Eddy impacts on the marine biogeochemistry of the California Current System

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Abstract

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Key Points:

• In the California Current, subduction by submesoscale eddies near the coast and mesoscale eddies offshore reduces surface nutrients.
• In the presence of submesoscale eddies, the non-linear nature of nutrient uptake decreases primary production by up to ∼ 50%.
• The amplitude and sign of eddy nutrient uptake is controlled by the covariance of temperature, nutrient and phytoplankton fluctuations.

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Abstract

Eddies play a crucial role in shaping ocean dynamics by affecting material transport, and generating spatio-temporal heterogeneity. However, how eddies at different scales modulate biogeochemical transformation rates remains an open question. Applying a multi-scale decomposition to a numerical simulation, we investigate the respective impact of mesoscale and submesoscale eddies on nutrient transport and biogeochemical cycling in the California Current System. First, the non-linear nature of biological nutrient uptake results in a 50% reduction in primary production in the presence of eddies. Second, eddies shape the vertical transport of nutrients with a strong compensation between mesoscale and submesoscale. Third, the eddy effect on nutrient uptake is controlled by the covariance of temperature, nutrient and phytoplankton fluctuations caused by eddies. Our results shed new light on the tight interaction between non-linear fluid dynamics and ecosystem processes in realistic eddy regimes, highlighting the importance of both mesoscale and submesoscale variability.

1 Introduction

Mesoscale and submesoscale eddies are ubiquitous in the ocean, and play a central role in its dynamics. Eddies directly influence transport of momentum and material properties, and generate spatial and temporal heterogeneity in biogeochemical tracers and transformation rates (McGillicuddy, 2016; Mahadevan, 2016; Lévy et al., 2018). In contrast with the mean oceanic circulation, eddy dynamics is generally described as that occurring on time scales shorter than a few months, and spatial scales of a hundred of kilometers or less. Transport of material properties at these scales (i.e., eddy-induced fluxes) arises from the covariance of tracer and momentum fluctuations around their large-scale means (Levy & Martin, 2013). Because of the prevalence of eddies in the oceanic kinetic energy spectrum (Chelton et al., 2007), eddy fluxes often represent major contributions to momentum and material exchanges, sometimes rivaling transport by the mean circulation (McGillicuddy et al., 2003; Lévy et al., 2012).

Circulation at eddy scales affects biogeochemistry in multiple ways. In the simplest way, eddy-induced physical-biogeochemical interactions occur via two main processes: eddy transport and eddy reaction rates (Levy & Martin, 2013). These are are similar in essence, but reflect different underlying mechanisms (Goodman & Robinson, 2008). Eddy transport arises from eddy-scale correlations between fluctuations in currents and tracer concentrations. This is an advective stirring process with both vertical (Falkowski et al., 1991; Oschlies & Garcon, 1998; Benitez-Nelson et al., 2007; F. Kessouri et al., 2020) and horizontal (Lathuilière et al., 2010; Gruber et al., 2011; Gaube et al., 2014) contributions. The effects of eddy transport depend on the circulation regime and large-scale biogeochemical gradients, and remain an active field of investigation (Lévy et al., 2018).

Eddy reaction rates consist of a “rectification” of large-scale, low-frequency biogeochemical transformation rates that arises from the non-linear nature of biogeochemical interactions (which include primary production, zooplankton grazing, remineralization) in a turbulent, heterogeneous environment. As a result, biogeochemical transformation rates estimated from a “mean field approximation”, i.e., estimated from properties averaged over scales greater than those of eddies, often fail to represent the biogeochemical dynamics of a turbulent ocean (Rovinsky et al., 1997; Brentnall et al., 2003). In analogy to eddy transport fluxes, a Reynolds decomposition can be applied to biogeochemical rates to separate mean from eddy contributions. This approach relies on appropriate spatial or temporal filters to separate the effects of the mean tracer distribution from fluctuations induced by eddies (Goodman & Robinson, 2008; Wallhead et al., 2008; Goodman, 2011).

Beyond theoretical and idealized studies (Brentnall et al., 2003; Goodman & Robinson, 2008; Wallhead et al., 2008), Levy and Martin (2013) showed that eddy contribu-
tions accounted for between 5 and 30% of primary production and grazing rates in an idealized, eddy-resolving simulation of the North Atlantic Ocean. Eddy effects were mostly attributed to mesoscale variability (with typical length scales of between 30 and 100 km). A somewhat weaker eddy contribution was confirmed by analysis of in situ and satellite observations in the same region (Martin et al., 2015), suggesting that, while non-negligible, eddy reaction rates may have only a minor impact on open-ocean biogeochemistry. However, these estimates focused mostly on mesoscale eddies, while submesoscales remained under-resolved and under-sampled. Thus, it is possible that, in region with vigorous sub-mesoscale activity — such as intense frontal regions and upwelling systems, eddy reactions may be more important than previously appreciated.

The California Current System (CCS) is ideally suited for studies of eddy-driven physical-biogeochemical interactions. In this coastal environment, wind-driven upwelling of nutrient-rich waters fuels intense biological productivity (Carr & Kearns, 2003; Messié et al., 2009) and generates a highly energetic field of mesoscale and submesoscale eddies (Marchesiello et al., 2003; Capet et al., 2008). Baroclinic instabilities of the alongshore current (Marchesiello et al., 2003) result in a cross-shore transport of nutrients and organic material followed by subduction along the CCS fronts. This so-called “eddy quenching” process (Gruber et al., 2011) reduces productivity in the coastal band, and supplies nutrients to remote open-ocean regions (Lovecchio et al., 2018; Yamamoto et al., 2018; Frenger et al., 2018). At the submesoscale, eddies enhance both nutrient removal in the coastal region, and nutrient entrainment and re-supply to the euphotic zone offshore (F. Kessouri et al., 2020). However, the contribution of submesoscale and mesoscale eddy transports in this upwelling system remains poorly characterized, and the impact of eddies on biogeochemical reactions rates has not been quantified yet.

Here, we evaluate the role of mesoscale and submesoscale eddies on nutrient transport and uptake rates by applying a multi-scale Reynolds decomposition to output from a submesoscale-permitting model of the CCS (F. Kessouri et al., 2020; Damien et al., 2023). Our analysis provides new insights on the different routes of nutrient supply and removal in the euphotic layer, and on the scale-dependent interplay between non-linear fluid and ecosystem dynamics in a highly heterogeneous environment.

2 Methods

2.1 Physical-biogeochemical model

We use the Regional Ocean Modeling System (ROMS, (Shchepetkin & McWilliams, 2005)) coupled online to the Biogeochemical Elemental Cycling model (BEC, (Moore et al., 2004; Deutsch et al., 2021)). ROMS solves the hydrostatic primitive equations for the tree-dimensional velocity, temperature, salinity and the transport of tracers in a terrain-following coordinate system. BEC represents the biogeochemical cycles of major elements (C, N, P, O, Fe, Si) resulting from the interaction of three phytoplankton and one zooplankton group.

We analyze output from two twin simulations for the northern and southern U.S. West Coast at 1 km resolution (Damien et al., 2023), sufficient to allow emergence of submesoscale dynamics (F. Kessouri et al., 2020), obtained by dynamical downscaling of a coastwide configuration at 4 km resolution (Renault et al., 2021; Deutsch et al., 2021). Because these simulations do not include tidal forcings, the highest frequencies captured by the model only include submesoscale circulation and internal waves generated within the domain. Output consists of physical and biogeochemical variables, transport fluxes, and biogeochemical rates calculated online by the model, and is saved as daily averages.

In the model, an arbitrary biogeochemical tracer $X_i$ obeys the conservation equation:

$$
\partial_t X_i = T(X_i) + \partial_z (\kappa \partial_z X_i) + J_i(X_{j=1,...,J}).
$$

(1)
The first term on the right hand side, $T(X_i) = -\nabla \cdot (u \, X_i)$, represents the divergence of the advective flux, with $u = (u, v, w)$ the velocity vector. It can be further decomposed into a horizontal $T_h(X_i)$ and vertical $T_v(X_i)$ component. The second term represents vertical mixing, with $\kappa$ the vertical eddy diffusivity. The third term, $J_i$, is the sum of all biogeochemical rates that affect the tracer $X_i$, which in turn depend on $J$ model state variables $X_j$.

We focus on the balance of nitrate ($\text{NO}_3^-$), the main limiting nutrient in the CCS (Deutsch et al., 2021). For this variable, the net biogeochemical reaction rate is:

$$J = J_{\text{Uptk}} + J_{\text{Nit}} + J_{\text{Denit}}$$  \hspace{1cm} (2)

Here, $J_{\text{Uptk}}$ is the rate of uptake by phytoplankton, $J_{\text{Nit}}$ production by nitrification, and $J_{\text{Denit}}$ consumption by denitrification. Note that here, $J_{\text{Uptk}}$ is a negative rate because it removes nutrient from sea water. Therefore, it is equivalent to net primary production, but with an opposite sign, and expressed in nitrogen units. In the CCS, denitrification only occurs in the deeper parts of anoxic basins and in the sediment, and is a minor term compared to nitrification and biological uptake. Hence, when discussing water column processes, we focus primarily on nitrification and uptake. The nitrification rate, $J_{\text{Nit}} = \tau_{\text{Nit}} \, \text{NO}_2^-$, is modeled as a linear function of nitrite ($\text{NO}_2^-$) concentration, with $\tau_{\text{Nit}}$ a constant timescale. Non-linearities in nitrification arise from limitation under high irradiance in the euphotic zone, and inhibition at vanishing oxygen and nitrite concentrations (Deutsch et al., 2021).

