from those of the open ocean (Regnier et al., 2022). Two such models, with reasonable agreement in regions where time series are available (0.2–0.5° resolution in Bourgeois et al. (2016); 0.4° resolution in Lacroix et al. (2021)), suggest that the global coastal CO₂ sink density has increased at a slightly slower rate than the open ocean CO₂ sink since the preindustrial era, even when accounting for increasing global nutrient sources via river and atmospheric transport (Lacroix et al., 2020). However, both models have important limitations and potential biases related to their representation of fine-scale hydrodynamics of shelf circulation and biophysical processes that impact biogeochemical cycling in the shallow ocean (Mathis et al., 2022; Rutherford & Fennel, 2018).

In this second phase of the Regional Carbon Cycle Assessment and Processes (RECCAP2), we aim to address gaps in our understanding of air-sea greenhouse gas fluxes for the global coastal ocean. Our objectives are threefold. First, we revisit the estimate of the net coastal ocean CO₂ flux and combine it with CH₄ and N₂O emissions to derive a global climatological coastal ocean budget of greenhouse gas fluxes (Section 3.1). Second, we

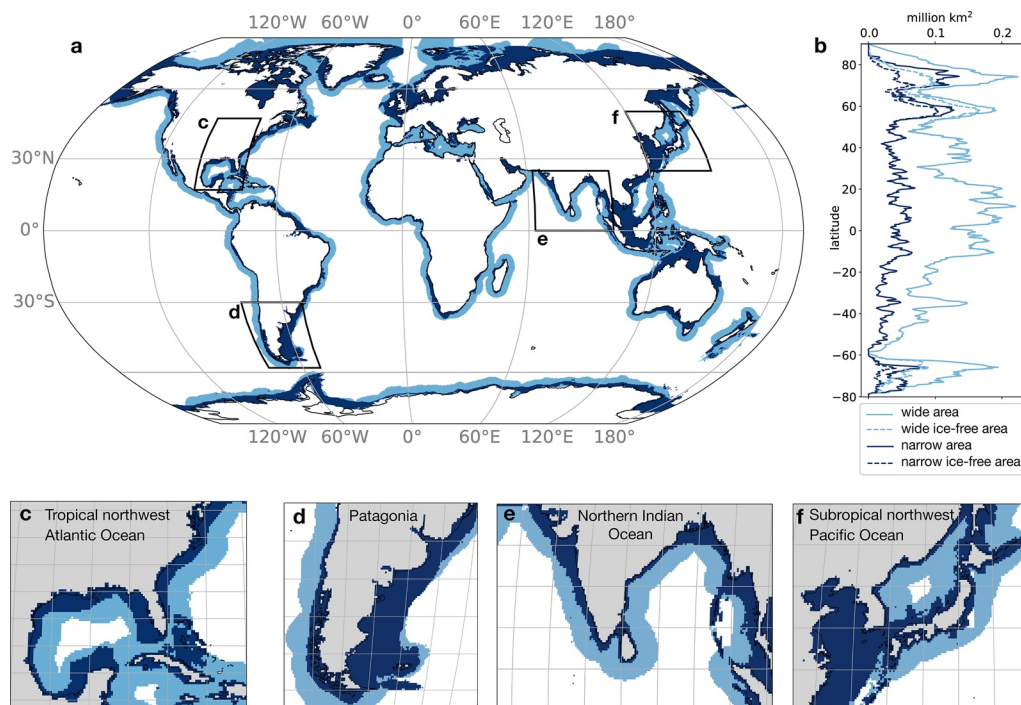


Figure 1. (a) Coastal masks used in this study for the wide (dark + light blue) and narrow (dark blue) coastal oceans, (b) Surface area (in km^2) at each latitude in the wide (light blue) and narrow (dark blue) coastal ocean masks (solid lines) and the 1998–2018 averaged sea-ice free surface area (dashed lines). (c–f) Insets showing the extent of the narrow and wide coastal oceans in four coastal regions. Sea ice coverage used in b is from NOAA OISST. See Methods for details.

analyze the spatial and seasonal variability in the CO_2 flux density and how it might differ from that of the open ocean and examine spatial patterns in coastal CH_4 and N_2O fluxes (Sections 3.2 and 3.3). Third, we investigate trends in the coastal CO_2 flux over the last four decades (Section 3.3). This synthesis complements the global ocean RECCAP2 chapter (DeVries et al., 2023), which includes the coastal ocean area, but does not specifically address the spatio-temporal dynamics of coastal CO_2 fluxes or present an integrated budget of CO_2 , N_2O and CH_4 fluxes. We consider the net contemporary air-sea fluxes (natural + anthropogenic) of CO_2 , N_2O and CH_4 using the 1998–2018 period (except if specified otherwise) over the coastal ocean but exclude estuaries and coastal vegetation, which are examined in the RECCAP2 synthesis of Rosentreter et al. (2023). Our approach combines observation-based and model-based estimates with different strengths and limitations discussed in the method and discussion sections.

2. Methods

2.1. Coastal Ocean Definition and Analysis Period

Different definitions of coastal oceans are used in the literature (Chen et al., 2013; Laruelle et al., 2017). We use two definitions of the coastal ocean. We primarily use a “wide” coastal ocean definition following Laruelle et al. (2017), where the seaward boundary is 300 km from shore or the 1,000-m isobath, whichever is further from shore, amounting to a total coastal ocean area of 77.2 million km^2 (Figure 1). This wide delineation of the shelf allows us to include the effect of upwelling systems and deep arctic shelves on the shelf greenhouse gas budget, which are only partly included in the narrow definition (Laruelle et al., 2017). We also use a “narrow” coastal ocean definition, which is delimited by the shelf break (defined as the isobath with maximum slope increase in the 0–1,000 m interval) and amounts to a total area of 28 million km^2 (see details in Laruelle et al., 2013, 2014). The landward boundary in the masks used to define the narrow and wide coastal oceans excludes estuaries and coastal vegetation, which are described in the RECCAP2 chapter of Rosentreter et al. (2023), but includes greenhouse gas uptake and emissions from large river plumes. The partitioning of coastal vegetation between estuarine systems and very nearshore shelf environments is poorly known, especially for submerged vegetation

Table 1
Description of Observation-Based Products Used in This Study, Including the Wind Speed Product and Gas Exchange Coefficient (k_w) Formulation Used to Compute the Fluxes

Product	Gases	Land-sea inputs	Domain	Frequency/period in this study	Horizontal resolution	Wide area (million km ²)	Wind speed and k_w	Reference
Carboscope-1	CO ₂	N/A	Global	Mon ^a 1998–2018	1°	73.9	JRA55-do v1.5.0 W92 ^b	Rödenbeck et al. (2022)
CMEMS*	CO ₂	N/A	Global ^c	Mon 1998–2018	1°	55.8	ERA5 W14 ^b	Chau et al. (2022)
Coastal-SOM-FFN	CO ₂	N/A	Global	Mon 1998–2015	0.25°	77.2	ERA-interim H11	Roobaert et al. (2019)
Coastal-SOM-FFN- k_w	CO ₂	N/A	Global	Mon 1998–2015	0.25°	77.2	JRAv1.3 W92	
Merged-SOM-FFN	CO ₂	N/A	Global	Mon 1998–2015	0.25°	77.2	ERA-interim H11	Landschützer et al. (2020)
Yang-N2O	N ₂ O	N/A	Global	Mon 1998–2015	0.25°	73.4	ERA-5 W14 ^b , L13 ^b	Yang et al. (2020)
Weber-CH4	CH ₄	N/A	Global	Ann 1999–2016	0.25°	73.7	ERA-5 W14 ^b , L13 ^b	Weber et al. (2019)
MARCATS-N2O & MARCATS-CH4	N ₂ O, CH ₄	N/A	Global	Ann 1980–2016	Regional ^d	77.2	NCEP II N00	Kock and Bange (2015)

Note. W92, H11, W14, L13, N00 stands for k_w -formulations from Ho et al. (2011), Liang et al. (2013), Nightingale et al. (2000), and Wanninkhof (1992, 2014) respectively. Mon and Ann stands for monthly and annual mean frequencies. Wide coastal areas are calculated after the products and models have been regridded on the 0.25° × 0.25° grid. Further details and references on observation-based products and models are provided in Supporting Information S1.

^aFrom originally daily. ^bScaled to global ocean mean value of 16.5 cm/hr. ^cMissing Arctic filled with Coastal-SOM-FFN climatology north of 75°N. ^dNo gap filling, one value per MARGins and CATCHments Segmentation (MARCATS).

such as seagrasses. In RECCAP2, their contribution to the greenhouse gas balance was estimated in the study by Rosentreter et al. (2023) and explicitly excluded from the present study to avoid double counting issues. The seagrass greenhouse gas flux, however, might slightly overlap with the present shelf estimate, as some of the greenhouse gas data used here might record the effect of the submerged vegetation on the air-water exchange. Note that the RECCAP2 global ocean chapter of DeVries et al. (2023) includes the surface area of the coastal ocean, that is, the CO₂ fluxes reported here should not be added to the global ocean CO₂ flux (global ocean chapter does not report CH₄ and N₂O fluxes). See Figure 1 for maps area latitudinal distribution of the narrow and wide coastal ocean waters.

The analysis is done over the 1998–2018 period to maximize the number of models and observation-based products available (see Tables 1–3 for periods covered by models and observation-based products). Note that this period differs from the one used in the open-ocean RECCAP2 studies that analyze oceanic CO₂ fluxes since 1985. All trends are calculated as linear trends over the 1998–2018 period.

2.2. Data Sets

We use observation-based and process model-based estimates because they have different strengths and limitations. Notably, gap-filled global observational-based products rely on machine learning algorithms or a mixed-layer model to fill the gaps of observations that are often too sparse to capture the full range of spatio-temporal variability in coastal regions (except in densely sampled regions such as major parts of the North American and European ocean margins), and are highly sensitive to the wind product and the choice of the gas exchange coefficient formulation (e.g., Roobaert et al., 2018). In contrast, ocean biogeochemical models can be associated with systematic biases. For instance, only some of the models used here include land-sea riverine

carbon inputs which sustain an oceanic CO₂ outgassing flux, and land-sea nutrient inputs which would yield an opposing biologically-driven oceanic CO₂ uptake in coastal ocean waters (Gao et al., 2023; Hauck et al., 2020; Regnier et al., 2022; Resplandy et al., 2018, see Tables 1–3).

2.2.1. Observation-Based pCO₂-Products

We use 4 global pCO₂-products that provide global monthly gridded surface ocean pCO₂ (noted pCO₂ here) and air-sea CO₂ flux fields based on observations from the SOCAT database, which compiles surface ocean pCO₂ observations and provides a subset after quality control (Bakker et al., 2016, 2022). Three of them use neural network-based interpolation methods: Coastal-SOM-FFN (Laruelle et al., 2017; Roobaert et al., 2019, 2022), merged-SOM-FFN (Landschützer et al., 2020) and CMEMS-LSCE-FFNN (which we refer as CMEMS, Chau et al., 2022), while the fourth product, Carboscope-1, uses a simple statistical representation of mixed-layer biogeochemistry fitted to the pCO₂ data (Rödenbeck et al., 2022). All these products are using SOCAT pCO₂ observations and are therefore not independent (see Supporting Information S1 for details on SOCAT versions and Figure S1 in Supporting Information S1 for SOCAT data coverage). In particular, the Merged-SOM-FFN product merged the Coastal-SOM-FFN (Laruelle et al., 2017) with an open ocean SOM-FFN product (Landschützer et al., 2014) to produce a global ocean product; the Coastal-SOM-FFN and Merged-SOM-FFN are therefore identical in the nearshore coastal region and only differ in the more offshore band of the wide coastal domain (see details in Landschützer et al., 2020). Coastal-SOM-FFN (and therefore also the near-shore product in Merged-SOM-FFN) was designed for the coastal ocean and uses coastal SOCAT data for their neural network training. In the three other products that use both open ocean and coastal ocean data (i.e., CMEMS, Carboscope-1 and offshore portion of Merged-SOM-FFN), the coastal estimate may be strongly influenced by open-ocean information extrapolated toward the coast. See Tables 1–3 and Supporting Information S1 for details on pCO₂-products (e.g., period, wind speed product, gas exchange formulation).

Carboscope-1 and CMEMS products resolve interannual variability over the whole 1998–2018 period and may be used to estimate decadal trends, while Coastal-SOM-FFN and Merged-SOM-FFN provide a 1998–2015 monthly climatology and do not resolve interannual variability. pCO₂-products often have unrealistic pCO₂ values under sea-ice (Laruelle et al., 2017). We therefore used the sea-ice fraction from the NOAA-OISST product (Reynolds et al., 2007) to mask pCO₂ and CO₂ flux values under sea-ice in the four products. We mask both to maintain consistency, but this method should not impact the flux dramatically since it is often inhibited by sea-ice in flux formulations. In this study, we also filled in the missing values north of 75N in CMEMS using the Coastal-SOM-FFN climatology. This approach only marginally impacts the results (adds –0.03 PgC year^{–1} to the wide coastal ocean net CO₂ flux) because the surface area north of 75N contributes 5 million km² to the wide coastal ocean (6% of the total wide area) but only 1.4 million km² is ice-free on average for the entire study period. This filled-in version of CMEMS is referred to as CMEMS* and we report no long-term trend in the Arctic for this product.

We also illustrate the sensitivity of the flux in pCO₂-products to the choice of the wind speed product and gas transfer coefficient (k_w) formulation (e.g., Roobaert et al., 2018) by presenting a second version of the Coastal-SOM-FFN flux product but with a different wind product and k_w (labeled Coastal-SOM-FFN- k_w) in which the CO₂ flux is calculated as $F = k_w \text{Ko} (\text{pCO}_{2a} - \text{pCO}_2)$ where Ko is the gas solubility and pCO_{2a} the atmospheric pCO₂. The default version of Coastal-SOM-FFN uses the ERA5 wind speeds and the k_w formulation from Ho et al. (2011), whereas Coastal-SOM-FFN- k_w uses JRA55v1.3 winds and the Wanninkhof (1992) k_w formulation (i.e., wind and formulation used in some ocean biogeochemical models, see Tables 1–3 for details on k_w parametrization and wind products used in models and products). The four pCO₂-products are used for the analysis of the wide and narrow coastal oceans, and the three pCO₂-products that extend outside the coastal domain are used for the open ocean (CMEMS*, Merged-SOM-FFN, and Carboscope-1). Coastal-SOM-FFN- k_w is shown only in the wide coastal ocean for discussion and is not used to compute the pCO₂-product median.

2.2.2. Observation-Based N₂O and CH₄ Flux Products

We used two observation-based estimates of the N₂O and CH₄ fluxes. In each case, we use an estimate based on simple extrapolation of the MEMENTO (MarinE MethanE and NiTrous Oxide) database to the 45 MARGins and CATCHments Segmentation (MARCATS, Figure S2 in Supporting Information S1) coastal regions (referred to as MARCATS-N2O and MARCATS-CH4 Kock & Bange, 2015), and an estimate that extrapolates MEMENTO and supplementary observations to global 0.25° climatology using supervised machine learning

models (Weber et al., 2019; Yang et al., 2020, referred to as Weber-CH₄ and Yang-N₂O). The MARCATS-N₂O and MARCATS-CH₄ products provide an annual mean value based on data from 1980 to 2016, Yang-N₂O provides monthly climatology for 1988–2017 and Weber-CH₄ an annual mean value for 1999–2016 (Table 1). In Yang-N₂O surface N₂O disequilibrium was extrapolated globally using an ensemble of 100 Random Regression Forest (RRF) models, and in Weber-CH₄ surface CH₄ disequilibrium was extrapolated using 1,000 RRF models and 1,000 Artificial Neural Network models. In both cases, diffusive fluxes were calculated and uncertainty propagated by coupling the mapped disequilibrium to multiple high-resolution wind reanalysis products (two in Yang-N₂O, four in Weber-CH₄), and multiple piston velocity parameterizations (two in Yang-N₂O and four in Weber-CH₄). These estimates for each gas are not independent as they use the same MEMENTO database. The Yang-N₂O and Weber-CH₄ products use interpolation techniques to fill observational gaps, but the lack of observations likely leads to large uncertainties in coastal regions.

For CH₄ emissions, the contribution from gas bubble plumes must be taken into account in addition to the diffusive flux (arising from the air-sea difference in partial pressure and a gas exchange coefficient). The MEMENTO database allows the calculation of the diffusive CH₄ flux only because CH₄ from bubble plumes is usually not captured by the conventional CH₄ measurements based on discrete samples or continuous underway measurement systems. However, an estimate of the ebullitive (i.e., bubbling) CH₄ fluxes is, however, included in Weber-CH₄ (but not in MARCATS-CH₄), by combining previous seafloor emissions estimates with models of bubble transfer to the surface (Weber et al., 2019). We evaluated the uncertainty on the net Weber-CH₄ flux in the narrow and wide coastal oceans from the quadrature of uncertainties on diffusive and ebullitive fluxes, using a 50% uncertainty on diffusive flux and a 60% uncertainty on ebullitive flux (Weber et al., 2019). More details on these products can be found in Supporting Information S1.

