

## MOTIVATION

### Previous work indicates:

- The North Atlantic is an important carbon sink, absorbing 23% of anthropogenic carbon stored in the oceans (Sabine et al., 2004), yet basin-integrated CO<sub>2</sub> flux variability is small (McKinley et al., 2004).
- Changes in uptake (Shuster and Watson, 2007) and interior carbon storage (Feely et al., 2005) may be significant.
- Variability in the spring bloom has a complex relationship to physical

variability (Follows and Dutkiewicz, 2002) and biological impacts on carbon fluxes are poorly understood.

### We ask:

- What is the sign and magnitude of recent variability and trends in the North Atlantic carbon sink? What are the dominant controls?
- Does biological variability have a significant impact on the N. Atlantic carbon sink?

## MODEL DESCRIPTION

### PHYSICAL MODEL:

- MITgcm (Marshall et al. 1997a,b) configured for North Atlantic domain
- Horizontal resolution of 0.5° x 0.5°, 23 vertical levels ranging in resolution from 10m near surface to 500 m at depth
- KPP mixing layer scheme, GM-Redi representation of isopycnal mixing

### BIOGEOCHEMICAL MODEL:

- Cycling of carbon, phosphorus, silica, iron, oxygen and alkalinity
- Ecosystem model includes 2 phytoplankton functional types, one zooplankton type and explicit DOM and POM (Dutkiewicz et al., 2005)
- Surface carbonate chemistry (Follows et al., 2006)
- Iron chemistry (Parekh et al., 2004)
- Air-sea exchange of carbon dioxide and oxygen (Wanninkhof, 1992)
- Atmospheric pCO<sub>2</sub> from Mauna Loa observations

## MODEL VERIFICATION

- Run 1 - Spin up: 80 years physics only (daily NCEP)  
26 years physical-biogeochemical model (daily NCEP)  
Interannual: 1992 to 2006 (daily NCEP, relaxation to satellite-derived SST (Reynolds et al, 2002))
- Run 2 - Spin up: 81 years physics only (daily NCEP)  
20 years physical-biogeochemical model (daily NCEP)  
Interannual: 1981 to 2006 (daily NCEP)
- In both runs, the model simulates the circulation, sea surface temperature, and mixed layers reasonably well.
- The ecosystem model captures the timing and magnitudes of the North Atlantic bloom. The biomass in the subtropics is, however, too low, as is typical of this class of models.
- Model pCO<sub>2</sub> compares favorably to that observed by Takahashi et al. (2002). The model also captures the pCO<sub>2</sub> components due to temperature and non-temperature effects (Figure 1).
- Comparison to observations at Bermuda (Bates, 2007) indicate the model is capturing variability in surface pCO<sub>2</sub> quite well.

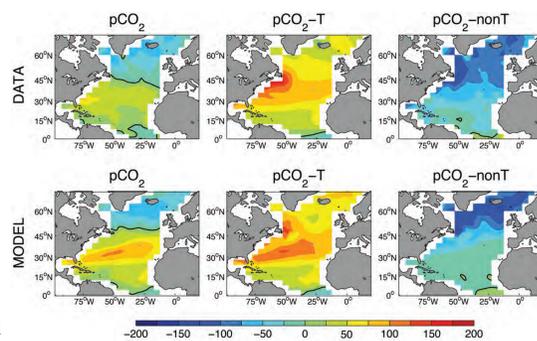


Figure 1: Amplitude of North Atlantic pCO<sub>2</sub> cycles (ASO-FMA). Takahashi et al. (2002) climatology (top) and MODEL (bottom). Temperature (pCO<sub>2</sub>-T) and non-temperature (pCO<sub>2</sub>-nonT) components are calculated following Takahashi et al., 2002. (Run 2)

## CO<sub>2</sub> FLUX TREND, 1992-2006

- The basin-scale air-sea flux is increasing. The flux is also increasing in each gyre.
- Basin-wide, for 1992 and 2006, we find a sink increase of 0.09 PgC/yr/decade (Run 1). For the same period, Run 2 has a larger trend of 0.2 PgC/yr/decade.
- Results are consistent with the latest data-based estimate (Takahashi et al., 2002).