Biological uptake depends on nutrient concentrations following a Michaelis-Menten kinetics and Liebig’s law of the minimum, phytoplankton biomass, and a temperature- and light-dependent growth rate (see Supporting Information T2, and Deutsch et al. (2021)). Thus, uptake is highly non-linear because of the presence of bilinear ($X_iX_j$), exponential ($e^{X_i}$), and hyperbolic (Michaelis-Menten) terms.

### 2.2 Triple decomposition of transport and biogeochemical rates

The non-linear nature of advection, nitrification, and uptake in the nitrate conservation equation (Equation 1) is at the root of eddy rectification effects that modulate the final rate of change of this tracer.

To separate the effects of mesoscale and submesoscale eddies, we apply a triple Reynolds decomposition based on two low-pass filters, $\tilde{\lambda}$ and $\tilde{\tau}$, with respective space/time scales ($\lambda, \tau$) and ($\tilde{\lambda}, \tilde{\tau}$) (Capet et al., 2008). Accordingly, a model variable $X_i$ is decomposed into mean and fluctuating mesoscale and submesoscale components as:

$$X_i = \overline{X_i} + X_i' + X_i''$$  \hspace{1cm} (3)

where

$$X_i' = \tilde{X_i} - \overline{X_i} \quad \text{and} \quad X_i'' = X_i - \tilde{X_i}.$$  \hspace{1cm} (4)

By definition, $\overline{X_i'} = 0$ and $\tilde{X_i''} = 0$. Here, ($\overline{X_i}, \overline{\tau}$) and ($\tilde{X_i}, \tilde{\tau}$), are chosen to separate mesoscale ($X_i'$) and submesoscale ($X_i''$) fluctuations from a large-scale, low-frequency mean ($\overline{X_i}$) that includes the seasonal cycle. The $X_i''$ component represents the smallest scales and highest frequencies allowed by the model, i.e., mostly submesoscales. The choice of the filter scales is dependent on the circulation regime, and may not always perfectly separate intrinsic variability from forced motions. For example, along the U.S. West Coast, wind-driven upwelling is generally considered part of the mean seasonal variability. However, short-term wind events can generate high frequency variability in circulation that overlaps with mesoscale and submesoscale motions. In our choice of filters, we were especially careful to attribute the main upwelling signal to large-scale regional variability (i.e., the mean term $\overline{X_i}$) and not higher frequency fluctuations. To this end, we found a reasonable combination of temporal and spatial filter scales, defined as follows:
\[ (\lambda, \tau) = (15 \text{ km}, 3 \text{ months}) \], with a centered averaging scheme,
\[ (\lambda, \tau) : (5 \text{ km}, 3 \text{ days}) \], with a centered averaging scheme.

We refer the readers to Supporting Information T1 that further discusses these filters and their performance, using surface temperature and vertical velocities as an illustration.

By applying these filters to model variables, biogeochemical transformation rates can be separated into mean and eddy components. For a nonlinear reaction rate \( J_i(X_j) \) with dependence on multiple variables \( X_j, j = 1, ..., J \) and the transport divergence \( T(X_i) \), the analogous Reynolds decomposition takes the form:

\[
J = J^{\text{mean}} + J^{\text{meso}} + J^{\text{subm}} \quad \text{and} \quad T = T^{\text{mean}} + T^{\text{meso}} + T^{\text{subm}}
\]

where the various terms are now calculated as:

\[
J^{\text{mean}} = \bar{J}_i(\bar{X}_j) \quad \text{and} \quad T^{\text{mean}} = \bar{T}(\bar{X}_i)
\]

\[
J^{\text{meso}} = J_i(\bar{X}_j) - \bar{J}_i(\bar{X}_j) \quad \text{and} \quad T^{\text{meso}} = T(X_i) - T(\bar{X}_i)
\]

\[
J^{\text{subm}} = J_i(X_j) - J_i(\bar{X}_j) \quad \text{and} \quad T^{\text{subm}} = T(X_i) - T(\bar{X}_i)
\]

By adopting the filtering approach discussed above, the three terms in Equation 5 can be respectively interpreted as the contribution to the total rate caused by the large-scale mean tracer distributions \((J^{\text{mean}} \text{ and } T^{\text{mean}})\); the contribution caused by heterogeneity at the scale of mesoscale eddies \((J^{\text{meso}} \text{ and } T^{\text{meso}})\); and the contribution caused by heterogeneity at submesoscale and smaller scales captured by the model \((J^{\text{subm}} \text{ and } T^{\text{subm}})\).

Specifically, the biogeochemical eddy contributions only exist as a rectification of biogeochemical rates that depend in non-linear ways on model variables. These contributions would vanish in the case of perfectly linear rates (Levy & Martin, 2013).

### 2.3 Amplitude and sign of the eddy rectification

Assuming high frequency fluctuations of small amplitudes relative to the mean, the low frequency and large scale advective transport \((T)\) and biogeochemical rates \((J)\) can be approximated by a Taylor series expansion (Levy & Martin, 2013):

\[
J(X_i) = \bar{J}(\bar{X}_i) + \sum_i \frac{\partial J}{\partial X_i} \bar{X}_i + \frac{1}{2} \sum_{i,j} \frac{\partial^2 J}{\partial X_i \partial X_j} \bar{X}_i \bar{X}_j + O(\bar{X}_i \bar{X}_j)^2
\]

An equivalent Taylor expansion can be written for the advection term \(T\), leading to a typical definition of eddy transport fluxes (Capet et al., 2008). Since the fluctuations have zero mean, the linear terms disappear. Ignoring the contribution of third-order terms, the amplitude and sign of the eddy rectified effect depend on the curvature of the functional dependencies (encapsulated by \(J\)) and eddy correlation terms between model variables \((\bar{X}_i \bar{X}_j)\):

\[
\frac{\partial^2 J}{\partial X_i \partial X_j} \bar{X}_i \bar{X}_j \quad (10)
\]

### 3 Results

We find that, along the CCS coast, the balance of nitrate in the surface layer (Equation 1) reflects a near compensation of two major terms: biological uptake \((U_{\text{ptk}})\), and the divergence of the vertical transport \((v)\) (Figure 1). The mean component of \(J^{\text{ptk}}\) increases towards the coast (Figure 1a), reflecting high nutrient concentrations following inputs by upwelling (Figure 1d). Both mesoscale and submesoscale contributions to \(J^{\text{ptk}}\) are opposite in sign and partially offset the mean component. The magnitude of
the submesoscale contribution is particularly large, reaching about -34% of the mean $J^{uptk}$, while the mesoscale contribution is more limited.

Supply of NO$_3^-$ by vertical transport (i.e., the divergence of the vertical flux; Figure 1d-f) shows noisier patterns, reflecting the high variability and large magnitude of advective fluxes. However, notable patterns emerge. The most significant is the positive mean $T_h$ (i.e., NO$_3^-$ supply) along the coastal band caused by upwelling. The submesoscale $T_v$ largely opposes the mean upwelling along the coast, reducing NO$_3^-$ supply by 50 to 70%. In contrast, mesoscale $T_h$ is weaker, and is characterized by upwelling close to the coast, and downwelling offshore, thus reinforcing the mean vertical transport.

Based on these patterns, we distinguish between the coastal region, where nutrients are upwelled into the euphotic layers, and the offshore region, where subduction by mesoscale eddies dominates (Fig. 1). This separation occurs at a distance of approximately 40 km from the coast, comparable with the width of the continental shelf (Damien et al., 2023). Over the coastal region, the main balance in the NO$_3^-$ budget is between NO$_3^-$ supplied to the euphotic layer by vertical advection and uptake by phytoplankton (Fig. 2). Offshore, horizontal transport ($T_h$) replaces vertical advection as the main source of NO$_3^-$. Vertical mixing is also significant offshore, accounting for 33% of the NO$_3^-$ inputs.

In the NO$_3^-$ balance, eddy reaction rates generally oppose mean reaction rates: eddy $J^{uptk}$ is positive and eddy $J^{Nit}$ is negative. The magnitude of the eddy $J^{uptk}$ is particularly large, accounting for ~45% of the mean rate in both the coastal and offshore regions. This eddy contribution is largely dominated by submesoscale.

Near the coast, mean upwelling $T_v$ is the largest source of NO$_3^-$, and is largely offset (-64%) by submesoscale subduction. The total horizontal advection is negligible, reflecting a balance between the mean $T_h$, which supply NO$_3^-$, and eddy $T_h$, which remove it. The picture is different offshore. Both mean and eddy currents supply NO$_3^-$ at similar rates. NO$_3^-$ delivery by the mean transport is equivalent to that in the coastal region, accounting for 64% of the horizontal NO$_3^-$ supply, while mesoscale and submesoscale components account for 26% and 10% respectively. Along the vertical direction, we observe a close balance between subduction at mesoscales and supply at submesoscales. The magnitude of NO$_3^-$ supply by vertical mixing is similar in both regions (~1.0 $10^{-5}$ mmol m$^{-2}$ s$^{-1}$), but its relative contribution is more significant offshore (33% of the total NO$_3^-$ supply). This mixing term is largely driven by deepening of the mixed layer in winter (not shown).

