Mesoscale modulation of air-sea CO$_2$ flux in Drake Passage

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Abstract We investigate the role of mesoscale eddies in modulating air-sea CO$_2$ flux and associated biogeochemical fields in Drake Passage using in situ observations and an eddy-resolving numerical model. Both observations and model show a negative correlation between temperature and partial pressure of CO$_2$ (pCO$_2$) anomalies at the sea surface in austral summer, indicating that warm/cold anticyclonic/cyclonic eddies take up more/less CO$_2$. In austral winter, in contrast, relationships are reversed: warm/cold anticyclonic/cyclonic eddies are characterized by a positive/negative pCO$_2$ anomaly and more/less CO$_2$ outgassing. It is argued that DIC-driven effects on pCO$_2$ are greater than temperature effects in austral summer, leading to a negative correlation. In austral winter, however, the reverse is true. An eddy-centric analysis of the model solution reveals that nitrate and iron respond differently to the same vertical mixing: vertical mixing has a greater impact on iron because its normalized vertical gradient at the base of the surface mixed layer is an order of magnitude greater than that of nitrate.

1. Introduction

The Southern Ocean (SO) is the largest sink of anthropogenic carbon dioxide (CO$_2$), taking up roughly one third of the global inventory [Mikaloff Fletcher et al., 2006; Gruber et al., 2009; Lenton et al., 2013]. At the same time, the SO releases natural CO$_2$ into the atmosphere, partially canceling out anthropogenic uptake [Gruber et al., 2009]. Several modeling and observational studies suggest that the SO has become less effective in taking up anthropogenic CO$_2$ over the last several decades [Metzl, 2009; Le Quéré et al., 2010; Lovenduski et al., 2013]. A large-scale analysis by Landschützer et al. [2015] and regional studies by Xue et al. [2015] and Munro et al. [2015a], indicate that SO CO$_2$ uptake may have increased over the past decade. Landschützer et al. [2015] suggest that this recent increase in SO CO$_2$ uptake follows a decade in which CO$_2$ uptake stagnated consistent with previous studies, while an analysis by Majkut et al. [2014] suggest that no decrease in the effectiveness of SO CO$_2$ uptake has occurred.

The air-sea CO$_2$ flux in the SO has marked seasonal variability [Takahashi et al., 2002; Lenton et al., 2006; Metzl et al., 2006; Brix et al., 2013; Lenton et al., 2013]. In austral summer, the ocean takes up CO$_2$ due to increased biological productivity. In austral winter, however, suppression of biological productivity and the upwelling of waters rich in dissolved inorganic carbon (DIC) increases surface ocean pCO$_2$. Over the last decade Landschützer et al. [2014] and Munro et al. [2015a] suggest that the sign of winter air-sea exchange may be into the ocean due to slower oceanic pCO$_2$ increase than that in the atmosphere. Surface pCO$_2$ is also sensitive to temperature through solubility effects and decreases/increases as temperature cools/warms. These two opposing effects (biogeochemistry and solubility) compensate one another reducing seasonal pCO$_2$ variability [Jiang et al., 2014; Munro et al., 2015b] relative to other parts of the ocean where these effects are not as well balanced [Takahashi et al., 2002]. Even in the presence of compensation, observations show a seasonal variability in oceanic pCO$_2$ which is larger than that of atmospheric pCO$_2$ in both the Indian and west Pacific sectors [Metzl et al., 2006; Brix et al., 2013]. This indicates that the ocean plays an active role in air-sea CO$_2$ flux in the SO by changing the difference between oceanic and atmospheric pCO$_2$.

Mesoscale eddies, often identified by anomalies in sea surface height (SSH), modulate oceanic physical fields such as temperature, salinity, heat flux, heat content and vertical mixing [Hausmann and Czaja, 2012;...
However, it is not straightforward to anticipate how mesoscale eddies change CO2 flux because the overall effects and an eddy-resolving ocean model. Both observations and model reveal mesoscale modulation of the mixed layer.

Mesoscale eddies can also impact biogeochemical fields such as nutrients and plankton in complex ways (i.e., stirring, trapping and isopycnel displacement, Ekman pumping from eddy-wind interaction, and anomalous vertical mixing, see Gaube et al. [2013]). In particular, the trapping mechanism can transport water parcels horizontally to other areas with different biogeochemical properties, and is more effective when there is a stronger lateral gradient. Isopycnel displacement changes the depth of the nutricline and affects the nutrient supply to the euphotic zone. Anticyclones/cyclones deepen/shoal isopycnals and the nutricline, resulting in less/more available nutrients. However deeper/shallower mixed layers in anticyclones/cyclones can provide more/less nutrients to the surface [Kunze, 1985; Kunze et al., 1995; Koszalka et al., 2010; Song et al., 2015; Dufois et al., 2016].

