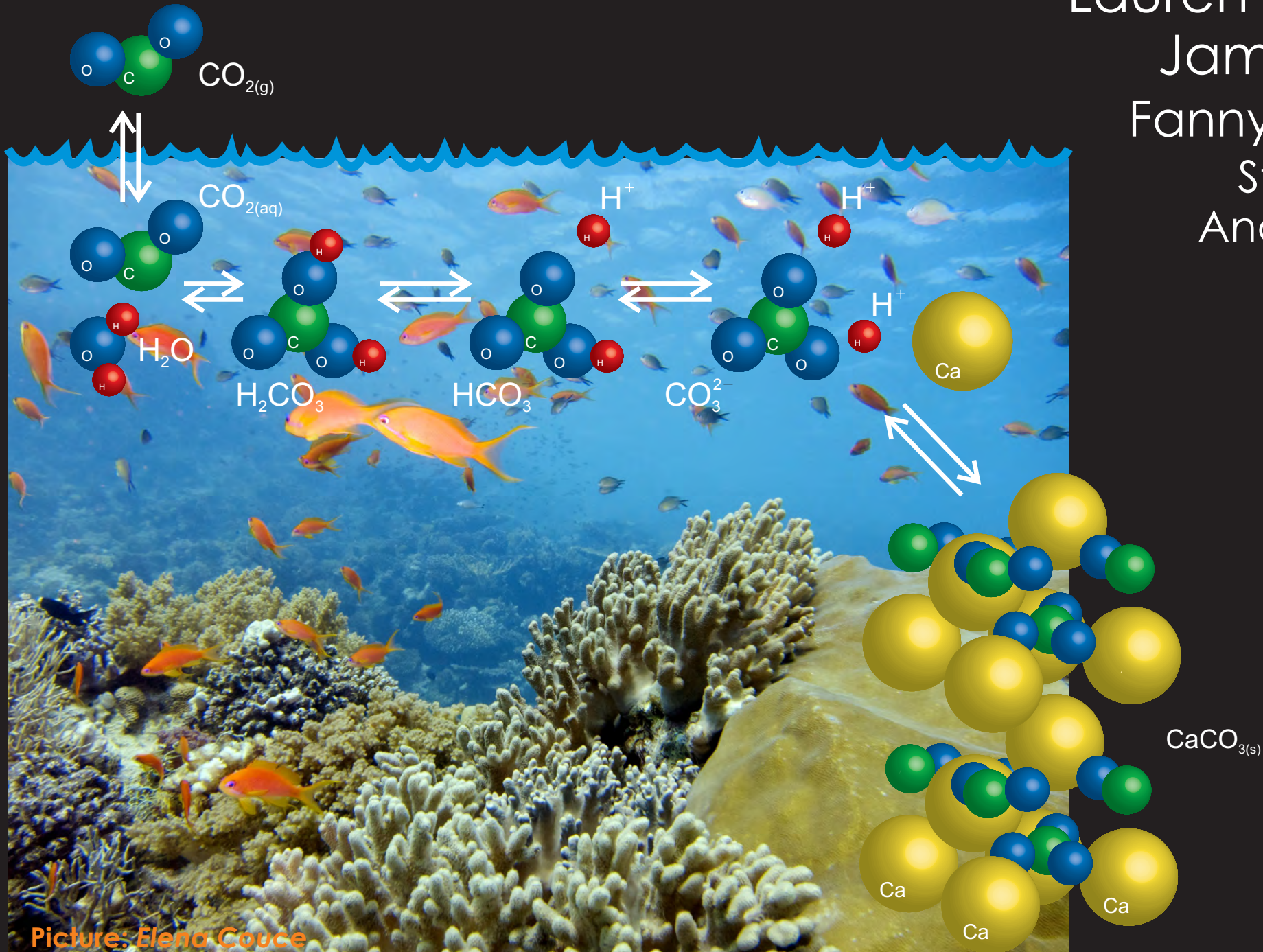


CO₂-Carbon Cycle-Climate-Interactions C⁴I

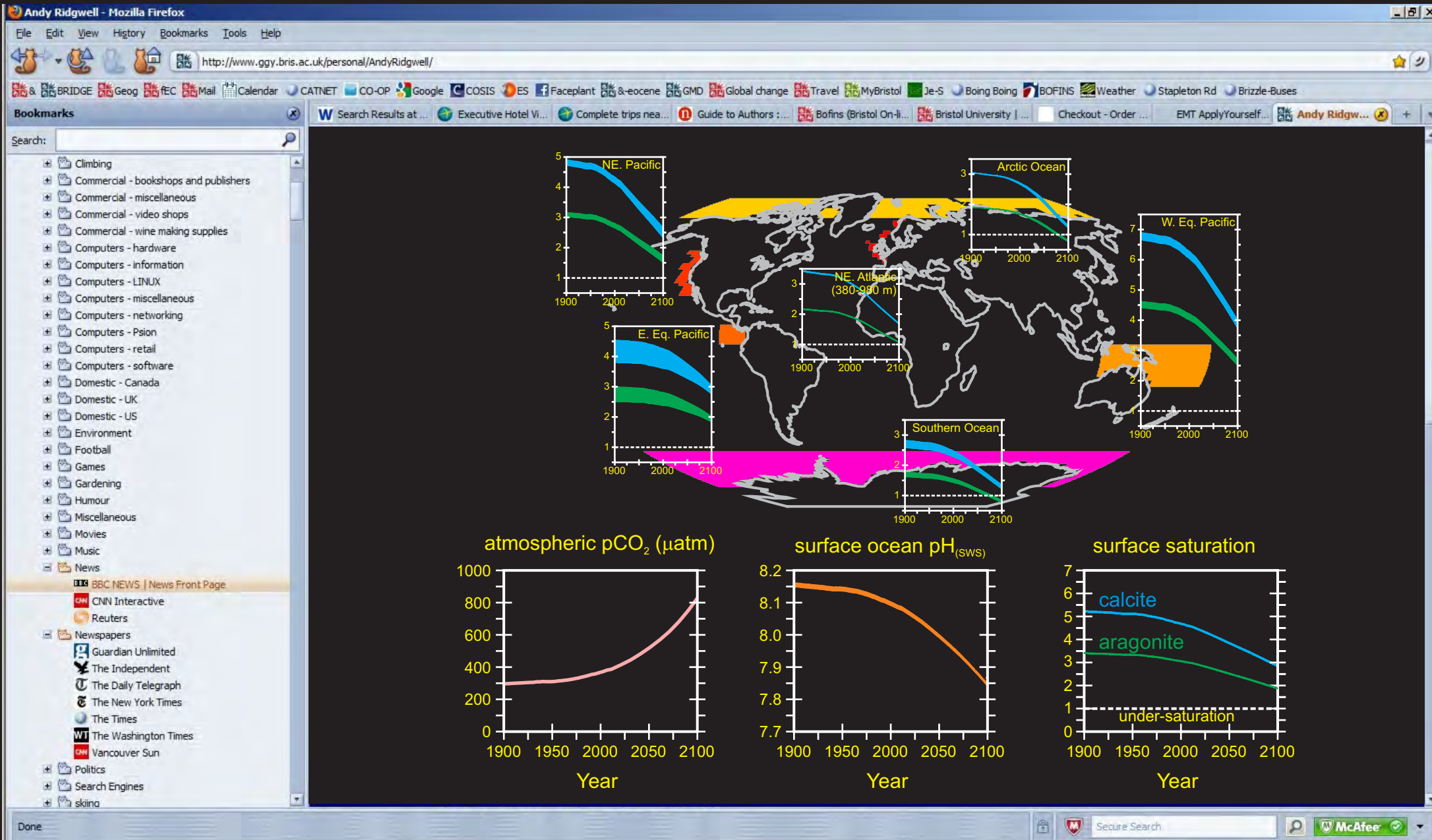
Lauren Gregoire
Jamie Wilson
Fanny Monteiro
Steve Barker
Andy Ridgwell



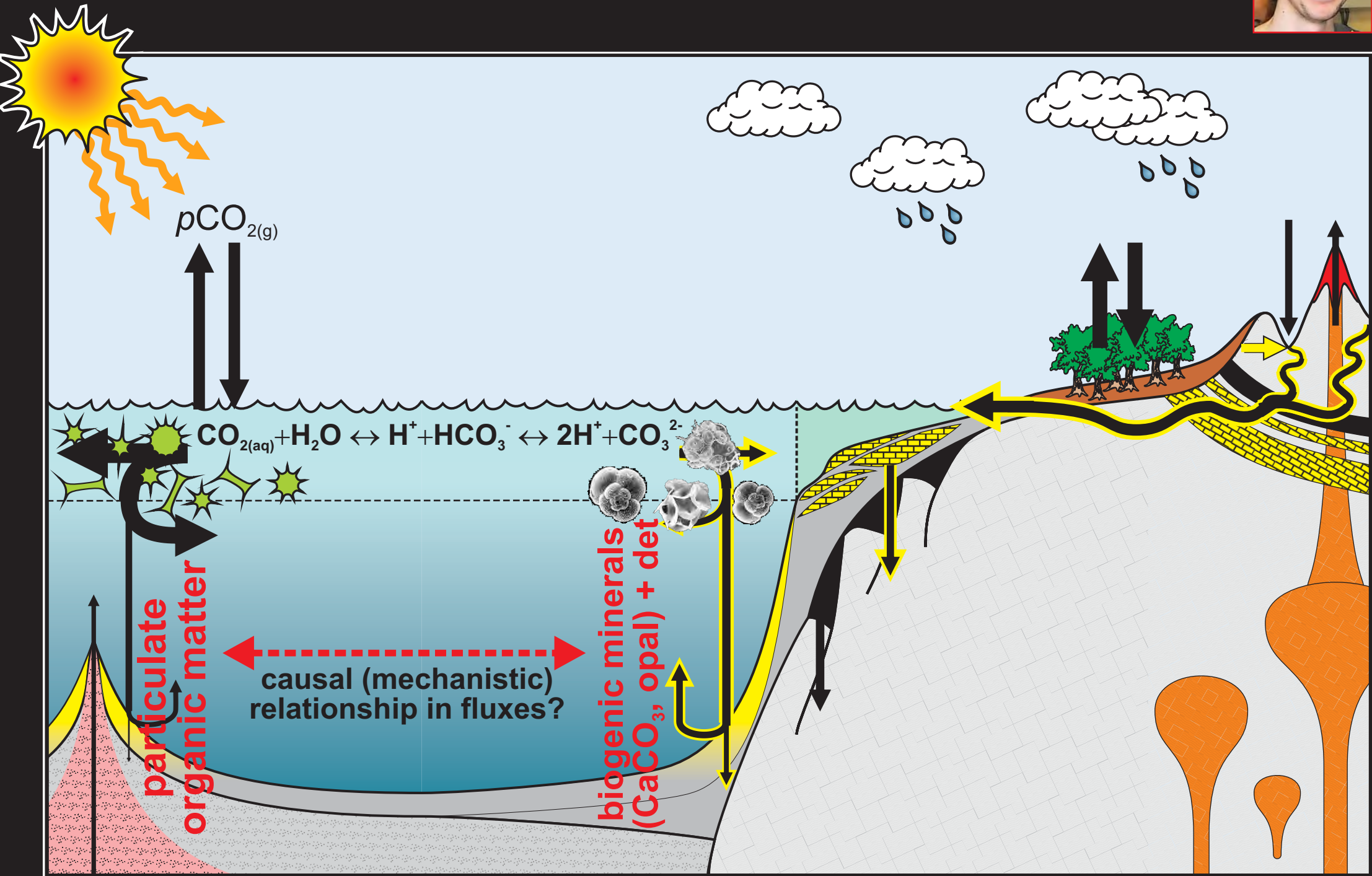
Picture: Elena Couce



projecting global (bio)geochemical impacts (and the 'ocean acidification viewer')



