

Inverse models of GEOTRACES datasets

New insights into trace metal scavenging

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GORDON AND BETTY
MOORE
FOUNDATION



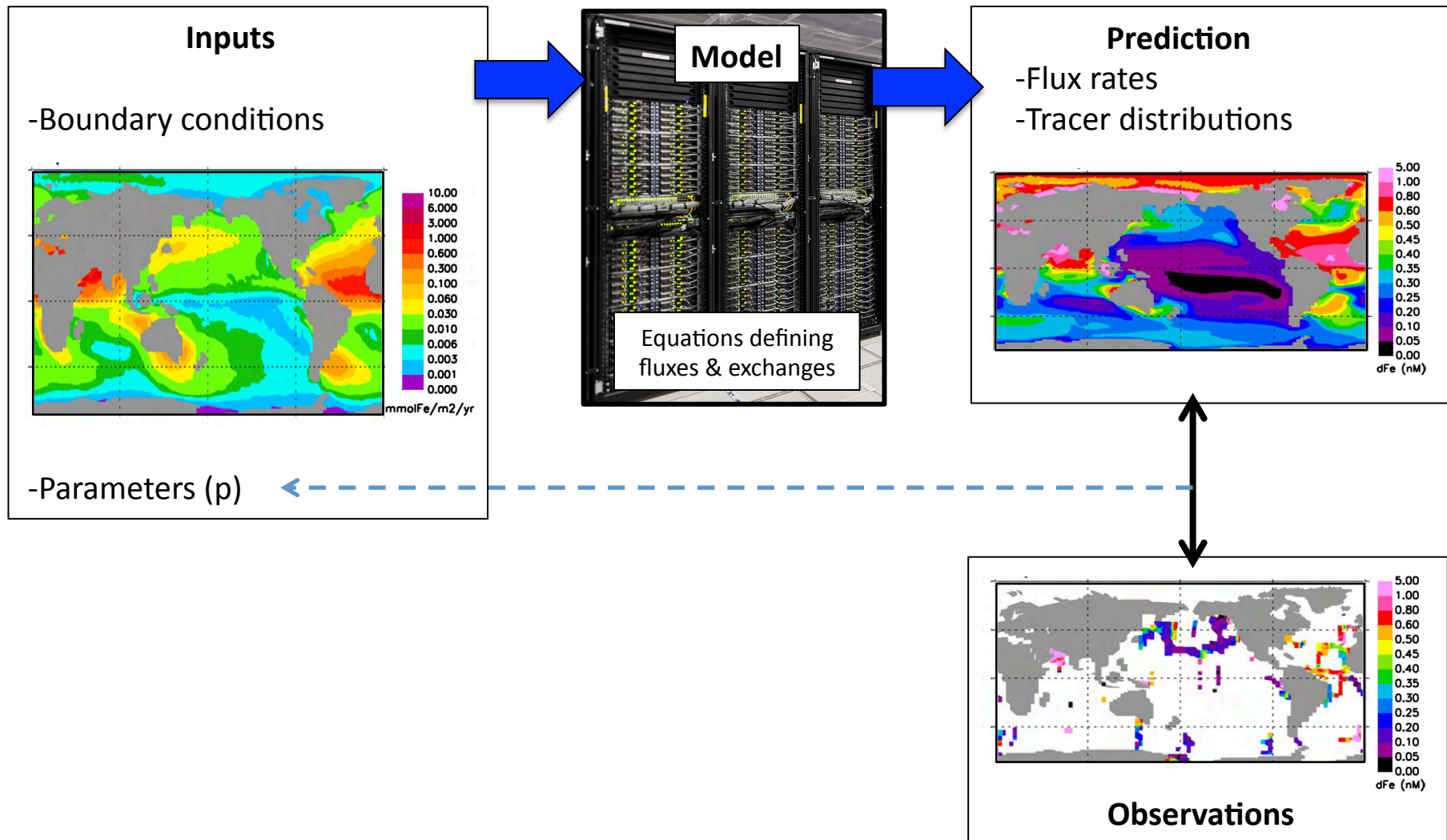
Objectives

- Concept of a data-assimilating (inverse) model
 - Differences/similarities to other model approaches
 - Important considerations
- How can they be used to extract new information from data?
 - Rates
 - Processes
 - Chemical mechanisms