The mean $J^{uptk}$ and its submesoscale rectification show a large seasonal cycle, with a maximum during upwelling in summer (Fig. 3a,b). $J^{Nit}$ follows a similar seasonal cycle, with a maximum following the peak in biological uptake, and is dominated by the mean component (Fig. 3c,d). In contrast, mesoscale eddy reactions show weak seasonality, and large fluctuations on time scales of weeks, especially in summer.

Over the course of the year, the vertical transport near the coast is shaped by mean upwelling (Fig. 3e), and balanced by submesoscale subduction. While mesoscale fluctuations cancel out when integrated over the annual cycle (Fig. 2), they drive the total transport at weekly timescales. Offshore, seasonal variability is less pronounced, and the period of maximum transport follows the upwelling season. Subduction by mesoscale eddies is larger from June to November, when re-supply by submesoscale eddies also increases.

The mean horizontal transport remains small relative to the mesoscale component, which dominates on weekly timescales. Over the year, the horizontal NO$_3^-$ flux from the coast to the open-ocean (Fig. 3i) is largely positive (~8.3 $10^4$ molN s$^{-1}$). This redistribution of nutrients occurs at all scales, with a major contribution from the mean circulation (56%), reflecting wind-driven Ekman transport, followed by mesoscale eddies.
(37%). The strong correlation between $T_{h}$ offshore (Fig. 3 h) and the cross-shore flux (Fig. 3 i) indicates that $\text{NO}_3^-$ variations in the open-ocean section of the CCS are mostly caused by transport from the region of active upwelling near the coast (Damien et al., 2023).

### 4 Discussion and conclusions

#### 4.1 Eddy transport

In the CCS, similar to other EBUS, nutrient subduction by eddies, or “quenching”, plays a major role in modulating primary production (Gruber et al., 2011; Nagai et al., 2015; Renault et al., 2016). Here, we show that eddy quenching reflects two contrasting regimes: subduction of freshly-upwelled nutrients by submesoscale eddies nearshore, and by mesoscale eddies further offshore (Figs. 1 and 2). Mesoscale eddies thus transport nutrient from the coast to the open-ocean, but also tend to “bury” them below the euphotic zone (Gruber et al., 2011). Offshore, we observe a near compensation between subduction at mesoscale and delivery at submesoscale (Fig. 2). This balance is particularly evident between July and October, following the large coastal nutrient injection caused by upwelling (Fig. 3). As recently-upwelled nutrients travel offshore and progressively sink along isopycnals, submesoscale eddies tend to resupply them back to the euphotic layer (F. Kessouri et al., 2020).

Globally, submesoscale eddies have been shown to enhance both nutrient delivery to the surface, in particular in the open ocean (Lévy et al., 2001; Mahadevan, 2016), and nutrient and organic matter subduction in regions of strong frontal activity (Omand et al., 2015; Hæck et al., 2023) and upwelling systems (Stukel et al., 2017; F. Kessouri et al., 2020). Here we show that both effects coexist along a gradient of surface nutrient concentrations in the CCS. Specifically, the direction of submesoscale nutrient transport depends on the balance between biological uptake and typical nutrient supply from below the euphotic layer. Relatively long nutrient residence times in surface layers associated with large nutrient concentrations and weak vertical gradients (as observed in nutrient-rich systems) favor nutrient removal by submesoscale currents. In contrast, short surface nutrient residence times associated with low nutrient concentrations and sharp nutrient gradients (typical of oligotrophic systems) favor submesoscale nutrient supply. This idea is supported by idealized (Freilich et al., 2022) and realistic (F. Kessouri et al., 2020) modelling studies.

#### 4.2 Eddy Reactions

In the California Current, eddies reduce the mean nutrient uptake, and thus net primary production, by about 50%. Most of this compensation (35%) occurs at submesoscale. This eddy rectification is significantly larger than suggested by previous studies, which mostly focused on open-ocean regions and mesoscale circulations (Levy & Martin, 2013; Martin et al., 2015). Our study is the first to directly assess the magnitude of eddy reaction rates using a submesoscale-permitting model and a scale-dependent separation of mesoscale and submesoscale (Capet et al., 2008) in a region with particularly vigorous eddies. At coarser resolution, eddy kinetic energy is likely damped (Capet et al., 2008), thus leading to an underestimate of eddy heterogeneity and its contribution to biogeochemical rates.

Mesoscales and submesoscales are highly advective regimes that favor the emergence of heterogeneity and variability in tracer fields, which cause an eddy rectification of the mean biogeochemical rates. Integrated over large scales and low frequencies, eddy contributions are consistently reducing the mean uptake (Fig. 1 and 2). The magnitude and sign of this eddy rectification result from the eddy covariance of model state variables and the functional dependencies that describe biogeochemical transformations (Equa-
tion 10, see also Levy and Martin (2013)). Because biogeochemical rates depend on several tracers in complex ways (see Supporting Information T2), eddy reaction rates generally involve contributions from the interaction of multiple tracer pairs.

Analysis of the mesoscale contributions to NO$_3^-$ uptake (fig. 4) shows that the dominant terms arise from the saturating response of uptake at high nutrient concentrations (fig. 4 b-d), and the negative correlation between NO$_3^-$ and phytoplankton (Fig. 4 i,j). Specifically, the negative curvature of the Michaelis-Menten saturation function implies that, in a heterogeneous environment, high-frequency events characterized by large NO$_3^-$ concentrations are not as important in boosting uptake, relative to low-NO$_3^-$ events that are instead more effective at reducing it.

Furthermore, assuming a small Damkohler number (i.e., the ratio of the reaction rate to the high-frequency transport rate), Equation 10 can be re-stated by invoking mean tracer gradients and high frequency fluctuations, here assumed to occur mostly along the vertical direction $z$:

$$J_{\text{eddy}} \approx \frac{1}{2} \sum_{i,j} \frac{\partial^2 J}{\partial X_i \partial X_j \left|_{X_i, X_j, \ldots} \frac{\partial X_i}{\partial z} \frac{\partial X_j}{\partial z} \delta z'^2 \right.}$$

with $\delta z'$ a small vertical fluctuation. Because vertical profiles of nutrients and phytoplankton show large and opposite gradients, in particular near the base of the euphotic zone, vertical fluctuations enhance the negative covariance between phytoplankton and NO$_3^-$ (Fig. 4 i), producing a sub-surface maximum in the eddy uptake rectification terms.

In contrast, the smaller amplitude of $J^{Uptk}$ rectification at the mesoscale likely reflects a larger influence of horizontal rather than vertical fluctuations, where negative correlations between nutrients and phytoplankton are more ambiguous. Furthermore, this argument is based on a small Damkohler number approximation. Considering a time scale of the order of 1.0 d$^{-1}$ for nutrient uptake (see Supporting Information T2), this approximation is more appropriate for submesoscale rather than mesoscale fluctuations.

When integrated over a full seasonal cycle, we obtain ratios between eddy and mean uptake rates that are remarkably constant (\sim -0.35 for submesoscale and \sim -0.10 for mesoscale) across the CCS. To what extent these ratios can be generalized to different regions and circulation regimes remains an open question.

4.3 Implications

We found a remarkable compensation between mean and submesoscale terms in the net balance of NO$_3^-$ over a seasonal cycle in the California Current System (Fig. 2). This suggests that, in the productive coastal region, nitrate supply occurs predominately at large scales and low frequencies, while removal occurs at small scales and high frequencies. This balance is reversed offshore. While mesoscale contributions tend to cancel out over the seasonal cycle, they generate large variability, producing extremes in both nutrient transport fluxes and uptake rates (Fig. 3 and 4).

The nutrient heterogeneity caused by eddies does not necessarily promote biological productivity. Indeed, it systematically reduces it when averaged over large scales and low frequencies, thus representing a different kind of productivity “quenching” associated with non-linear ecosystem dynamics. The reasons are twofold. First, phytoplankton uptake quickly saturates at high nutrient concentrations. Second, high nutrient concentrations are often associated with low phytoplankton biomass, which limits the potential for increased productivity. We also note that changes in productivity caused by correlations involving temperature (which modulates uptake rates with an exponential dependence) are negligible in the open ocean, but become more important along the continental margin (Fig. 4 g, m, and p).
More generally, we find that eddy terms are far from negligible compared to mean biogeochemical rates. This result questions the ability of coarser models to adequately represent nutrient fluxes and biogeochemical transformations. For example, a non eddy-resolving global model would likely overestimate the vertical nutrient supply and biological uptake along upwelling systems. Physical parameterizations of eddy transport (Gent & McWilliams, 1990; Fox-Kemper et al., 2008) can partially alleviate this issue. However, analogous parameterizations for eddy biogeochemical rates are in early stages of development (Wallhead et al., 2013) and are not yet applied to biogeochemical models. Historically, biases in ocean circulation have been addressed by tuning biogeochemical parameters, which thus implicitly depend not only on the choice of model equations, but also on the resolution at which models are run and evaluated against observations. Our finding of a constant ratio between eddy and mean nutrient uptake rates across a range of circulations (~ -0.35 for submesoscale and ~ -0.10 for mesoscale), and our analysis of the different contributions of tracer covariance terms to eddy rates, offer new insights for the development of eddy parameterizations of biogeochemical transformations.