These complex impacts of mesoscale eddies in the SO have been reported in the literature. Kahru et al. [2007] observed that cyclones have enhanced chlorophyll biomass while anticyclones are marked with lower chlorophyll biomass in the Scotia Sea of the southwest Atlantic. However, Meredith et al. [2003] found an increased chlorophyll biomass in an anticyclonic eddy found above the northwest Georgia Rise in the Scotia Sea. Korb and Whitehouse [2004] argue that this is because of anomalously high iron supply from shelf sediments. Thomalla et al. [2011] report a complex interplay between mixed-layer depth and chlorophyll in the SO, which suggests that anomalies in vertical mixing associated with mesoscale eddies can have different impacts on the surface chlorophyll biomass depending on the region. Indeed, satellite observations of SSH and chlorophyll reveal complex relationships between mesoscale eddies and chlorophyll biomass in the SO, with both positive and negative cross correlations [Gaube et al., 2014].

Mesoscale eddies modulate air-sea CO2 flux through anomalous physical and biogeochemical properties; however, it is not straightforward to anticipate how mesoscale eddies change CO2 flux because the overall sign and magnitude are dependent on a complicated balance of physical and biological processes with sometimes opposing effects. In the high-nutrient low-chlorophyll SO, primary production is generally regulated by iron (Fe) [Cooper et al., 1996; Boyd et al., 2000; Watson et al., 2000] and light. Thus higher levels of biologically available Fe will result in enhanced primary production, leading to lower DIC, pCO2 and either increased uptake or reduced outgassing of CO2. Anticyclones may have less Fe if the downward displacement of isopycnals pushes Fe-rich water to a deeper level, leading to less primary production and CO2 uptake. In contrast, anticyclones potentially lead to more CO2 uptake if they trap Fe-rich water near the sediment sources and are transported offshore [Korb and Whitehouse, 2004]. If vertical mixing in the open ocean is deep enough to entrain Fe-rich subsurface water in anticyclones, we can also expect more CO2 drawdown by enhanced primary production. Even in the presence of elevated Fe, the biological effect may be muted if anomalously deep mixed layers lead to light limitation, complicating the role of mesoscale eddies in modulating air-sea CO2 flux.

In this study, we attempt to gain a deeper understanding of the role of cyclonic and anticyclonic eddies in modulating CO2 flux in the region of Drake Passage over the seasonal cycle using repeat shipboard transects and an eddy-resolving ocean model. Both observations and model reveal mesoscale modulation of the CO2 flux which varies seasonally. This is studied by separating the oceanic pCO2 into temperature-driven and DIC-driven components. An eddy-centric analysis of the model simulation reveals a differing response of iron and nitrate to the same vertical mixing, which we link to the size of the vertical gradients at the base of the mixed layer.

Our paper is organized as follows. In section 2, we describe the observations of pCO2 across the Drake Passage. Eddy-resolving simulations of the CO2 flux and associated biogeochemical fields in the Drake Passage region are presented in section 3 using cross correlation and eddy-centric analysis. The influence of mesoscale eddies on the CO2 flux and nutrient supply are summarized and discussed in section 4. We conclude in section 5.
2. Analysis of Observations in Drake Passage

2.1. Data From the Laurence M. Gould

We use the sea surface temperature (SST) and the pCO$_2$ in surface waters sampled from the Antarctic Research Supply Vessel (ARSV) Laurence M. Gould across the Drake Passage during the period 2002–2015 [Jiang et al., 2014; Munro et al., 2015a, 2015b] (Figure 1a). This data set includes 246 crossings of Drake Passage with at least some measurements in every month of the year; crossings are most frequent in October (30 times in the austral spring) and least frequent in July (6 times in the austral winter) over the length of the time series (Figure 1b). The cross correlation between SST and pCO$_2$ anomalies are of interest in the investigation of the role of mesoscale structures on pCO$_2$. Hence, the linear trend of the surface data along the crossing is removed to reveal anomaly patterns. The result is qualitatively the same if other plausible methods are used in the computation of anomalies (Appendix A).

2.2. Observed Correlations Between SST and pCO$_2$ Anomalies

We investigate how the anomalies of physical and biogeochemical variables change associated with mesoscale eddies by examining the correlation near Drake Passage as shown in Figure 1c. The correlation between SST and pCO$_2$, T (red) and pCO$_2$, DIC (green) anomalies is plotted.
takes up CO₂ in summer due to increased primary production, the negative correlation suggests that more CO₂ enters the ocean in anticyclones than surrounding areas because of lower pCO₂. In winter, when the ocean emits CO₂ to the atmosphere, anticyclones release more CO₂ owing to their higher pCO₂. Cyclones exhibit the opposite behavior; less CO₂ uptake in summer and less outgassing in winter.

Such relationships are also seen in the analysis of expendable bathythermograph (XBT) data across the Drake Passage. For example Sprintall et al. [2012] found that warmer waters have greater ΔpCO₂ (air-sea) in spring north of 57°S associated with the decrease of oceanic pCO₂. In winter, XBT data show a negative correlation between temperature and ΔpCO₂ south of the Polar Front, corresponding to higher oceanic pCO₂ in warmer water. Sprintall et al. [2012] attribute this seasonal change in correlation to variation in the relative impacts of temperature-driven solubility and biological uptake, similar to what we will describe in section 4.1. In order to probe the mechanisms linking hydrodynamic variability with air-sea CO₂ flux, we now turn to a high resolution model of a section of the Antarctic Circumpolar Current.