2.2.3. Ocean Models for CO₂ and N₂O Fluxes

For CO₂, we used 15 ocean general circulation models coupled with biogeochemical modules: 11 are global and 4 are regional models, all covering the study period of 1998–2018 except CCSM-WHOI, which ends in 2017 (see details in Tables 2 and 3). Most global models have native horizontal grid resolutions varying between 0.25° and 1° in the coastal domain, except FESOM-HR which has an unstructured mesh that reaches a higher resolution (see Figure S3 in Supporting Information S1) and MPIOM-HAMMOC, NEMO-PlankTOM12 and CCSM-WHOI which have a coarser resolution of ~1.5°, ~2° and ~3° respectively (Table 2). The regional models covering the Indian Ocean (NYUAD-ROMS-Indian) and Northwest Atlantic Ocean (NW-Atl) have horizontal resolutions of approximately 10 km. The regional models covering the Atlantic (ETHZ-ROMS-Atl) and the Pacific Ocean (ETHZ-ROMS-Pac) have resolution varying in space between 4 and 120 km: the ETHZ-ROMS-Atl telescopes focus on the Amazon outflow region where the resolution is higher and the ROMS-ETHZ-Pac grid focuses on the California Current region (Table 3). We note that some of these models include land-sea nutrient and carbon inputs by rivers, while others do not. Details on these models can be found in Tables 2 and 3 and Supporting Information S1.

For N₂O, we use five models: three of them are also used for CO₂ (CNRM-HR, CNRM-LR, and NEMO-PlankTOM5) and cover the full study period (1998–2018), while the two other models are from the ECCO family (ECCO-Darwin and ECCO2-Darwin) in which the circulation is optimized to capture the distribution of tracers such as temperature and salinity in the ocean but cover shorter periods (ECCO-Darwin for 1997–2013 and ECCO2-Darwin for 2006–2013). See Table 2 and Supporting Information S1 for further details and references of each model.

Model-based analyses in this study use all global models available for the wide coastal ocean (i.e., 11 models for CO₂ and 5 for N₂O), but subsets of models that are eddy-permitting due to their higher native horizontal resolution are used for the narrow coastal ocean (4 models for CO₂: CNRM-HR, FESOM-HR, MOM6, MRI-ESM2.1, and 3 models for N₂O: CNRM-HR, ECCO-Darwin and ECCO2-Darwin, see Table 2). Global averages and integrated fluxes are based on the global models, while regional models were used in addition to the global models for the analysis at the grid-point scale (e.g., maps). Note that we did not examine the seasonal and interannual variability in N₂O and CH₄ fluxes, as these temporal scales are either unresolved (CH₄) or have not yet been analyzed (N₂O) in the coastal ocean.

2.3. Grid Harmonization and Coastal Ocean Waters Area Rescaling

All models and data products were re-gridded from their native grid onto the same 1/4° grid for analysis. However, due to differences in horizontal resolution and ocean-land mask definition, observational products and

Table 2
Description of Global Ocean Biogeochemical Models Used in This Study, Including the Wind Speed Product and Gas Exchange Coefficient (k_w) Formulation Used to Compute the Fluxes

Model	Gases	Land-sea inputs	Domain	Frequency/period in this study	Horizontal resolution	Wide area (million km ²)	Wind speed and k_w	Reference
CCSM-WHOI	CO ₂	No	Global	Mon 1998–2017	3.6°lon 0.8–1.8°lat	34.5	NCEP W92	Doney et al. (2009)
CNRM-LR	CO ₂ , N ₂ O	Yes	Global	Mon 1998–2018	1°lon 0.3–1°lat	64.8	JRA55-do W14	Séférián et al. (2019)
CNRM-HR	CO ₂ , N ₂ O	Yes	Global	Mon 1998–2018	0.25°	71.3	JRA55-do W14	Berthet et al. (2019)
FESOM-LR	CO ₂	No	Global	Mon 1998–2018	Unstructured ~1°	75.5	JRA55-do W14	Hauck et al. (2020)
FESOM-HR	CO ₂	No	Global	Mon 1998–2018	Unstructured ~0.25°	76.4	JRA55-do W14	Hauck et al. (2020)
IPSL	CO ₂	Yes	Global	Mon 1998–2018	1°lon 0.3–1°lat	65	JRA55-do W14	Bopp et al. (2015)
MOM6-Princeton	CO ₂	Yes ^a	Global	Mon 1998–2018	0.5°lon 0.25–0.5°lat	63.8	JRA55-do v1.3 W92	Liao et al. (2020)
MPIOM-HAMMOC	CO ₂	Yes	Global	Mon 1998–2018	1.5°	44.5	NCEP W14	Ilyina et al. (2013)
MRI-ESM2.1	CO ₂	No	Global	Mon 1998–2018	1°lon 0.3–0.5°lat	66.3	JRA55-do W14	Yukimoto et al. (2019)
NEMO-PlankTOM12	CO ₂	Yes	Global	Mon 1998–2018	2°lon 0.3–1.5°lat	62.8	NCEP W92	Wright et al. (2021)
NEMO-PlankTOM5	N ₂ O	Yes	Global	Mon 1998–2018	2°lon 0.3–1.5°lat	62.8	NCEP W92	Buitenhuis et al. (2018)
NorESM-OC2.0	CO ₂	Yes	Global	Mon 1998–2018	Nominal 1°	63.9	JRAv1.3 W14	Tjiputra et al. (2020)
ECCO-Darwin	N ₂ O	No	Global	Mon 1997–2014	1/3°lon	66.5	ERA-Interim W92	Carroll et al. (2020)
ECCO2-Darwin	N ₂ O	No	Global	Mon 2006–2013	1/6°lon	90.5	ECMWF & JRA-55 W92	Manizza et al. (2019)

Note. W92 and W14 stand for k_w -formulations from Wanninkhof (1992, 2014) respectively. Mon stands for monthly frequency. Wide coastal areas are calculated after the products and models have been regrided on the 0.25° × 0.25° grid. Further details and references on observation-based products and models are provided in Supporting Information S1.

^aCarbon inputs are only partial and calculated to roughly balance burial.

ocean biogeochemical models can have different coastal ocean areas even after they have been re-gridded to the same 1/4° grid (e.g., wide coastal ocean areas resolved by the models range from 34 to 76 million km² vs. 77.2 million km² in the mask of Laruelle et al. (2017), see Tables 1–3 and Figure 1). To minimize the effect of this common issue, most results are presented as area-weighted averages of CO₂, N₂O and CH₄ flux densities (per m²) and surface ocean pCO₂ masked using time varying ice-free surface to account for fractional sea ice coverage (in μatm). We used the ice fraction from the NOAA-OISST product for pCO₂-products and the ice fraction of each individual model for models. For the globally integrated CO₂ flux (in PgC year⁻¹), we used the globally averaged CO₂ flux densities found in each pCO₂-product and model for the narrow and wide coastal oceans and multiplied them by the corresponding coastal area of Laruelle et al. (2017, narrow area = 28 million km²; wide area = 77

Table 3
Description of Regional Ocean Biogeochemical Models Used in This Study, Including the Wind Speed Product and Gas Exchange Coefficient (k_w) Formulation Used to Compute the Fluxes

Model	Gases	Land-sea inputs	Domain	Frequency/period in this study	Horizontal resolution	Wind speed and k_w	Reference
ETHZ-ROMS-Atl	CO ₂	Yes	Regional Atlantic Ocean	Mon 1998–2018	~4–120 km ^a	ERA5 W14	Louchard et al. (2021)
ETHZ-ROMS-Pac	CO ₂	Yes	Regional Pacific Ocean	Mon 1998–2018	~4–60 km ^b	ERA5 W14	Desmet et al. (2022)
ACM-NWAtl	CO ₂	Yes	Regional Northwest Atlantic (36.3–53.8N; –74.6 to –45.2 E)	Mon 1998–2018	~9.5 km	ERA-interim H06, W14	Rutherford et al. (2021)
NYUAD-ROMS-Indian	CO ₂	Nutrients No carbon	Regional Indian Ocean (31.5°S to 31°N; 30°E to 120°E)	Mon 1998–2018	1/10°	ERA-interim W14	Lachkar et al. (2021)

Note. W92, H06 and W14 stand for k_w -formulations from Ho et al. (2006) and Wanninkhof (1992, 2014) respectively. Mon stands for monthly frequency. Further details and references on observation-based products and models are provided in Supporting Information S1.

^aHighest resolution in Amazon plume and western Africa (2 poles). ^bHighest resolution in California Current (1 pole).

million km²). We did not apply this area rescaling to the globally integrated fluxes of N₂O and CH₄ given the smaller number of products/models available.

2.4. Calculation of the Global Coastal Radiative Balance

We computed the coastal contribution to the radiative balance by converting global N₂O and CH₄ fluxes (i.e., spatially integrated annual net air-sea flux of greenhouse gases) to a mass of CO₂ equivalent (PgCO₂-e). Our analysis is based on contemporary greenhouse gas sinks and sources (not on the historical changes in those sinks/sources from a pre-industrial baseline), and therefore informs on the contribution of coastal oceans to the radiative balance but does not provide any information on the radiative forcing (e.g., perturbations of the pre-industrial radiative balance, Neubauer, 2021).

We used the Intergovernmental Panel on Climate Change (IPCC) Assessment Report 6 (Arias et al., 2021) updated 100-year global warming potential for N₂O (GWP_{N2O} = 273, i.e., the 100-year time integrated radiative forcing from the instantaneous release of 1 kg of N₂O is 273 times larger than the forcing of 1 kg of CO₂) and for CH₄ of non-fossil fuel origin (GWP_{CH4} = 27.2). We calculated two versions of this radiative balance for the wide coastal ocean: one using observation-based flux products only and one using mostly models. The observation-based budget uses the global gap-filled observational products, that is, the 4 pCO₂-product median flux for CO₂ (CMEMS*, Carboscope-1, Coastal-SOM-FFN and Merged-SOM-FFN), the Yang-N2O flux for N₂O and the Weber-CH4 flux for CH₄. Uncertainty bars presented for this observation-based balance give the ranges of all products presented in this study, that is, the 4 pCO₂-product range for CO₂, the 2 observational-product range for N₂O (Yang-N2O and MARCATS-N2O) and the 2 observational-product range for CH₄ (i.e., the low bound corresponds to the low uncertainty bound of Weber-CH4 and the high bound to the value of MARCATS-CH4). The model-based balance uses the 11 global model median flux for CO₂, the 4 global model median flux for N₂O, and the product-based Weber-CH4 flux for CH₄ as no model is available. Uncertainty bars presented for this model-based balance are the 11-model range for CO₂, the 4-model range for N₂O, and the 2-observational product range for CH₄ (same as the product-based balance described above).

3. Results

3.1. Global Coastal Ocean Greenhouse Gas Fluxes

In this section, we present a compilation of gap-filled observation-based and modeled net air-sea fluxes of CO₂ (4 pCO₂-products and 11 global ocean models), N₂O (2 observation-based products and 4 global ocean models) and CH₄ (2 observation-based products) in the global coastal ocean (Figure 2), and assess the contribution of the coastal ocean to the atmospheric greenhouse gas budget by combining the three gases using a single CO₂-equivalent flux (Figure 3).

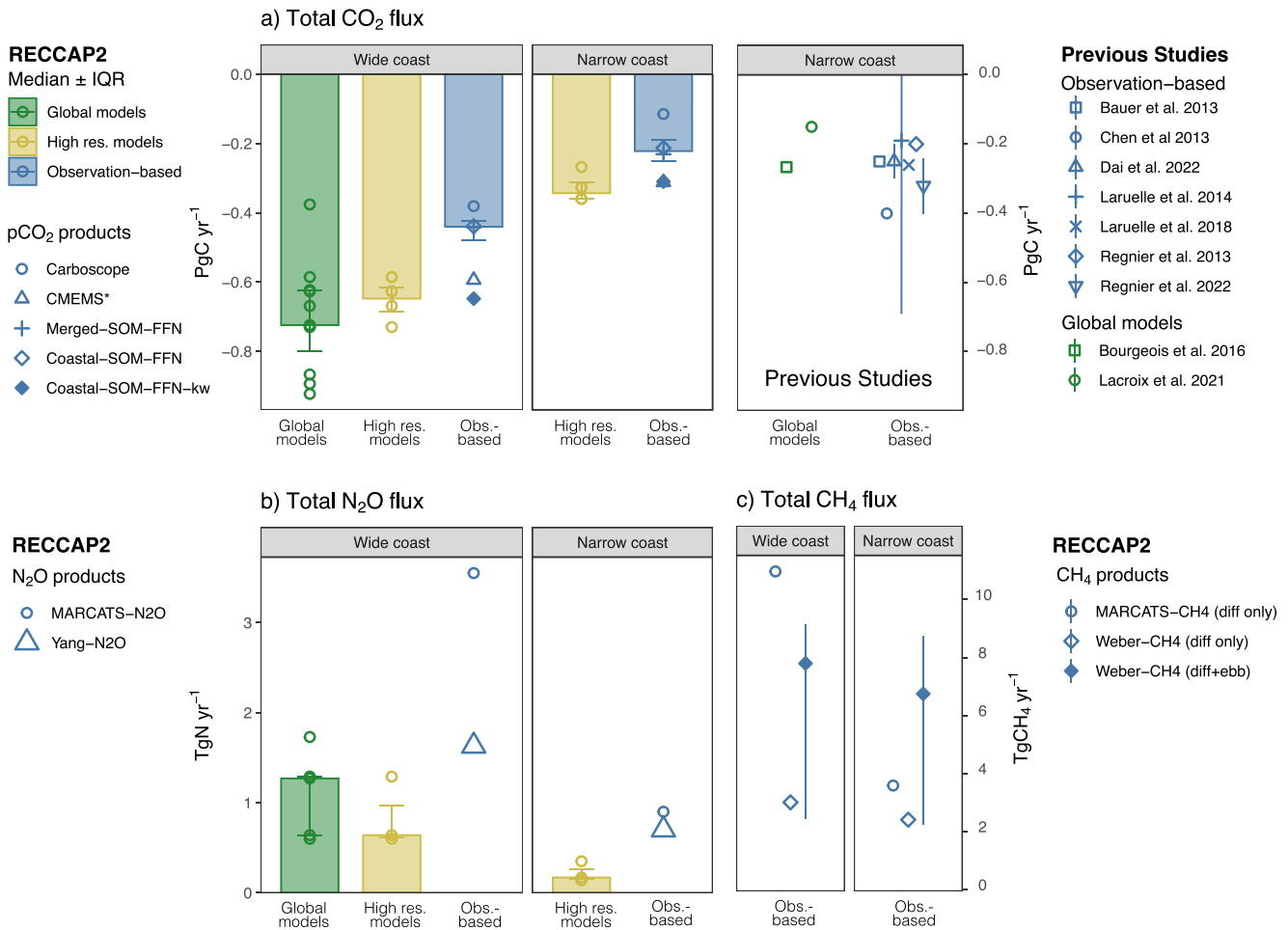


Figure 2. Net globally-integrated coastal fluxes of (a) CO₂ [PgC year⁻¹], (b) N₂O [Tg N year⁻¹] and (c) CH₄ [Tg CH₄ year⁻¹] over the wide and narrow coastal oceans. Figure shows individual products and models (symbols) and their median and interquartile ranges. Models are shown for the full ensemble available (11 models for CO₂ and 4 for N₂O) and a subset of higher resolution models (4 models for CO₂ and 2 for N₂O, see Methods and Table 2 for details). Previous estimates available for the narrow coastal ocean are shown on the right of panel a (see list in Table S2 in Supporting Information S1). Coastal-SOM-FFN-k_w, which is a second version of Coastal-SOM-FFN computed using different wind speed and k_w formulation (filled diamond, see Methods), is not used in the calculation of the pCO₂-product median. Weber-CH₄ total flux (diffusive + ebullitive) and diffusive contribution (comparable to MARCATS-CH₄ flux) are shown in panel (c).