Finally, we focused on biological nutrient uptake as the dominant biogeochemical transformation in the highly productive CCS. However, the dynamics of pelagic ecosystems is characterized by many non-linear processes, from food web interactions, to remineralization and microbial dynamics under low oxygen conditions, which remain untouched here. In environments naturally sensitive to multiple stressors, such ocean acidification, warming, and oxygen loss, eddy rectification of ecological processes could greatly alter ecosystem dynamics and marine habitats. Analysis of these processes requires a shift in emphasis from nutrients to carbon and oxygen balances, and from biogeochemical to ecological interactions.

Data Availability Statement

The model code used to generate the simulation is openly available in Kessouri et al. (2020) (https://doi.org/10.5281/zenodo.398861). The simulations are reproducible using the setup and forcing described in Damien et al. (2023).

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Figure 1. Triple scale decomposition (mean, mesoscale and submesoscale) of (a-c) NO$_3^-$ biological uptake ($J_{\text{uptk}}$), and (d-f) NO$_3^-$ vertical transport divergence ($T_v$, equal to the flux at the base of the layer) averaged over a full seasonal cycle and integrated over the euphotic layer ($\sim$ 0-50 m depth). Units are mmol N m$^{-2}$s$^{-1}$. Black lines highlight the mean $J_{\text{uptk}}$ isolines of -0.5, -1, -1.5, -2, and -2.5 in the upper panels and the mean $T_v$ isolines of 1, 2 and 5 in the lower panels. A companion figure showing the other terms of the NO$_3^-$ balance is provided in the Supporting Information, Fig. S5, S6.
Figure 2. Separation into mean, mesoscale and submesoscale components of the NO$_3^-$ balance terms (Equation 1) integrated in time over a seasonal cycle, in depth over the euphotic layer, and in space over two distinct regions of the U.S. West Coast: a coastal region, from Point Conception to Cape Blanco, up to 40km from the coast, and an offshore region up to 250km from the coast. Terms representing negligible component of the fluxes are omitted. Summed up by scales, the NO$_3^-$ balance represents +5.0, +0.8, and -6.8 at respectively mean, mesoscale, and submesoscale in the coastal region, and -4.2, +0.1, and +3.0 offshore. The total adds to 0 when the vertical mixing is included.
**Figure 3.** Daily averaged time-series of the (blue) mean, (green) mesoscale, and (purple) sub-mesoscale terms of the NO$_3^-$ balance integrated over the (left panels) coastal and (right panels) offshore regions. In each panel, the red line shows the total rate (calculated online), which equals to the sum of the 3 components. Units are mmol N m$^{-2}$ s$^{-1}$. The light shaded area shows the +/- standard deviation over the region. This is not included for the transport divergence because it is an order of magnitude larger than the regional average. Note that the y-axis of the transport divergence use a different scale on the left and right panels. Panel (i) shows the time series of the horizontal NO$_3^-$ flux from the coastal to offshore region in mmol N s$^{-1}$.
Figure 4. Cross sections, as a function of the distance from the coast and depth, of (a) the annual mean mesoscale eddy uptake, (b,e,h,k,n) the second derivative terms that modulate the (c) nutrient and (f) temperature eddy variance, (i) nutrient-phytoplankton eddy covariance, (l) nutrient-temperature eddy covariance, and (o) temperature-biomass eddy covariance at mesoscale. Following the Taylor series expansion (Equation 10, also shown at the top), the (a) mesoscale eddy uptake is approximated by the sum of the (d,g,j,m,p) second-order terms. Units of the uptake rate are mmol N m\(^{-3}\) s\(^{-1}\). The thick black contour represents the nutricline, defined by a nitrate concentration of 1 mmol N m\(^{-3}\). A companion figure comparing eddy covariance at mesoscale and submesoscale is provided in the Supporting Information, Fig. S3.
References


Eddy impacts on the marine biogeochemistry of the California Current System

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Key Points:

• In the California Current, subduction by submesoscale eddies near the coast and mesoscale eddies offshore reduces surface nutrients.
• In the presence of submesoscale eddies, the non-linear nature of nutrient uptake decreases primary production by up to ∼ 50%.
• The amplitude and sign of eddy nutrient uptake is controlled by the covariance of temperature, nutrient and phytoplankton fluctuations.

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Abstract
Eddies play a crucial role in shaping ocean dynamics by affecting material transport, and generating spatio-temporal heterogeneity. However, how eddies at different scales modulate biogeochemical transformation rates remains an open question. Applying a multi-scale decomposition to a numerical simulation, we investigate the respective impact of mesoscale and submesoscale eddies on nutrient transport and biogeochemical cycling in the California Current System. First, the non-linear nature of biological nutrient uptake results in a 50% reduction in primary production in the presence of eddies. Second, eddies shape the vertical transport of nutrients with a strong compensation between mesoscale and submesoscale. Third, the eddy effect on nutrient uptake is controlled by the covariance of temperature, nutrient and phytoplankton fluctuations caused by eddies. Our results shed new light on the tight interaction between non-linear fluid dynamics and ecosystem processes in realistic eddy regimes, highlighting the importance of both mesoscale and submesoscale variability.

1 Introduction
Mesoscale and submesoscale eddies are ubiquitous in the ocean, and play a central role in its dynamics. Eddies directly influence transport of momentum and material properties, and generate spatial and temporal heterogeneity in biogeochemical tracers and transformation rates (McGillicuddy, 2016; Mahadevan, 2016; Lévy et al., 2018). In contrast with the mean oceanic circulation, eddy dynamics is generally described as that occurring on time scales shorter than a few months, and spatial scales of a hundred of kilometers or less. Transport of material properties at these scales (i.e., eddy-induced fluxes) arises from the covariance of tracer and momentum fluctuations around their large-scale means (Lévy & Martin, 2013). Because of the prevalence of eddies in the oceanic kinetic energy spectrum (Chelton et al., 2007), eddy fluxes often represent major contributions to momentum and material exchanges, sometimes rivaling transport by the mean circulation (McGillicuddy et al., 2003; Lévy et al., 2012).

Circulation at eddy scales affects biogeochemistry in multiple ways. In the simplest way, eddy-induced physical-biogeochemical interactions occur via two main processes: eddy transport and eddy reaction rates (Lévy & Martin, 2013). These are similar in essence, but reflect different underlying mechanisms (Goodman & Robinson, 2008). Eddy transport arises from eddy-scale correlations between fluctuations in currents and tracer concentrations. This is an advective stirring process with both vertical (Falkowski et al., 1991; Oschlies & Garcon, 1998; Benitez-Nelson et al., 2007; F. Kessouri et al., 2020) and horizontal (Lathuilière et al., 2010; Gruber et al., 2011; Gaube et al., 2014) contributions. The effects of eddy transport depend on the circulation regime and large-scale biogeochemical gradients, and remain an active field of investigation (Lévy et al., 2018).

Eddy reaction rates consist of a “rectification” of large-scale, low-frequency biogeochemical transformation rates that arises from the non-linear nature of biogeochemical interactions (which include primary production, zooplankton grazing, remineralization) in a turbulent, heterogeneous environment. As a result, biogeochemical transformation rates estimated from a “mean field approximation”, i.e., estimated from properties averaged over scales greater than those of eddies, often fail to represent the biogeochemical dynamics of a turbulent ocean (Rovinsky et al., 1997; Brentnall et al., 2003). In analogy to eddy transport fluxes, a Reynolds decomposition can be applied to biogeochemical rates to separate mean from eddy contributions. This approach relies on appropriate spatial or temporal filters to separate the effects of the mean tracer distribution from fluctuations induced by eddies (Goodman & Robinson, 2008; Wallhead et al., 2008; Goodman, 2011).

Beyond theoretical and idealized studies (Brentnall et al., 2003; Goodman & Robinson, 2008; Wallhead et al., 2008), Levy and Martin (2013) showed that eddy contribu-
tions accounted for between 5 and 30% of primary production and grazing rates in an idealized, eddy-resolving simulation of the North Atlantic Ocean. Eddy effects were mostly attributed to mesoscale variability (with typical length scales of between 30 and 100 km). A somewhat weaker eddy contribution was confirmed by analysis of in situ and satellite observations in the same region (Martin et al., 2015), suggesting that, while non-negligible, eddy reaction rates may have only a minor impact on open-ocean biogeochemistry. However, these estimates focused mostly on mesoscale eddies, while submesoscales remained under-resolved and under-sampled. Thus, it is possible that, in region with vigorous submesoscale activity — such as intense frontal regions and upwelling systems, eddy reactions may be more important than previously appreciated.

The California Current System (CCS) is ideally suited for studies of eddy-driven physical-biogeochemical interactions. In this coastal environment, wind-driven upwelling of nutrient-rich waters fuels intense biological productivity (Carr & Kearns, 2003; Messié et al., 2009) and generates a highly energetic field of mesoscale and submesoscale eddies (Marchesiello et al., 2003; Capet et al., 2008). Baroclinic instabilities of the alongshore current (Marchesiello et al., 2003) result in a cross-shore transport of nutrients and organic material followed by subduction along the CCS fronts. This so-called “eddy quenching” process (Gruber et al., 2011) reduces productivity in the coastal band, and supplies nutrients to remote open-ocean regions (Lovecchio et al., 2018; Yamamoto et al., 2018; Frenger et al., 2018). At the submesoscale, eddies enhance both nutrient removal in the coastal region, and nutrient entrainment and re-supply to the euphotic zone offshore (F. Kessouri et al., 2020). However, the contribution of submesoscale and mesoscale eddy transports in this upwelling system remains poorly characterized, and the impact of eddies on biogeochemical reactions rates has not been quantified yet.