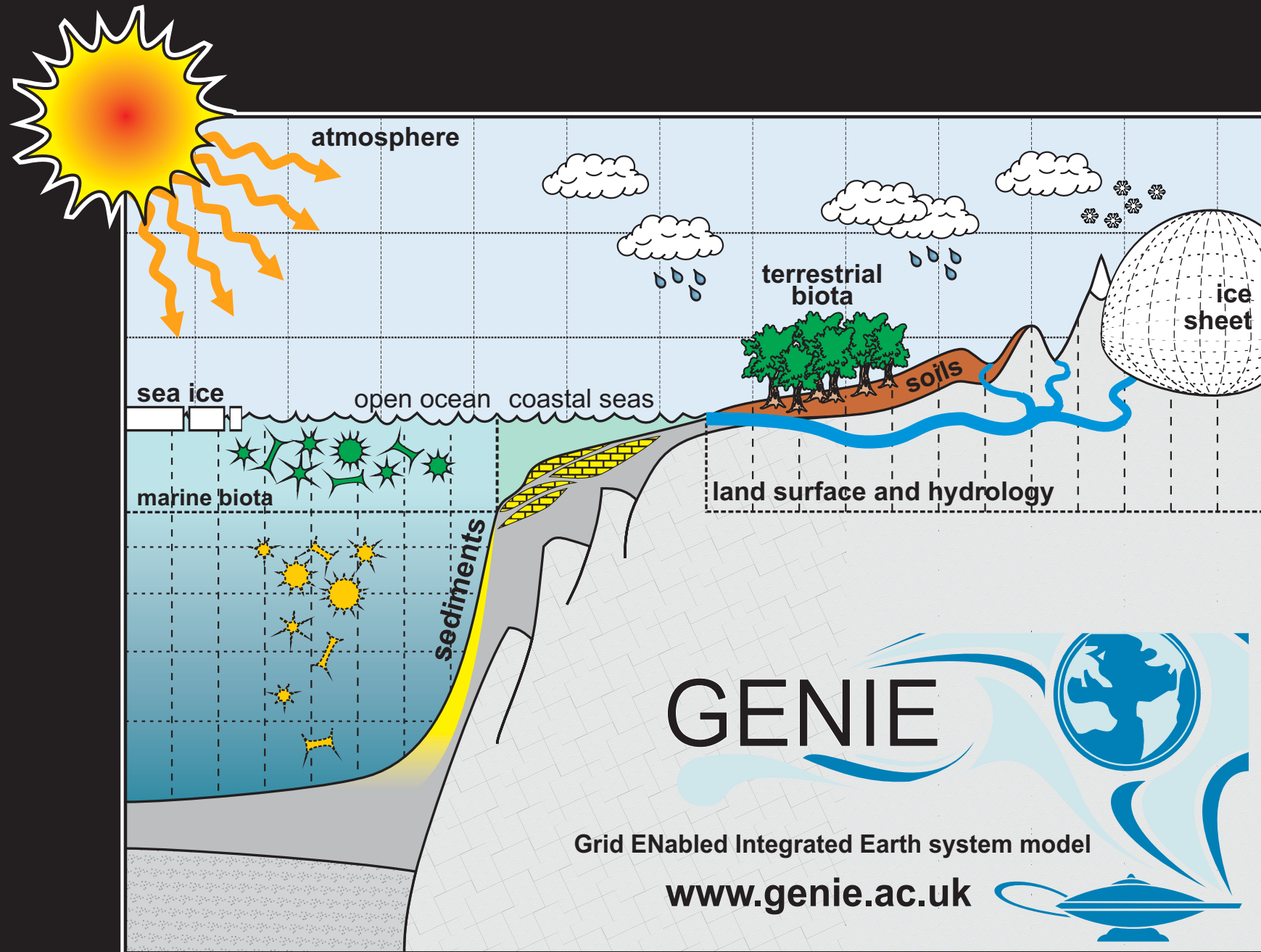
The biological pump and mineral 'ballasting'



'Hands on' OA (Earth system) modelling workshop(s)

mid/late summer 2012/13? ... TELL ME!

(also: carbonate chemistry calculation tutorials?)



UKOA-AVA activity:

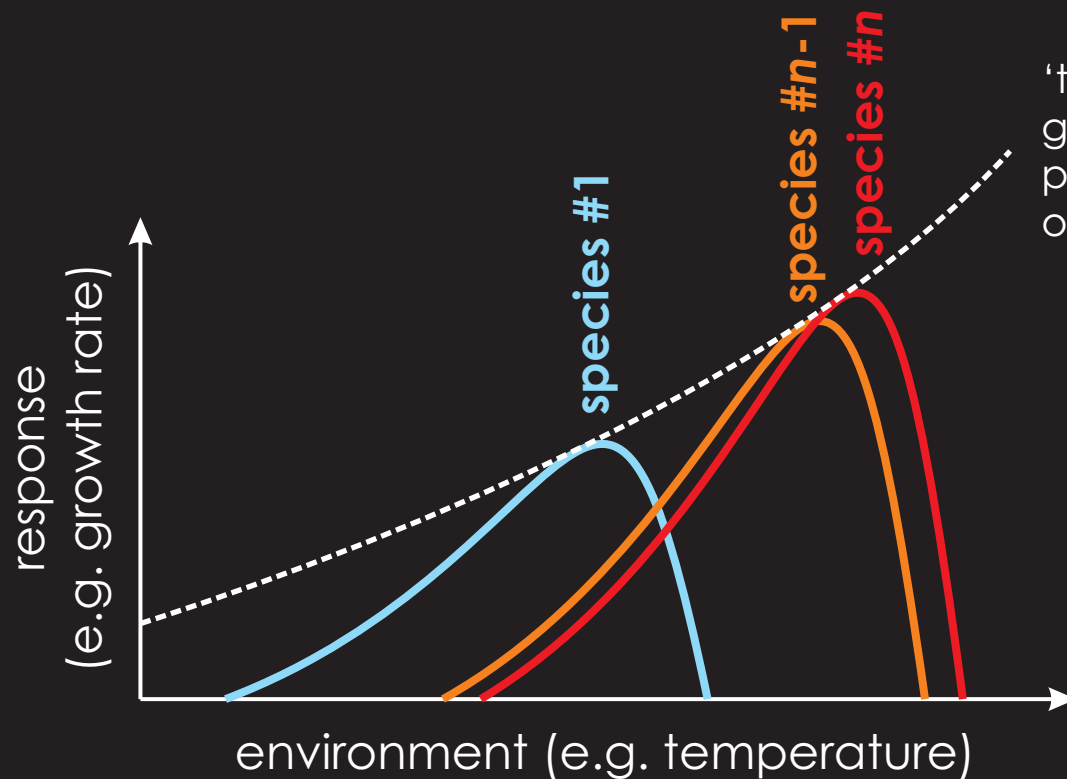
'Modelling the ecology of coccolithophores in the ocean'



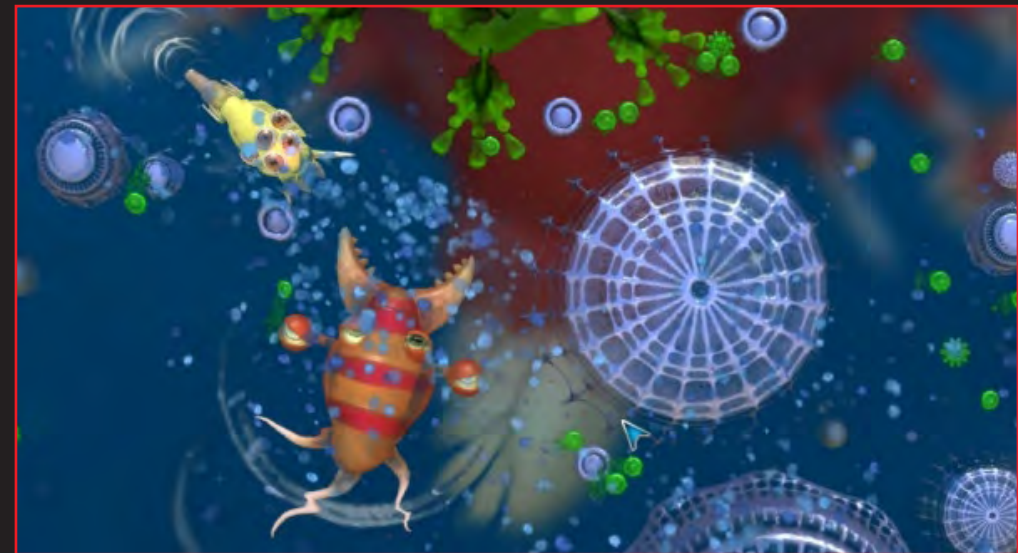
2-day workshop to bring together:
observationalists+experimentalists+geneticists+evolutionary 'historians'
(geologists?) and ... computer modellers, in conjunction with explicit
hypothesis-testing in an advanced global ecosystem model 'framework'

13+14th September 2012

(organiser: Dr. Fanny Monteiro)



'traditional' (ca. AR5) GCM
generic parameterized
physiological response
of bulk plankton biomass