3.1.1. Net Coastal Ocean CO₂ Uptake

The gap-filled pCO₂-products yield a weaker net CO₂ uptake than the global ocean biogeochemical models in the wide coastal ocean during the 1998–2018 period (Figure 2a). The pCO₂-product estimates (−0.59 to −0.37 PgC year⁻¹) fall at the upper (less negative) end of the model range (−0.92 to −0.38 PgC year⁻¹), and the pCO₂-product flux median (−0.44 PgC year⁻¹) is about two thirds of the model median (−0.72 PgC year⁻¹). Most of this model-product mismatch can be attributed to differences in ocean pCO₂ seasonality at mid- and high-latitudes (poleward of 25°N and 25°S), which tend to reinforce the northern hemisphere winter uptake in models compared to pCO₂-products (see details in Section 3.2.3). These differences in pCO₂ seasonality are likely amplified by differences in wind speed and gas exchange coefficient formulation (see Methods and Tables 1–3). For instance, the net CO₂ uptake in the Coastal-SOM-FFN product increases by about 50% and falls closer to the model median when changing the wind speed product (from ERA5 to JRA55) and gas exchange coefficient formulation (from Ho et al., 2011 to Wanninkhof, 1992) used to compute the flux (from −0.44 PgC year⁻¹ in Coastal-SOM-FFN to −0.65 PgC year⁻¹ in Coastal-SOM-FFN-k_w, blue dot vs. blue circle in Figure 2a, see further details in Section 3.2.3). We also note that using the subset of four global eddy-permitting models (CNRM-HR, FESOM-HR, MOM6, MRI-ESM2.1 with nominal horizontal resolution of 0.5° or higher) yields a weaker net CO₂ uptake (median of −0.65 PgC year⁻¹ for only four models vs. −0.72 PgC year⁻¹ for all global models), slightly closer to the pCO₂-products median (−0.44 PgC year⁻¹) and in relatively good agreement with

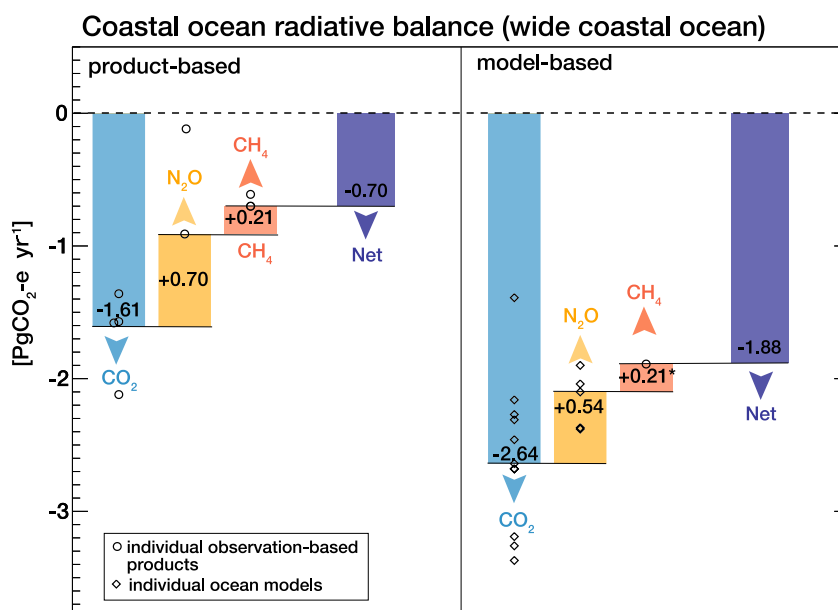


Figure 3. Wide coastal ocean atmospheric radiative balance (using $\text{PgCO}_2\text{-e year}^{-1}$) based on observational products and models of contemporary CO_2 , N_2O and CH_4 fluxes. Observation-based central values are from 4 pCO_2 -products, Yang-N2O and Weber-CH4. Model-based central values are from the 11 global models for CO_2 and 4 global models for N_2O , but the Weber-CH4 product is used for CH_4 as indicated by the asterisk (no model available for CH_4 , hence minimizing the difference between the two assessments). Individual models and observation-based product estimates are shown by symbols. The net greenhouse gas flux in $\text{PgCO}_2\text{-e year}^{-1}$ corresponds to the sum of the three gases' contributions.

one of the pCO_2 -products ($-0.59 \text{ PgC year}^{-1}$ in CMEMS*, Figure 2a). Other factors that could contribute to the differences between ocean biogeochemical models and pCO_2 -products (e.g., land-sea carbon and nutrient inputs) are discussed in Section 4.1.3.

We can compare the net CO_2 flux estimates presented here to prior work using the narrower definition of the coastal ocean ending at the shelf break (28 million km^2), a domain more aligned with the definition used in past studies (Table S2 in Supporting Information S1). For this comparison, we include all pCO_2 -products but use only the subset of four global eddy-permitting models with higher horizontal resolution. We find that the narrow coastal ocean accounts for about half of the wide coastal ocean CO_2 uptake (-0.22 out of $-0.44 \text{ PgC year}^{-1}$ for the 4- pCO_2 -product median and $-0.34 \text{ PgC year}^{-1}$ out of the $-0.65 \text{ PgC year}^{-1}$ for the 4-model median), while only accounting for about a third of the surface area. The pCO_2 -product median in the narrow coastal ocean ($-0.22 \text{ PgC year}^{-1}$) is consistent with the most recent observation-based estimates (Dai et al., 2022; Regnier et al., 2022; Roobaert et al., 2019), but the four pCO_2 -products span a relatively large range with differences in the order of a factor 2 ($-0.12 \text{ PgC year}^{-1}$ in Carboscope-1 and $-0.31 \text{ PgC year}^{-1}$ in CMEMS*, see Table S2 in Supporting Information S1 for estimates). The 4-model median simulates a slightly stronger sink ($-0.34 \text{ PgC year}^{-1}$) than these most recent estimates (although it is similar to the estimate of Regnier et al., 2022) but again differences in pCO_2 seasonality, and potentially in wind speed and gas exchange formulation could explain part of this discrepancy. Similar to the wide coastal ocean, the net CO_2 sink increases by nearly 50% in the narrow coastal ocean from Coastal-SOM-FFN to Coastal-SOM-FFN- k_w (from -0.21 to $-0.31 \text{ PgC year}^{-1}$, blue dot vs. circle, Figure 2a).

3.1.2. Net N_2O and CH_4 Coastal Ocean Emissions

Estimates of the global coastal emissions of N_2O range from 0.14 to 0.90 Tg N year^{-1} in the narrow coastal ocean and from 0.60 to 3.56 Tg N year^{-1} in the wide coastal ocean (Figure 2b). Part of this considerable variability comes from differences between model-based and observation-based estimates, but also from systematic differences between the two observation-based products (MARCATS-N2O and Yang-N2O). In the wide coastal ocean, the Yang-N2O estimate (1.63 Tg N year^{-1}) falls at the high end of the model estimates (0.60–1.73 Tg N year^{-1}), while MARCATS-N2O yields N_2O emissions that are more than twice the emissions of Yang-N2O (3.56 Tg N year^{-1} , Figure 2b). This finding implies that global ocean biogeochemical model emission estimates are overall lower

than those of the observation-based products. Furthermore, the subset of 3 high-resolution models generally simulates N_2O emissions that are lower than the full set of 5 models and therefore lower than both observation-based estimates (3-model median of $0.64 \text{ Tg N year}^{-1}$ vs. 5-model median of $1.27 \text{ Tg N year}^{-1}$ for the wide coastal ocean, Figure 2b). In the narrow coastal ocean, the two observation-based estimates are in relatively good agreement ($0.90 \text{ Tg N year}^{-1}$ in MARCATS- N_2O and $0.70 \text{ Tg N year}^{-1}$ in Yang- N_2O), while the subset of 3 eddy-permitting high resolution global ocean models simulate emissions that are again about 2–4 times lower ($0.14\text{--}0.35 \text{ Tg N year}^{-1}$). The Yang- N_2O product suggests that the narrow coastal ocean accounts for about 50% of the emissions of the wide coastal ocean, while in the subset of 3 global ocean models and MARCATS- N_2O it only accounts for about 25% (Figure 2b). We note, however, that the particularly low model values in both the 5-model ensemble and the 3-model high resolution subset are from ECCO-Darwin and ECCO2-Darwin ($0.60\text{--}0.64 \text{ Tg N year}^{-1}$ in the wide and $0.14\text{--}0.17 \text{ Tg N year}^{-1}$ in the narrow coastal ocean) which are based on the same model and are therefore not independent.

Global CH_4 emissions in Weber- CH_4 include both the diffusive and ebullitive (bubbling) components, and are estimated to be $6.80 [2.30\text{--}8.8] \text{ Tg CH}_4 \text{ year}^{-1}$ for the narrow coastal ocean and $7.85 [2.50\text{--}9.20] \text{ Tg CH}_4 \text{ year}^{-1}$ for the wide coastal ocean (Figure 2c). Note that the flux estimates presented here are observation-based only because no model-based estimates are available. The CH_4 flux from Weber- CH_4 is dominated by the ebullitive flux, which occurs mostly in shallow waters of the narrow coastal ocean (accounting for $4.33 \text{ Tg CH}_4 \text{ year}^{-1}$ in the narrow and $4.79 \text{ Tg CH}_4 \text{ year}^{-1}$ in the wide coastal ocean). Subtracting the ebullitive flux from the total Weber- CH_4 fluxes results in a CH_4 diffusive flux of $2.46 [1.23\text{--}3.69] \text{ Tg CH}_4 \text{ year}^{-1}$ in the narrow coastal ocean, which is in relatively good agreement with the diffusive flux estimated from MARCATS- CH_4 ($3.64 \text{ Tg CH}_4 \text{ year}^{-1}$). In contrast, the diffusive flux of $3.06 [1.53, 4.59] \text{ Tg CH}_4 \text{ year}^{-1}$ obtained in the wide coastal ocean in Weber- CH_4 has a central value ~ 3.5 times smaller than the diffusive flux of MARCATS- CH_4 ($11.02 \text{ Tg CH}_4 \text{ year}^{-1}$).

3.1.3. Coastal Ocean Radiative Balance

We combined coastal greenhouse gas emissions of CO_2 , N_2O and CH_4 to evaluate the coastal contribution to the radiative balance (in CO_2 -equivalent, Figure 3). We find that from a net radiative perspective, N_2O and CH_4 coastal emissions offset much of the coastal CO_2 sink, by $\sim 60\%$ in the product-based balance and $\sim 30\%$ in the model-based balance. As a result, the net greenhouse gas flux into the coastal ocean is $-0.66 \text{ PgCO}_2\text{-e year}^{-1}$ in the product-based balance ($-1.58 \text{ PgCO}_2\text{-e year}^{-1}$ CO_2 flux offset by $+0.70$ and $+0.21 \text{ PgCO}_2\text{-e year}^{-1}$ of N_2O and CH_4) and $-1.81 \text{ PgCO}_2\text{-e year}^{-1}$ in the model-based balance ($-2.57 \text{ PgCO}_2\text{-e year}^{-1}$ CO_2 flux offset by $+0.54$ and $+0.21 \text{ PgCO}_2\text{-e year}^{-1}$ of N_2O and CH_4 , Figure 3). Most of the difference between the product- and model-based radiative balances presented here comes from the stronger CO_2 uptake in the models mentioned above. However, there are very few global coastal N_2O and CH_4 estimates and the spread amongst the products and models are large ($1\text{--}2 \text{ PgCO}_2\text{-e year}^{-1}$), indicating that the compensation of the coastal carbon sink could be substantially different from 30%–60% found here.

3.2. Coastal Ocean CO_2 Dynamics

3.2.1. Contrast Between Coastal Ocean and Open Ocean

When averaged globally, models and pCO_2 -products show lower mean surface ocean pCO_2 and more negative CO_2 flux densities (i.e., more uptake) in narrow and wide coastal oceans than in the open ocean (Figure 4). This coastal to open ocean difference is found in the median of the four pCO_2 products, which shows an increase in global mean sea surface pCO_2 from the narrow coastal ocean to the wide coastal ocean ($+15 \mu\text{atm}$ from 350 to $365 \mu\text{atm}$) and from the wide coastal ocean to the open ocean ($+7 \mu\text{atm}$ from 365 to $372 \mu\text{atm}$ for the 1998–2018 period, Figure 4a). The only pCO_2 -product among the four without this apparent gradient is Carboscope-1, likely because of potential biases in the coastal Arctic Ocean (pCO_2 values generally higher in Carboscope-1 than in other pCO_2 -products and models).

The 11-model median simulates slightly higher ocean pCO_2 than the product median but it also captures an increase in global mean pCO_2 from a wide coastal ocean to open ocean ($+6 \mu\text{atm}$ from 369 to $375 \mu\text{atm}$) similar to the pCO_2 -products. The 4-model median (subset of eddy-permitting higher resolution models) also shows a consistently lower mean pCO_2 in the narrow coastal ocean ($363 \mu\text{atm}$), compared to the wide coastal ocean ($370 \mu\text{atm}$) and the open ocean ($373 \mu\text{atm}$). Thus, although observation-based and modeled pCO_2 values show

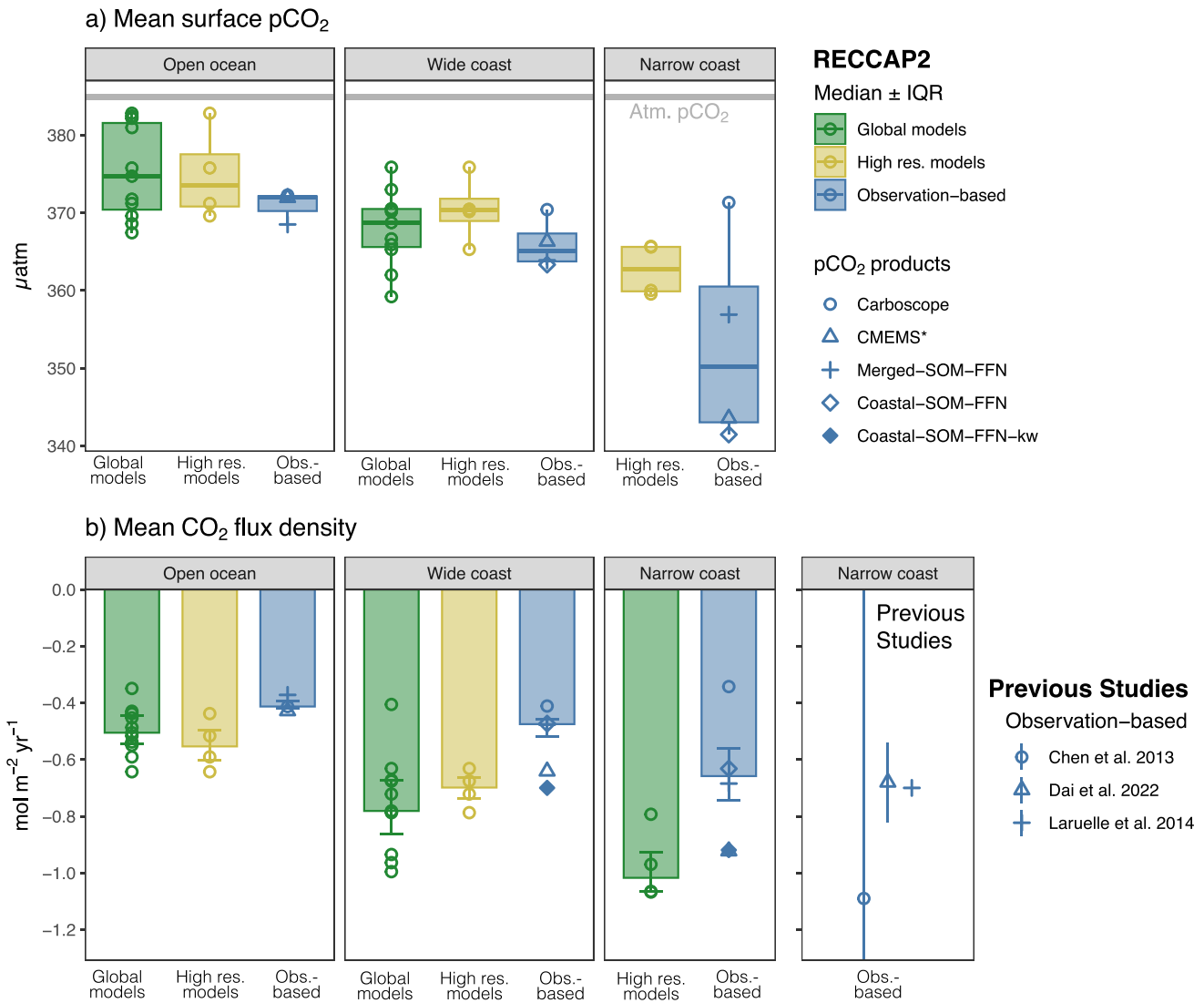


Figure 4. Comparison of globally-averaged (a) sea surface pCO₂ [µatm] and (b) flux density [mol C m² year⁻¹] averaged over open ocean, wide coastal and narrow coastal ocean waters in pCO₂-products and ocean models. Figure shows individual products and models (symbols) and their median and interquartile ranges. Models are shown for the full ensemble available (11 models) and a subset of higher resolution models (4 models, see Methods and Table 2 for details). Previous estimates available for the narrow coastal ocean are shown on the right of panel (b). Coastal-SOM-FFN-k_w, which is a second version of Coastal-SOM-FFN computed using different wind speeds and k_w formulation (filled diamond, see Methods), is not used in the calculation of the pCO₂-product median.

discrepancies in the mean within each domain, the narrow coastal to open ocean differences derived from observations and models are in remarkable agreement, and amount to about 10–15 µatm. This apparent coastal to open ocean gradient could be interpreted as decreasing pCO₂ landward, but should be interpreted carefully. As shown previously for the coastal-SOM-FFN product (Roobaert et al., 2019), this coastal to open ocean difference is attributable to the increasing contribution of polar waters, characterized by lower flux densities and stronger sinks, to the total surface area from open ocean to narrow coastal domains (polar coastal ocean waters account for 29% of the narrow coastal ocean, 17% of the wide coastal ocean and 2% of open ocean waters, contributions calculated as the percentage of ice-free surface area located poleward of 50° based on NOAA's OISST ice product, Figure 1).