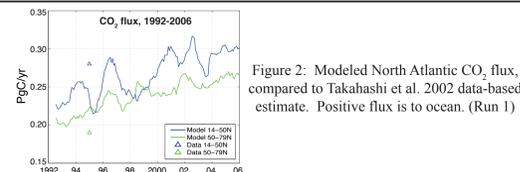


Figure 2: Modeled North Atlantic CO<sub>2</sub> flux, compared to Takahashi et al. 2002 data-based estimate. Positive flux is to ocean. (Run 1)

## CONTROLS ON pCO<sub>2</sub> VARIABILITY

### Subpolar Gyre: DIC MIXING

- The negative correlation between pCO<sub>2</sub> and SST indicates colder SST is associated with more mixing, and vice-versa.
- Mixing modifies DIC, driving pCO<sub>2</sub> change.
- ↑SST → ↓MLD → ↓ upwelling DIC → ↓ pCO<sub>2</sub>

### Subtropical Gyre: Temperature

- The positive correlation between pCO<sub>2</sub> and SST indicates that SST anomalies directly influence pCO<sub>2</sub> via the thermodynamic effect on gases.
- ↑SST → ↑ pCO<sub>2</sub> (thermodynamic effect on gases)

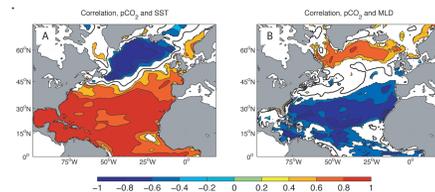


Figure 3: Correlation plots (a) pCO<sub>2</sub> and SST, (b) pCO<sub>2</sub> and MLD. Based on fields detrended and 12-month smoothed. Shaded regions are significant at the 95% confidence level. (Run 1)

## DOMINANT PATTERNS OF pCO<sub>2</sub> TREND

- The first EOF of the pCO<sub>2</sub> (Figure 4a) explains 73% of the interannual variance, and its PC1 (Figure 4, inset) is characterized by a steadily increasing trend.
- EOF1, the dominant pattern of pCO<sub>2</sub> trend, illustrates greater increase of pCO<sub>2</sub> in the east and south (5-10 μatm), and slower increase in the west (0-5 μatm). These changes are consistent in sign with recent changes observed in-situ (Schuster and Watson, 2007) though smaller in magnitude.
- Regression of pCO<sub>2</sub> components (pCO<sub>2</sub>-ALK, pCO<sub>2</sub>-DIC, pCO<sub>2</sub>-T) onto PC1 unveils localized regions of high component variability
- Warming and freshening in the Labrador Sea waters increase pCO<sub>2</sub>, while the associated decrease in DIC decreases pCO<sub>2</sub>.
- There is evidence of modified advection of high-DIC and high-ALK eastern subtropical waters along the NAC into the western subpolar region, consistent with observed spin-down of the subpolar gyre (Häkkinen and Rhines, 2004).

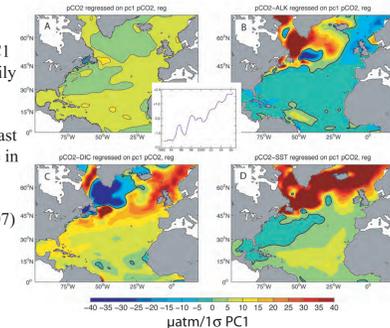


Figure 4: (a) EOF1 of pCO<sub>2</sub>, and regression of (b) pCO<sub>2</sub>-ALK, (c) pCO<sub>2</sub>-DIC, and (d) pCO<sub>2</sub>-T onto PC1 (inset). (Run 1)

## BLOOM DOMINATES pCO<sub>2</sub> VARIABILITY

- Interannual variability of pCO<sub>2</sub> is highest during the spring/early summer in the subpolar gyre. This variability is driven by year-to-year changes in pCO<sub>2</sub>-nonT (Figure 5) associated with timing of the spring bloom.

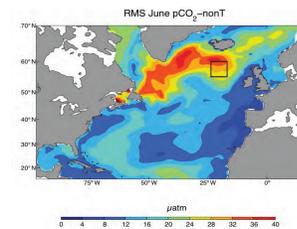


Figure 5: Standard deviation of June pCO<sub>2</sub>-nonT between 1982 and 2006. (Run 2)

- The seasonal cycle of pCO<sub>2</sub> at Iceland (boxed region in Figure 5) has strong, opposing impacts from temperature and DIC-driven influences (Figure 6).