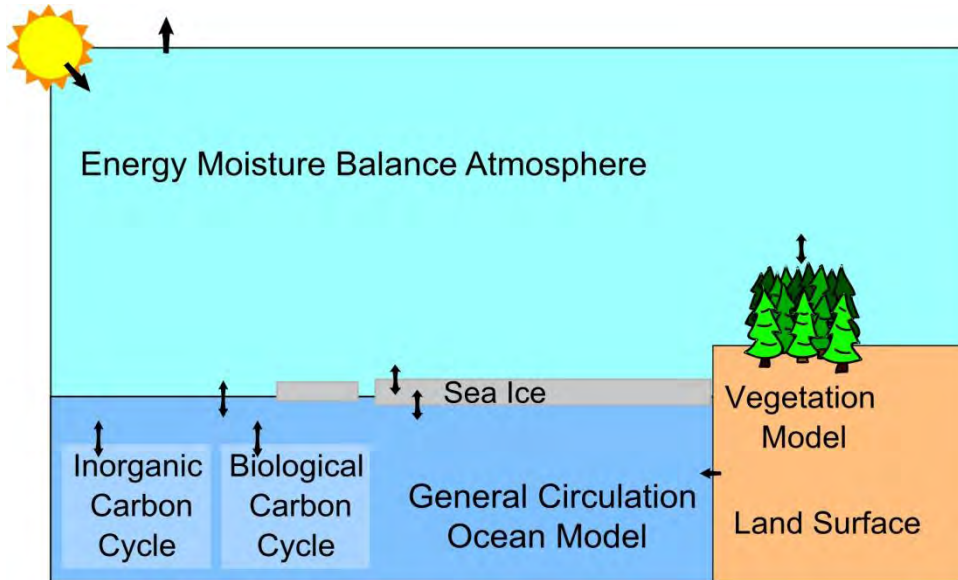


‘Climatology’ of ocean acidification for a range of CO₂ emission scenarios

Lauren J Gregoire, Andy Ridgwell, Chris Vernon, Michael Eby



Uvic-ESCM



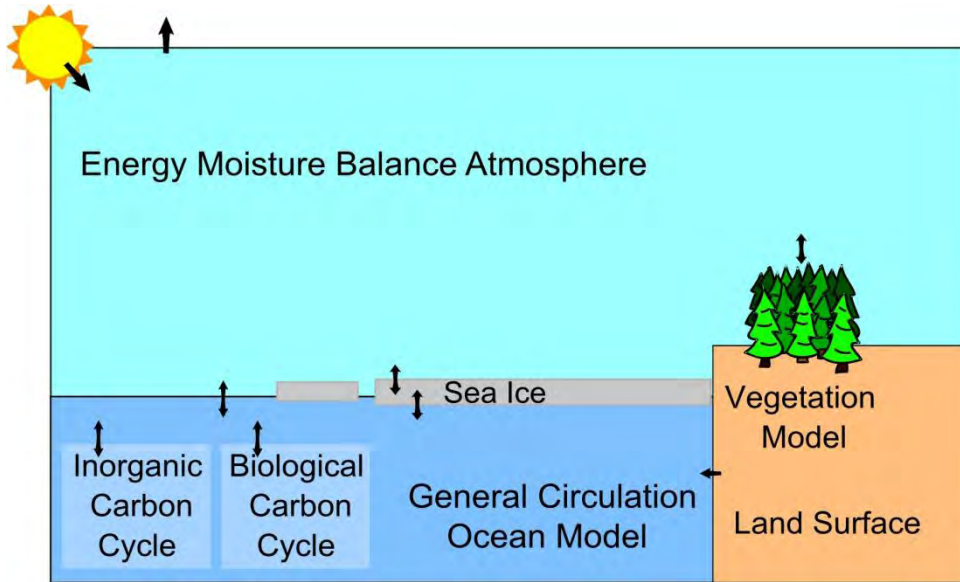
Global model of the Earth system

Resolution: 3.6° longitude x 1.8° latitude

19 levels in the ocean

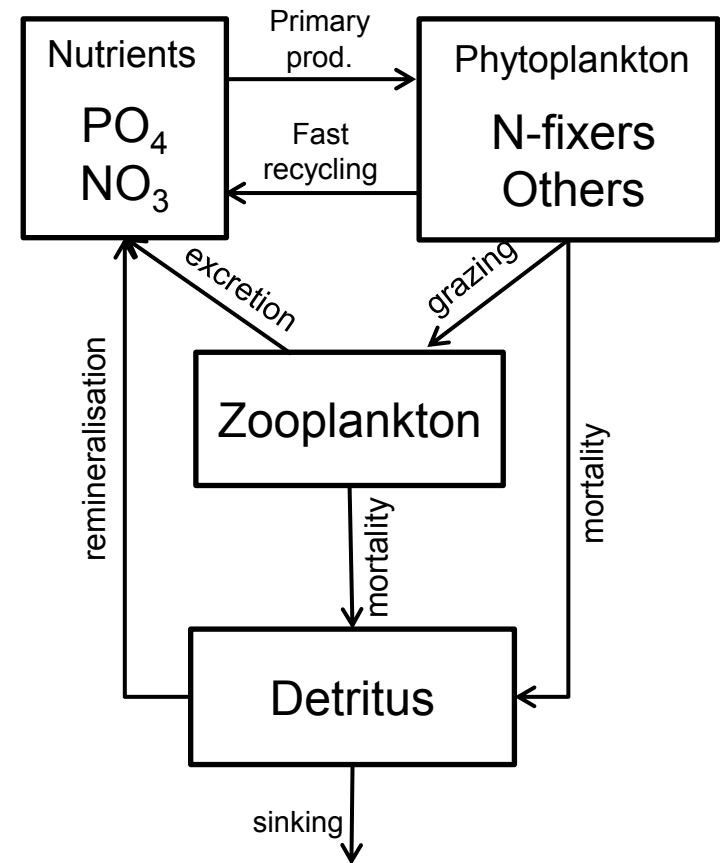
Can solve seasonal cycle but not interannual variability

Uvic-ESCM

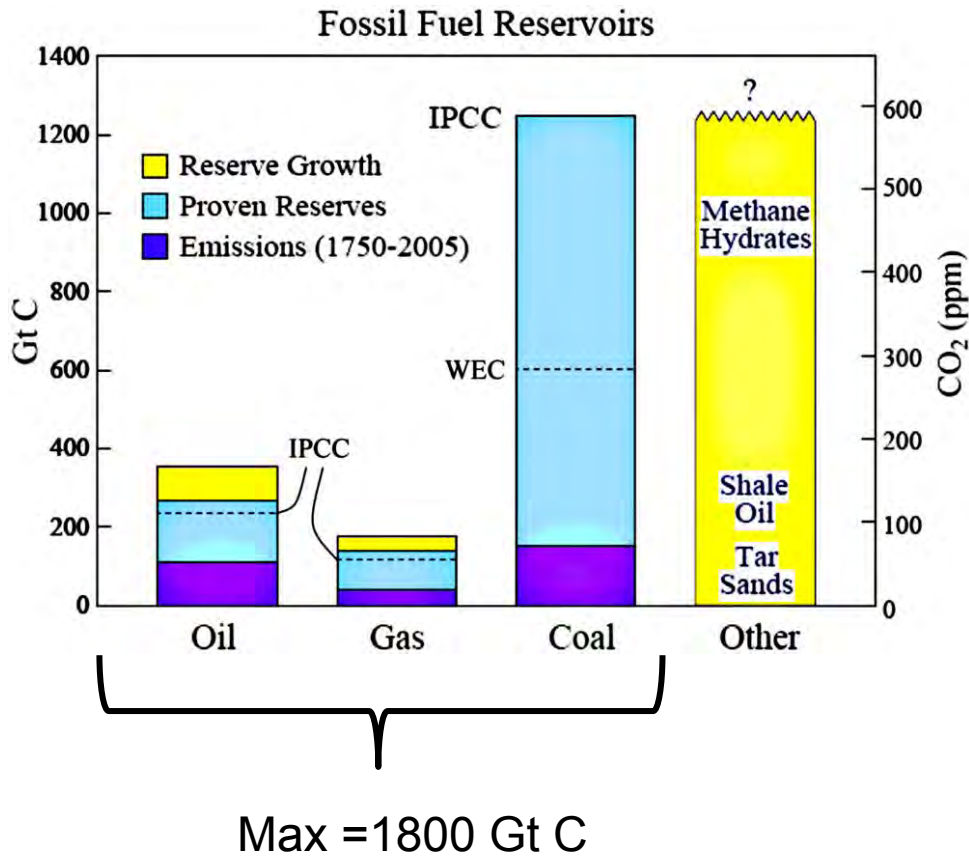


Global model of the Earth system
Resolution: 3.6° longitude x 1.8° latitude
19 levels in the ocean
Can solve seasonal cycle but not interannual variability