A consequence of these coastal-to-open ocean differences in sea surface pCO₂ is that the mean partial pressure difference with the atmosphere (mean pCO_{2a} of 385 µatm for 1998–2018) is higher in the coastal ocean than in the open ocean. As a result, air-sea CO₂ flux densities are lower (stronger uptake) in the narrow coastal ocean (–1.02 and –0.66 mol m² year⁻¹ for 4-model and 4-product medians) than in open ocean waters (–0.55 and

$-0.41 \text{ mol m}^{-2} \text{ year}^{-1}$ for 4-model and 4-product medians, Figure 4b). In between, the wide coastal ocean shares characteristics of narrow coastal ocean and open ocean waters and is characterized by intermediate CO_2 flux densities (-0.70 and $-0.48 \text{ mol m}^{-2} \text{ year}^{-1}$ for 4-model and 4-product medians, Figure 4b).

3.2.2. Spatial Variability in Coastal Ocean CO_2 Sources and Sinks

Coastal air-sea CO_2 flux densities are characterized by latitudinal gradients captured by both pCO_2 -products and models (Figure 5). Mid- and high-latitude regions (poleward of 25° of latitude) are characterized by annual mean surface ocean pCO_2 lower than the atmosphere ($\text{pCO}_{2a} = 385 \text{ ppm}$ for 1998–2018) and thus by oceanic CO_2 uptake, whereas tropical coastal oceans (equatorward of 25° of latitude) are generally associated with pCO_2 similar or slightly higher than the atmospheric level and weak or near-zero CO_2 outgassing (Figure 5 and Figure S4 in Supporting Information S1). When averaged latitudinally over the wide coastal ocean, models and products follow a similar pattern, with most negative flux densities ($< -1 \text{ mol m}^{-2} \text{ year}^{-1}$, i.e., strongest sinks) at mid-latitudes in both hemispheres (50°S – 25°S and 25°N – 50°N) and high latitudes in the northern hemisphere (50°N – 80°N), and weak sources in the tropical band (typically between 0 and $0.5 \text{ mol m}^{-2} \text{ year}^{-1}$ in 25°S – 25°N , Figure 6a).

Largest departures between pCO_2 -products and models are found in the northern mid- and high latitudes, where the model median flux densities are often more negative (stronger sink) than the pCO_2 -product median (down to $-4 \text{ mol m}^{-2} \text{ year}^{-1}$ in models vs. $-2 \text{ mol m}^{-2} \text{ year}^{-1}$ in pCO_2 -products, Figure 6a). These systematically more negative flux densities in the models extend over large coastal areas of the northern hemisphere, including the shelves of western boundary currents (Gulf Stream and Kuroshio), the Norwegian Sea and the southern Greenland basin (blue colors in Figure 5c), and therefore largely explain the stronger globally integrated coastal sink found in the model median (Figure 2a). These northern hemisphere regions are relatively well sampled by the SOCAT pCO_2 database (Figure S1 in Supporting Information S1). In regions such as the coasts of Japan and eastern US, the comparison of SOCAT pCO_2 data to the pCO_2 -product and model medians suggests that models are indeed underestimating ocean surface pCO_2 and that the model-product difference might be largely attributable to model biases (Figure S5 in Supporting Information S1).

Models and pCO_2 -products also differ on Antarctic shelves, in particular at the tip of the Antarctic Peninsula (around 60°S which is also relatively well sampled compared to the rest of the coastal ocean) where models simulate a weak sink (about $-1 \text{ mol m}^{-2} \text{ year}^{-1}$) but pCO_2 -products show a weak source (about $+1 \text{ mol m}^{-2} \text{ year}^{-1}$, Figures 5 and 6a) supported by the comparison to the raw SOCAT data indicating that the model median underestimate surface ocean pCO_2 in this region (Figure S6 in Supporting Information S1). In the Antarctic Peninsula, however, the model-product mismatch is confined to a relatively small surface area and the impact on the net global flux is smaller compared to the mismatch found in the northern extratropics. Finally, we note that the model median yields less negative or more positive flux densities (i.e., weaker sinks or stronger sources) in some coastal regions, such as the California Current, Peruvian margin, Sea of Okhotsk, or Hudson Bay (red colors in Figure 5c), which offsets part of the stronger sinks simulated in northern and southern extratropical latitudes in the latitudinal mean and global integral. Comparison to SOCAT pCO_2 in the well-sampled California Current suggests that the model median overestimates ocean near-shore pCO_2 in the northern part of the region, likely due to the poor representation of the upwelling system, but that the pCO_2 -product median underestimates pCO_2 in the southern part of the region (Figure S6 in Supporting Information S1). This suggests that both pCO_2 -product bias and model bias might contribute to the model-product difference in this region. In poorly sampled regions, these model-product differences could be attributable to model bias, pCO_2 -product bias, or both.

3.2.3. Seasonal Variability in Coastal Ocean CO_2 Sources and Sinks

The global coastal ocean is a sink of CO_2 in all seasons (Figures 6c–6f). In the pCO_2 -products, the seasonal amplitude of the air-sea CO_2 flux is similar in both hemispheres and shows a strong latitudinal contrast between (a) the tropics (25°S – 25°N) where the flux is weak and the seasonal amplitude is small (absolute values $< 1 \text{ mol m}^{-2} \text{ year}^{-1}$); (b) the mid-latitudes (50°S – 25°S and 25°N – 50°N) where the seasonal amplitude is relatively large (absolute values of 1 – $2.5 \text{ mol m}^{-2} \text{ year}^{-1}$) and the sink is stronger in winter and spring; and (c) high-latitudes (poleward of 50°N and 50°S) where the seasonal amplitude is also large (similar to mid-latitudes) but the CO_2 sink is stronger in summer (except in the Arctic, north of 80°N , where the seasonal amplitude is small, Figures 6b–6f).

Global ocean biogeochemical models largely agree with the pCO_2 -products on the latitudinal patterns of seasonality but differences emerge in the seasonal phasing and amplitude, in particular north of 25°N (Figures 6c–6f).

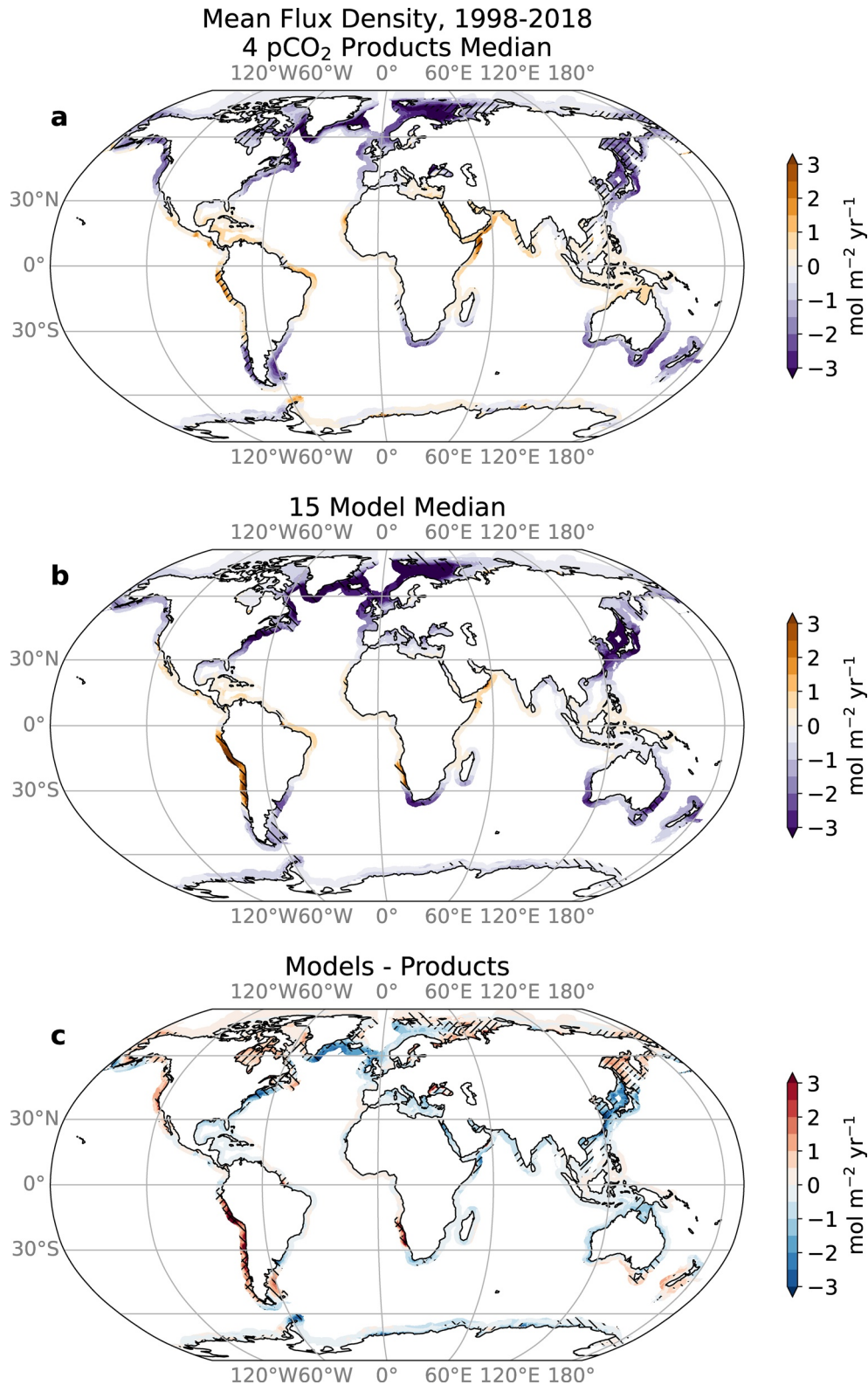


Figure 5. Annual-mean CO₂ flux density [mol C m² year⁻¹] in the wide coastal ocean for (a) the median across the 4 pCO₂-products, (b) the median across the 15 models, and (c) the difference between model and pCO₂-product medians. The model-median is calculated using the 11 global models and the 4 regional models, where available. Hatching indicates the coastal area with root mean square difference (RMSD) greater than 0.60 mol C m² year⁻¹ across pCO₂ products (panels a and c) or 0.95 mol C m² year⁻¹ across models (panels b and c) (in both cases the RMSD values correspond to the 20% of coastal area with highest RMSD).

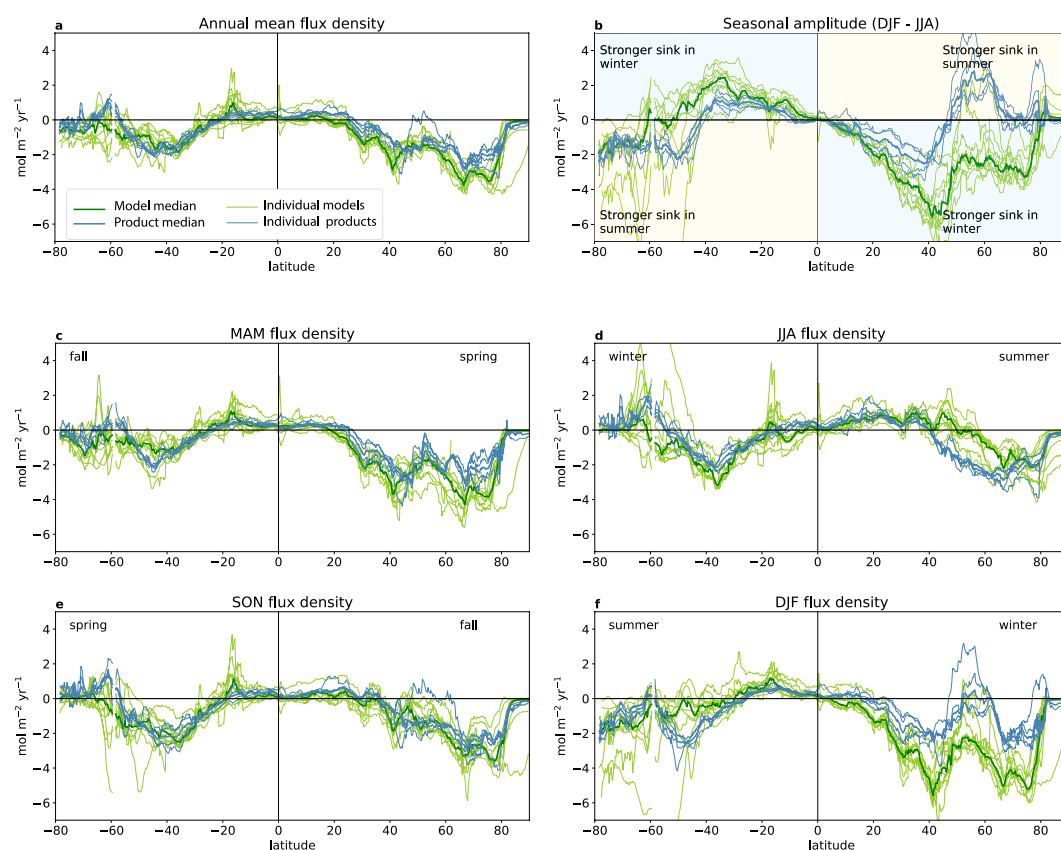


Figure 6. The latitudinal distribution of coastal ocean CO_2 flux and seasonal amplitude (using the wide coastal ocean). The latitudinal distribution of (a) annual mean flux density, (b) seasonal flux density amplitude, calculated as December–February (DJF) minus June–August (JJA), the blue (orange) quadrants indicate when the ocean uptake is stronger in winter (summer). (c–f) Seasonal mean flux density for March–May, JJA, September–November and DJF. Product and model medians are shown with thick lines and the individual 11 global models and 4 products with thin lines. Units are in $\text{mol C m}^{-2} \text{ year}^{-1}$ in all panels for consistency, converting from per month to per year also for the 3-month periods.

The seasonal amplitude of the CO_2 flux is 50%–100% larger in the models at mid-latitudes (despite having a similar phasing, i.e., stronger sink in winter and spring, Figure 6b). In addition, the CO_2 sink is systematically stronger in winter than in summer at all latitudes (except around Antarctica) and does not reproduce the latitudinal change in seasonal phasing obtained in the pCO_2 -products (from stronger winter uptake in the tropics to stronger summer uptake at high-latitudes, Figure 6b).

The products show little seasonality when averaged globally across coastal ocean waters (net median flux of $-0.35 \text{ PgC year}^{-1}$ for December–February [DJF] vs. $-0.32 \text{ PgC year}^{-1}$ for June–August [JJA], Figure 7a). This is largely explained by compensations between mid-latitudes (stronger uptake in winter) and high-latitudes (strong uptake in summer) within each hemisphere (Figure 6), which results in a relatively weak seasonality in both the northern (-0.24 in DJF and -0.22 in JJA) and southern ($-0.11 \text{ PgC year}^{-1}$ in DJF and $-0.10 \text{ PgC year}^{-1}$ in JJA) hemispheres (Figures 7b and 7c). In the case of the 11-model median, however, this compensation between the mid- and high-latitudes is much weaker and the seasonal cycle is stronger, especially in the northern hemisphere (-0.73 in DJF and $0.00 \text{ PgC year}^{-1}$ in JJA, Figures 7b and 7c). As a result, the global coastal ocean in the model median displays a marked seasonality controlled by the seasonality of the northern hemisphere, resulting in a net global coastal sink for DJF ($-1.15 \text{ PgC year}^{-1}$) that is about four times the sink for JJA ($-0.29 \text{ PgC year}^{-1}$, Figure 7). This model-product difference in CO_2 flux seasonality and specifically the extremely large boreal winter uptake explains the stronger annual mean global CO_2 sink found in the model median compared to the pCO_2 -products (Figures 2a, 6a, and 7b).