3. Coupled Eddy Biogeochemical-Physical Model of the Drake Passage Region

3.1. High Resolution Model Including Drake Passage

The oceanic carbon cycle is simulated by accounting for oceanic transport as well as biological and carbonate sources/sinks. Here physical fields taken from the MIT Ocean General Circulation Model (MITgcm) [Marshall et al., 1997a, 1997b; Adcroft et al., 1997; Marshall et al., 1998; Adcroft et al., 1999] are used to drive a carbon cycle model. The model is configured in a sector of the Antarctic Circumpolar Current including the Drake Passage with a horizontal resolution of 1/20° and 50 vertical levels. The model domain also encompasses the region of the Laurence M. Gould transects. Forcing data sets for the model are the corrected normal year Common Ocean-Ice Reference Experiments version 2 (CORE-II) [Large and Yeager, 2009] at the surface and Ocean Comprehensible Atlas (OCCA) product [Forget, 2010] at the open boundaries. This configuration is adopted from that described in Tulloch et al. [2014] where the simulated physical solution is intensively evaluated against observations.

The biogeochemical sources and sinks for the carbon cycle are air-sea CO₂ flux, carbonate flux and primary production. They are computed using a simple biogeochemical model [Dutkiewicz et al., 2005; Parekh et al., 2006; Verdy et al., 2007] which simulates 6 biogeochemical variables: DIC, alkalinity, oxygen, nitrate (NO₃), and other relevant parameters.
3.2. Simulations of Air-Sea CO2 Flux

The air-sea CO2 flux \( F_{CO2} \) is parameterized as

\[
F_{CO2} = k_w (1 - A_B) (p_{CO2}^{atm} - p_{CO2})
\]

where \( k_w \) is the gas transfer velocity \((m s^{-1})\) estimated using wind speed squared and SST [Wanninkhof, 1992]. \( k_w \) is then inversely scaled by the sea-ice coverage \( A_B \) in each grid cell and becomes zero when the grid cell is completely covered by sea-ice. The atmospheric \( p_{CO2} \) \( (p_{CO2}^{atm}) \) is set to a pre-industrial level \( (278 \text{ ppmv}) \), and the oceanic \( p_{CO2} \) at the surface \( (p_{CO2}) \) is calculated following Follows et al. [2006]. A positive \( F_{CO2} \) means that the ocean receives CO2 from the atmosphere, and vice versa.

3.3. Analysis of Model Results

3.3.1. Eddy Correlations Over the Seasonal Cycle

The correlation between simulated SSH and \( p_{CO2} \) anomalies (solid black line in Figure 3) exhibits similar behavior to that found in the data described in section 2 (compare to Figure 1c). In summer, the \( p_{CO2} \) anomaly is negatively correlated with the SSH anomaly, indicating that anticyclones/cyclones have lower/higher oceanic \( p_{CO2} \). However, the correlation changes the sign to become positive in winter: anticyclones have higher \( p_{CO2} \) and cyclones lower \( p_{CO2} \). Higher \( p_{CO2} \) indicates either less CO2 uptake in summer, or more outgassing in winter, explaining the correlation between anomalies of SSH and \( p_{CO2} \) flux (orange line in Figure 3). A positive correlation in summer suggests that warm/cold anticyclonic/cyclonic eddies have higher/lower CO2 flux into the ocean. In winter, anticyclones/cyclones release more/less CO2 into the atmosphere than their surroundings.

Correlations between SSH anomaly and other biogeochemical variables are robust throughout the year. DIC and NO3 anomalies are negatively correlated with SSH anomalies (light blue and red lines, respectively, in Figure 3), suggesting that anticyclones/cyclones have lower/higher DIC and NO3. On the other hand, Fe and community productivity have positive correlations with SSH anomalies (purple and green lines, respectively, in Figure 3), indicating that enhanced levels of Fe and more community productivity occurs in anticyclones. Since the Fe is the limiting nutrient in most of the model domain in the SO (see Appendix D), it is understandable that Fe and community productivity anomalies show a similar relationship with mesoscale eddies.

3.3.2. Biogeochemical Fields Using Eddy-Centric Analysis

We compute the mean anomalies of biogeochemical variables using an eddy-centric coordinate. Using total SSH fields, we identified 1268 anticyclones and 1624 cyclones with lifetimes longer than 20 days over 2 years (Appendix C). The eddy-centric composite average of the CO2 flux anomalies changes sign with SSH and its dissolved organic form, and Fe. The model simulates net community production in the ocean using a combination of light and nutrient (Fe and NO3) limitation. This is similar, but distinct from primary production which only includes the impact of autotrophs (described in Appendix B). The biogeochemical model was integrated for 3 years from an initial state taking monthly mean boundary conditions from a global model. Model parameters are tuned to yield biogeochemical states close to the climatology. A list of parameter values is given in Table 1. The simulated surface \( p_{CO2} \) captures seasonal changes of the meridional \( p_{CO2} \) gradient in the climatology: increasing/decreasing \( p_{CO2} \) toward the equator in summer/winter (Figure 2). More evaluation of the modeled biogeochemistry against the climatology and observation is provided in Appendix D.