NPZD model

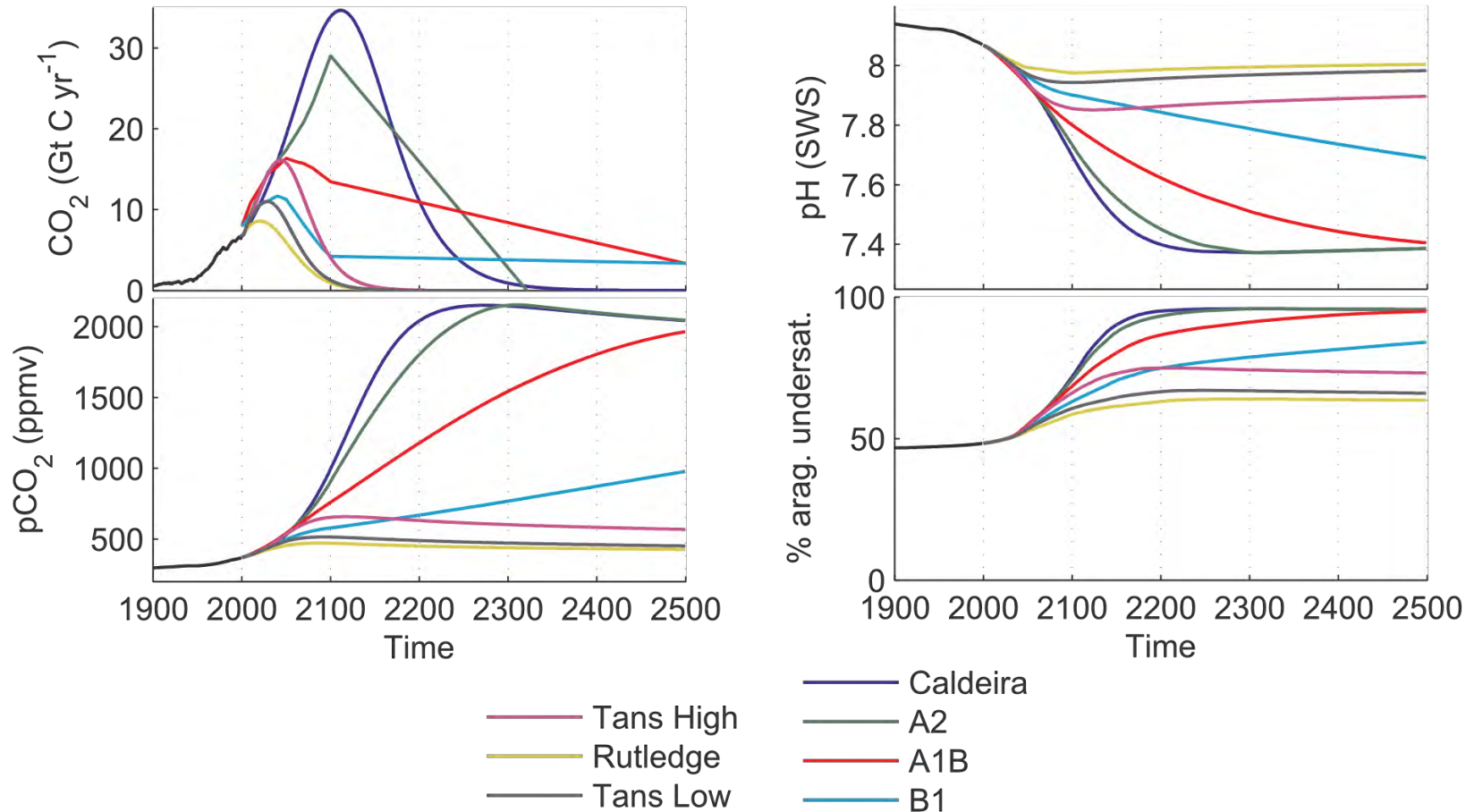


CO₂ emission scenarios

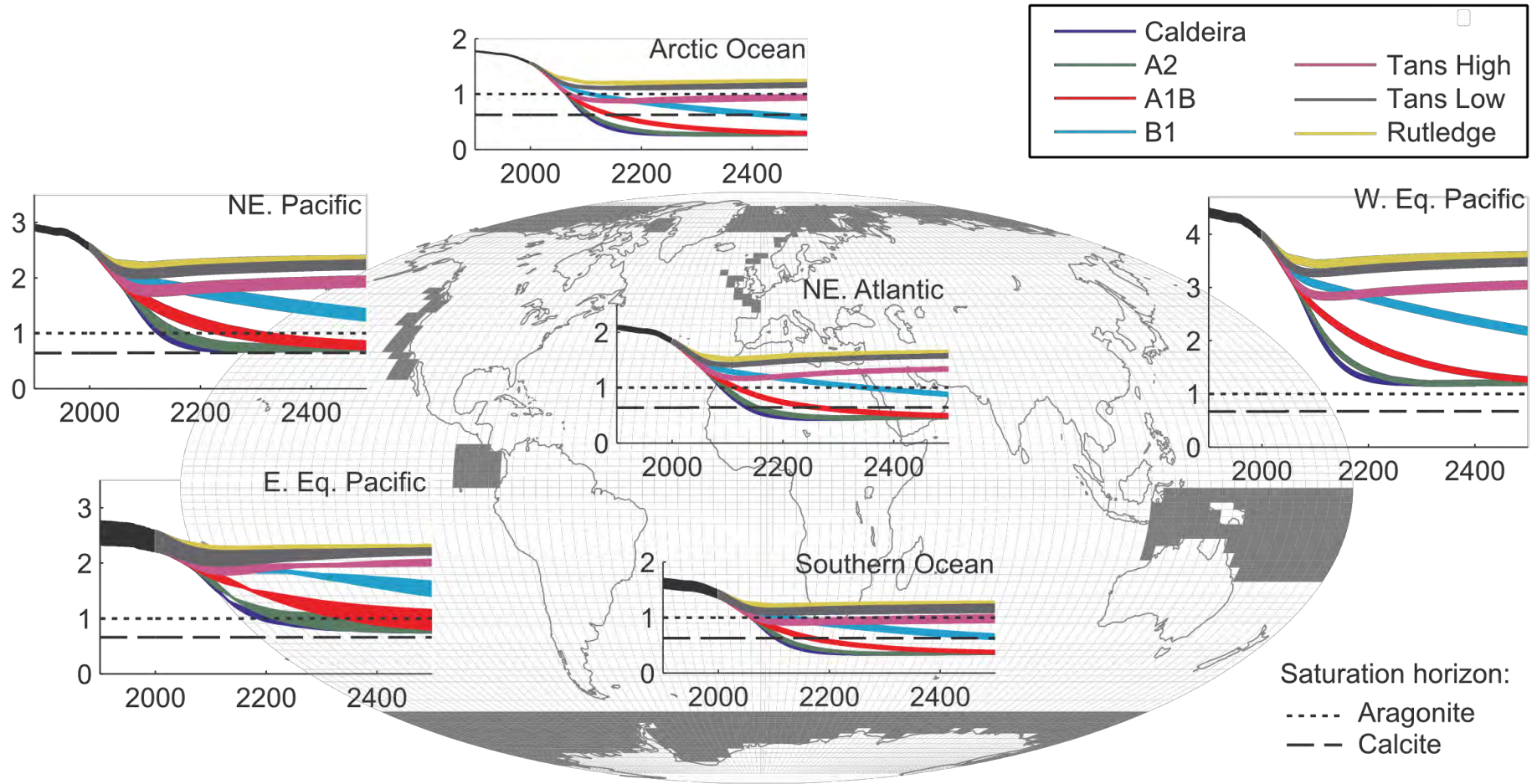


Scenarios	Cummulative emissions / Gt C	
Demand-driven	Caldeira	5000
	A1B	5000
	A2	5000
Supply-limited	B1	5000
	Tans High	1225
	Tans low	725
	Rutledge	550

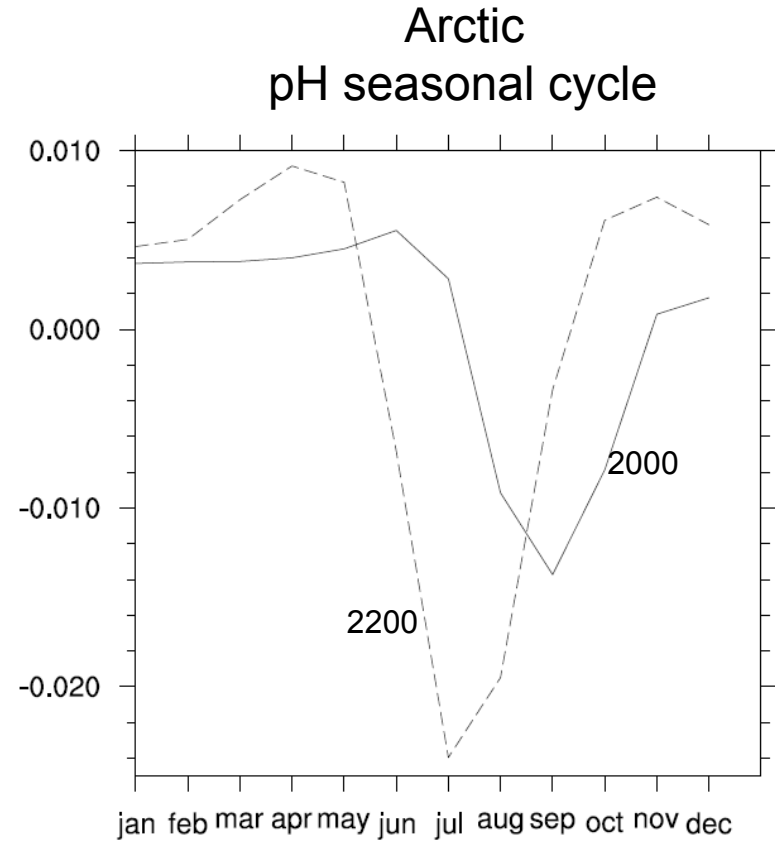
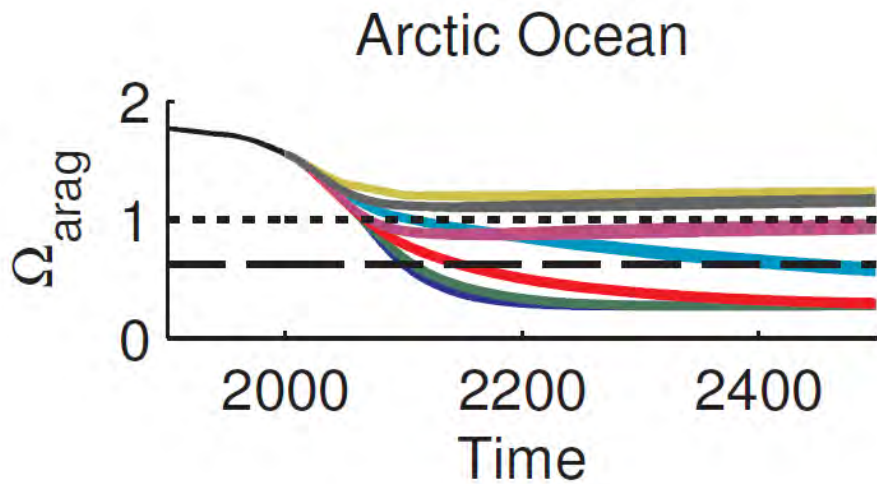
Projections of ocean acidification