Model-product differences in CO_2 flux seasonality are largely tied to differences in the surface ocean pCO_2 . The stronger flux seasonality at mid-latitudes in models and the opposed flux seasonality at high latitudes (i.e., stronger uptake in winter in models vs. stronger uptake in summer in products) are both explained by the higher summer ocean pCO_2 (leading to weaker summer uptake) and the lower winter ocean pCO_2 (leading

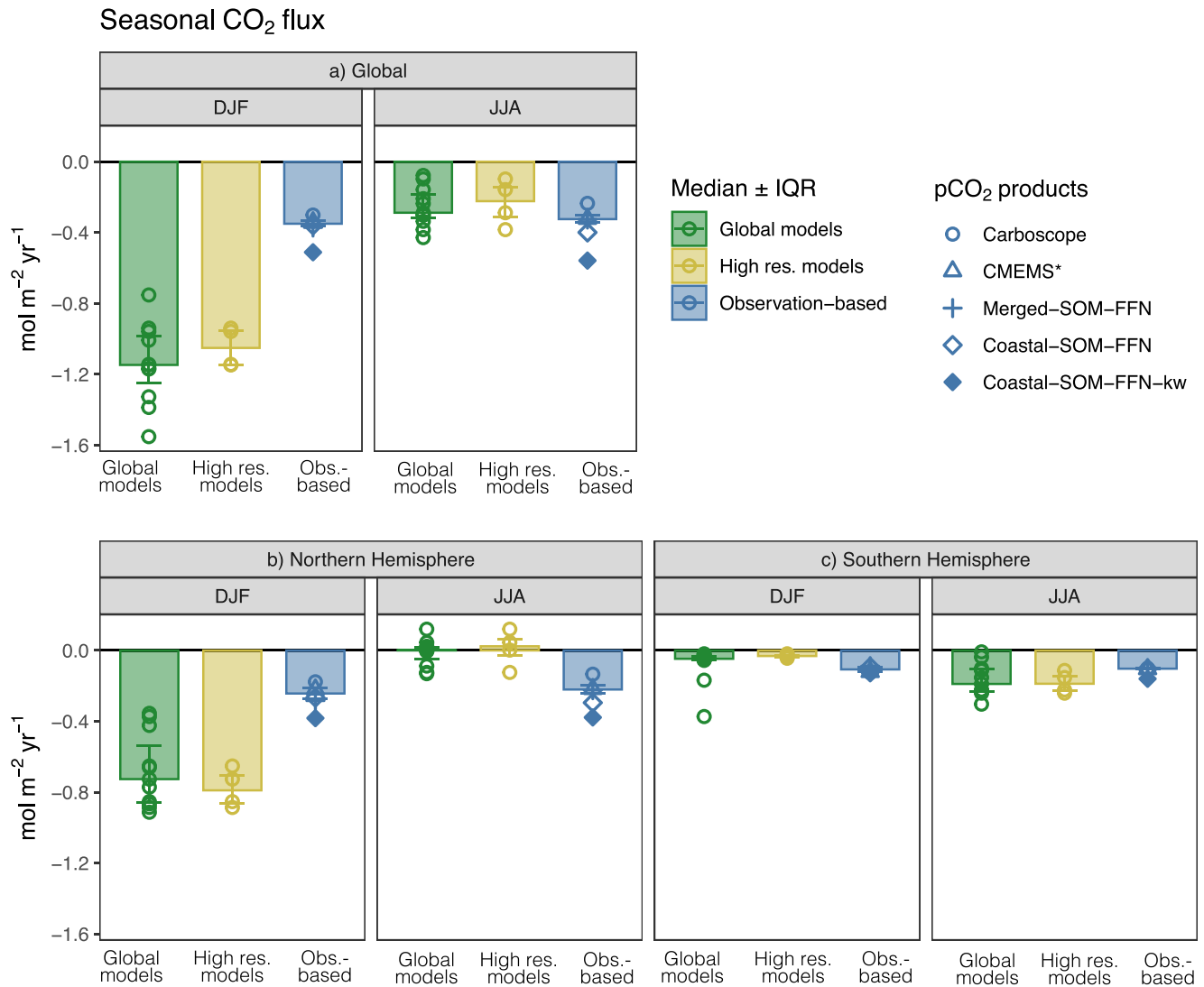


Figure 7. Net air-sea CO_2 flux in December–February and June–August for (a) the global coastal ocean and (b, c) the northern and southern hemispheres. Figure shows individual products and models (symbols) and their median and interquartile ranges. Models are shown for the full ensemble available (11 models) and a subset of higher resolution models (4 models, see Methods and Table 2 for details). Coastal-SOM-FFN- k_w , which is a second version of Coastal-SOM-FFN computed using different wind speeds and k_w formulations (filled diamond, see Methods), is not used in the pCO_2 -product median. Units are in PgC year^{-1} in all panels for consistency, using a 12-month scale up value for the 3-month periods.

to stronger winter uptake) found in the model median compared to the pCO_2 -product median (Figure S7 in Supporting Information S1). This systematic model/product difference in seasonality is also found in the open ocean but the amplitude of this mismatch is amplified in the coastal ocean (see Figures S9 and S10 in Supporting Information S1).

These differences in ocean pCO_2 seasonality can be amplified by the choice of wind speed and gas exchange coefficient formulation. The comparison of the two Coastal-SOM-FFN versions reveals that both the high-latitude summer uptake and the mid-latitude winter uptake are enhanced in Coastal-SOM-FFN- k_w compared to Coastal-SOM-FFN (Figure 7 and Figure S8 in Supporting Information S1). This enhancement occurs in both hemispheres, but the impact of the northern hemisphere on the global coastal annual mean uptake is larger due to the larger coastal surface area. Finally, we note that some models reproduce better the latitudinal pattern expected from the pCO_2 -products, in particular the stronger summer uptake at high-latitude in the northern hemisphere (e.g., MOM6-Princeton and MPIOM-HAMOCC, see individual models in Figure S9 in Supporting Information S1 and thin green lines overlapping with thin blue lines in Figure 6).

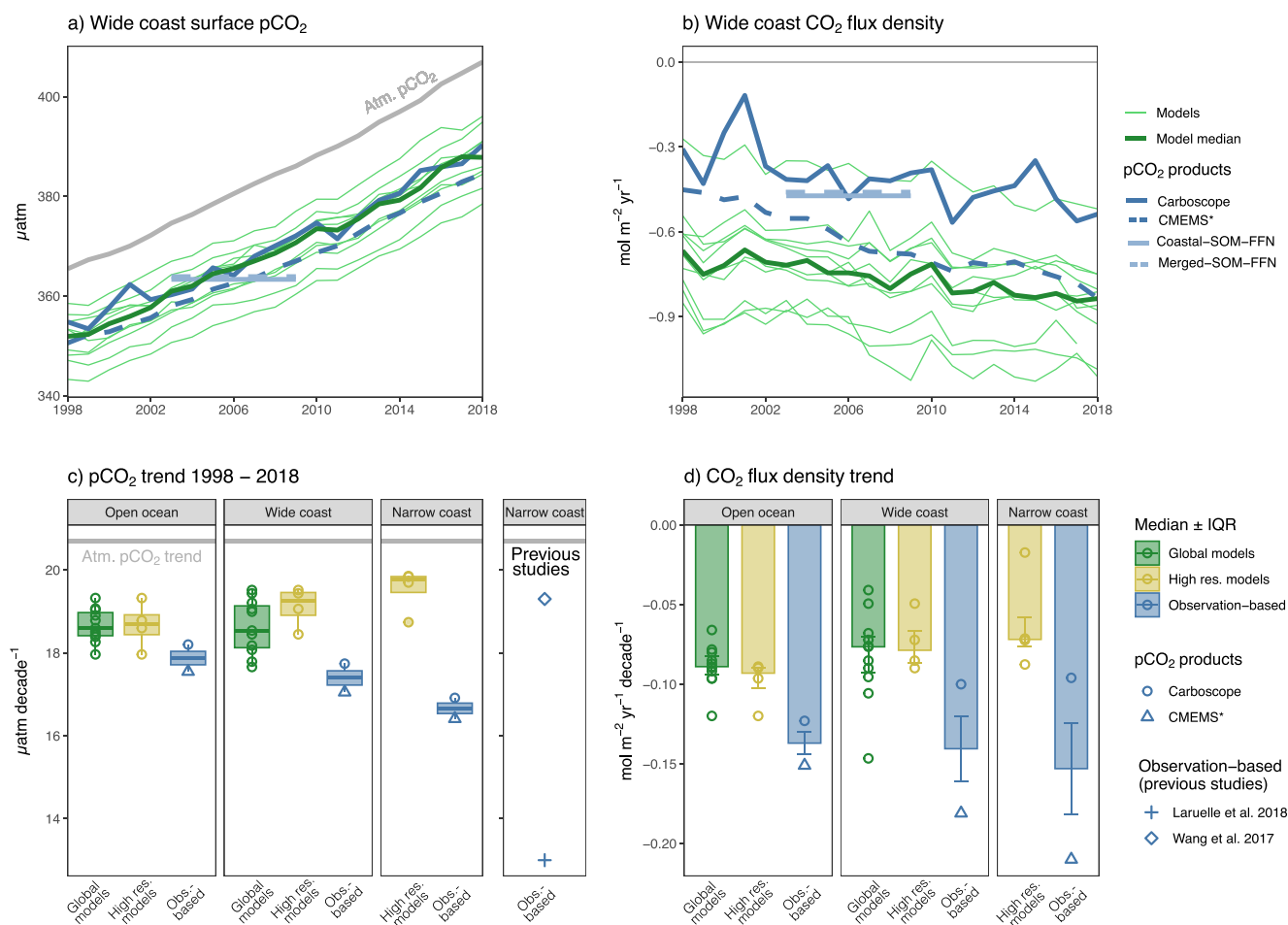


Figure 8. Temporal evolution of the global annual mean (a) surface ocean $p\text{CO}_2$ [μatm] and (b) net air-sea CO_2 flux density [$\text{mol C m}^{-2} \text{ year}^{-1}$] in the four $p\text{CO}_2$ -products (blue lines) and the 11 global ocean models (thin green lines) and model median (thick green lines) in the wide coastal ocean. Trends in (c) ocean surface $p\text{CO}_2$ [$\mu\text{atm decade}^{-1}$] and (d) flux density [$\text{mol C m}^{-2} \text{ year}^{-1}$] for the open ocean, wide coastal and narrow coastal waters. Panels b and c show individual products and models (symbols), the model median and interquartile range for the full model ensemble (11 models) and a subset of higher resolution models (4 models, see Methods and Table 2 for details). Note that only two time-varying $p\text{CO}_2$ -based products are available to calculate trends (Carboscope-1 and CMEMS*, other products are monthly 1998–2015 climatologies). Prior estimates are shown on the right of panel c (see Table S1 in Supporting Information S1).

3.2.4. Trends in Coastal Ocean CO_2 Flux and Surface $p\text{CO}_2$

For the 1998–2018 period, global coastal $p\text{CO}_2$ trends are slightly weaker than the atmospheric $p\text{CO}_2$ trend ($+20.7 \mu\text{atm/decade}$) in the two $p\text{CO}_2$ -products that are time-varying (about $+17$ – $18 \mu\text{atm/decade}$ in the wide coastal ocean) and in the models ($+17$ – $20 \mu\text{atm/decade}$ in the wide coastal ocean; see Figures 8a and 8c). In the narrow coastal ocean, the $p\text{CO}_2$ trends from the $p\text{CO}_2$ -products are lower than in the wide coastal ocean, and fall halfway between the two central values published in previous observation-based estimates ($+16$ – $17 \mu\text{atm/decade}$ vs. $+19.3 \mu\text{atm/decade}$ in Wang et al., 2017 and $+13 \mu\text{atm/decade}$ in Laruelle et al., 2018). In contrast, the $p\text{CO}_2$ trends found in the subset of four high resolution ocean biogeochemical models are higher in the narrow coastal ocean ($+19.8 \mu\text{atm/decade}$) than in the wide coastal ocean, and in good agreement with the highest of the previous observation-based estimate (Wang et al., 2017).

The trend difference between atmospheric and oceanic $p\text{CO}_2$ leads to an increase in the coastal carbon sink from 1998 to 2018 in $p\text{CO}_2$ -products and models (flux density trends between -0.15 and $-0.04 \text{ mol m}^{-2} \text{ year}^{-1}$ per decade in the wide coastal ocean, Figures 8b and 8d). Yet, because the rate of increase in coastal $p\text{CO}_2$ is lower in the $p\text{CO}_2$ -products than in the models, their respective CO_2 uptake trend is larger (Figure 8c). This is consistent with the expectation that a slower increase in sea surface $p\text{CO}_2$, which does not closely follow the atmospheric $p\text{CO}_2$ trend, should result in a stronger increase in the flux density (e.g., Laruelle et al., 2018).

Our results show, however, that $p\text{CO}_2$ trends and flux trends are not directly proportional, suggesting that factors other than $p\text{CO}_2$ variability are at play. These include trends in sea-ice cover (e.g., sea-ice retreat influence on flux trends in the Arctic Ocean) and/or in surface winds (via their effect on the gas exchange transfer velocity). For instance, the Carboscope-1 $p\text{CO}_2$ trends are slightly weaker than the CMEMS* $p\text{CO}_2$ trends in the narrow and wide coastal oceans, and yet the increase in the coastal sink is lower in Carboscope-1 than in CMEMS* (Figures 8c and 8d). Another example of the decoupling between $p\text{CO}_2$ trends and flux trends is found in the coastal to open ocean difference. The global ocean biogeochemical model ensemble simulates smaller differences between atmospheric and oceanic $p\text{CO}_2$ trends in the coastal ocean than in the open ocean, resulting in a weaker increase in the carbon sink in the coastal ocean (following here the expected link between $p\text{CO}_2$ and flux trends, Figures 8c and 8d). In contrast to the models, both time-resolving $p\text{CO}_2$ -products reveal higher differences between atmospheric and oceanic $p\text{CO}_2$ trends in the coastal ocean than in the open ocean (Figure 8c), which would suggest a stronger trend in the flux density in the coastal ocean (i.e., a stronger increase in the uptake). This expected increase in the uptake is, however, only found in CMEMS* and not in Carboscope-1 (Figure 8d).

Inconsistencies between $p\text{CO}_2$ trends and flux trends arise from the complex and uncertain interplay between the spatio-temporal changes in ocean $p\text{CO}_2$, wind speed and sea-ice coverage. Trends in ocean $p\text{CO}_2$ and therefore in $\Delta p\text{CO}_2$ (difference between coastal ocean surface ocean $p\text{CO}_2$ and atmospheric $p\text{CO}_2$) strongly differ between the two time-varying $p\text{CO}_2$ -products (Figure S11 in Supporting Information S1). CMEMS*, as well as the multi-model median, show more negative $\Delta p\text{CO}_2$ trends (potentially stronger uptake or weaker sources with time) in mid-to-high latitudes, but less negative or even positive $\Delta p\text{CO}_2$ trends in the tropics and in the Arctic (Figure S12 in Supporting Information S1). In contrast, the Carboscope-1 product shows strongly negative $\Delta p\text{CO}_2$ trends in the Arctic, and much larger variability in trends at other latitudes. These differences in $\Delta p\text{CO}_2$ trends explain, to the first order, the differences in flux density trends (negative $\Delta p\text{CO}_2$ trends generally yield negative flux trends, i.e., stronger uptake or weaker sources with time, Figure 9 and Figure S12 in Supporting Information S1).

The $\Delta p\text{CO}_2$ trends can be amplified or dampened by trends in wind speed and sea-ice coverage, which are also strongly spatially heterogeneous (see sea-ice trends in Figure S13 in Supporting Information S1). This effect is highlighted by the spatial differences and sometimes even a switch in sign between $\Delta p\text{CO}_2$ trends and air-sea CO_2 flux trends in the model median in sea-ice regions (hatching in Figure 9). This is true, for instance, in the Arctic where the ocean models tend to simulate an increase in ocean CO_2 uptake despite a positive trend in $\Delta p\text{CO}_2$ (i.e., ocean $p\text{CO}_2$ increases at a high rate than atmospheric $p\text{CO}_2$ which would reduce ocean uptake with constant sea-ice coverage and winds, Figure 9 and Figure S12 in Supporting Information S1). This decoupling between CO_2 flux and $\Delta p\text{CO}_2$ in the Arctic is indeed associated with a decrease in sea ice coverage in most models (Figure S13 in Supporting Information S1) and an increase in wind speed in two of the wind products that are widely used in these models (JRA-55 and ERA-5, Figure S14 in Supporting Information S1 and Tables 2 and 3), both effects inducing an increase in the flux with time despite the reduction in $\Delta p\text{CO}_2$. These results clearly indicate that the global coastal sinks are increasing. However, the magnitude of this increase, its spatial patterns and how it compares to the open ocean are still uncertain.