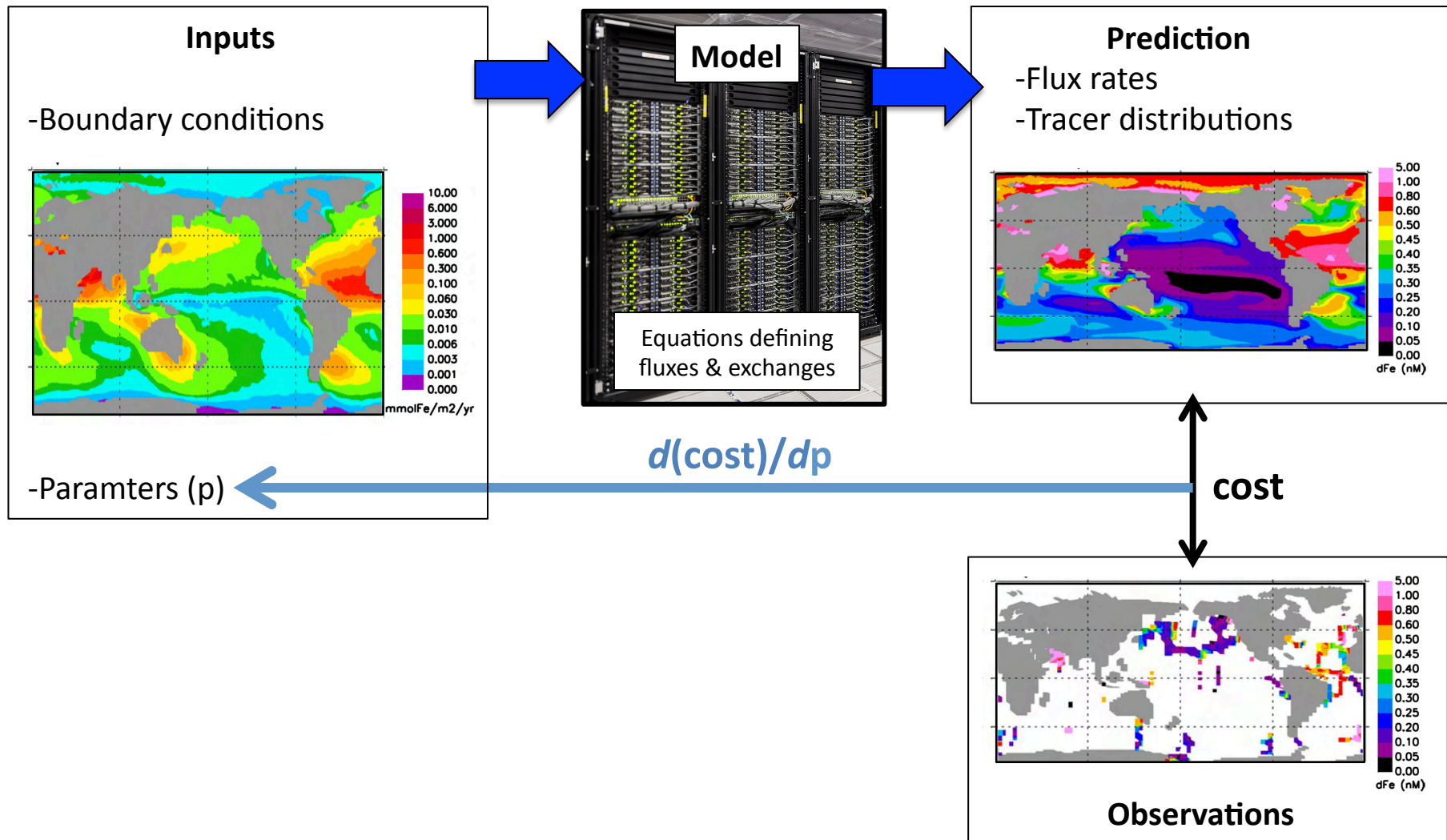
Objectives

- Concept of a data-assimilating (inverse) model
 - Differences/similarities to other model approaches
 - Important considerations
- How can they be used to extract new information from data (about trace element scavenging)?
 - Rates (lifetime of hydrothermal Fe)
 - Processes (global Zn distribution)
 - Chemical mechanisms (Cd scavenging in low O₂)