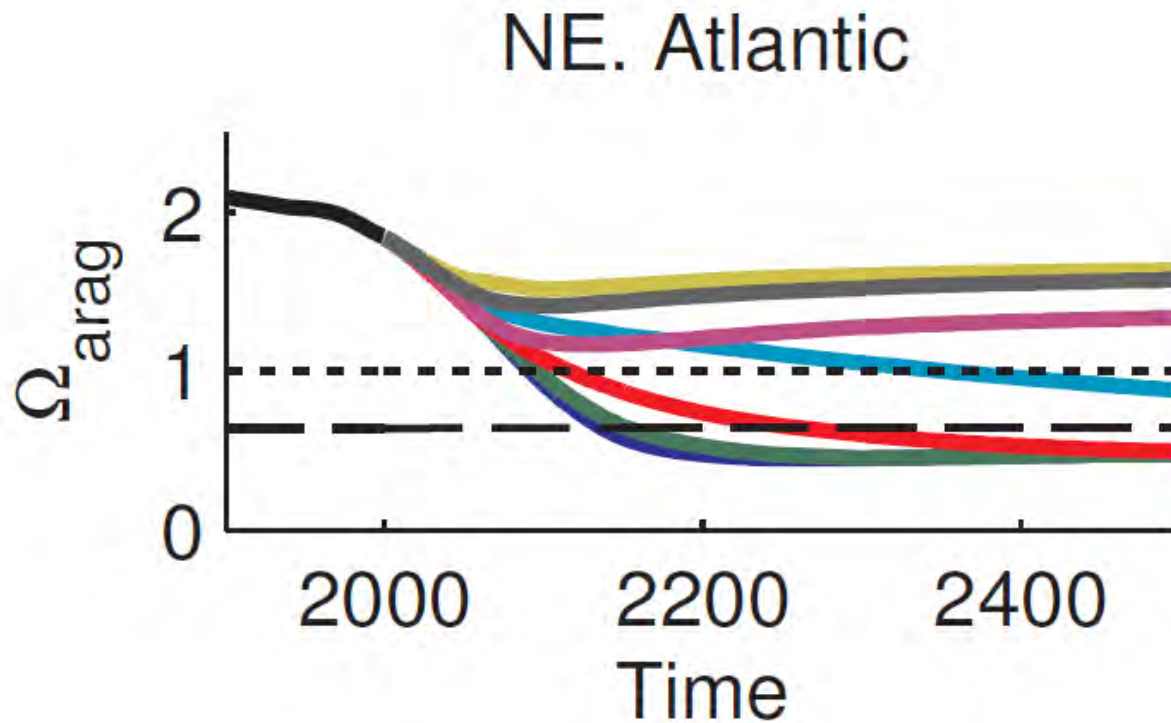
🌿 Aragonite saturation by region



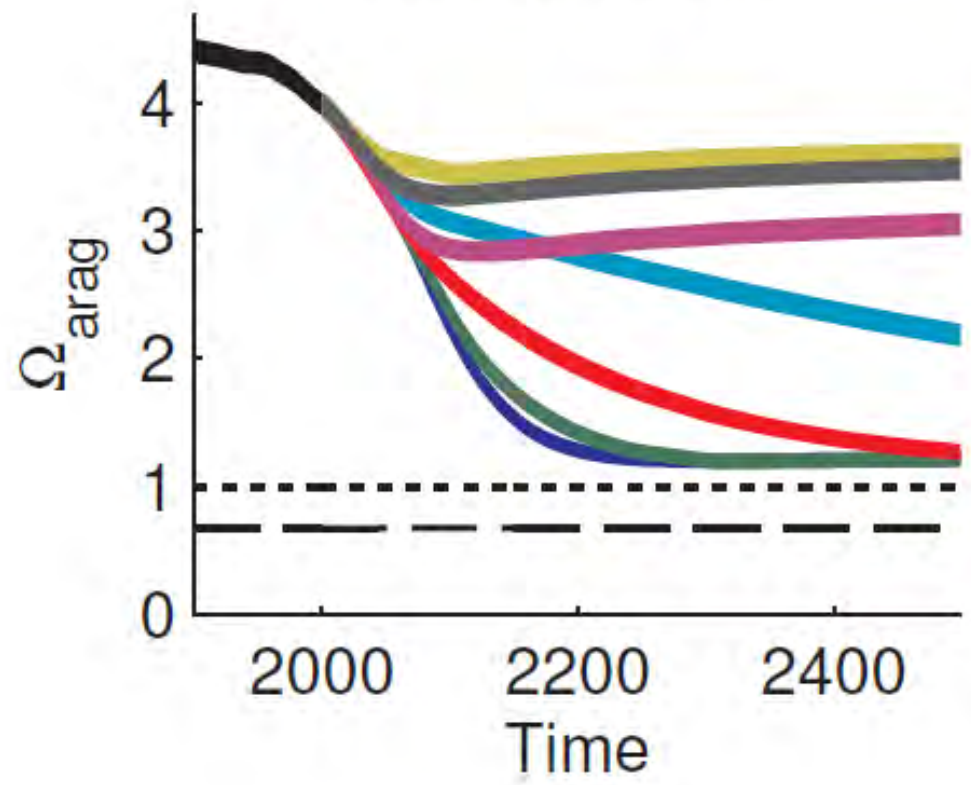
High latitudes



NE Atlantic ocean floor

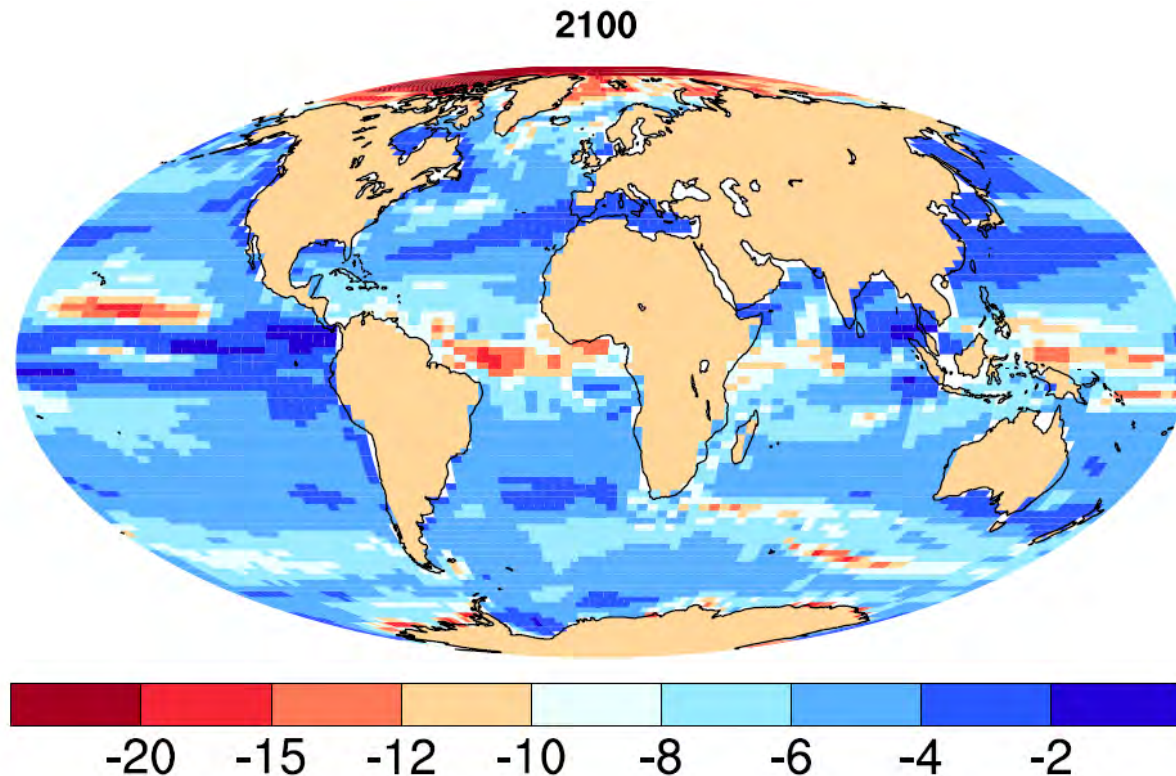


W. Eq. Pacific



pH change / pH variability

Scenario A1B



Ocean acidification viewer

- Functionalities:
 - Plot :
 - Time series: global mean or local
 - Maps of annual mean or specific month
 - Depth profiles at specific location
 - Download 4-D data (lat, lon, depth, time) in netcdf format
 - Download subset of data in text format
- Feedback wanted



Summary

- Future projections of ocean acidification simulated by UVic model, with a range of future scenarios
- Arctic and Southern ocean reach calcite undersaturation in all but two scenarios.
- East equatorial pacific subject to large changes in carbonate chemistry compared to the current variability
- Shifts in seasonality of carbonate chemistry
- Ocean acidification viewer





Assessing the spatial variability in ballasting statistics

Jamie Wilson

(wilsonjd@cardiff.ac.uk)

School of Earth and Ocean Sciences, Cardiff University

Stephen Barker & Andy Ridgwell



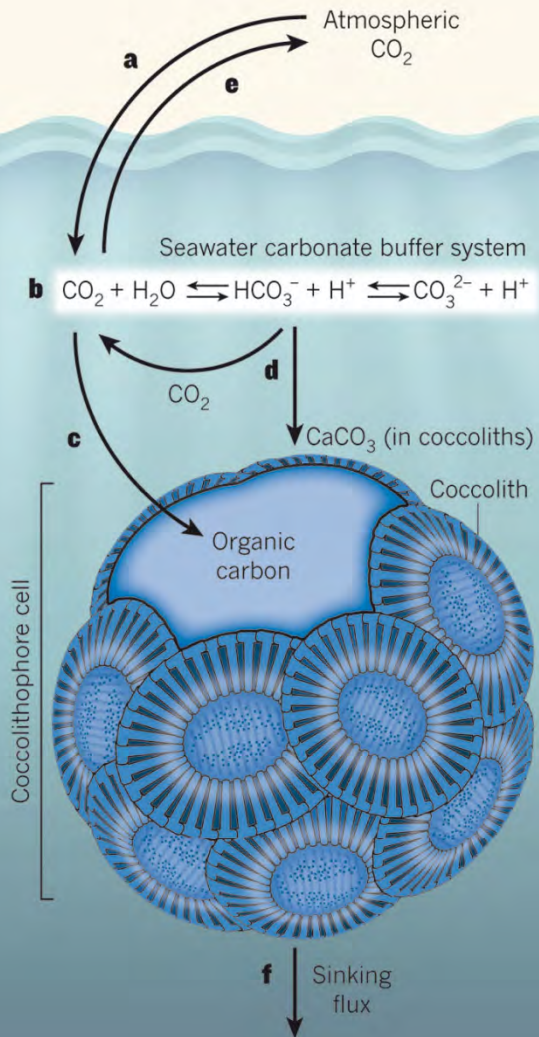
UK Ocean Acidification
Research Programme

Outline



- Introduction: Ballasting
- Assessing Spatial Variability: Geographically weighted regression
- Implications for the ballast hypothesis
- Questions/Comments
- Poster

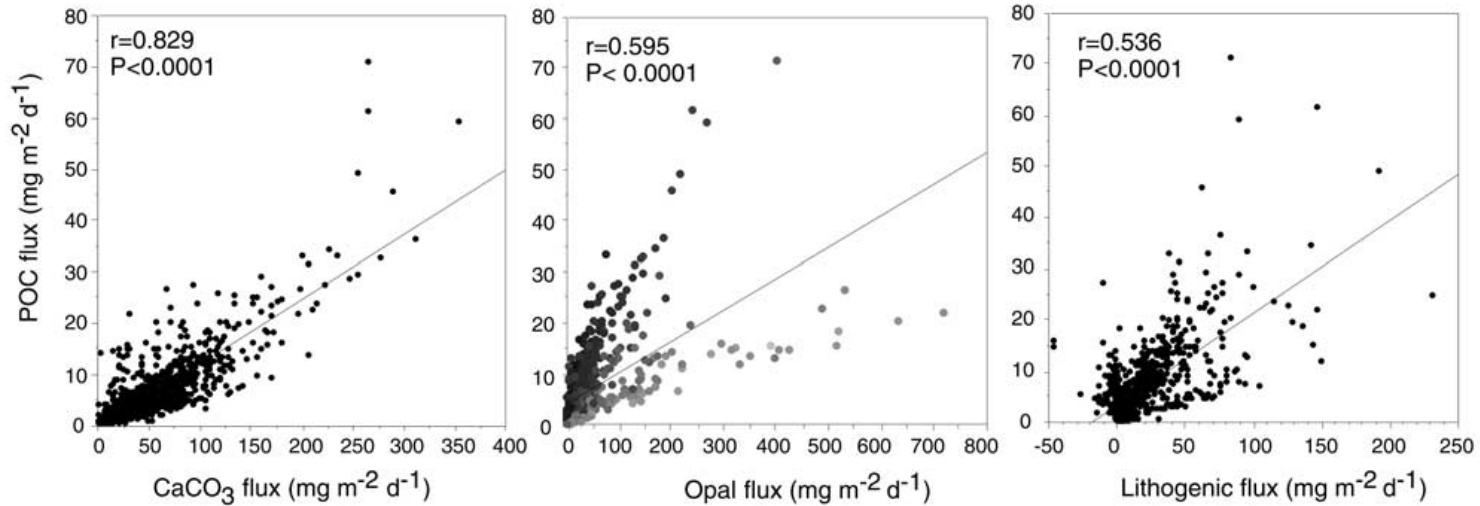
Introduction



Hutchins (2011)

- Biological Pump – transfers carbon from the surface ocean to the deep ocean and sediments
- Efficiency of the Bio. Pump is an important factor in the ocean carbon cycle
- Ballasting is suggested to be a major control on this efficiency
- The ballast hypothesis states that the sinking of POC is aided by the presence of minerals, particularly CaCO_3 , through excess density
- Could potentially act as a positive feedback to atmospheric CO_2

Defining Ballasting



Klaas and Archer (2002)

$$F_{POC} = \beta_1 \cdot F_{CaCO_3} + \beta_2 \cdot F_{Opal} + \beta_3 \cdot F_{Litho}$$