3.3. Coastal Ocean Nitrous Oxide and Methane Spatial Variability

The spatial distribution of the coastal N_2O fluxes computed with the observation-based (i.e., Yang-N2O) and the mean of the model-based approaches are shown in Figure 10. Coastal N_2O fluxes are generally positive, indicating that coastal areas are a source of atmospheric N_2O . Flux densities vary considerably, from 0 (=equilibrium with the atmosphere) to about $10 \text{ g N m}^{-2} \text{ year}^{-1}$. The results from Yang-N2O reveal hotspots of N_2O emissions in eastern boundary upwelling systems, the upwelling areas of the northwestern Indian Ocean, the subpolar North Pacific, the Baltic Sea, the Black Sea and the shallow marginal seas of Southeast and East Asia. These are generally characterized by high surface productivity, low subsurface oxygen, and shallow oxyclines. High N_2O emissions from these regions thus likely reflect subsurface water-column production by a combination of nitrification and denitrification pathways, both of which are enhanced in the presence of low O_2 and high remineralization rates, and subsequent transport to the surface by upwelling and mixing processes. Similar hotspot regions are detected in the model-ensemble median, although with somewhat reduced magnitude relative to the observational products, and with the notable exception of marginal seas in Asia and Europe, suggesting

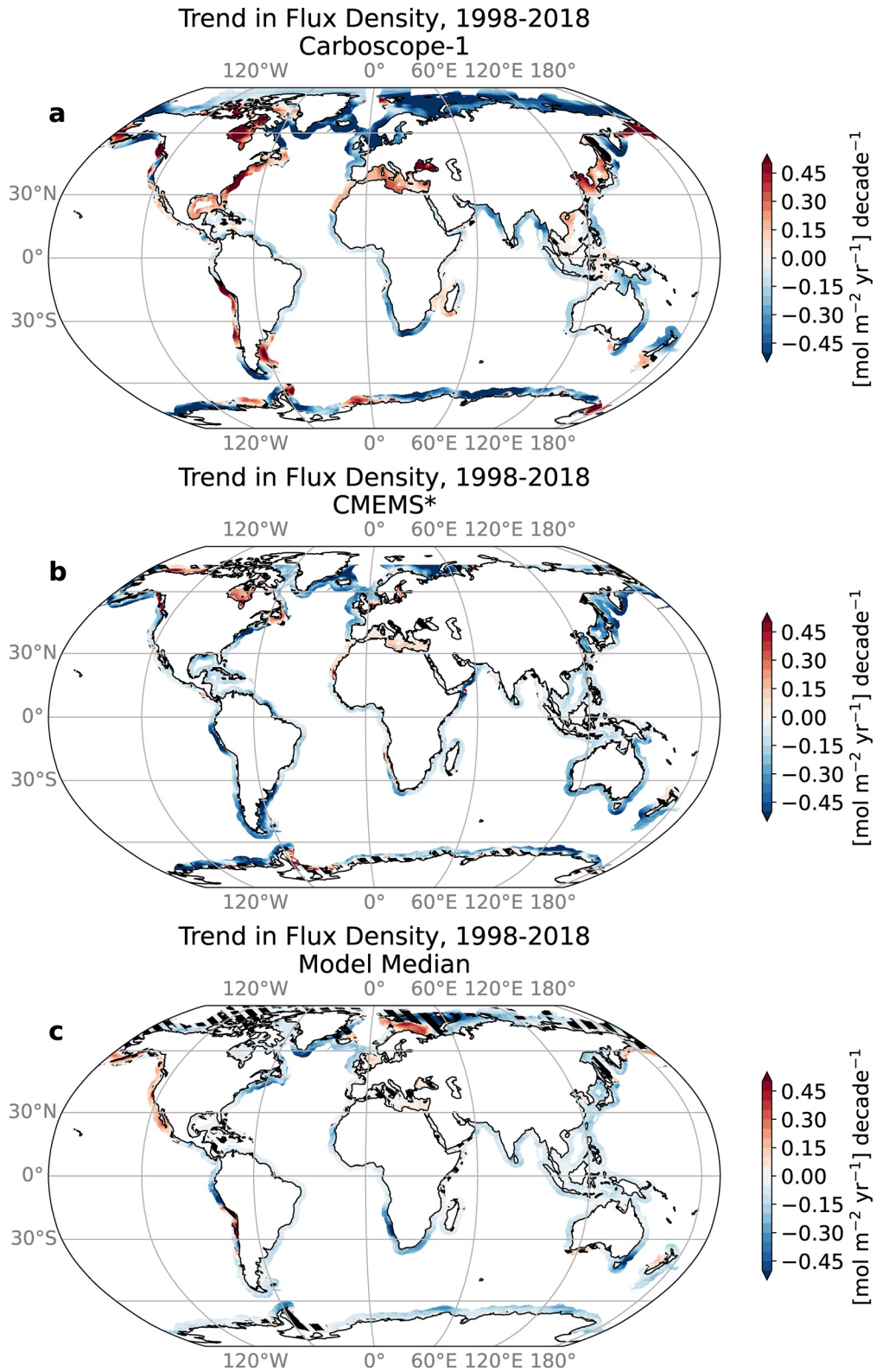


Figure 9.

that global models might not fully capture the nitrogen cycle in these regions, or the mechanisms transporting N_2O -laden waters to the surface.

The model-ensemble also identifies mid-latitude western boundary systems, including the US East Coast, the North Pacific east of Japan, the southeast coast of Australia, and the southeastern tip of Africa, as additional areas of intense N_2O emissions that are not captured by the Yang- N_2O product. Notably, these regions are not generally characterized by high surface productivity and low subsurface O_2 as coastal upwelling systems, although vigorous mixing along western boundary currents may favor local N_2O outgassing in the models. Most of these regions are also not densely sampled by observations in the MEMENTO database, in particular along the US, South Africa, and Japan eastern coasts, and thus the Yang- N_2O observational extrapolation may be poorly constrained there. The magnitude of the flux in these hotspots often differs among the data products and model-ensemble (Figure 10c). The N_2O flux distributions shown in Figure 10 likely reflect the fact that enhanced coastal N_2O concentrations—and thus enhanced N_2O emissions fluxes—are associated with enhanced land-sea inputs of nitrogen (as nitrate or ammonium) or with upwelling of N_2O -enriched subsurface water masses in upwelling systems.

The spatial distribution of the coastal CH_4 fluxes computed with the observation-based Weber- CH_4 product are shown in Figure 11. Coastal CH_4 fluxes are generally positive and range from 0 to $0.4 \text{ g } CH_4 \text{ m}^{-2} \text{ year}^{-1}$ indicating that coastal areas are a source of atmospheric CH_4 . Patterns in CH_4 emissions in Weber- CH_4 are largely correlated with water depth, with the most intense emissions cooccurring at depth shallower than 50 m (Figure 11). Indeed, coastal emissions of CH_4 are largely fueled by benthic-sourced biogenic methane, which is produced via methanogenesis in anoxic sediments and diffusively released into the overlying water column (Arndt et al., 2013; Bourgeois et al., 2016; Reeburgh, 2007). The benthic CH_4 source is enhanced in coastal waters where the rapid organic matter flux to the seafloor drives sediment anoxia and rapid sediment accumulation inhibits the growth of methane oxidizing microbes (e.g., Egger et al., 2016). Furthermore, aerobic respiration acts as an efficient sink of CH_4 in the water column (Mao et al., 2022), meaning that transfer from the seafloor to the surface must be extremely rapid if CH_4 is to be emitted to the atmosphere. Ebullition (bubbling) from CH_4 -enriched sediments can provide an important alternative pathway for CH_4 to surface (Rehder et al., 1998), but CH_4 is rapidly stripped from rising bubbles (McGinnis et al., 2006) and a small fraction reaches the surface only in shallow water depths. This further strengthens the coastal-offshore gradient in CH_4 emissions and explains why total emissions differ very little between the narrow and wide coast regions in Weber- CH_4 (Figure 2c).

We find that coastal CH_4 emissions are further enhanced in hotspots under the significant influence of freshwater discharge (Rosentreter et al., 2021), which, due to their low sulfate concentration, promotes the degradation of organic matter through the methanogenesis pathway. In addition to the biogenic CH_4 production pathway, CH_4 emissions can also be driven by geologically sourced methane originating from shallow seafloor seeps fed by hydrocarbon reservoirs or high-latitude hydrates (Puglini et al., 2020; Ruppel & Kessler, 2017). Overall, the distribution of coastal CH_4 emissions (Figure 11) can largely be understood in terms of water depth, organic matter production and delivery to sediments, and freshwater inputs.

4. Discussion

4.1. Coastal Ocean CO_2 Fluxes

4.1.1. Net CO_2 Uptake and Challenges Tied to Spatial Variability

This study presents a synthesis of the global coastal ocean air-sea CO_2 fluxes combining observational pCO_2 -based products and an ensemble of ocean biogeochemical models. The global ocean biogeochemical models yield a net median CO_2 uptake in the wide coastal ocean that is about 0.28 PgC stronger than the one inferred from pCO_2 -products for the 1998–2018 period, equivalent to a $\sim 60\%$ stronger sink ($-0.44 \text{ PgC year}^{-1}$ for products vs. $-0.72 \text{ PgC year}^{-1}$ for models in the wide coastal ocean). This mismatch of model- and product-based work in the net coastal CO_2 sink arises from a combination of factors, including strong differences in the coastal CO_2 flux

Figure 9. 1998–2018 trend in CO_2 flux [$\text{mol C m}^{-2} \text{ year}^{-1}$] for (a) Carboscope-1; (b) CMEMS* (area north of 75°N removed) (c) multi-model median (global and regional models). Hatching indicates regions where the flux trend has a different sign than the ΔpCO_2 trend (shown in Figure S12 in Supporting Information S1), highlighting the influence of wind and/or sea-ice trends. Negative values indicate that the ocean uptake increases or the ocean outgassing decreases with time (both leading to more carbon accumulation in the ocean). Trends are shown for the only two time-varying pCO_2 -based products available (Carboscope-1 and CMEMS* in blue, other pCO_2 -based products are monthly climatologies).

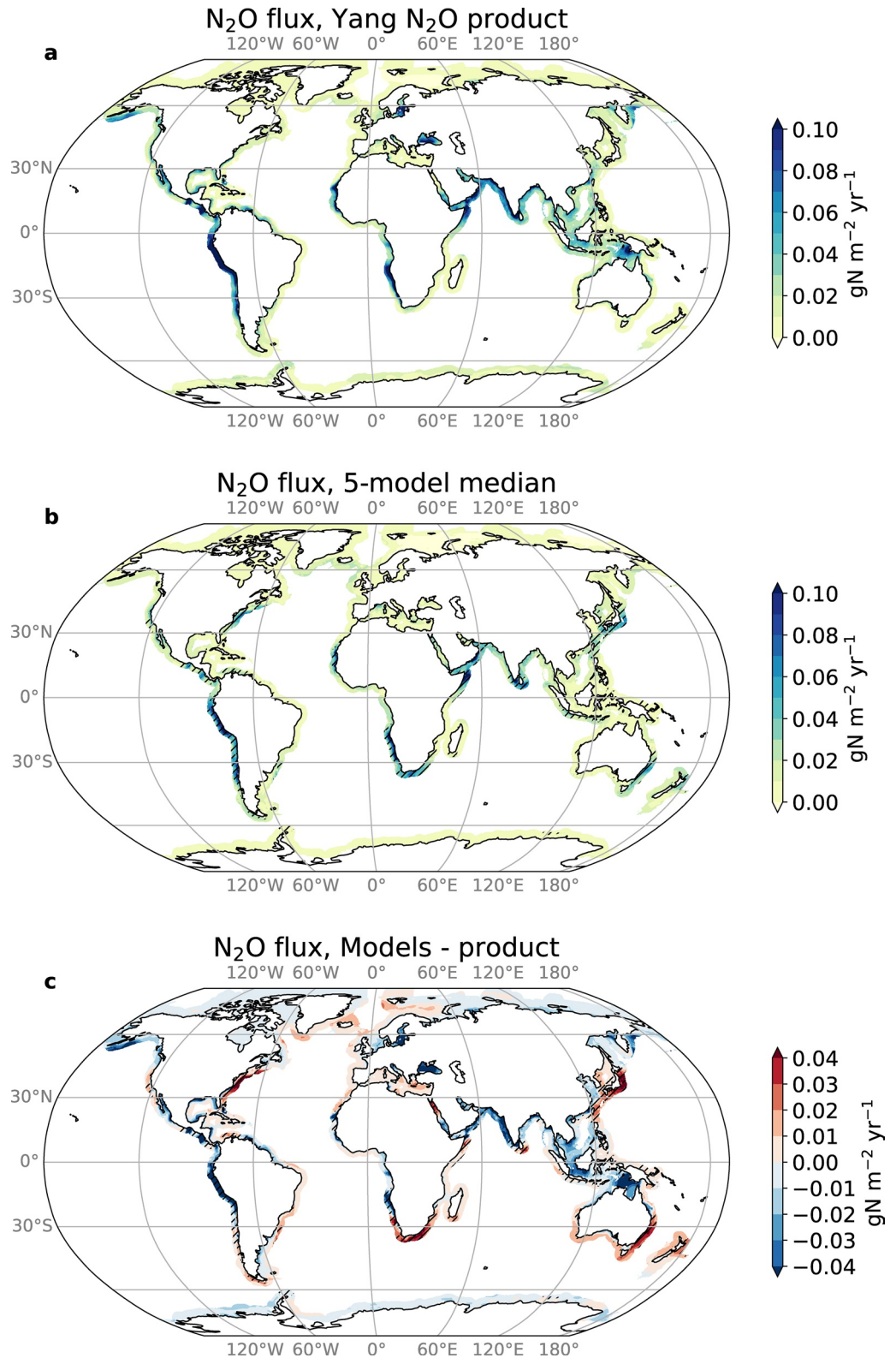


Figure 10. Maps of coastal N₂O flux (in g N m⁻² year⁻¹) from (a) Yang-N₂O product, (b) the mean of the 5 global ocean models that simulate N₂O (CNRM-LR, CNRM-HR, ECCO-Darwin, ECCO2-Darwin, and NEMO-PlankTOM5), and (c) the difference between models and the Yang-N₂O product. Hatching in panels (b, c) shows where root mean square difference (RMSD) among models exceeds 0.016 g N m⁻² year⁻¹ (RMSD threshold corresponds to the 20% of coastal area with highest RMSD).

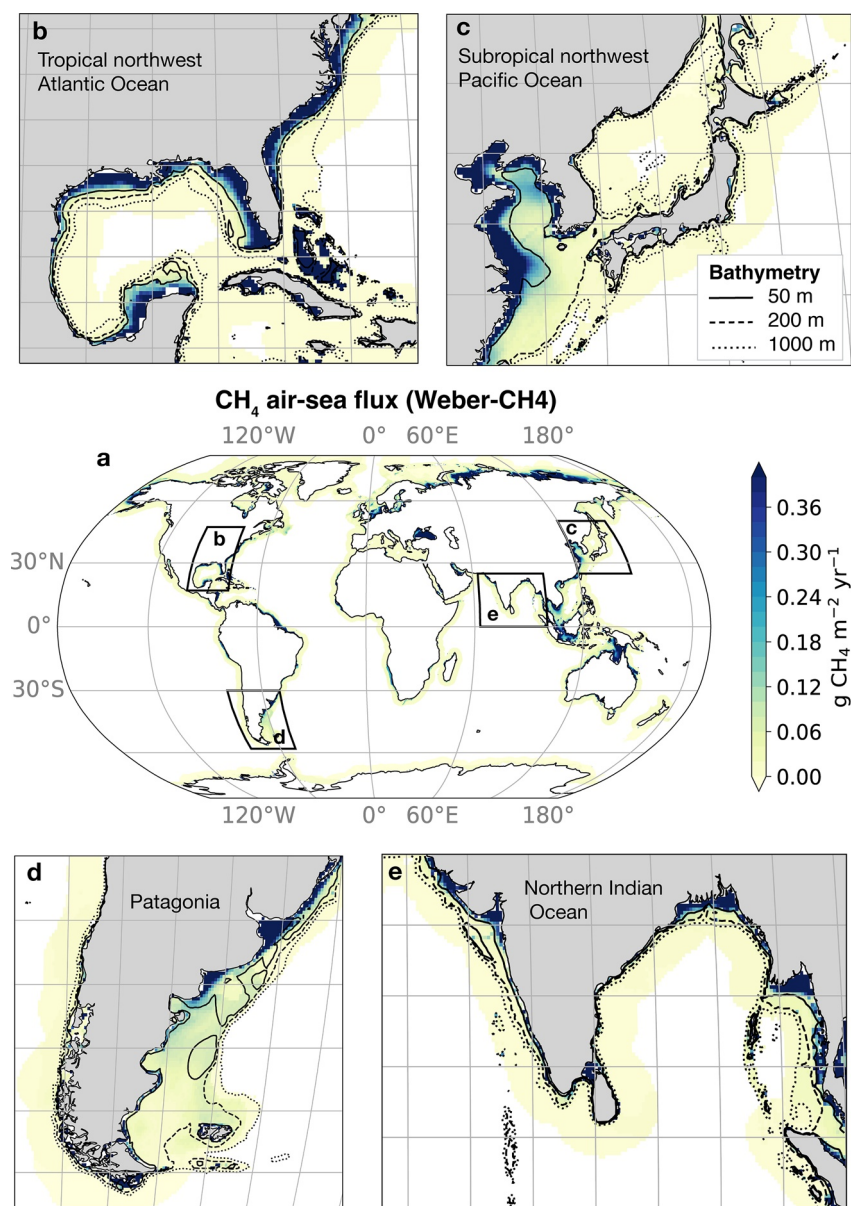


Figure 11. (a) Global maps of coastal CH₄ flux from the Weber-CH4 product (includes diffusive and ebullitive flux, in g CH₄ m⁻² year⁻¹); (b–e) insets show the CH₄ flux in four coastal regions along with 50, 200 and 1,000 m bathymetry contours. CH₄ emissions are most intense in shallow coastal environments.

seasonality (themselves attributed to differences in ocean pCO₂ seasonality and potentially wind speed and gas exchange transfer coefficient formulation) resulting in a stronger wintertime CO₂ uptake in northern subpolar and polar coastal systems in models (see Section 4.1.2).