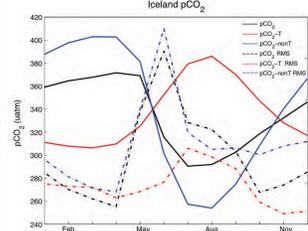


Figure 6: Climatology and RMS of pCO<sub>2</sub> and its temperature and non-temperature components within the 5°x5° box at Iceland (57.5°N, 17.5°W). (Run 2)

- The dominance of biological variability to surface ocean pCO<sub>2</sub> variance suggests that bloom variability may be a first order control on annual CO<sub>2</sub> flux variability in the subpolar gyre.

## ... BUT DOES NOT TRANSLATE TO CO<sub>2</sub> FLUX

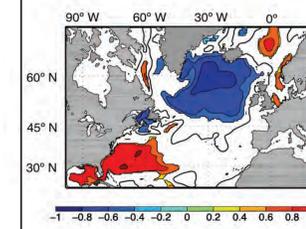


Figure 7: Correlation between anomalies of annual surface DIC and annual CO<sub>2</sub> fluxes. Shaded regions are significant at the 95% confidence level. (Run 2)

- DIC controls interannual variability of pCO<sub>2</sub> in the subpolar gyre and is significantly anti-correlated with annual CO<sub>2</sub> flux variability (Figure 7, see also Figure 3).

- However, change in annual carbon export is not associated with CO<sub>2</sub> flux changes in most of the subpolar gyre (Figure 8); i.e. biological variability is not driving CO<sub>2</sub> flux variability.

- In the small region centered at 60°N, 30°W, there is a correlation between CO<sub>2</sub> flux and export, but it is indirect. Here, enhanced winter mixing increases surface DIC and increases pCO<sub>2</sub>, lowering the flux; and at the same time weakens the bloom.

- Correlations between export and CO<sub>2</sub> flux in the subtropical region are also indirect. Figure 3 illustrates that SST drives pCO<sub>2</sub> change in the subtropics. Colder surface temperatures are associated with more mixing, and thus greater nutrient supply, and vice-versa.

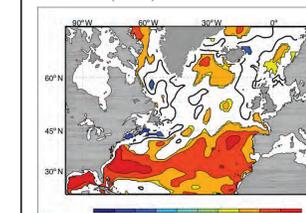


Figure 8: Correlation between anomalies of annual export of POC below 100m and CO<sub>2</sub> fluxes. (Run 2)

## FLUX IMPACT SMALL BECAUSE BLOOM STRENGTH VARIES ONLY 10-25%

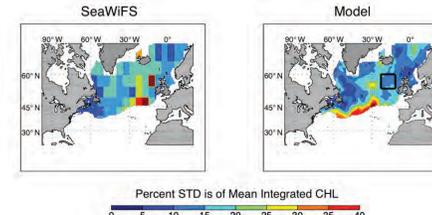
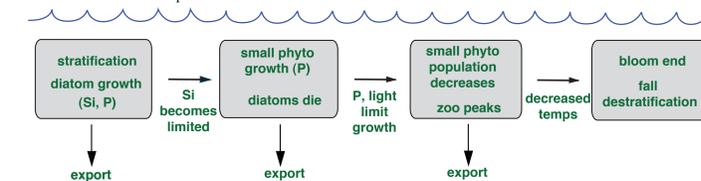


Figure 9: Integrated chlorophyll variability 1998-2006 as a percent of the annual mean chlorophyll SeaWiFS (left) and MODEL (right). (Run 2)

- Annually integrated chlorophyll in the subpolar North Atlantic varies by an average of 15% in the model and 14% in SeaWiFS satellite data between 1998 and 2006 (Figure 9).
- Blooms that start earlier tend to end earlier, as seen in the 25-year timeseries of POC export at Iceland (Figure 10).
- Modeled variability of annually-integrated export is smaller than chlorophyll variability (5-20 %).
- Thus, we find that variability in the biological pump of carbon is only 5-20% across the North Atlantic.