β_1 (CaCO₃) 0.075 ± 0.011

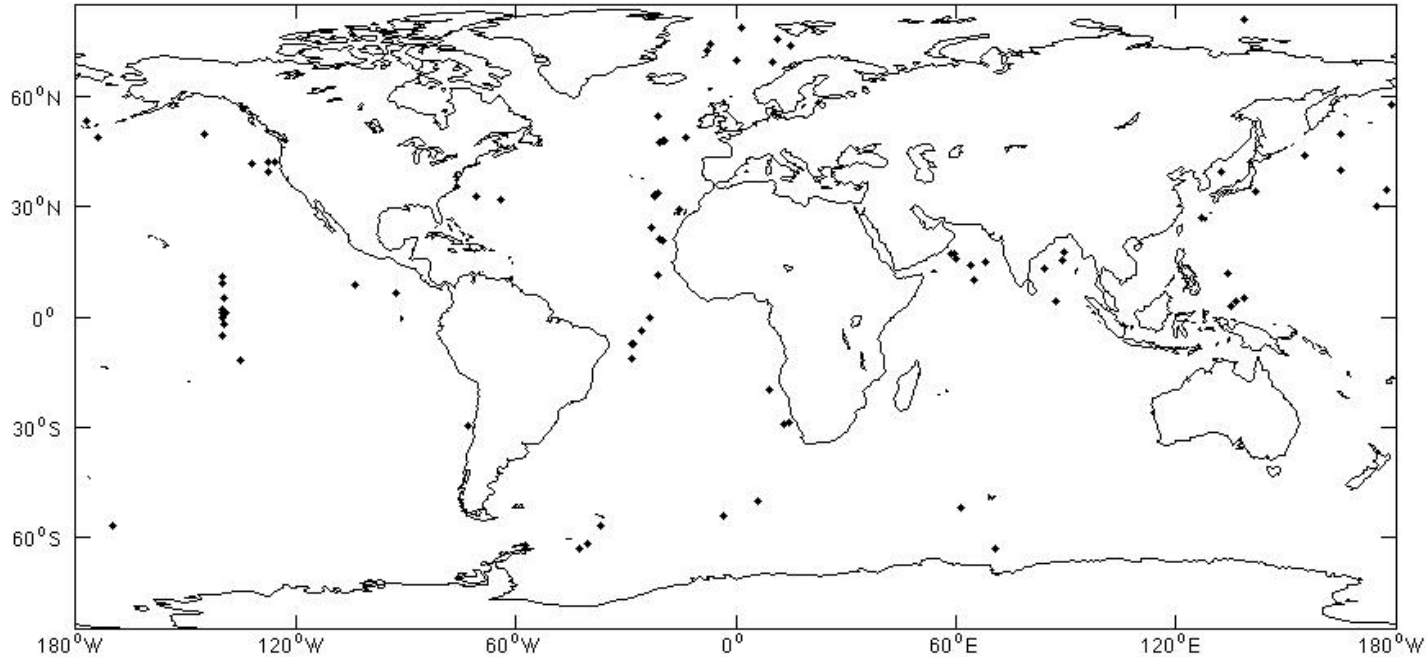
β_2 (Opal) 0.029 ± 0.008

β_3 (Litho.) 0.052 ± 0.018

'Carrying Coefficients'

$$R^2 = 0.94$$

Reassessing Ballasting



160 data points ($\geq 1500\text{m}$) *c.f.* 62-78
Increased spatial coverage
Increased temporal coverage

$$\beta_1 \quad (\text{CaCO}_3) \quad 0.090 \pm 0.012$$

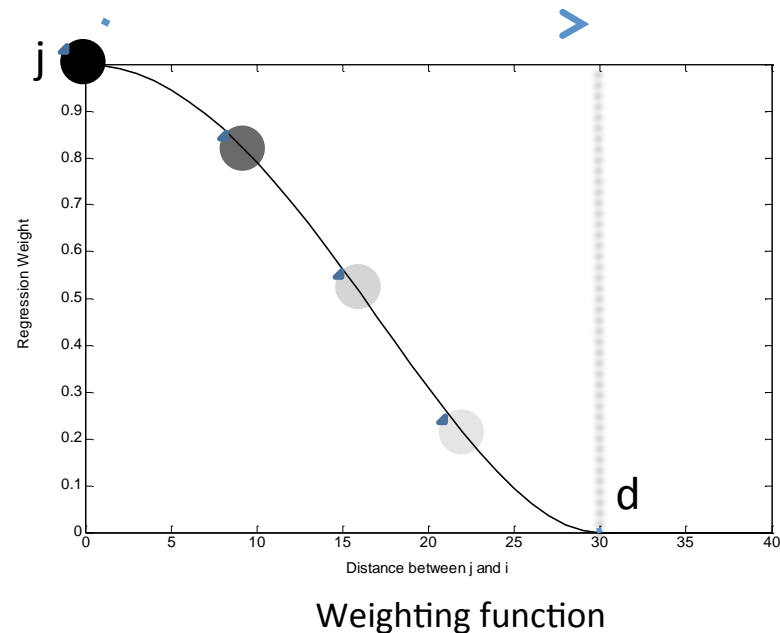
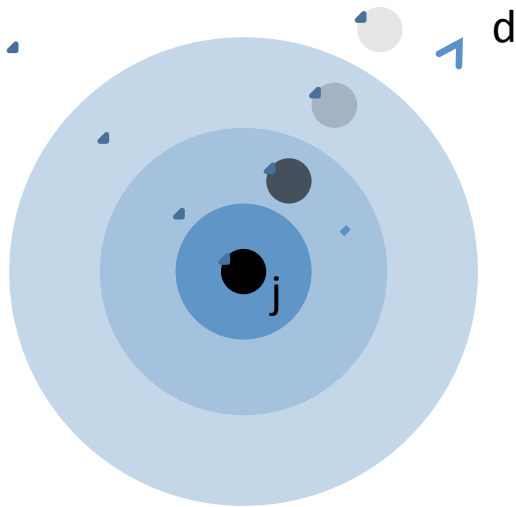
$$\beta_2 \quad (\text{Opal}) \quad 0.023 \pm 0.011 \quad R^2 = 0.66$$

$$\beta_3 \quad (\text{Litho.}) \quad 0.036 \pm 0.019$$

Geographically Weighted Regression



- a technique that allows regression coefficients to vary in space
- defines a subset of data centred at each data point by a calibrated parameter ('bandwidth')
- weights the data points by distance to the central point
- performs ordinary least squares regression with weighted dataset
- objective approach