We find that the subset of global ocean biogeochemical models with the highest spatial resolution yields a slightly weaker net CO₂ uptake (−0.65 PgC year⁻¹) in better agreement with the pCO₂-products than the full model ensemble. The small number of models in that subset (4) makes any statistical argument about resolution difficult. Yet, this result suggests that a better representation of fine scale coastal dynamics improves the representation of the CO₂ flux, likely by improving the representation of the physical and biogeochemical processes controlling CO₂ seasonality in the northern hemisphere (Laurent et al., 2021; Rutherford et al., 2021; Rutherford & Fennel, 2022). The horizontal resolution of the global ocean biogeochemical models used in this synthesis (about 1/4° or coarser) is, however, still too coarse to fully capture coastal ocean dynamics (coastal oceans require resolutions of 1/16° or higher, Hallberg, 2013).

This synthesis confirms the hypothesis of prior work that when averaged globally, CO₂ flux densities are more negative (stronger sinks) in the coastal ocean than in the open ocean waters (Dai et al., 2022; Laruelle et al., 2010, 2014; Roobaert et al., 2019). As put forward by Roobaert et al. (2019), we find that the differences between coastal and open ocean flux densities are largely explained by the disproportionate contribution of high latitude systems (generally strong sinks) to the coastal ocean surface area. Global ocean biogeochemical models and pCO₂-products agree relatively well on this coast-to-open ocean contrast in CO₂ flux densities, but recent syntheses of discrete observations (Cao et al., 2020; Dai et al., 2022) find stronger heterogeneity than the global pCO₂-products and global ocean models presented here, suggesting that gap-filling approaches might smooth some of the coastal ocean spatial variability and supporting the fact that higher horizontal resolution is needed to resolve coastal oceans in models (Hallberg, 2013).

4.1.2. Differences in Seasonality Influence the Net Coastal Ocean CO₂ Uptake

The seasonality in the four pCO₂-products used here falls into three latitudinal regimes. Tropical coastal waters (25°S–25°N) are characterized by small seasonal amplitudes and a stronger sink or weaker source in winter, both attributed to the weak seasonal thermal changes that slightly reduce surface ocean pCO₂ in winter (Laruelle et al., 2014; Roobaert et al., 2019). Mid-latitude coastal waters (50°S–25°S and 25°N–50°N) are characterized by larger seasonal amplitudes and a stronger CO₂ sink in winter and spring, likely due to the combined effect of thermal changes which lowers ocean pCO₂ in winter, biological drawdown of dissolved inorganic carbon (DIC) which further lowers pCO₂ during the spring bloom, and the influence of stronger winds in winter (Laruelle et al., 2014; Roobaert et al., 2022). High latitude coastal waters (poleward of 50°N and 50°S) are characterized by seasonal variations similar in magnitude to mid-latitudes, but where the maximum CO₂ uptake occurs in summer in response to intense biological drawdown. The biologically-driven uptake in high-latitude systems peaks a few months later than in mid-latitude systems because of the poleward propagation of the bloom (Ouyang et al., 2020, 2022; Roobaert et al., 2019, 2022; Siegel et al., 2002).

This marked seasonality in CO₂ fluxes contrasts with the RECCAP synthesis, which found very little seasonality in global coastal CO₂ flux densities, although the results were deemed inconclusive because of the sparse data and averaging process required to analyze the data available at the time (Chen et al., 2013). The results found here are, however, consistent with more recent work. In particular, the transition from thermally driven systems in the tropics (stronger winter sinks) to biologically driven systems at high latitudes (stronger summer sinks), and the increase in seasonal amplitude from tropical to high-latitude systems found in the pCO₂-products, are consistent with the global seasonal patterns in the coastal ocean described by Roobaert et al. (2019), the global open ocean seasonality patterns assessed in the framework of RECCAP2 (Rodgers et al., submitted to the RECCAP2 special issue) and supported by field and remote sensing studies at regional scale (Ouyang et al., 2022; Signorini et al., 2013; Tu et al., 2021).

Our synthesis reveals, however, strong differences in seasonality between pCO₂-products and global ocean biogeochemical models. The model median simulates a weak CO₂ flux seasonality in tropical coastal oceans similar to the pCO₂-products, but yields a CO₂ uptake that is stronger in winter at mid- and high-latitudes. This is likely due to a weaker contribution of biologically induced seasonality compared to thermal changes in the models, which would explain the lower surface ocean pCO₂ simulated in winter (due to an underestimated upward transport of remineralized DIC) and the higher pCO₂ simulated in spring/summer (due to a weaker biological drawdown). Part of these systematic differences compensate for the global mean coastal flux (winter vs. summer, northern vs. southern hemisphere), but because the model-product difference is larger in winter in the northern hemisphere, the net CO₂ uptake in the wide coastal ocean is about 60% larger in the model median.

The RECCAP2 chapter on open ocean seasonality (Rodgers et al., submitted to the RECCAP2 special issue) finds a similar systematic bias in model winter-to-summer pCO₂, which they attribute to a generally too small surface DIC seasonal cycle in models compared to observation-based reconstructions. This bias is particularly evident in the subpolar North Atlantic and North Pacific Oceans, where it manifests itself not only as a difference in amplitude but also in phasing. In these regions, the simulated too low DIC seasonality results in a thermal control of the pCO₂ seasonality in the global ocean biogeochemical models and thus in a phase shift of the seasonal pCO₂ cycle compared to the observation-based estimate dominated by non-thermal forcing. This suggests that the systematically stronger winter sink and weaker summer sink found in northern coastal waters in the models are at least partly attributable to general biases in the biogeochemical (e.g., bloom dynamics) or physical (e.g., vertical mixing) components of the ocean models, rather than a characteristic of the models that is specific to the

coastal ocean. See details in the RECCAP2 studies of (Rodgers et al., submitted to the RECCAP2 special issue). Nevertheless, we find that the amplitude of this systematic model/product difference in seasonality is amplified in the coastal ocean (see Figures S9 and S10 in Supporting Information S1).

Differences in ocean $p\text{CO}_2$ seasonality between models and $p\text{CO}_2$ products can be amplified by differences in gas exchange coefficient k_w , either through the influence of winds or the gas exchange coefficient formulation (which are different across the different ocean biogeochemical models and $p\text{CO}_2$ -products, Tables 1–3), and maybe to a lesser extent spatio-temporal differences in sea-ice cover (e.g., lower ice cover in some products/models could yield stronger fluxes). In models, the surface $p\text{CO}_2$ and k_w are tightly coupled in the sense that a larger k_w drives down the air-sea $p\text{CO}_2$ disequilibrium and therefore the air-sea CO_2 flux. In contrast, the calculation of the flux in $p\text{CO}_2$ products (except for Carboscope-1 which links fluxes and $p\text{CO}_2$ changes in a mixed-layer carbon budget equation) is done offline without any compensatory effect between k_w and air-sea $p\text{CO}_2$ disequilibrium. Therefore, the observation-based flux assessments are even more sensitive to the choice of the wind and k_w parameterization. For instance, we find that the net global coastal CO_2 uptake in the Coastal-SOM-FFN product is increased by nearly 50% in the wide and the narrow coastal oceans when changing the wind product (from ERA-interim to JRAv1.3) and gas exchange parametrization (from Ho et al., 2011 to Wanninkhof, 1992, see Table 1). These results are in line with published literature that assessed the impact of k_w parametrizations on global air-sea CO_2 fluxes (Boutin et al., 2009; Reichl & Deike, 2020; Roobaert et al., 2018) but highlight that its influence is also crucial in the coastal ocean because of the disproportionate contribution of mid- to high-latitude/high-wind systems in the total coastal area. The sensitivity of CO_2 flux estimates to the k_w parameterizations and their disparate implementation in the different assessments (including among ocean biogeochemical models, Tables 1–3) calls for more consistent approaches in future research. Furthermore, global wind-based gas exchange parameterization might not capture the complexity of the coastal ocean processes, such as the influence of bubbles entrained by wave breaking (Deike & Melville, 2018; Woolf et al., 2019), the presence of high surfactant concentrations (Pereira et al., 2018), or fine scale water-side convection (Gutiérrez-Loza et al., 2022).

4.1.3. Land-Sea Carbon and Nutrient Fluxes

An additional factor that could explain part of the difference in the net CO_2 uptake between $p\text{CO}_2$ -products and models is the presence of systematic bias in global ocean biogeochemical models, in particular the contribution of carbon land-sea riverine inputs or the models' horizontal resolution and ability to resolve coastal dynamics. At pre-industrial times (and assuming steady-state consistent with stable ice-core atmosphere CO_2 values; Elsig et al., 2009), the supply of carbon from land must have been balanced by burial in sediments and an outgassing of CO_2 from the ocean to the atmosphere. This land-driven outgassing flux, recently estimated to be $0.65 \pm 0.3 \text{ PgC year}^{-1}$ (mean ± 2 -sigma) for the global open ocean (Regnier et al., 2022, note that this outgassing of $0.65 \text{ PgC year}^{-1}$ is quantified for the open ocean outside of the narrow coastal ocean and thus include part of the wide coastal ocean), is still active today and therefore partially offsets the ingassing CO_2 flux that is directly driven by anthropogenic CO_2 emissions to the atmosphere (e.g., Friedlingstein et al., 2022; Regnier et al., 2022; Resplandy et al., 2018). Observation-based $p\text{CO}_2$ -products estimate the net contemporary flux of CO_2 , and therefore implicitly include the fluxes of natural and anthropogenic carbon, as well as the outgassing fluxes of carbon from land origin (e.g., Hauck et al., 2020). Most models, however, do not or only partially include these land-sea carbon inputs (see Tables 2 and 3) and are therefore likely to overestimate the net CO_2 ocean uptake, in particular in coastal waters adjacent to the land (Lacroix et al., 2020).

In globally integrated estimates, such as analyzed in the Global Carbon Budget (e.g., Friedlingstein et al., 2022) or the IPCC (Arias et al., 2021), the net air-sea CO_2 flux can in principle be adjusted for the outgassing of carbon from land to isolate the oceanic net sink, or it can be used to shed light on differences between modeled and observation-based flux estimates (e.g., Friedlingstein et al., 2022; Hauck et al., 2020). The RECCAP2 open ocean chapters estimated the spatial distribution of this land-driven CO_2 outgassing by upscaling the spatial distribution from an open-ocean model (Lacroix et al., 2020) to match an independent bottom up constraint on its global magnitude ($0.65 \pm 0.3 \text{ PgC/year}$; Regnier et al., 2022). This estimate suggests that $0.12 \text{ PgC year}^{-1}$ out of the $0.65 \text{ PgC year}^{-1}$ of land-driven CO_2 outgassing occurs in the wide coastal ocean, which could explain part of the model-product discrepancy. It is important to recognize, however, that the spatial distribution of this land-driven outgassing and contribution to the coastal ocean air-sea flux are very poorly constrained. In particular, we note that the model used to estimate the land-driven outgassing pattern (Lacroix et al., 2020) lacks some of the processes that control the magnitude (hence the upscaling to match the global number of $0.65 \text{ PgC year}^{-1}$ from Regnier et al. (2022) but also the spatial distribution of this outgassing (e.g., CO_2 uptake by coastal vegetation).

Another factor to consider is the land-sea input of nutrients, which could promote biological CO₂ uptake in coastal waters downstream of the river mouth (e.g., Gao et al., 2023; Louchard et al., 2021; Terhaar et al., 2021). However, we find no clear relationship between the strength of the simulated net coastal CO₂ uptake and the presence or absence of land-sea inputs in the global ocean biogeochemical models used here (i.e., models with weaker coastal CO₂ uptake more in line with pCO₂-products are not systematically the ones with land-sea inputs), suggesting that land-driven inputs are likely not the main factor in this discrepancy. This result is in agreement with prior work showing a relatively modest impact of riverine nutrient inputs on coastal productivity compared to the physical supply by cross-shelf and along-shelf exchanges on the shelf (e.g., Cotrim Da Cunha et al., 2007; Wollast, 1998). In addition, the models considered here either include both carbon and nutrient land-sea inputs or neither (Tables 2 and 3), leading to a potential offset between land-driven CO₂ outgassing associated with carbon runoffs and biological CO₂ uptake associated with nutrient runoffs (although we do not expect the patterns of the CO₂ outgassing and biological CO₂ uptake to match). This might explain why models with land-sea carbon inputs did not systematically yield weaker CO₂ uptake in the coastal ocean compared to the one without land-sea inputs.

4.1.4. Uncertain Trends in Coastal Ocean CO₂ Flux and Decoupling From pCO₂

This synthesis indicates that the coastal ocean CO₂ sink has increased between 1998 and 2018, in line with the expectation from previous work that showed surface pCO₂ in the narrow coastal ocean increasing at a smaller rate than in the atmosphere (Laruelle et al., 2018; Wang et al., 2017). The rate at which the coastal sink has increased is, however, poorly constrained by the models and products presented here (flux density trend varies by a factor 2 between the two time-varying pCO₂-products and by a factor 3 between the 11 models). In addition, it is still unclear if this increase in the global coastal CO₂ sink is comparable, slower, or faster than in the open ocean due to the inconsistent responses found in models and the two time-varying pCO₂-products but also in prior modeling and observation-based work (Bourgeois et al., 2016; Lacroix et al., 2021; Laruelle et al., 2018; Wang et al., 2017). The CMEMS* pCO₂-product suggests that the CO₂ uptake increases faster in the coastal ocean than in the open ocean, which is in line with the prior observation-based results of Laruelle et al. (2018). In contrast, the ensemble of 11 global ocean models and the CarboScope-1 pCO₂-product suggest that the coastal ocean sink is increasing at a slightly smaller rate than the open ocean, a result in line with another prior work based on pCO₂ observations (Wang et al., 2017) and global ocean biogeochemical models (Bourgeois et al., 2016; Lacroix et al., 2021).

Bourgeois et al. (2016) explained the weaker increase in the coastal carbon sink compared to the open ocean by a bottleneck in offshore transport, which leads to anthropogenic carbon accumulation and limits the ability of coastal waters to take up anthropogenic carbon. Although we did not quantify surface residence time or off-shelf transport in this study, our finding that the modeled CO₂ sink increases at a lower rate in the coastal region than in the open ocean lends support for this interpretation. Other processes at play could explain this behavior. For instance, relatively shallow waters in coastal oceans might limit the exchanges with deep (free of anthropogenic CO₂) waters, such that the coastal ocean surface layer saturates more quickly with additional CO₂ added to the atmosphere. In models, this slower rate is associated with regions of increased outgassing or reduced uptake, but the specific regions at play vary across models (e.g., North Pacific, Mediterranean Sea and Parts of the Arctic in the model median in this study vs. tropical ocean and parts of the Arctic in Lacroix et al., 2021), highlighting further the uncertainties that remain in constraining coastal trends.