### WHAT LIMITS VARIABILITY IN THE BIOLOGICAL PUMP?

- Blooms in the subpolar region begin with stratification. Diatoms bloom first, using silicate and phosphorous.
- When silicate becomes depleted, the ecosystem becomes dominated by small phytoplankton, using the remaining phosphorous.
- When nutrients and/or light becoming limiting, potential growth is overcome by mortality and grazing. The bloom ends.
- This succession was observed during NABE (Sieracki et al., 1993; Lochte et al., 1993), over one seasonal cycle. We find that the same sequence limits interannual variations.



## SUMMARY

- Interannual variability in pCO<sub>2</sub> and CO<sub>2</sub> flux is controlled by temperature in the subtropics and by DIC mixing in the subpolar gyre (Fig. 3, 7).
- For 1992-2006, pCO<sub>2</sub> across the gyre has increased 0-5 μatm in the west and 5-10 μatm in the east and south. Over the same period, changes in pCO<sub>2</sub>-T, pCO<sub>2</sub>-ALK and pCO<sub>2</sub>-DIC can exceed ±40μatm north of 45° N. (Figure 4)
- Large summertime variability in subpolar gyre pCO<sub>2</sub> (Figure 5, 6) is driven by changes in bloom timing. But, because integrated bloom

strength variability is small (Figure 9, 10), this does not translate to large changes in carbon export (Figure 8).

- We conclude that biological variability is not a first order control on variability in the North Atlantic carbon sink.
- Modeled patterns of pCO<sub>2</sub> trend (Figure 3a), driven physically, help explain trends observed in-situ (Shuster and Watson, 2007).
- Future work: Study of multi-decadal variability over the entire NCEP/NCAR Reanalysis period.

## REFERENCES

Dutkiewicz, S., et al. 2005. Global Biogeochem. Cycles, 19, GB1021, doi:10.1029/2004GB002342.  
 Bates, N. 2007. J. Geophys. Res., 112, C09013, doi:10.1029/2006JC003759.  
 Bennington, V., G. A. McKinley, S. Dutkiewicz and D. Ullman, 2008. submitted to Global Biogeochem. Cycles.  
 Feely, R., L. Talley, G. Johnson, C. Sabine and R. Wanninkhof, 2005. EOS, 86, 42, 404-405.  
 Follows, M.J., and S. Dutkiewicz, 2002. Deep-Sea Res. II, 49, 321-344.  
 Follows, M., T. Ito and S. Dutkiewicz, 2006. Ocean Modeling, 12, 290-301.  
 Feely, R.A. et al., 2005. EOS 86, 399, 404-405.  
 Häkkinen, S. and P.B. Rhines, 2004. Science, 304, 555-559.  
 Lochte, K., H.W. Ducklow, M.J.R. Fasham and C. Stienen, 1993. Deep Sea Research II, 40, 91-114.  
 Marshall, J. C., C. Hill, L. Perelman, and A. Adcroft, 1997a. J. Geophys. Res., 102, 5733 - 5752.  
 Marshall, J. C., A. Adcroft, C. Hill, L. Perelman, and C. Heisey, 1997b. J. Geophys. Res., 102, 5753 - 5766.  
 McKinley, G.A., M.J. Follows, and J. Marshall, 2004. Global Biogeochem. Cycles, 18, GB2011, doi:10.1029/2003GB002179.  
 Parekh, P., M.J. Follows and E.A. Boyle, 2004. Global Biogeochem. Cycles, 18, GB1002, doi:10.1029/2003GB002061.  
 Reynolds, R.W., N.A. Rayner, T.M. Smith, D.C. Stokes, and W. Wang, 2002. J. Climate, 15, 1609-1625.  
 Sabine, C., et al., 2004. Science, 305, 365-371.  
 Schuster, U., and A.J. Watson, 2007. J. Geophys. Res., 112 (C11), C11006, doi:10.1029/2006JC003941.  
 Sieracki, M.E., P.G. Verity and D.K. Stoeker, 1993. Deep Sea Research II, 40, 213-225.  
 Takahashi, T., et al., 2002. Deep-Sea Res. II, 49, 1601-1622.  
 Ullman, D., 2008. Masters Thesis, University of Wisconsin-Madison.  
 Wanninkhof, R., 1992. J. Geophys. Res., 97, 7373-7382.