<table>
<thead>
<tr>
<th>Parameter Name</th>
<th>Value</th>
<th>Units</th>
</tr>
</thead>
<tbody>
<tr>
<td>Light</td>
<td>0.02</td>
<td>m(^{-1})</td>
</tr>
<tr>
<td>Self-shading coefficient ( k_{cdw} )</td>
<td>0.02</td>
<td>m(^{2}) mg(^{-1})</td>
</tr>
<tr>
<td>Photosynthetically active radiation</td>
<td>0.4</td>
<td>Dimensionless</td>
</tr>
<tr>
<td>Half saturation light constant</td>
<td>30</td>
<td>W m(^{-2})</td>
</tr>
<tr>
<td>Maximum consumption rate ( s )</td>
<td>0.84</td>
<td>( \mu ) M month(^{-1})</td>
</tr>
<tr>
<td>Half saturation NO3 constant ( K_{NO3} )</td>
<td>8</td>
<td>( \mu ) M</td>
</tr>
<tr>
<td>Half saturation Fe constant ( K_{Fe} )</td>
<td>0.12</td>
<td>nM</td>
</tr>
<tr>
<td>Fraction of new production to DON pool</td>
<td>0.7</td>
<td>Dimensionless</td>
</tr>
<tr>
<td>Time scale for DON remineralization</td>
<td>3</td>
<td>month</td>
</tr>
<tr>
<td>N:Fe stoichiometry</td>
<td>0.007</td>
<td>Dimensionless</td>
</tr>
<tr>
<td>Scavenging rate</td>
<td>0.2 \times 10^{-7}</td>
<td>s(^{-1})</td>
</tr>
<tr>
<td>Ratio of sediment Fe to NO3 flux ( \ell )</td>
<td>1.153</td>
<td>Dimensionless</td>
</tr>
<tr>
<td>Minimum Fe flux from sediment ( F_{Fe,M} )</td>
<td>1.0 \times 10^{-3}</td>
<td>pm s(^{-1})</td>
</tr>
</tbody>
</table>

Table 1. Parameter Names, Values, and Units for the Biogeochemical Model
anomaly in different seasons (Figures 4a–4d), as revealed by the cross correlation in section 3.3.1. The composite average of CO$_2$ flux has a monopole structure and the maximum anomaly can be found within a circle of radius $L_s$ whose size is the same as the one enclosed by SSH contour with the maximum mean current speed. The CO$_2$ flux anomalies are the largest in winter. Anomalous pCO$_2$ in mesoscale eddies influences the CO$_2$ flux anomalies in all seasons. A positive anomaly of pCO$_2$ leads to less CO$_2$ uptake in summer and more outgassing in winter and thus a negative anomaly of CO$_2$ flux. On the other hand, a negative pCO$_2$ anomaly results in a positive CO$_2$ flux anomaly. Since the ocean receives CO$_2$ in summer but emits in winter, the CO$_2$ flux anomaly indicates that anticyclones/cyclones take up more/less CO$_2$ in summer and releases more/less CO$_2$ in winter. If one takes the modification of wind speed by warm/cold eddies into account [Frenger et al., 2013], the CO$_2$ flux anomaly can be amplified. It should be noted that the interaction between SST and wind speed is not simulated in this study.

Variability of DIC on the mesoscale does not show a seasonal dependence (Figures 4i–4l). Anticyclones and cyclones have lower and higher DIC levels, respectively, than the spatial mean in both summer and winter. The DIC concentration generally increases with depth (Figure 5) and so the downward/upward displacement of isopycnals in anticyclones/cyclones results in a negative/positive DIC anomaly in both seasons. DIC has a negative meridional gradient, decreasing toward the north (Figure 5). Because the gradient of DIC across the Antarctic Circumpolar Current is negative, eddy fluxes across the front can yield negative correlation between SSH and DIC anomalies through the so-called “trapping” mechanism. Specifically, this effect is produced by lateral cross-frontal transport of warm anticyclonic eddies (cold cyclonic eddies) with low (high) DIC from the north (south) to the south (north).

Net community productivity is enhanced in anticyclones and reduced in cyclones in all seasons (Figures 6a–6d). The composite averages of the net community productivity anomaly have a sign opposite to that of the NO$_3$ anomaly (Figures 6e–6h), but the same sign as the Fe anomaly (Figures 6i–6l), reflecting the Fe-limited environment. The NO$_3$ spatial distribution is similar to that of DIC (Figure 5), suggesting that vertical isopycnal displacement and the trapping mechanism are both potentially at work in shaping the NO$_3$ anomalies. The Fe spatial distribution is quite different from those of DIC and NO$_3$ and its anomalies are not entirely determined by these two mechanisms, as will be discussed in section 4.2.

4. Proposed Mechanisms

4.1. Seasonal Cycle of CO$_2$ Flux Anomalies

Both observed and simulated mesoscale variabilities of pCO$_2$ show the same seasonal dependence as presented in section 3.3.1. Why does the correlation between anomalies of SST (or SSH) and pCO$_2$ reverse with
Figure 4. Eddy-centric composite averages for anomalies of (a–d) CO2 flux (e–h) pCO2 and (i–l) DIC. From the left to right, columns represent anticyclones in summer, cyclones in summer, anticyclones in winter and cyclones in winter. Eddies are mapped on the eddy-centric coordinate spanning ±2Ls, hence the dashed circle with a unit radius corresponds to one with a radius of Ls. Red, blue and black contours represent positive, negative and zero SSH anomalies with a 5 cm interval, respectively. The numbers on the lower right corner are the percentage of eddies whose anomaly have the same sign as the composite average.

season? We partitioned pCO2 into the temperature-driven (pCO2,T) and DIC-driven (pCO2,DIC) components for both observation and model following Takahashi et al. [2002], Jiang et al. [2014] and Munro et al. [2015b] and investigate the seasonality of mesoscale variability of pCO2.