Here, we evaluate the role of mesoscale and submesoscale eddies on nutrient transport and uptake rates by applying a multi-scale Reynolds decomposition to output from a submesoscale-permitting model of the CCS (F. Kessouri et al., 2020; Damien et al., 2023). Our analysis provides new insights on the different routes of nutrient supply and removal in the euphotic layer, and on the scale-dependent interplay between non-linear fluid and ecosystem dynamics in a highly heterogeneous environment.

2 Methods

2.1 Physical-biogeochemical model

We use the Regional Ocean Modeling System (ROMS, (Shchepetkin & McWilliams, 2005)) coupled online to the Biogeochemical Elemental Cycling model (BEC, (Moore et al., 2004; Deutsch et al., 2021)). ROMS solves the hydrostatic primitive equations for the tree-dimensional velocity, temperature, salinity and the transport of tracers in a terrain-following coordinate system. BEC represents the biogeochemical cycles of major elements (C, N, P, O, Fe, Si) resulting from the interaction of three phytoplankton and one zooplankton group.

We analyze output from two twin simulations for the northern and southern U.S. West Coast at 1 km resolution (Damien et al., 2023), sufficient to allow emergence of submesoscale dynamics (F. Kessouri et al., 2020), obtained by dynamical downscaling of a coastwide configuration at 4 km resolution (Renault et al., 2021; Deutsch et al., 2021). Because these simulations do not include tidal forcings, the highest frequencies captured by the model only include submesoscale circulation and internal waves generated within the domain. Output consists of physical and biogeochemical variables, transport fluxes, and biogeochemical rates calculated online by the model, and is saved as daily averages.

In the model, an arbitrary biogeochemical tracer $X_i$ obeys the conservation equation:

$$\partial_t X_i = T(X_i) + \partial_z (\kappa \partial_z X_i) + J_i(X_{j=1,...,J}).$$

(1)
The first term on the right hand side, \( T(X_i) = -\nabla \cdot (u X_i) \), represents the divergence of the advective flux, with \( u = (u, v, w) \) the velocity vector. It can be further decomposed into a horizontal \( T_\theta(X_i) \) and vertical \( T_v(X_i) \) component. The second term represents vertical mixing, with \( \kappa \) the vertical eddy diffusivity. The third term, \( J_i \), is the sum of all biogeochemical rates that affect the tracer \( X_i \), which in turn depend on \( J \) model state variables \( X_j \).

We focus on the balance of nitrate (\( \text{NO}_3^- \)), the main limiting nutrient in the CCS (Deutsch et al., 2021). For this variable, the net biogeochemical reaction rate is:

\[
J = J_{\text{Uptk}} + J_{\text{Nit}} + J_{\text{Denit}}
\]

Here, \( J_{\text{Uptk}} \) is the rate of uptake by phytoplankton, \( J_{\text{Nit}} \) production by nitrification, and \( J_{\text{Denit}} \) consumption by denitrification. Note that here, \( J_{\text{Uptk}} \) is a negative rate because it removes nutrient from sea water. Therefore, it is equivalent to net primary production, with an opposite sign, and expressed in nitrogen units. In the CCS, denitrification only occurs in the deeper parts of anoxic basins and in the sediment, and is a minor term compared to nitrification and biological uptake. Hence, when discussing water column processes, we focus primarily on nitrification and uptake. The nitrification rate, \( J_{\text{Nit}} = \tau_{\text{Nit}} \text{NO}_2^- \), is modeled as a linear function of nitrite (\( \text{NO}_2^- \)) concentration, with \( \tau_{\text{Nit}} \) a constant timescale. Non-linearities in nitrification arise from limitation under high irradiance in the euphotic zone, and inhibition at vanishing oxygen and nitrite concentrations (Deutsch et al., 2021). Biological uptake depends on nutrient concentrations following a Michaelis-Menten kinetics and Liebig’s law of the minimum, phytoplankton biomass, and a temperature- and light-dependent growth rate (see Supporting Information T2, and Deutsch et al. (2021)). Thus, uptake is highly non-linear because of the presence of bilinear \( (X_i X_j) \), exponential \( (e^{X_i}) \), and hyperbolic (Michaelis-Menten) terms.

### 2.2 Triple decomposition of transport and biogeochemical rates

The non-linear nature of advection, nitrification, and uptake in the nitrate conservation equation (Equation 1) is at the root of eddy rectification effects that modulate the final rate of change of this tracer.

To separate the effects of mesoscale and submesoscale eddies, we apply a triple Reynolds decomposition based on two low-pass filters, \( \tau \) and \( \tau' \), with respective space/time scales \( (\lambda, \tau) \) and \( (\lambda', \tau) \) (Capet et al., 2008). Accordingly, a model variable \( X_i \) is decomposed into mean and fluctuating mesoscale and submesoscale components as:

\[
X_i = \overline{X_i} + X'_i + X''_i,
\]

where

\[
X'_i = \overline{X_i} - \overline{X_i}
\]
and

\[
X''_i = X_i - \overline{X_i}.
\]

By definition, \( \overline{X_i} = 0 \) and \( \overline{X''_i} = 0 \). Here, \( (\lambda, \tau') \) and \( (\lambda', \tau) \), are chosen to separate mesoscale \( (X'_i) \) and submesoscale \( (X''_i) \) fluctuations from a large-scale, low-frequency mean \( (\overline{X_i}) \) that includes the seasonal cycle. The \( X''_i \) component represents the smallest scales and highest frequencies allowed by the model, i.e., mostly submesoscales. The choice of the filter scales is dependent on the circulation regime, and may not always perfectly separate intrinsic variability from forced motions. For example, along the U.S. West Coast, wind-driven upwelling is generally considered part of the mean seasonal variability. However, short-term wind events can generate high frequency variability in circulation that overlaps with mesoscale and submesoscale motions. In our choice of filters, we were especially careful to attribute the main upwelling signal to large-scale regional variability (i.e., the mean term \( \overline{X_i} \)) and not higher frequency fluctuations. To this end, we found a reasonable combination of temporal and spatial filter scales, defined as follows:
• \((\lambda, \tau) = (15 \text{ km}, 3 \text{ months})\), with a centered averaging scheme,
• \((\lambda, \tau) : (5 \text{ km}, 3 \text{ days})\), with a centered averaging scheme.

We refer the readers to Supporting Information T1 that further discusses these filters and their performance, using surface temperature and vertical velocities as an illustration.

By applying these filters to model variables, biogeochemical transformation rates can be separated into mean and eddy components. For a nonlinear reaction rate \(J_i(X_j)\) with dependence on multiple variables \(X_j, j = 1, \ldots, J\) and the transport divergence \(T(X_i)\), the analogous Reynolds decomposition takes the form:

\[
J = J^{\text{mean}} + J^{\text{meso}} + J^{\text{subm}} \quad \text{and} \quad T = T^{\text{mean}} + T^{\text{meso}} + T^{\text{subm}}
\]

where the various terms are now calculated as:

\[
\begin{align*}
J^{\text{mean}} &= J_i(\overline{X}_j) \quad \text{and} \quad T^{\text{mean}} = T(\overline{X}_i) \\
J^{\text{meso}} &= J_i(\tilde{X}_j) - J_i(\overline{X}_j) \quad \text{and} \quad T^{\text{meso}} = T(\tilde{X}_i) - T(\overline{X}_i) \\
J^{\text{subm}} &= J_i(X_j) - J_i(\tilde{X}_j) \quad \text{and} \quad T^{\text{subm}} = T(X_i) - T(\tilde{X}_i)
\end{align*}
\]

By adopting the filtering approach discussed above, the three terms in Equation 5 can be respectively interpreted as the contribution to the total rate caused by the large-scale mean tracer distributions (\(J^{\text{mean}}\) and \(T^{\text{mean}}\)); the contribution caused by heterogeneity at the scale of mesoscale eddies (\(J^{\text{meso}}\) and \(T^{\text{meso}}\)); and the contribution caused by heterogeneity at submesoscales and smaller scales captured by the model (\(J^{\text{subm}}\) and \(T^{\text{subm}}\)). Specifically, the biogeochemical eddy contributions only exist as a rectification of biogeochemical rates that depend in non-linear ways on model variables. These contributions would vanish in the case of perfectly linear rates (Levy & Martin, 2013).