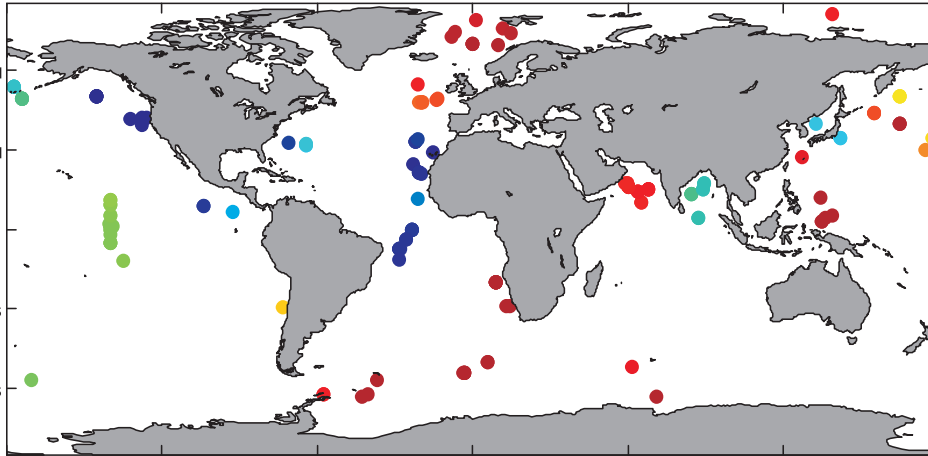
Choosing the Bandwidth



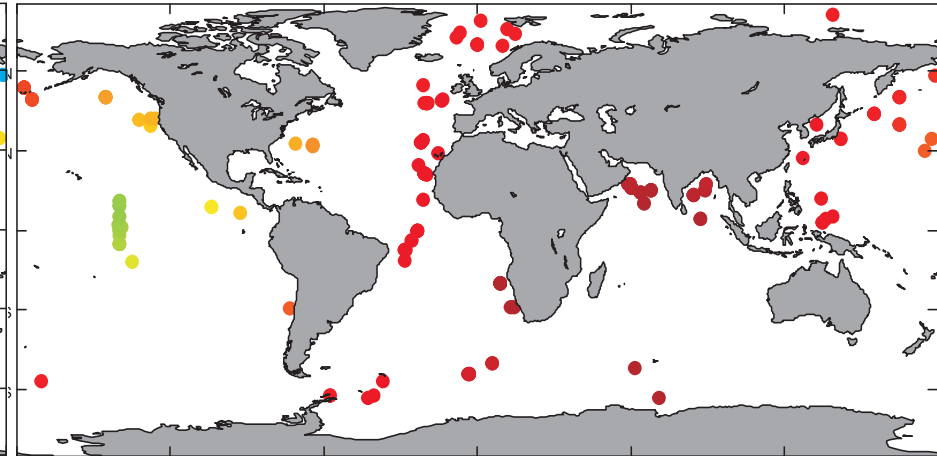
Smaller

Bandwidth Size

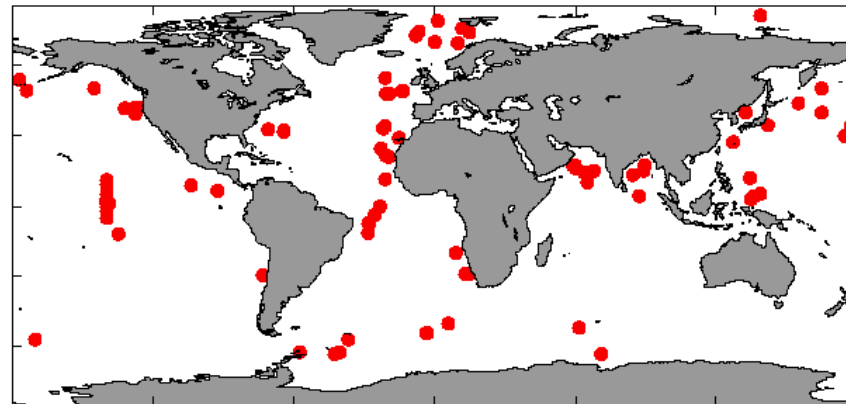
Larger



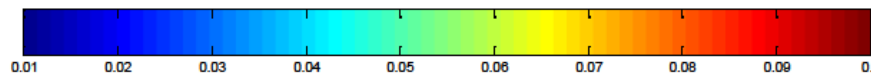
20 nearest neighbours



160 nearest neighbours



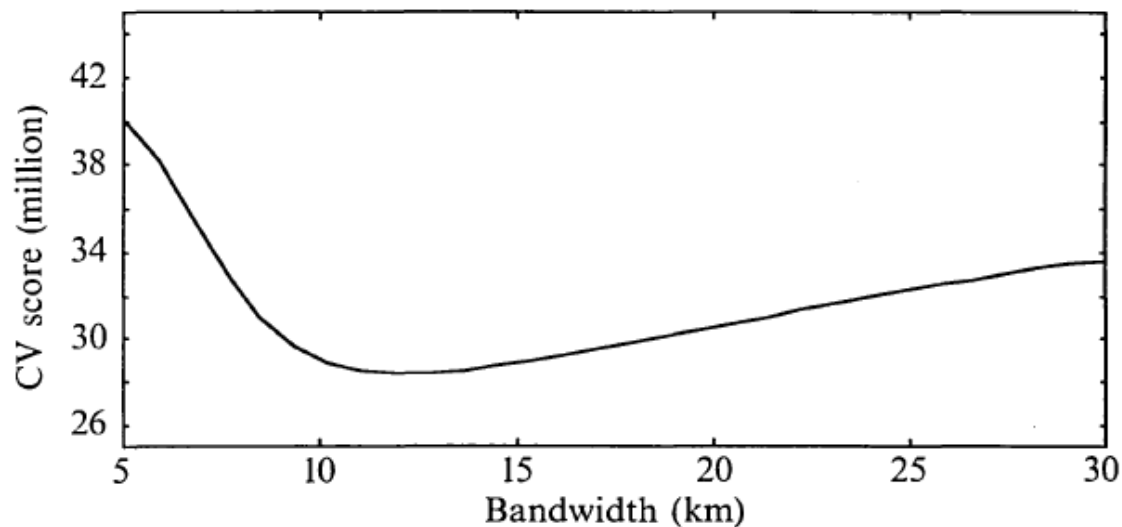
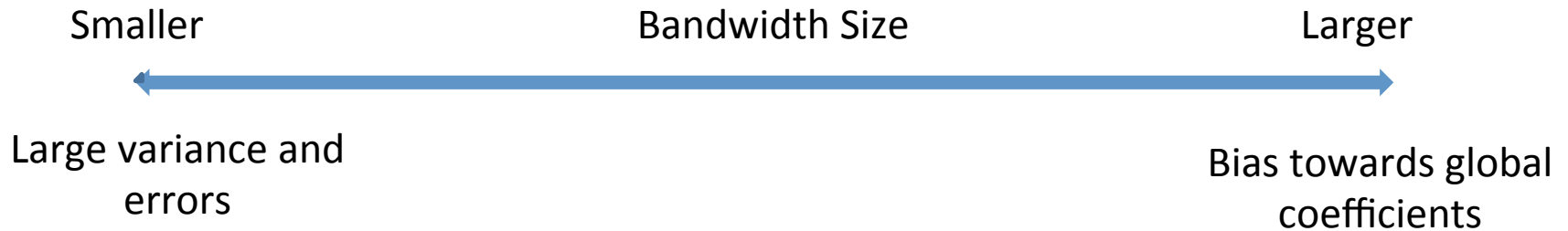
Global (0.090)



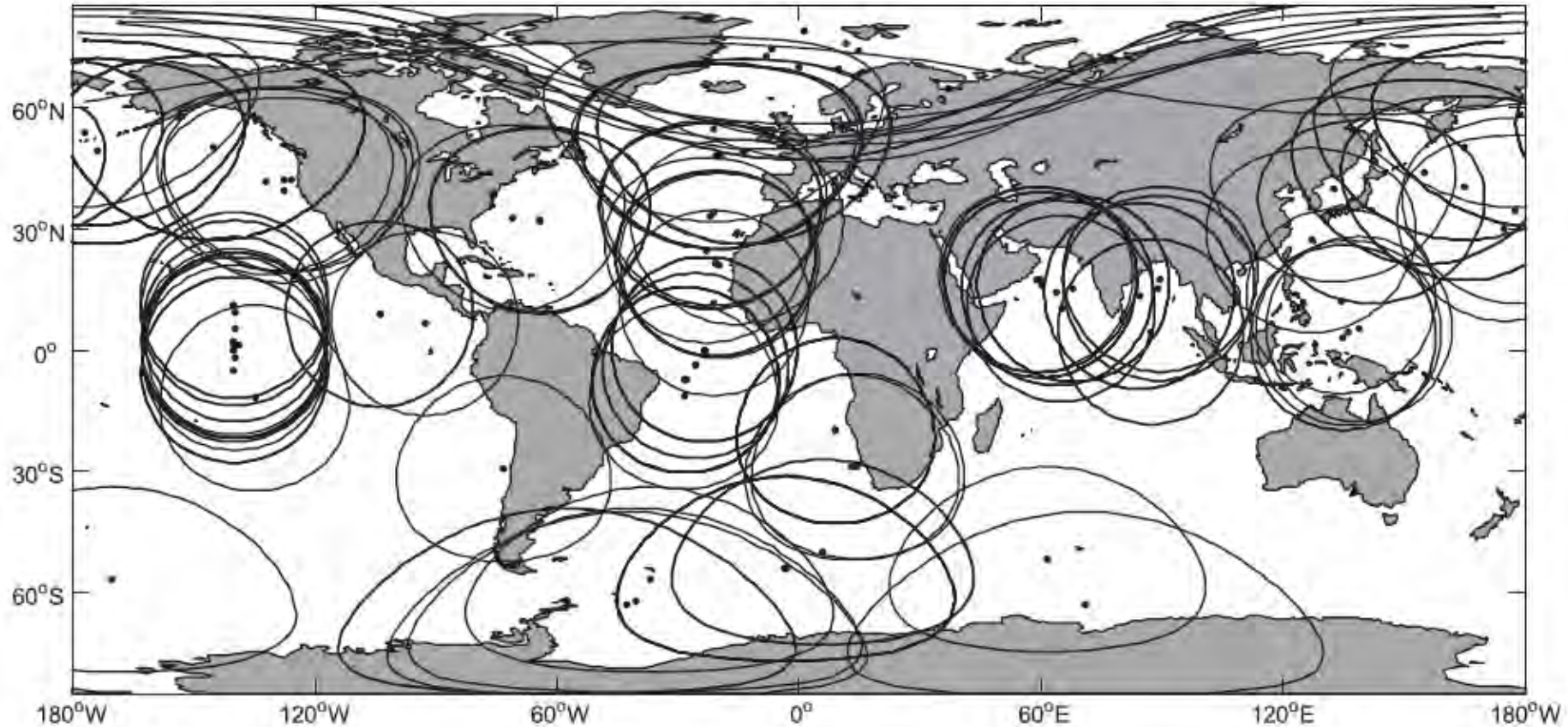
Choosing the Bandwidth



- calibrated using a metric of regression model performance
- requires consideration of bias and variance:



Choosing the Bandwidth



- illustration of the bandwidth and weighting function
- bandwidth = 50 nearest neighbours
- weighting function shown ≥ 0.75

Results: General



- Bandwidth of 50 nearest neighbours is calibrated
- Increase in R^2 (goodness of fit) from 0.66 to 0.84
- Significant improvement in model skill scores
- ANOVA results rejects null hypothesis that the regional model is no better improvement on the global model.
- Monte Carlo results reject null hypothesis that there is no spatial variability in coefficients

Median carrying coefficients:

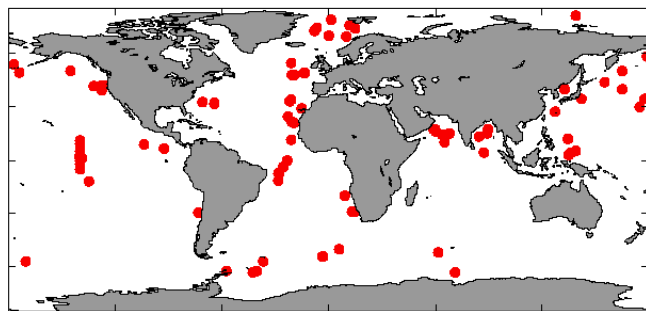
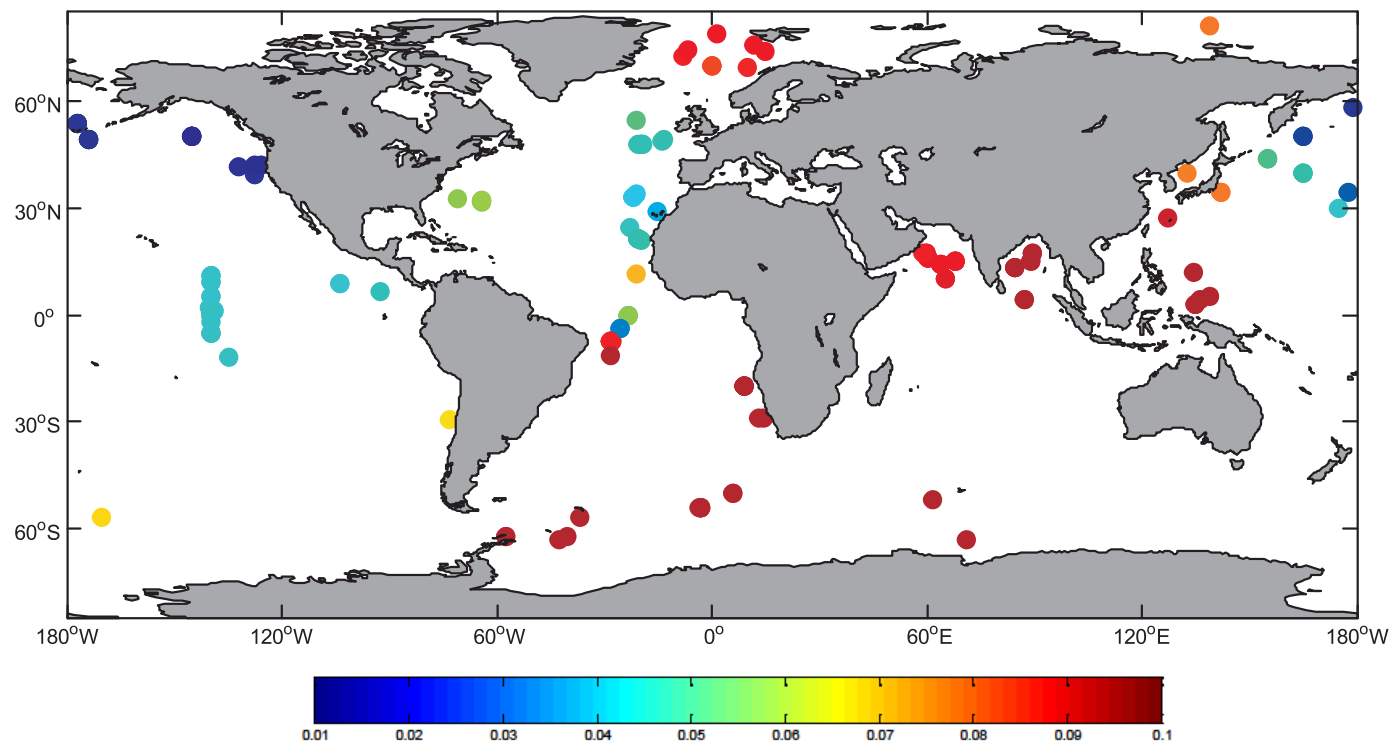
- CaCO_3 0.056
- Opal 0.038
- Litho. 0.022

c.f.