Observations suggest that anomalies of pCO2,T and SST are positively correlated throughout the year (red line in Figure 1c), indicating that warmer temperature in anticyclones increases pCO2 because of reduced solubility. In contrast, cyclones with colder temperature have higher solubility and lower pCO2. On the other hand, pCO2,DIC are negatively correlated with the SST anomaly (green line in Figure 1c). This suggests that anomalously low and high pCO2 are expected in anticyclones and cyclones, respectively, owing to changes in DIC. Hence the observed seasonal variability of the correlation between anomalies of SST and pCO2 results from a change in the balance between pCO2,T and pCO2,DIC. In summer, the modulation of pCO2 by anomalous temperature in eddies is weaker than that by anomalous DIC, and the mesoscale variabilities of pCO2 and CO2 flux are governed by anomalous aqueous CO2 of DIC. In winter, however, the temperature-driven component dominates the pCO2 and CO2 flux anomalies.

Model simulations confirm relationships seen in the observations. The pCO2,T and SSH anomalies are positively correlated in both summer and winter (Figures 7a–7d). Warm eddies have low solubility, hence pCO2 increases, leading to a positive correlation between pCO2,T and SSH. The pCO2,DIC is negatively correlated with SSH anomaly (Figures 7e–7h). The inverse relationship between pCO2,DIC and SSH can be anticipated from the negative correlation between DIC and SSH anomalies (Figures 3 and 4i–4l)).

The eddy-centric analysis clearly shows that the balance between temperature and DIC tendencies varies (Figure 7). In summer, the pCO2,DIC anomaly is greater and more robust than the pCO2,T anomaly, resulting...
Figure 5. The zonal mean DIC and NO₃ from (top) the World Ocean Atlas 2009 climatology and (bottom) biogeochemical model.

Figure 6. Eddy-centric composite averages for anomalies of (a–d) surface net community productivity (e–h) NO₃ and (i–l) Fe. From the left to right, columns represent anticyclones in summer, cyclones in summer, anticyclones in winter and cyclones in winter. Eddies are mapped on the eddy-centric coordinate spanning ±2Lₑ, hence the dashed circle with a unit radius corresponds to one with a radius of Lₑ. Red, blue and black contours represent positive, negative and zero SSH anomalies with a 5 cm interval, respectively. The numbers on the lower right corner are the percentage of eddies whose anomaly have the same sign as the composite average.
in a negative correlation between $pCO_2$ and SSH anomalies. In winter, on the other hand, the $pCO_2_T$ anomaly dominates the DIC effect on $pCO_2$, leading to a positive correlation. The changes in the balance can be understood using anomalies in the eddy centric composite averages. One can estimate effects on $pCO_2$ due to SST, DIC and alkalinity (Alk) changes thus:

$$\frac{\Delta pCO_2}{pCO_2} \approx \frac{1}{pCO_2} \frac{\partial pCO_2}{\partial T} \Delta T + \frac{1}{pCO_2} \frac{\partial pCO_2}{\partial DIC} \Delta DIC + \frac{1}{pCO_2} \frac{\partial pCO_2}{\partial Alk} \Delta Alk$$

(2)

when salinity changes are small [Takahashi et al., 1993; Mahadevan et al., 2004; Sarmiento and Gruber, 2006; Lovenduski et al., 2007; Merlivat et al., 2015]. $\beta$ is close to 0.0423 C$^{-1}$, and $\gamma_{DIC}$ and $\gamma_{Alk}$ are the Revelle factor and alkalinity factor that are approximately 12 ± 0.5 and −11 ± 0.5 in the Drake Passage, respectively [Takahashi et al., 2014a]. In the open Southern Ocean, changes driven by alkalinity perturbation are smaller than the other two terms and have an insignificant contribution to $\Delta pCO_2$ while its contribution can be significant in other ocean regions (e.g., North Atlantic) [Takahashi et al., 2014a]. Our analysis also shows that the impact of alkalinity on $pCO_2$ is insignificant (not shown). If we drop off the last term on the right-hand-side, equation (2) indicates that the sign of $\Delta pCO_2$ in mesoscale eddies is determined by the sum of weighted SST and DIC anomalies. As shown in Figures 4i–4l, anticyclonic eddies are characterized by warmer SST and lower DIC concentrations. Cyclonic eddies, by contrast, have colder SSTs and higher DIC concentrations. On the mesoscale, SST and DIC anomalies averaged over the circular area with the radius $L_s$ vary by season (Table 2). SST anomalies are larger in winter while DIC anomalies are larger in summer. Inserting these values in equation (2) for anticyclones in Table 2 results in negative

**Figure 7.** Eddy-centric composite averages for anomalies of (a–d) temperature-driven $pCO_2$ and (e–h) DIC-driven $pCO_2$. From the left to right, columns represent anticyclones in summer, cyclones in summer, anticyclones in winter and cyclones in winter. Eddies are mapped on the eddy-centric coordinate spanning ± 2$L_s$, hence the dashed circle with a unit radius corresponds to one with a radius of $L_s$. Red, blue and black contours represent positive, negative and zero SSH anomalies with a 5 cm interval, respectively. The numbers on the lower right corner are the percentage of eddies whose anomaly have the same sign as the composite average.