### 2.3 Amplitude and sign of the eddy rectification

Assuming high frequency fluctuations of small amplitudes relative to the mean, the low frequency and large scale advective transport \(T\) and biogeochemical rates \(J\) can be approximated by a Taylor series expansion (Levy & Martin, 2013):

\[
\overline{J}(X_i) = J(X_i) + \sum_i \frac{\partial J}{\partial X_i} \overline{X}_i + \frac{1}{2} \sum_{i,j} \left( \frac{\partial^2 J}{\partial X_i \partial X_j} \right) \overline{X}_i \overline{X}_j + O(\overline{X}_i \overline{X}_j)^2
\]

An equivalent Taylor expansion can be written for the advection term \(T\), leading to a typical definition of eddy transport fluxes (Capet et al., 2008). Since the fluctuations have zero mean, the linear terms disappear. Ignoring the contribution of third-order terms, the amplitude and sign of the eddy rectified effect depend on the curvature of the functional dependencies (encapsulated by \(J\)) and eddy correlation terms between model variables \((\overline{X}_i, \overline{X}_j)\):

\[
\overline{J}^{\text{eddy}} \approx \frac{1}{2} \sum_{i,j} \frac{\partial^2 J}{\partial X_i \partial X_j} \overline{X}_i \overline{X}_j
\]

### 3 Results

We find that, along the CCS coast, the balance of nitrate in the surface layer (Equation 1) reflects a near compensation of two major terms: biological uptake \(J^{U_{\text{ptk}}}\), and the divergence of the vertical transport \(T_v\) (Figure 1). The mean component of \(J^{U_{\text{ptk}}}\) increases towards the coast (Figure 1a), reflecting high nutrient concentrations following inputs by upwelling (Figure 1d). Both mesoscale and submesoscale contributions to \(J^{U_{\text{ptk}}}\) are opposite in sign and partially offset the mean component. The magnitude of
the submesoscale contribution is particularly large, reaching about -34% of the mean $J_{U^{ptk}}$, while the mesoscale contribution is more limited.

Supply of NO$_3^-$ by vertical transport (i.e., the divergence of the vertical flux; Figure 1d-f) shows noisier patterns, reflecting the high variability and large magnitude of advective fluxes. However, notable patterns emerge. The most significant is the positive mean $T_v$ (i.e., NO$_3^-$ supply) along the coastal band caused by upwelling. The submesoscale $T_v$ largely opposes the mean upwelling along the coast, reducing NO$_3^-$ supply by 50 to 70%. In contrast, mesoscale $T_v$ is weaker, and is characterized by upwelling close to the coast, and downwelling offshore, thus reinforcing the mean vertical transport.

Based on these patterns, we distinguish between the coastal region, where nutrients are upwelled into the euphotic layers, and the offshore region, where subduction by mesoscale eddies dominates (Fig. 1). This separation occurs at a distance of approximately 40 km from the coast, comparable with the width of the continental shelf (Damien et al., 2023). Over the coastal region, the main balance in the NO$_3^-$ budget is between NO$_3^-$ supplied to the euphotic layer by vertical advection and uptake by phytoplankton (Fig. 2). Offshore, horizontal transport ($T_h$) replaces vertical advection as the main source of NO$_3^-$. Vertical mixing is also significant offshore, accounting for 33% of the NO$_3^-$ inputs.

In the NO$_3^-$ balance, eddy reaction rates generally oppose mean reaction rates: eddy $J_{U^{ptk}}$ is positive and eddy $J_{Nit}$ is negative. The magnitude of the eddy $J_{U^{ptk}}$ is particularly large, accounting for ~45% of the mean rate in both the coastal and offshore regions. This eddy contribution is largely dominated by submesoscale.

Near the coast, mean upwelling $T_v$ is the largest source of NO$_3^-$, and is largely offset (~64%) by submesoscale subduction. The total horizontal advection is negligible, reflecting a balance between the mean $T_h$, which supply NO$_3^-$, and eddy $T_h$, which remove it. The picture is different offshore. Both mean and eddy currents supply NO$_3^-$ at similar rates. NO$_3^-$ delivery by the mean transport is equivalent to that in the coastal region, accounting for 64% of the horizontal NO$_3^-$ supply, while mesoscale and submesoscale components account for 26% and 10% respectively. Along the vertical direction, we observe a close balance between subduction at mesoscales and supply at submesoscales. The magnitude of NO$_3^-$ supply by vertical mixing is similar in both regions (~1.0 $10^{-5}$ mmol m$^{-2}$ s$^{-1}$), but its relative contribution is more significant offshore (33% of the total NO$_3^-$ supply). This mixing term is largely driven by deepening of the mixed layer in winter (not shown).

The mean $J_{U^{ptk}}$ and its submesoscale rectification show a large seasonal cycle, with a maximum during upwelling in summer (Fig. 3a,b). $J_{Nit}$ follows a similar seasonal cycle, with a maximum following the peak in biological uptake, and is dominated by the mean component (Fig. 3c,d). In contrast, mesoscale eddy reactions show weak seasonality, and large fluctuations on time scales of weeks, especially in summer.

Over the course of the year, the vertical transport near the coast is shaped by mean upwelling (Fig. 3e), and balanced by submesoscale subduction. While mesoscale fluctuations cancel out when integrated over the annual cycle (Fig. 2), they drive the total transport at weekly timescales. Offshore, seasonal variability is less pronounced, and the period of maximum transport follows the upwelling season. Subduction by mesoscale eddies is larger from June to November, when re-supply by submesoscale eddies also increases.

The mean horizontal transport remains small relative to the mesoscale component, which dominates on weekly timescales. Over the year, the horizontal NO$_3^-$ flux from the coast to the open-ocean (Fig. 3 i) is largely positive (~8.3 $10^4$ molN s$^{-1}$). This redistribution of nutrients occurs at all scales, with a major contribution from the mean circulation (56%), reflecting wind-driven Ekman transport, followed by mesoscale eddies.
The strong correlation between $T_h$ offshore (Fig. 3 h) and the cross-shore flux (Fig. 3 i) indicates that NO$_3^-$ variations in the open-ocean section of the CCS are mostly caused by transport from the region of active upwelling near the coast (Damien et al., 2023).

4 Discussion and conclusions

4.1 Eddy transport

In the CCS, similar to other EBUS, nutrient subduction by eddies, or “quenching”, plays a major role in modulating primary production (Gruber et al., 2011; Nagai et al., 2015; Renault et al., 2016). Here, we show that eddy quenching reflects two contrasting regimes: subduction of freshly-upwelled nutrients by submesoscale eddies nearshore, and by mesoscale eddies further offshore (Figs. 1 and 2). Mesoscale eddies thus transport nutrient from the coast to the open-ocean, but also tend to “bury” them below the euphotic zone (Gruber et al., 2011). Offshore, we observe a near compensation between subduction at mesoscale and delivery at submesoscale (Fig. 2). This balance is particularly evident between July and October, following the large coastal nutrient injection caused by upwelling (Fig. 3). As recently-upwelled nutrients travel offshore and progressively sink along isopycnals, submesoscale eddies tend to resupply them back to the euphotic layer (F. Kessouri et al., 2020).

Globally, submesoscale eddies have been shown to enhance both nutrient delivery to the surface, in particular in the open ocean (Lévy et al., 2001; Mahadevan, 2016), and nutrient and organic matter subduction in regions of strong frontal activity (Omand et al., 2015; Haéck et al., 2023) and upwelling systems (Stukel et al., 2017; F. Kessouri et al., 2020). Here we show that both effects coexist along a gradient of surface nutrient concentrations in the CCS. Specifically, the direction of submesoscale nutrient transport depends on the balance between biological uptake and typical nutrient supply from below the euphotic layer. Relatively long nutrient residence times in surface layers associated with high nutrient concentrations and weak vertical gradients (as observed in nutrient-rich systems) favor nutrient removal by submesoscale currents. In contrast, short surface nutrient residence times associated with low nutrient concentrations and sharp nutrient gradients (typical of oligotrophic systems) favor submesoscale nutrient supply. This idea is supported by idealized (Freilich et al., 2022) and realistic (F. Kessouri et al., 2020) modelling studies.

4.2 Eddy Reactions

In the California Current, eddies reduce the mean nutrient uptake, and thus net primary production, by about 50%. Most of this compensation (35%) occurs at submesoscale. This eddy rectification is significantly larger than suggested by previous studies, which mostly focused on open-ocean regions and mesoscale circulations (Levy & Martin, 2013; Martin et al., 2015). Our study is the first to directly assess the magnitude of eddy reaction rates using a submesoscale-permitting model and a scale-dependent separation of mesoscale and submesoscale (Capet et al., 2008) in a region with particularly vigorous eddies. At coarser resolution, eddy kinetic energy is likely damped (Capet et al., 2008), thus leading to an underestimate of eddy heterogeneity and its contribution to biogeochemical rates.

Mesoscales and submesoscales are highly advective regimes that favor the emergence of heterogeneity and variability in tracer fields, which cause an eddy rectification of the mean biogeochemical rates. Integrated over large scales and low frequencies, eddy contributions are consistently reducing the mean uptake (Fig. 1 and 2). The magnitude and sign of this eddy rectification result from the eddy covariance of model state variables and the functional dependencies that describe biogeochemical transformations (Equa-
tion 10, see also Levy and Martin (2013)). Because biogeochemical rates depend on several tracers in complex ways (see Supporting Information T2), eddy reaction rates generally involve contributions from the interaction of multiple tracer pairs.

Analysis of the mesoscale contributions to NO$_3^-$ uptake (fig. 4) shows that the dominant terms arise from the saturating response of uptake at high nutrient concentrations (fig. 4 b-d), and the negative correlation between NO$_3^-$ and phytoplankton (Fig. 4 i,j). Specifically, the negative curvature of the Michaelis-Menten saturation function implies that, in a heterogeneous environment, high-frequency events characterized by large NO$_3^-$ concentrations are not as important in boosting uptake, relative to low-NO$_3^-$ events that are instead more effective at reducing it.