Global carrying coefficients:

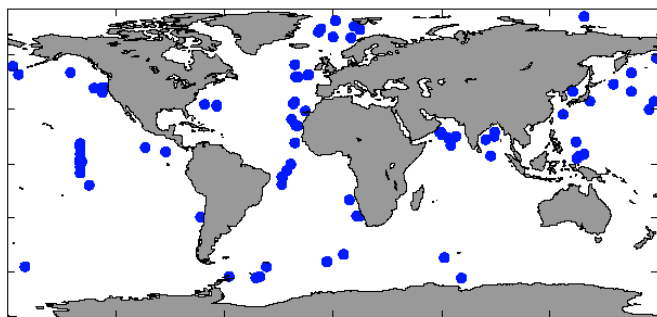
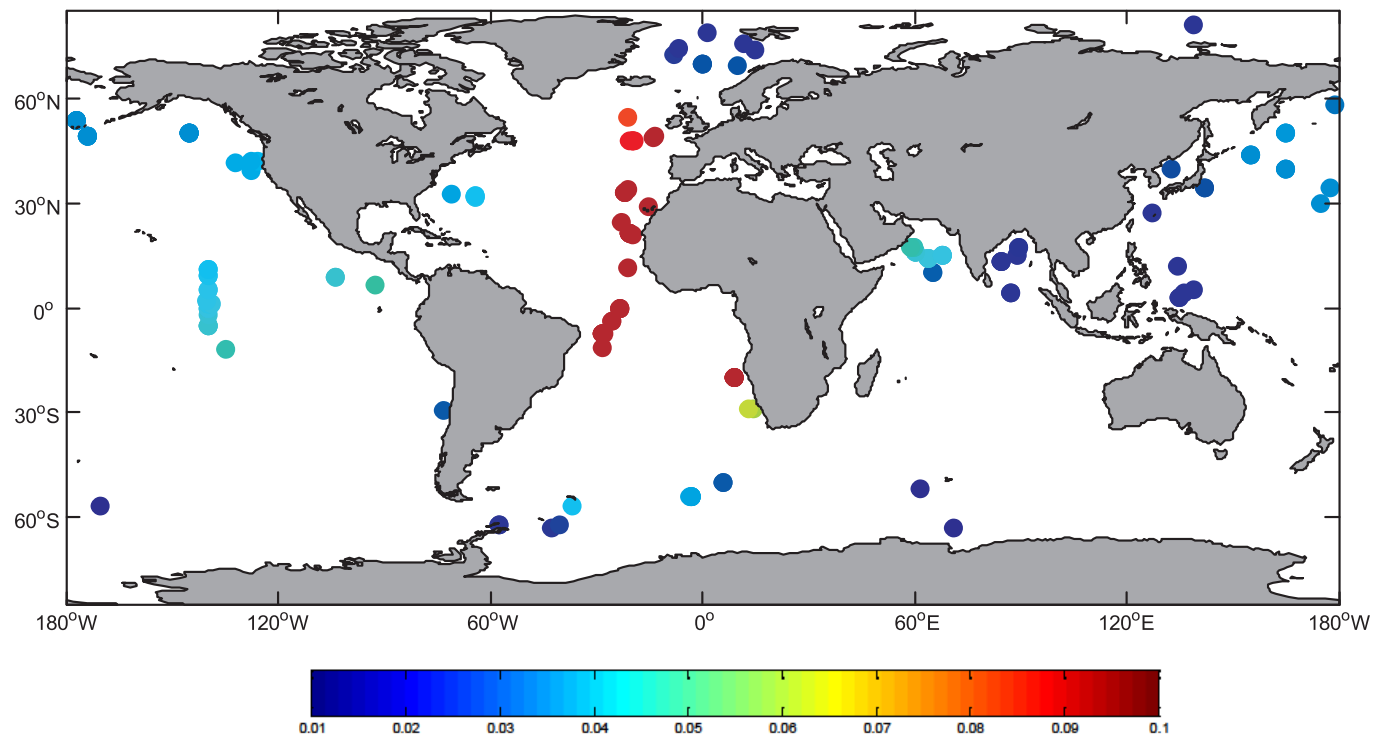
- CaCO_3 0.090
- Opal 0.023
- Litho. 0.036

Results: CaCO_3



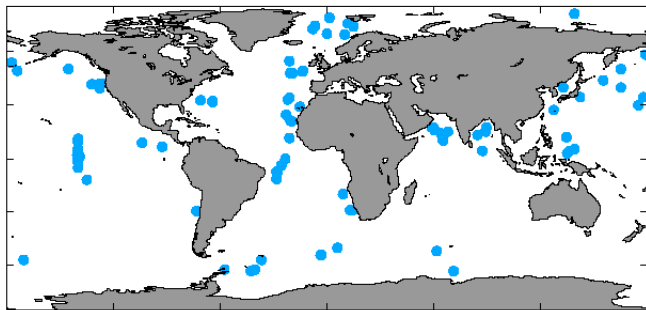
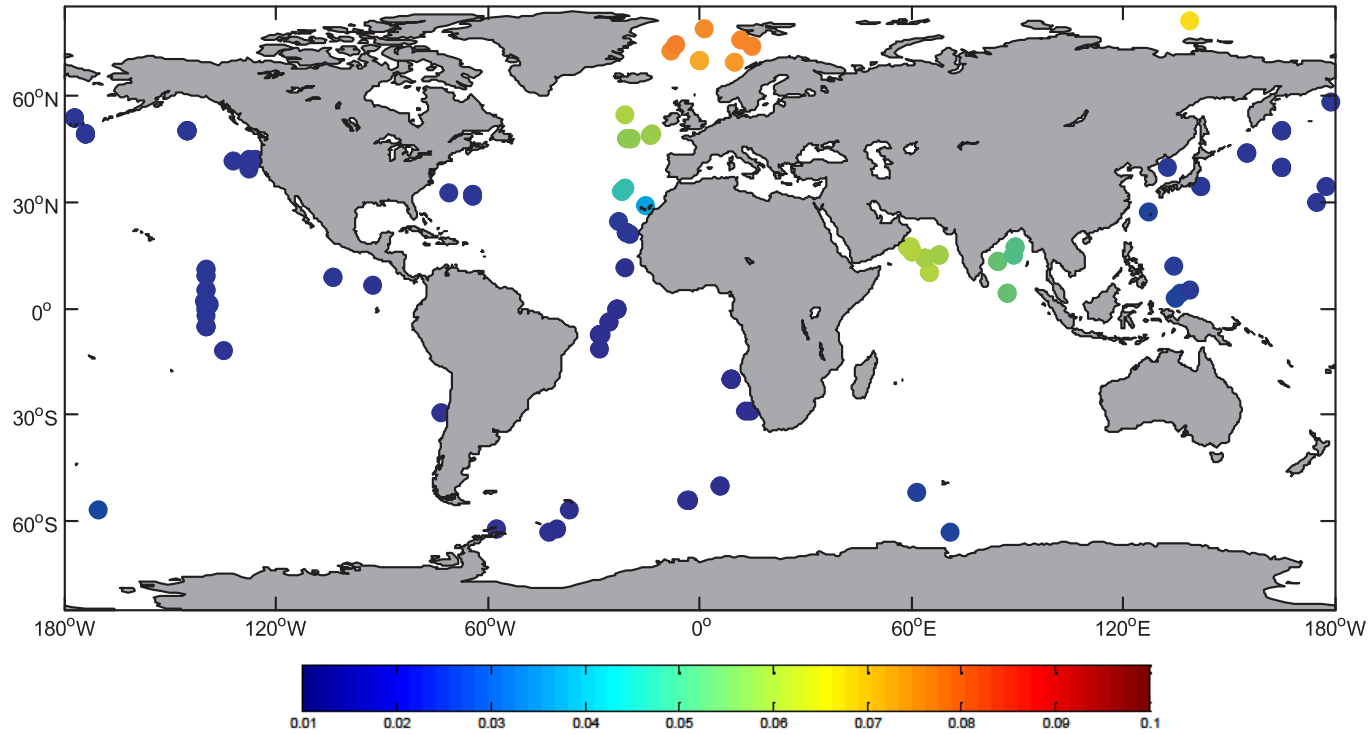
Global: 0.090

Results: Opal



Global: 0.023

Results: Lithogenic

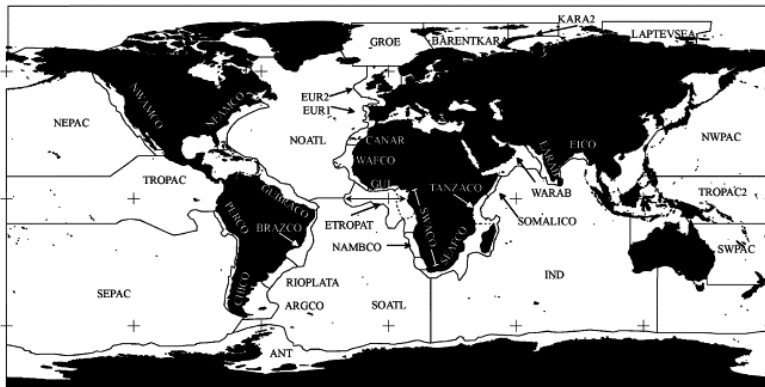


Global: 0.036

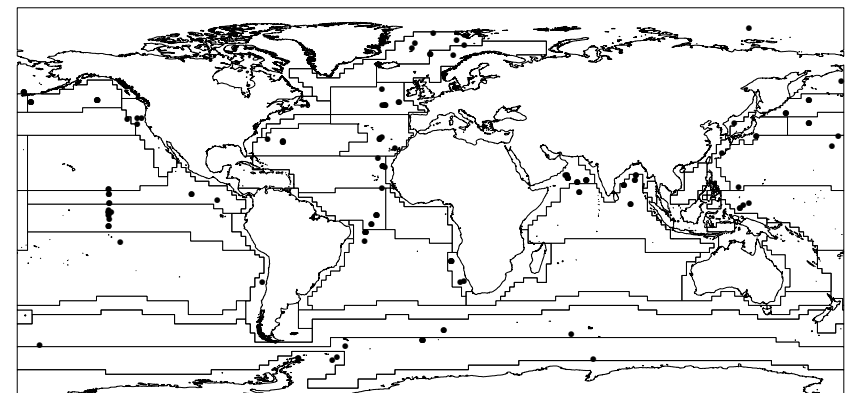
Implications for Ballasting



- The strong global relationship between POC and CaCO_3 hides significant spatial variability
- There are no consistent relationships between minerals and POC
- Unclear on the mechanism(s) resulting in the spatial patterns
 - May reflect biogeochemical provinces and ecosystem characteristics
- Implications for use of ballasting as a parameterisation in modelling
 - May suggest different feedback mechanisms



Degradability of POC (Seiter *et al.* 2004)



Biogeochemical Provinces (Vichi *et al.* 2011)



wilsonjd@cardiff.ac.uk

Does the ballast hypothesis still pull its own weight?: Understanding marine particulate fluxes and their effect on atmospheric CO₂

References:

Fotheringham, A.S., Charlton, M.E., and Brunson, C., (1998) Geographically weighted regression: a natural evolution of the expansion method for spatial data analysis. *Environment and Planning A*. 30 (11), 1905 – 1927

Fotheringham, A.S., Brunson, C., and Charlton, M.E., (2002) *Geographically Weighted Regression: the analysis of spatially varying relationships*. Wiley: Chichester

Hutchins D.A., (2011) Oceanography: Forecasting the rain ratio. *Nature*. 476, 41 - 42

Klaas, C., and Archer, D.E., (2002) Association of sinking organic matter with various types of mineral ballast in the deep sea: Implications for the rain ratio. *Global Biogeochemical Cycles*. 3 (4)

Seiter, K., Hensen, C., Schroter, J., and Zabel, M., (2004) Organic carbon content in surface sediments – defining regional provinces. *Deep Sea Research Part I: Oceanographic Research Papers*. 51 (12), 2001 - 2026

Vichi, M., Allen, J.I., Masina, S., and Hardman-Montford, N.J., (2011) The emergence of ocean biogeochemical provinces: a quantitative assessment and a diagnostic for model evaluation. *Global Biogeochemical Cycles*. 25