**Table 2.** Spatially Averaged Eddy Composite Averages of SST Anomaly, DIC Anomaly and DIC Concentration in Anticyclones and Cyclones in Summer and Winter

<table>
<thead>
<tr>
<th></th>
<th>Summer</th>
<th>Winter</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Anticyclone</td>
<td>Cyclone</td>
</tr>
<tr>
<td>$\Delta$SST (°C)</td>
<td>0.37</td>
<td>0.20</td>
</tr>
<tr>
<td>$\Delta$DIC (mmol C/m$^3$)</td>
<td>−5.91</td>
<td>4.46</td>
</tr>
<tr>
<td>DIC (mol C/m$^3$)</td>
<td>2.09</td>
<td>2.12</td>
</tr>
</tbody>
</table>

*The circular area with the radius $L_s$ is considered in the spatial average.*
In contrast, $\Delta pCO_2$ in cyclones becomes positive in summer but negative in winter. Hence, the seasonal change of correlation coefficient between SSH and $pCO_2$ anomaly is the consequence of the seasonally varying magnitude in the anomaly of SST and DIC.

### 4.2. Responses in Nitrate and Iron

It is interesting to note that simulated NO$_3$ and Fe have opposite-signed anomalies in both summer and winter (Figures 6e–6l): anticyclones have anomalously high Fe but low NO$_3$, and cyclones have anomalously low Fe and high NO$_3$. Why do they display different responses driven by the same physical processes? Mesoscale eddies displace vertical isopycnals and modulate the surface NO$_3$. Doming of isopycnals in cyclones shoals subsurface water abundant with NO$_3$ and creates a positive anomaly. The opposite is true for anticyclones. Downward displacement of isopycnal in anticyclones results in a negative NO$_3$ anomaly. The trapping mechanism can also be applied to explain the NO$_3$ anomalies seen in the eddy-centric composite average in Figures 6e–6h. The meridional overturning circulation and the Fe-limited environment in the SO set up warm water with low NO$_3$ in the north and cold water with high NO$_3$ in the south. Mesoscale eddies can trap water parcels and transport them horizontally to other regions with different environments. If this trapping mechanism is at work, warm anticyclonic eddies should carry waters with low NO$_3$ and cold cyclonic eddies waters with high NO$_3$ across the Antarctic Circumpolar Current.

Similar to NO$_3$, Fe concentration increases with depth near the surface. As such, isopycnal displacements by mesoscale eddies would tend to produce the same sign of Fe anomalies as for NO$_3$. However, Figures 6i–6l show the anomalies with the opposite sign, indicating other processes are at work.

In order to evaluate the potential role of trapping, we examined the meridional gradient of Fe using the data described in Tagliabue et al. [2014]. Although Fe concentration can be elevated near the shelf with sedimentary sources, Fe is mostly depleted at the surface in the open ocean. Dust input increases Fe concentration in the lee of the continent while intense vertical mixing at the north of ACC in winter supplies Fe to the surface ocean. However these sources are limited to specific regions and Fe is quickly used up by biological activity when light is available. All these complex sources and sinks make it difficult to anticipate the effect of trapping on the composite average of Fe anomaly, which leads us to consider the role of vertical mixing.

The vertical diffusion of a tracer depends on the vertical diffusivity and the vertical gradient of that tracer. Since the same vertical diffusivity is applied to both NO$_3$ and Fe, it is the vertical gradient of the nutrients that creates the different responses.

| Table 3. Normalized Vertical Gradient ($\times 10^{-2}$) at the Base of the Mixed Layer |
|----------------------------------------|---------------------------------|---------------------------------|---------------------------------|
|                                       | Summer                          | Winter                          |                                  |
|                                       | Anticyclone                     | Cyclone                         | Anticyclone                     | Cyclone                         |
| NO$_3$                                | -0.31                           | -0.47                           | -1.68                           | -1.29                           |
| Fe                                    | -9.86                           | -9.31                           | -7.38                           | -7.83                           |

In contrast, $\Delta pCO_2$ in summer but positive in winter. In contrast, $\Delta pCO_2$ in cyclones becomes positive in summer but negative in winter. Hence, the seasonal change of correlation coefficient between SSH and $pCO_2$ anomaly is the consequence of the seasonally varying magnitude in the anomaly of SST and DIC.