Furthermore, assuming a small Damkohler number (i.e., the ratio of the reaction rate to the high-frequency transport rate), Equation 10 can be re-stated by invoking mean tracer gradients and high frequency fluctuations, here assumed to occur mostly along the vertical direction $z$:

$$J^\text{eddy} \approx \frac{1}{2} \sum_{i,j} \frac{\partial^2 J}{\partial X_i \partial X_j} \bigg|_{X_i, X_j} \frac{\partial X_i}{\partial z} \frac{\partial X_j}{\partial z} \delta z'$$

(11)

with $\delta z'$ a small vertical fluctuation. Because vertical profiles of nutrients and phytoplankton show large and opposite gradients, in particular near the base of the euphotic zone, vertical fluctuations enhance the negative covariance between phytoplankton and NO$_3^-$ (Fig. 4 i), producing a sub-surface maximum in the eddy uptake rectification terms.

In contrast, the smaller amplitude of $J^{U^\text{uptk}}$ rectification at the mesoscale likely reflects a larger influence of horizontal rather than vertical fluctuations, where negative correlations between nutrients and phytoplankton are more ambiguous. Furthermore, this argument is based on a small Damkohler number approximation. Considering a time scale of the order of 1.0 d$^{-1}$ for nutrient uptake (see Supporting Information T2), this approximation is more appropriate for submesoscale rather than mesoscale fluctuations.

When integrated over a full seasonal cycle, we obtain ratios between eddy and mean uptake rates that are remarkably constant ($\sim$ -0.35 for submesoscale and $\sim$ -0.10 for mesoscale) across the CCS. To what extent these ratios can be generalized to different regions and circulation regimes remains an open question.

### 4.3 Implications

We found a remarkable compensation between mean and submesoscale terms in the net balance of NO$_3^-$ over a seasonal cycle in the California Current System (Fig. 2). This suggests that, in the productive coastal region, nitrate supply occurs predominately at large scales and low frequencies, while removal occurs at small scales and high frequencies. This balance is reversed offshore. While mesoscale contributions tend to cancel out over the seasonal cycle, they generate large variability, producing extremes in both nutrient transport fluxes and uptake rates (Fig. 3 and 4).

The nutrient heterogeneity caused by eddies does not necessarily promote biological productivity. Indeed, it systematically reduces it when averaged over large scales and low frequencies, thus representing a different kind of productivity “quenching” associated with non-linear ecosystem dynamics. The reasons are twofold. First, phytoplankton uptake quickly saturates at high nutrient concentrations. Second, high nutrient concentrations are often associated with low phytoplankton biomass, which limits the potential for increased productivity. We also note that changes in productivity caused by correlations involving temperature (which modulates uptake rates with an exponential dependence) are negligible in the open ocean, but become more important along the continental margin (Fig. 4 g, m, and p).
More generally, we find that eddy terms are far from negligible compared to mean biogeochemical rates. This result questions the ability of coarser models to adequately represent nutrient fluxes and biogeochemical transformations. For example, a non eddy-resolving global model would likely overestimate the vertical nutrient supply and biological uptake along upwelling systems. Physical parameterizations of eddy transport (Gent & McWilliams, 1990; Fox-Kemper et al., 2008) can partially alleviate this issue. However, analogous parameterizations for eddy biogeochemical rates are in early stages of development (Wallhead et al., 2013) and are not yet applied to biogeochemical models. Historically, biases in ocean circulation have been addressed by tuning biogeochemical parameters, which thus implicitly depend not only on the choice of model equations, but also on the resolution at which models are run and evaluated against observations. Our finding of a constant ratio between eddy and mean nutrient uptake rates across a range of circulations (~0.35 for submesoscale and ~0.10 for mesoscale), and our analysis of the different contributions of tracer covariance terms to eddy rates, offer new insights for the development of eddy parameterizations of biogeochemical transformations.

Finally, we focused on biological nutrient uptake as the dominant biogeochemical transformation in the highly productive CCS. However, the dynamics of pelagic ecosystems is characterized by many non-linear processes, from food web interactions, to remineralization and microbial dynamics under low oxygen conditions, which remain untouched here. In environments naturally sensitive to multiple stressors, such ocean acidification, warming, and oxygen loss, eddy rectification of ecological processes could greatly alter ecosystem dynamics and marine habitats. Analysis of these processes requires a shift in emphasis from nutrients to carbon and oxygen balances, and from biogeochemical to ecological interactions.

**Data Availability Statement**

The model code used to generate the simulation is openly available in Kessouri et al. (2020) ([https://doi.org/10.5281/zenodo.398861](https://doi.org/10.5281/zenodo.398861)). The simulations are reproducible using the setup and forcing described in Damien et al. (2023).

**Acknowledgments**

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Figure 1. Triple scale decomposition (mean, mesoscale and submesoscale) of (a-c) NO$_3^-$ biological uptake ($J_{Uptk}$), and (d-f) NO$_3^-$ vertical transport divergence ($T_v$, equal to the flux at the base of the layer) averaged over a full seasonal cycle and integrated over the euphotic layer ($\sim$ 0-50 m depth). Units are mmol N m$^{-2}$ s$^{-1}$. Black lines highlight the mean $J_{Uptk}$ isolines of -0.5, -1, -1.5, -2, and -2.5 in the upper panels and the mean $T_v$ isolines of 1, 2 and 5 in the lower panels. A companion figure showing the other terms of the NO$_3^-$ balance is provided in the Supporting Information, Fig. S5, S6.
### Figure 2.

Separation into mean, mesoscale and submesoscale components of the NO$_3^-$ balance terms (Equation 1) integrated in time over a seasonal cycle, in depth over the euphotic layer, and in space over two distinct regions of the U.S. West Coast: a coastal region, from Point Conception to Cape Blanco, up to 40km from the coast, and an offshore region up to 250km from the coast. Terms representing negligible component of the fluxes are omitted. Summed up by scales, the NO$_3^-$ balance represents +5.0, +0.8, and -6.8 at respectively mean, mesoscale, and submesoscale in the coastal region, and -4.2, +0.1, and +3.0 offshore. The total adds to 0 when the vertical mixing is included.
Figure 3. Daily averaged time-series of the (blue) mean, (green) mesoscale, and (purple) submesoscale terms of the NO$_3^-$ balance integrated over the (left panels) coastal and (right panels) offshore regions. In each panel, the red line shows the total rate (calculated online), which equals to the sum of the 3 components. Units are mmol N m$^{-2}$ s$^{-1}$. The light shaded area shows the +/- standard deviation over the region. This is not included for the transport divergence because it is an order of magnitude larger than the regional average. Note that the y-axis of the transport divergence use a different scale on the left and right panels. Panel (i) shows the time series of the horizontal NO$_3^-$ flux from the coastal to offshore region in mmol N s$^{-1}$. 
Figure 4. Cross sections, as a function of the distance from the coast and depth, of (a) the annual mean mesoscale eddy uptake, (b,e,h,k,n) the second derivative terms that modulate the (c) nutrient and (f) temperature eddy variance, (i) nutrient-phytoplankton eddy covariance, (l) nutrient-temperature eddy covariance, and (o) temperature-biomass eddy covariance at mesoscale. Following the Taylor series expansion (Equation 10, also shown at the top), the (a) mesoscale eddy uptake is approximated by the sum of the (d,g,j,m,p) second-order terms. Units of the uptake rate are mmol N m$^{-3}$ s$^{-1}$. The thick black contour represents the nutricline, defined by a nitrate concentration of 1 mmol N m$^{-3}$. A companion figure comparing eddy covariance at mesoscale and submesoscale is provided in the Supporting Information, Fig. S3.
References


Supporting Information for ”Eddy impacts on the marine biogeochemistry of the California Current System”

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1. Texts T1 to T3
2. Figures S1 to S6

Introduction

This file contains information on the computation of the triple eddy decomposition (T1), the mathematical formulation of the nutrient uptake in the biogeochemical model used in the study (T2), and additional figures complementing the results section of the main paper (T3).

1. Technical details for the triple decomposition

The proposed decomposition method relies on two filters that aim to separate the mean field associated with regional and seasonal variations from the mesoscale and submesoscale fields associated with turbulence in these dynamical regimes. This decomposition is made delicate by the absence of clear boundaries between these regimes. In fact, a certain degree of overlapping exists making the choice of the filters partly subjective. Practically, we tested several space, time, and combined space-time filters, and finally opt for the combination of space and time box-averaging filters defined as follows:
• $\tau$: 3 months and 15 km center averaging,

• $\tilde{\tau}$: 3 days and 5 km center averaging.

Although not providing an excellent precision in the selected cut-off scale, this method is widely used in ocean dynamics to decompose mean and eddy flows. Figures S1 and S2 expose the performance of this decomposition on temperature $T$ and vertical velocity $w$ fields at 25m in the CCS. As expected, the upwelling signal, characterized by cold water along the coast, is largely contained in $T$. $T'$ exhibits the large positive and negative anomalies with moderate gradients expected at mesoscale while sharp filamentary and frontal anomalies characterized $T''$ at submesoscales. Submesoscale is also associated with the large majority of vertical motions, as ageostrophic current start to be significant at submesoscale. Figure S3 presents the mesoscale and submesoscale eddy variance of several tracers and momentum at 25m depth. They both reach large magnitude compared to their low-frequency state. Except for vertical velocities, variance at mesoscale is significantly larger than at submesoscale.