Discrepancies between the different estimates of the CO₂ flux trends at least partly arise from the sparse temporal pCO₂ observational coverage. For instance, the prior studies of Laruelle et al. (2018) and Wang et al. (2017) only covered a small portion of the coastal surface area and might not be representative of the global ocean. This is supported by regional studies that identified coastal ocean pCO₂ trend weaker than the atmospheric pCO₂ trend (i.e., potentially yielding intensified CO₂ uptake or decreased outgassing) such as the northern Gulf Stream margin, the South China Sea, the Sea of Japan, the North Sea and the Antarctic Peninsula (Bauer et al., 2013; Dai et al., 2022; Laruelle et al., 2018; Wang et al., 2017), but also regions where coastal ocean pCO₂ increases at a similar rate (i.e., near-zero changes in the flux) or even higher rates (i.e., reduced CO₂ uptake or intensified outgassing) than atmospheric pCO₂, such as in the Baltic Sea (Schneider & Müller, 2018), the California Current or along the eastern US coast (Dai et al., 2022; Laruelle et al., 2018; Reimer et al., 2017; Salisbury & Jönsson, 2018; Xu et al., 2020).

Another source of discrepancy is the decoupling found between global coastal pCO₂ trends and flux trends, suggesting that the CO₂ flux trends are sensitive to trends in winds and sea-ice (via the gas exchange coefficient), and how they combine with the pCO₂ trends. This sensitivity to sea-ice and winds is likely more pronounced in

the observation-based estimates, which rely on an “offline” calculation of the flux (no mechanistic link between $p\text{CO}_2$ disequilibrium, wind and sea-ice, except for CarboScope-1), or even more simply assume that slower trends in coastal ocean $p\text{CO}_2$ translate into faster growing coastal CO_2 flux (e.g., Laruelle et al., 2018), an assumption that is not fulfilled in the 2 $p\text{CO}_2$ -products used in this study (although it does work in the multi-model median).

4.1.5. Confidence in $p\text{CO}_2$ -Products and Ocean Biogeochemical Models

The systematic differences found between the ensemble median of global ocean models and $p\text{CO}_2$ -products (including the larger net annual mean CO_2 uptake found in global ocean models, the different timing of mid- and high-latitude seasonality and the large range found in flux density trends) should be interpreted with caution. First, some models capture better than others the patterns reconstructed by the $p\text{CO}_2$ -products. In particular, some models are able to reproduce the stronger summer sink found at high-latitudes, or simulate a net annual mean CO_2 flux that better matches the product-based estimates. In addition, differences between products and models do not necessarily equate to model bias. Observation-based products rely on gap-filling techniques (mixed layer model or machine learning), and regions of largest product-model mismatch often correspond to regions where the observational sampling is sparse (68% of the wide coastal ocean surface area was never sampled, and of the sampled area, 33% has data for only 1 month in a single year, Figure S1 in Supporting Information S1) and where the spread across the observation-based products and across the global models is the highest (hatching on Figures 5a and 5b). In contrast, coastal regions that are relatively well sampled by observations and well constrained by the products generally correspond to regions of agreement between the observation-based and model-based estimates (Roobaert et al., 2022). Thus, while we have overall more confidence in the observation-based estimates of the ocean carbon sink, the uncertainties associated with these reconstructed estimates remain high. This precludes a clear conclusion about whether the observation- or model-based estimates are closer to the truth.

4.2. Coastal Ocean N_2O and CH_4

The coastal ocean is a substantial source of atmospheric N_2O (Yang et al., 2020) and a minor source of atmospheric CH_4 (Saunois et al., 2020; Weber et al., 2019). The N_2O flux estimates presented here for the narrow coastal ocean (0.14–0.75 Tg N year⁻¹) is lower than a previous estimate of the mean global N_2O fluxes from coastal waters (including upwelling and marginal seas) in the range of 1.9–3.0 Tg N year⁻¹ (Bange et al., 1996). The mean CH_4 flux estimates for the narrow coastal ocean (2.46–3.19 Tg CH_4 year⁻¹ for the diffusive flux and up to 6.79 Tg CH_4 year⁻¹ when accounting for the ebullitive flux in the narrow coastal ocean) are also in good agreement with a recently published mean CH_4 flux from shelves (0–200 m water depth) of 5.7 Tg CH_4 year⁻¹ (Rosentreter et al., 2021). Nevertheless, quantitative estimates of N_2O and CH_4 emissions remain highly uncertain, in particular due to the presence of poorly sampled or unresolved spatio-temporal variability.

We find that observation-based estimates of the N_2O emissions and the diffusive flux of CH_4 vary by about 20%–30% in the narrow coastal ocean and by about a factor 2 to 3.5 in the wide coastal ocean. The increase in the spread amongst these observational products (which use the same data sets and are therefore not independent) reflects the low number of oceanic N_2O and CH_4 measurements to date, in particular in many coastal regions, as compared to CO_2 . Specifically, the observation density decreases by about a factor 3 from narrow to wide (number of observations per million km² three times lower in the wide coastal ocean in more than 30 of the 45 regions used for the interpolation, see Table S1 in Supporting Information S1). Furthermore, significant differences between the observation-based estimates (MARCATS- N_2O , MARCATS- CH_4 on the one hand, and Yang- N_2O and Weber- CH_4 on the other hand) can result from (a) applying different approaches for estimating the air-sea gas exchange in combination with using different wind speed products (e.g., Garbe et al., 2014) and (b) applying different inter- and extrapolation techniques which can introduce significant uncertainties when applied to sparse data. The increase in discrepancy from narrow to wide coastal waters suggests that MARCATS- N_2O and MARCATS- CH_4 may extrapolate local observations over spatial domains where they are not representative anymore. In contrast, the neural networks of Yang- N_2O and Weber- CH_4 , albeit also relying on the same MEMENTO data set, may better capture spatial patterns, such as the overall decrease in CH_4 emissions as the shelf water depth increases.

Current observational products only provide a climatological view of N_2O and an annual mean view of CH_4 emissions, and the global models used here are limited by their relatively coarse horizontal resolution. Both observation-based products and models have therefore limited or missing information on (a) seasonal and

inter-annual variability, (b) fine-scale (i.e., few 10s of km or less) land-ocean gradients, (c) the effects of mesoscale and submesoscale features such as eddies (Grundle et al., 2017), and (d) extreme events such as storms and marine heat waves (Borges et al., 2019; Gindorf et al., 2022). Our study reveals, for instance, that while coastal N_2O flux emissions from observational products and models generally agree in terms of main patterns and magnitude, emission hotspots in productive low- O_2 upwelling systems appear to be underestimated by models. In contrast, there are regions where models point to coastal N_2O flux hotspots along mid-latitude western boundaries that are not evident in observational reconstructions. In addition, the observation-based CH_4 product or the N_2O models presented here do not capture features evidenced by field data in prior work, such as the persistent CH_4 under-saturation observed in the Ross Sea (Ye et al., 2023) and Weddell Sea (Heeschen et al., 2004) or the N_2O under-saturation observed in the Arctic Ocean (Wu et al., 2017; Zhang et al., 2015). The reasons for these mismatches remain unclear but likely reflect the presence of unresolved (e.g., complex microbial production/consumption, sedimentary processes, production in estuarine and coastal vegetation systems transported to the coastal ocean) or spatially under-resolved processes (e.g., high production and remineralization in shallow shelves, and shallow coastal oxygen minimum zones where N_2O emissions take place) in ocean biogeochemical models. In addition, the lack of observations in these regions (see Table S1 in Supporting Information S1) could limit the ability of reconstructions to reflect mean coastal conditions or capture undersaturation and emission hotspots. The recently proposed Global N_2O Ocean Observation Network (N_2O -ON) (Bange, 2022; Bange et al., 2019) might help to better constrain and understand temporal and spatial variability as well as reduce uncertainties in current global N_2O oceanic emission estimates.

Aspects of air-sea gas exchange that remain poorly understood are the effects of surface micro-layers on these gases (Kock et al., 2012). In parallel, commonly adopted model parameterizations greatly simplify complex source and sink processes that are the focus of ongoing research. For example, there remain significant uncertainty in the relative importance of the various (micro)biological and photochemical processes driving the production and consumption of N_2O and CH_4 in coastal waters and sediments, and their potential responses to changing oceanic conditions (Bange, 2022). The relationship between CH_4 concentration and floor depth shown by Weber et al. (2019) suggests that sediments are the major source of CH_4 to the water column over continental shelves, but the contribution of other biological and photochemical CH_4 sources remains poorly understood.

Methane can be produced aerobically in situ in surface waters, providing the most direct route to the atmosphere. Pathways that have been identified in the marine environment include (a) methanogenesis by phytoplankton during primary production (e.g., Bižić et al., 2020), which is potentially driven by reactive oxygen species (Ernst et al., 2022); (b) egestion by zooplankton (Schmale et al., 2018); and (c) decomposition of dissolved organic matter compounds including methylphosphonate (MPn, Karl et al., 2008; Repeta et al., 2016; von Arx et al., 2023) and dimethylsulfoniopropionate (DMSP, Damm et al., 2010). These processes have mostly been studied in the open ocean, where they explain the ubiquitous weak supersaturation of CH_4 in remote surface waters, referred to as the “Marine Methane Paradox” (Reeburgh, 2007). Recent evidence suggests that the MPn pathway is also active in some coastal waters (Mao et al., 2022), but its importance relative to benthic-sourced CH_4 in maintaining the strong supersaturation of coastal waters seems to be low (Kanwischer et al., 2023). In the North Sea, seasonal fluctuations in coastal CH_4 that align with temperature (a driver of benthic production and degassing) rather than chlorophyll or DMSP suggest that sedimentary sources of CH_4 overwhelm the biological production in the water column (Borges et al., 2018). Additional sources of N_2O and CH_4 remain poorly characterized and are not represented by models, including submarine groundwater discharge (Arévalo-Martínez et al., 2023) and production associated with marine microplastic (Royer et al., 2018; Su et al., 2022) and submerged aquatic vegetation (Hilt et al., 2022; Rosentreter et al., 2021, 2023; Roth et al., 2023).

Ongoing environmental changes such as ocean warming, decreasing pH, loss of dissolved oxygen, and eutrophication might significantly alter the production and consumption of both N_2O and CH_4 as well as their distribution patterns in coastal waters and, consequently, their release to the atmosphere (e.g., Rees et al., 2022; Zhou et al., 2023). However, our knowledge of recent trends on which future emission scenarios of N_2O and CH_4 from the coastal ocean rely is still far from complete. In particular, hydrate dissolution due to ocean warming may enhance this flux at the seafloor, but only at the feather-edge of the hydrate stability zone, which occurs in ~ 400 m deep water in mid-latitudes—which could be too deep for the methane to make it to the surface and escape to the atmosphere (Joung et al., 2022). Shallow hydrocarbon-fed seep fields allow for more efficient methane release to the atmosphere (Hovland et al., 1993), but their impact appears to be highly localized (Joung et al., 2020), and the global-scale contribution of geological CH_4 to marine emissions remains highly uncertain

(Etiopie et al., 2019). Understanding CH₄ oxidation dynamics in coastal environments is therefore an important focus area for future research. Although N₂O-ON was originally designed for N₂O only, adding measurements of CH₄ will be facilitated by deploying instruments on the basis of the same technique used for N₂O measurements (i.e., cavity-enhanced absorption spectroscopy), providing new opportunities to establish long-term time-series for these two greenhouse gases.

4.3. Coastal Ocean Greenhouse Gas Atmospheric Influence

This synthesis provides an estimate of the coastal contribution to the atmospheric radiative balance using an ensemble of observation-based products and global ocean biogeochemical models (in CO₂-equivalent). In both products and models, we find that a significant proportion of the coastal CO₂ uptake (~30%–60%) is offset by the radiative contribution of N₂O and CH₄ emissions, despite large uncertainties in the magnitude of the mean CO₂ uptake (large uptake in models) and relatively limited numbers of observation-based products and models available for N₂O and CH₄ fluxes. This offset is significantly larger than in the global ocean, for which a value of about 10% can be calculated based on the CO₂ (Le Quéré et al., 2018), N₂O (Tian et al., 2020), and CH₄ (Saunois et al., 2020) global budgets by the GCP. A smaller offset value of the order of 10%–20% has also been reported for estuaries and coastal vegetated ecosystems (Rosentreter et al., 2023), highlighting that the radiative balance on the shelf results from a significant contribution of the 3 greenhouse gases. Such an offset does not occur in inland waters either (rivers, lakes and reservoirs), as freshwater aquatic systems are a net source of CO₂, CH₄ and N₂O (Battin et al., 2023; Lauerwald et al., 2023), with CO₂ and CH₄ contributing roughly 75% and 25% to radiative balance, respectively, while N₂O is only a marginal contributor.

Integrating the three compartments of the land-to-ocean aquatic continuum (LOAC) from streams to the coastal oceans (i.e., inland waters, estuaries and coastal vegetation, and coastal ocean waters (Regnier et al., 2013; Regnier et al., 2022)), we find that the LOAC has a net-positive contribution to the radiative balance. Indeed, the 8.3 (range of 5.8–12.7) PgCO₂-e year⁻¹ emitted by inland waters (Lauerwald et al., 2023) are only partly compensated by the net uptakes of 0.4 (range 0.2–0.7) PgCO₂-e year⁻¹ from estuaries and coastal vegetation (Rosentreter et al., 2023) and 1.3 (range 0.7–1.8) PgCO₂-e year⁻¹ from wide coastal waters. For the 100 year time horizon, the LOAC as a whole thus contributes about 6.6 PgCO₂-e year⁻¹ globally to the radiative balance. These studies focus on the LOAC contemporary fluxes of greenhouse gases and their contribution to the radiative balance. An assessment of the historical changes in these fluxes compared to the pre-industrial baseline has been attempted in particular for CO₂ (e.g., Bourgeois et al., 2016; Lacroix et al., 2021; Mackenzie et al., 1998). Process-based model assessments and dedicated observation-based coastal products with a global coverage that resolve recent trends are, however, still missing informing on the LOAC contribution to the anthropogenic radiative forcing.

5. Conclusion

The main findings of the RECCAP2 global coastal ocean synthesis are:

- *Net global CO₂ uptake*—wide coastal ocean CO₂ uptake for the 1998–2018 period is estimated to –0.44 PgC year⁻¹ from pCO₂-products (median of 4 products) and –0.72 PgC year⁻¹ from global ocean biogeochemical models (median of 11 models). This 60% mismatch is attributed to the stronger wintertime CO₂ uptake in northern mid- and high-latitude systems in models due to a weak biologically induced seasonality compared to thermal changes in models. Biologically induced biases can be amplified by the choice of the wind product and gas exchange coefficient parameterization.
- *Trends in CO₂ uptake*—coastal ocean CO₂ uptake has increased during the 1998–2018 period but this increase is poorly constrained, with trends varying by a factor two across pCO₂-products and by a factor three across models (trends between –0.15 and –0.04 mol m⁻² year⁻¹ per decade) due to differences in ocean pCO₂, wind and sea-ice coverage trends.
- *N₂O and CH₄ emissions*—wide coastal ocean N₂O emissions are estimated to +1.63 Tg N year⁻¹ in the Yang et al. (2020) observation-based product, and to +1.27 Tg N year⁻¹ from global ocean biogeochemical models (median of 5 models). Emissions of CH₄ are estimated to +7.85 [2.50–9.20] TgCH₄ year⁻¹ from the Weber et al. (2019) observation-based product. N₂O and CH₄ flux estimates are fewer and more uncertain than CO₂ fluxes.
- *Net coastal radiative balance*—despite uncertainties, N₂O and CH₄ emissions strongly offset the CO₂ uptake in the net radiative balance of the coastal ocean (offset of 30%–60%), emphasizing the need to consider N₂O and CH₄ when examining the influence of coastal ocean on climate.

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Based on the main challenges identified in this synthesis, we suggest a focus on the following points:

- Expand the number of observation-based products and models specifically designed for coastal oceans (higher spatial resolution, account for land imprint, etc.).
- Develop pCO₂-based products that resolve interannual and decadal changes (only two out of four pCO₂-products provided time-varying information, while others were climatologies).
- Constrain changes in wind and sea-ice, which have strong implications for flux trend estimation.
- Better constrain the biological pump in the mid- and high-latitude coastal oceans in ocean biogeochemical models.
- Refine coastal N₂O and CH₄ emission estimates from observations (very few estimates available) and models (few N₂O estimates, no CH₄ estimates) to gain an understanding of their dynamics and close the gap with CO₂ estimates.

Conflict of Interest

The authors declare no conflicts of interest relevant to this study.

Data Availability Statement

The data used in this chapter are available in Resplandy et al. (2023) <https://zenodo.org/record/8326096> doi: 10.5281/ZENODO.8326096 and Müller (2023) <https://zenodo.org/record/7990823> doi: 10.5281/ZENODO.7990823.

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