We computed the vertical gradient of nutrients at the base of the surface mixed layer and normalized it by their concentration below the mixed layer. The normalized vertical gradients are all negative, suggesting that vertical mixing supplies nutrients to the surface mixed layer (Table 3). However, the effect of vertical mixing for NO$_3$ is not strong enough to overcome the influence of the trapping and isopycnal displacement mechanisms. The normalized vertical gradient of Fe is one order of magnitude greater than that of NO$_3$. Fe is the limiting nutrient in most of the eddy areas and becomes close to depletion when the productivity is high. As a result, the vertical gradient for Fe is greater than NO$_3$ and so vertical mixing of Fe is likely to have greater impact. Although the simulated vertical gradients of nutrients can slightly differ from the climatology, the observed vertical gradient of Fe [Tagliabue et al., 2014] is still greater than that of NO$_3$ from World Ocean Atlas 2009 climatology at the upstream of Drake Passage (not shown).

Even though the normalized vertical gradients of Fe are similar between anticyclones and cyclones, the simulated vertical diffusivity is much greater in anticyclones than in cyclones [see Song et al., 2015, Figure 3]. A variety of processes may contribute to differences in the mixing environments within cyclonic and anticyclonic eddies. For example, warm anticyclonic eddies become less stable after losing buoyancy to the atmosphere more quickly [e.g., Williams, 1988]. Also, the trapping of near-inertial waves promotes more vertical mixing [Kunze, 1985; Kunze et al., 1995]. The weaker stratification also helps the wind energy to penetrate deeper.
5. Conclusion

We investigated the mesoscale modulation of CO$_2$ flux and biogeochemical variables using both observations across the Drake Passage and an eddy-resolving numerical ocean model configured in the Drake Passage region. The 13 years (2002–2015) of SST and oceanic pCO$_2$ observations with 246 crossings are grouped by months after removing the linear trend across the Drake Passage. We computed the cross correlation between SST and pCO$_2$ anomalies to examine the observed mesoscale modulation. Using the eddy detection algorithm, more than 2800 eddies with lifetimes longer than 20 days were identified from the numerical model output and averaged after projecting to eddy-centric coordinate.

The cross correlation between SST and pCO$_2$ anomalies reveals a marked seasonal variability: negative correlation in summer and positive correlation in winter in both observations and model. The change in correlation over the seasons suggests that anticyclones are more active in both taking up CO$_2$ in summer and releasing CO$_2$ into the atmosphere in winter. Cyclones are less active than surrounding waters and behave in the opposite sense to anticyclones.

pCO$_2$ in the ocean changes with both physical and biogeochemical processes. The mesoscale modulation of CO$_2$ flux is thus determined by the balance between temperature-driven and DIC-driven effects. Warming of the surface ocean reduces the solubility, hence the pCO$_2$ increases. Warmer SST in anticyclones thus leads to the pCO$_2$ increase and either to less CO$_2$ uptake in summer or more outgassing in winter. Anomalously low DIC in anticyclones reduces pCO$_2$ leading to either more uptake in summer or less outgassing in winter. In summer, the DIC-driven effect is greater than the temperature-driven effect, resulting in a negative correlation between SST and pCO$_2$ anomalies. In winter, however, the oceanic pCO$_2$ is governed by the temperature-driven effect, yielding a positive correlation.

Another interesting finding from the model analysis is the different responses of nutrients to the same physical processes associated with mesoscale eddies. The vertical displacement of isopycnals and trapping mechanism result in higher/lower NO$_3$ concentration in cyclones/anticyclones with increasing NO$_3$ concentration toward the pole and depth. Although Fe concentration increases with depth, its anomalies have the sign opposite to NO$_3$, suggesting that the vertical displacement of isopycnals does not shape the anomalies. The meridional gradient of Fe is not as clear as NO$_3$, making the response of trapping mechanism complicated.

A hydrodynamic model simulation similar to that described here shows anomalous vertical mixing in mesoscale eddies [Song et al., 2015]. From the present solutions we infer that stronger vertical mixing in deeper mixed layers in anticyclones supplies Fe from the subsurface, resulting in a positive anomalies. Cyclones are characterized by weaker vertical mixing, shallower mixed layers, and negative Fe anomalies. The footprint of vertical mixing modulation by eddies, however, is not apparent in NO$_3$ anomalies. The two nutrients differ in their responses to the same vertical mixing as their vertical gradients at the base of the surface mixed layer are significantly different. As Fe is the limiting nutrient in the most of Southern Ocean, surface depletion creates a greater vertical gradient than for NO$_3$. NO$_3$ on the other hand, is not fully utilized near the surface, which reduces its vertical gradient. As a result, the same vertical mixing creates vastly different fluxes for these two nutrients, bringing about mesoscale anomalies of opposite sign.

The observations show that temperature-driven and DIC-driven pCO$_2$ are out of phase and compensate one-another, resulting in smaller seasonal variability of pCO$_2$. The compensation also occurs at the mesoscale and shapes the seasonality of mesoscale modulation on CO$_2$ flux. The strong seasonality of mesoscale variability of CO$_2$ flux suggests that an important mesoscale signal may be lost in annual averages.

Appendix A: Anomaly Computation

The anomalies of SST and pCO$_2$ were obtained after removing the linear trend of each crossing across the Drake Passage in this study. This separates an anomaly pattern from the sampled SST data if the meridional trend can be depicted by a linear function (Figure A1a). However, this technique may not be optimal for pCO$_2$ whose meridional gradient is not always linear. Indeed, the pCO$_2$ climatological mean distribution across the Drake Passage is not linear (Figure A1b). We therefore experimented by removing a quadratic trend and monthly mean.