The major downside of the spatial filtering method is the question of the boundary. Close to the coast, it induces a "shadow zone" of half the filter width. We excluded this region for the analysis and leave a dedicated assessment to future studies. A way to overcome this issue could rely on the use of degraded filters or exclusively time-based (space or frequency) filters along the coast.

The decomposition of any biogeochemical equation requires the online computation of daily averages of the equation trend terms and of the tracer fields. The online averaging of the rates and fluxes allows to capture a signal frequency as high as the model can provide, i.e. corresponding to the temporal and spatial resolution of the model. Then, mean terms are computed offline applying...
Figure S1. Triple decomposition applied to a temperature field at 25 m depth: (Upper panels from left to right) temperature snapshot and its decomposition into mean, mesoscale, and submesoscale components; (lower panels) cross-sections of the filter products at 35.4N (black line on the upper panels).

The flux formulation on filtered tracer fields. The eddy terms are computed by differences:

\[ J^{\text{mean}} = J(\bar{X}_i) \]  
(1)

\[ J^{\text{meso}} = J(\tilde{X}_i) - J(\bar{X}_i) \]  
(2)

\[ J^{\text{subm}} = J(X_i) - J(\tilde{X}_i) \]  
(3)

In a Reynolds decomposition, the eddy terms arise from the eddy-scale correlations between momentum and biogeochemical tracers. Since the chosen filters are not strictly orthogonal, the
Figure S2. Same as Fig. S1 except for vertical velocity.

necessary condition $\overline{X}_i = X_i$ for the cross-terms to vanish, is not necessarily satisfied. As a
results, the Reynolds decomposition does not strictly apply here. However, the chosen method
has the major advantage to retrieve an eddy turbulent flux by differences between total and mean
flux. The drawback is that it might also account for non-negligible cross terms arising from the
correlation between the mean and eddy components of momentum and tracers.
Figure S3. Mesoscale and submesoscale eddy variance of temperature, nitrate, phytoplankton, radiation, horizontal and vertical velocity at 25m depth. The mesoscale and submesoscale standard deviations are defined as respectively
\[ \hat{x} = \sqrt{\sum_{i=1}^{\tau} (\tilde{x}_i - \bar{x})/\tau} \] and \[ \tilde{x} = \sqrt{\sum_{i=1}^{\tau} (x_i - \tilde{x})/\tilde{\tau}}. \] Except for vertical velocities, variability at mesoscale tends to be larger than at submesoscale. Since the eddy transport and reaction grow on the variability of tracers and momentum at eddy scales, the variability in determined frequencies range gives an indication on the magnitude of the eddy transport and reactions.

Assuming \( \bar{X}_i = X_i \) and \( \tilde{X}_i = \tilde{X}_i \), \( J_{\text{meso}} \) and \( J_{\text{subm}} \) relate to the eddy \( X_i \) as follows:
\[
\partial_t \bar{x} = \tilde{J}(x_k=1,\ldots,n) \\
= J(x_k=1,\ldots,n) + J'(x_k'=1,\ldots,n) \\
\partial_t \tilde{x} = \tilde{J}(x_k=1,\ldots,n) + \tilde{J}(x_k''=1,\ldots,n) \\
= J(x_k=1,\ldots,n) + J'(x_k'=1,\ldots,n) + J''(x_k''=1,\ldots,n)
\]
$J(x_{k=1,...,n})$ stands for physical and biogeochemical fluxes. Since $\tau_{filt} > > \tau_{\tilde{filt}}$, we assume $\tilde{x} \approx x$, implying:

$$\partial_t \tilde{x} = J(\tilde{x}_{k=1,...,n}) + J(x'_{k=1,...,n}) + J(x''_{k=1,...,n})$$

(5)

2. Eddy uptake

In BEC, the mathematical formulation of the nitrate biological uptake $J^{Uptk}$ expressed as:

$$J^{Uptk} = -Q_{N:C} \frac{V_{NO_3}}{V_{NO_3} + V_{NH_4}} J^{photo}_C$$

(6)

$$V_{NO_3} = \frac{NO_3/k_{NO_3}}{1 + NO_3/k_{NO_3} + NH_4/k_{NH_4}}$$

(7)

$$V_{NH_4} = \frac{NH_4/k_{NH_4}}{1 + NO_3/k_{NO_3} + NH_4/k_{NH_4}}$$

(8)

$$V_{Fe} = \frac{Fe}{Fe + k_{Fe}}$$

(9)

$$V_{PO_4} = \frac{PO_4}{PO_4 + k_{PO_4}}$$

(10)

$$V_{SiO_2} = \frac{SiO_2}{SiO_2 + k_{SiO_2}}$$

(11)

$$J^{photo}_C = PC_{ref} f_{nut} T_{func} \left(1 - e^{-\frac{\alpha_{Chl:C} PAR}{T_{ref} f_{nut} T_{func}}} \right) C_{phyto}$$

(12)

$$f_{nut} = \min(V_{NO_3} + V_{NH_4}, V_{Fe}, V_{SiO_2}, V_{PO_4})$$

(13)

$$T_{func} = 2^{0.1 T - 3}$$

(14)

with $T_{func}$ a temperature dependency, $Q_{N:C}$ the constant stochiometric ratio of nitrogen over carbon, $PC_{ref}$ the constant maximum phytoplankton C-specific growth rate at given temperature set to 3.0 d$^{-1}$, $f_{nut}$ the nutrient limitation function, $Q_{Chl:C}$ the variable ratio of chlorophyll over carbon.
carbon in phytoplankton, and $\alpha_{ch}$ the chlorophyll-specific initial slope of $P$ vs. $I$ curve. $k_{NO_3}$, $k_{NH_4}$, $k_{Fe}$, $k_{PO_4}$, $k_{SiO_2}$ are the half saturation constant for nutrient uptake.

This formulation beholds multiple sources of non-linearities that allows an eddy rectification to emerge. The most evident ones are the covariance between nutrient ($N$) and phytoplankton ($P$) concentrations, the exponential temperature ($T$) dependency that also co-varies with nutrient and phytoplankton concentrations, and the exponential growth with light ($L$). These multiple eddy correlations ($N-P-T-L$) coupled to functional dependencies contribute to the magnitude of the eddy uptake and define its sign. However, we found that, averaged over the high frequency fluctuation period, the eddy uptake is largely negative (Fig. ?? and ??).

Assuming high frequency fluctuations of small amplitudes, we can approximate the rectified effect by a Taylor series expansion:

$$J_{eddy} \approx \sum_i \frac{\partial J}{\partial X_i} \bigg|_{X_i, X_j, \ldots} X'_i + \frac{1}{2} \sum_{i,j} \frac{\partial^2 J}{\partial X_i \partial X_j} \bigg|_{X_i, X_j, \ldots} X'_i X'_j$$

(15)

Because the fluctuations have zero average, the linear terms disappear and the sign and amplitude of the eddy rectification depend on the curvature of the functional dependencies and the eddy correlation term. A comparison of the different contribution for the uptake rectification (Fig. S4) evidences that $N'N'$ and $N'P'$ are dominant at mesoscale and submesoscale. This is largely due to the larger magnitudes of the second uptake derivatives that modulate the quadratic terms. The negative sign of the uptake rectification mainly arises from the product of $N'^2$, positive by definition, and the Michaelis-Menten nutrient growth dependencies with negative curvature. This rectification is increased at subsurface by the $N'P'$ where nutrients and phytoplankton are negatively correlated, and is partly compensated at surface by the covariance of the same vari-
ables. The other terms are overall less significant with the exception of the highly productive coast.

3. Additional figures completing the mean-eddy decomposition

This section includes additional figures that complement the result section and support the discussion.
Figure S4. Cross sections, as a function of off-coast distance and depth, of the annual mean (a) mesoscale and (e) submesoscale eddy uptake, (c,f,i,l,o) the second derivative terms that modulate the (d,e) nutrient and (g,h) temperature autocorrelation, (j,k) nutrient-phytoplankton eddy covariance, (m,n) nutrient-temperature eddy covariance, and (p,q) temperature-biomass eddy covariance at mesoscale and submesoscale. Units for uptake rates are mmol m\(^{-3}\) s\(^{-1}\). The thick black contour represents the nutricline defines as a nitrate concentration of 1 mmol.m\(^{-3}\). Note that the computation of the submesoscale eddy covariance was performed on daily averaged variables for numerical storage reason. Consequently, this approach leads to a significant underestimation of submesoscale variability, resulting in an underestimation of the magnitude of the submesoscale terms. Due to numerical storage limitation, we can only accurately diagnose the eddy reactions for the mesoscale. However, we remark that a partial diagnosis of submesoscale reactions based on daily average variables lead to similar result than the mesoscale diagnosis, but with different amplitude.
Figure S5. Triple decomposition of yearly averaged (upper panels) nitrification, and (lower panels) NO$_3^-$ transport divergence integrated over the euphotic layer: (from left to right) mean, mesoscale, and submesoscale. Units are mmol m$^{-2}$ s$^{-1}$
Figure S6. Cross-sections of the (blue) mean, (green) mesoscale, and (purple) submesoscale terms of the NO$_3^-$ balance averaged over a full seasonal cycle and integrated meridionally over the upwelling region, from Point Concepcion to Cape Blanco. The light shaded area is the meridonal standard deviation. The cross-sections of flux divergence is divided into coast and offshore with different y-axis scales. Units are mmol m$^{-2}$ s$^{-1}$. 