The seasonality of cross correlation coefficients is robust regardless the methods for anomaly computation. Correlation coefficients between SST and pCO$_2$ remain negative in austral summer and positive in austral
winter (Figure A1c). The seasonality becomes weak when the monthly mean was removed to compute the anomaly, but the timing when the coefficient changes sign remains the same. The anomaly computation for model data is explained in Appendix C.

Appendix B: Biogeochemical Model

The carbon cycle is simulated using a simple biogeochemical model, the DIC package in MITgcm. The CO₂ air-sea flux \( F_{CO2} \) acts as a source/sink term in the equation for dissolved inorganic carbon (DIC) in the model.

\[
\frac{\partial {\text{DIC}}}{\partial t} = - \nabla \cdot (u {\text{DIC}}) + \kappa \frac{\partial^2 {\text{DIC}}}{\partial z^2} + F_{CO2} + R_{CN} S_{NO3} + S_C. \tag{B1}
\]

where \( u \) is a vector that represents the three-dimensional velocity field, \( \kappa \) is the vertical diffusivity, \( R_{CN} \) is the Redfield ratio between carbon and nitrogen, \( S_{NO3} \) is the contribution from biological production and remineralization, and \( S_C \) is the source term associated with calcium carbonate flux. The vertical mixing of the biogeochemical variables is estimated by the nonlocal K-Profile Parameterization (KPP) scheme [Large et al., 1994].

The biological uptake of inorganic nutrients is parameterized as a function of light, NO₃ and Fe.

\[
B = \alpha \frac{l}{l + k_i} \min \left( \frac{NO_3}{NO_3 + k_{NO3}}, \frac{Fe}{Fe + k_{Fe}} \right) \tag{B2}
\]

where \( \alpha \) is maximum uptake rate, \( k_i \), \( k_{NO3} \) and \( k_{Fe} \) are the half saturation values for light \( l \), NO₃ and Fe, respectively. The uptake of inorganic matter is subsequently divided between a dissolved organic pool in the surface waters and the fraction exported to depth and remineralized. \( B \) can thus be likened to “community production” as it includes the impact of herbivores as well as primary producers. This is similar, but distinct from primary production which only includes the impact of autotrophs.

The light decays exponentially with the e-folding scale of \((k + Chl/k_{Chl})\) where \( k \) is the light attenuation coefficient by water molecules, \( Chl \) is the chlorophyll concentration and \( k_{Chl} \) accounts for the absorption of
light by chlorophyll [Song et al., 2016]. We used the annual mean of SeaWiFS chlorophyll concentration for Chl. Fe has two sources: aeolian dust from the data set by Luo et al. [2008], and the sediments parameterized following Elrod et al. [2004]. A list of parameter values is given in Table 1.

Appendix C: Eddy Centric Analysis

The first step for the eddy centric analysis is to detect eddies. We closely followed the procedure presented in Chelton et al. [2011] and Song et al. [2015] for the identification of well-formed eddies using the SSH field. The detection algorithm searches for the closed SSH contours and determines whether they are eddies by going through the criteria given in Song et al. [2015]. The detected eddies and their trajectory are plotted in Figure C1a.

For each eddy, we estimate the radius $L_s$ whose circle has the same size as the closed SSH contour with the maximum mean current speed. Then $V$, the subset of a variable of interest ranging from $-2.5 L_s$ to $2.5 L_s$ around the eddy center, is extracted and its anomaly $V_0$ is computed as $V_0 = V - \bar{V}$, where $\bar{V}$ represents the regression plane. The mean $L_s$ is approximately 45 km, hence the mean subset size is $225 \text{ km} \times 225 \text{ km}$. The anomaly $V_0$ is mapped onto the eddy-centric coordinate spanning $\pm 2 L_s$, and 4 composite averages were then computed: anticyclones and cyclones in austral summer (December - March) and austral winter (June - September).

Figure C1. (a) Biological production simulated by the biogeochemical model and (b) the chlorophyll biomass from the SeaWiFS data set averaged over December and January. The Patagonian shelf region with black dots in Figure C1a is where biological production is limited by NO3 in the model. Light blue and red dots represent the initial location of cyclonic and anticyclonic eddies who live more than 20 days, respectively, and lines are their trajectories.
Appendix D: Model Evaluation

Although the biogeochemical model for carbon flux is relatively simple, it captures the observed large scale features of biogeochemical ocean states. The zonal mean of DIC and NO3 are close to World Ocean Atlas 2009 climatology (Figure 5). The simulated Fe level is also comparable to the observations upstream and across the Drake Passage (not shown).

Net community productivity displays a similar spatial pattern as that of the chlorophyll biomass measured from satellite. The model shows high production over the Patagonian Shelf and near the Antarctic in austral summer, which is in a good agreement with observed chlorophyll biomass spatial distribution (Figure C1). The simulated community activity is mainly limited by Fe in most areas of the model domain, except over the Patagonian shelf where NO3 is the limiting nutrient. This is consistent with the observations presented in Paparazzo et al. [2010].

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