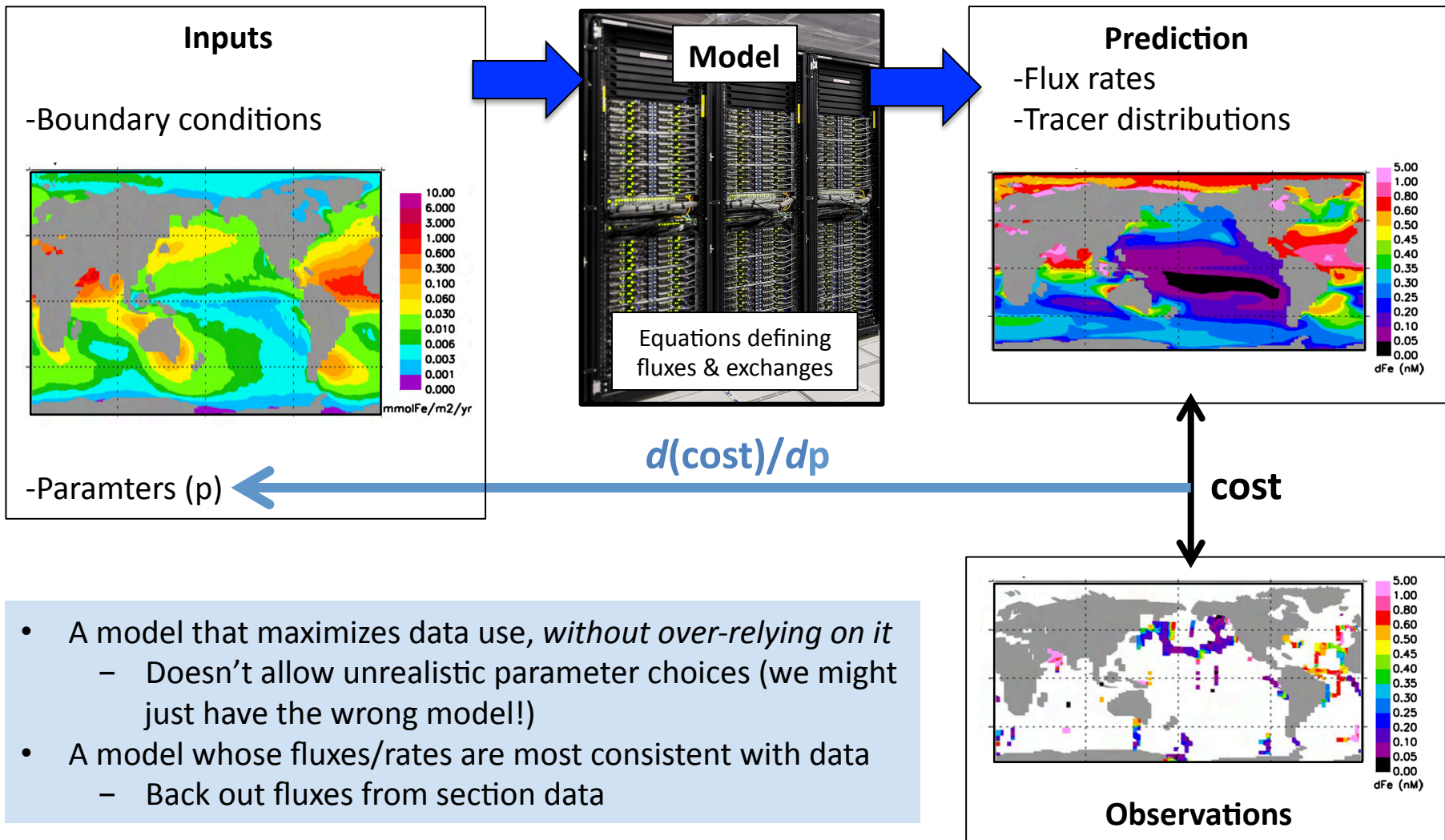
“Traditional” models



Data-assimilating models

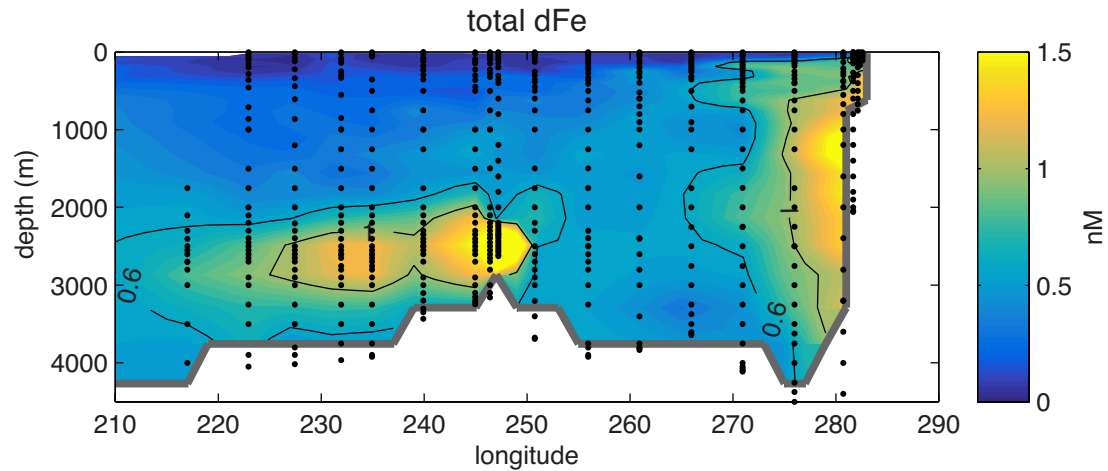


Data-assimilating models



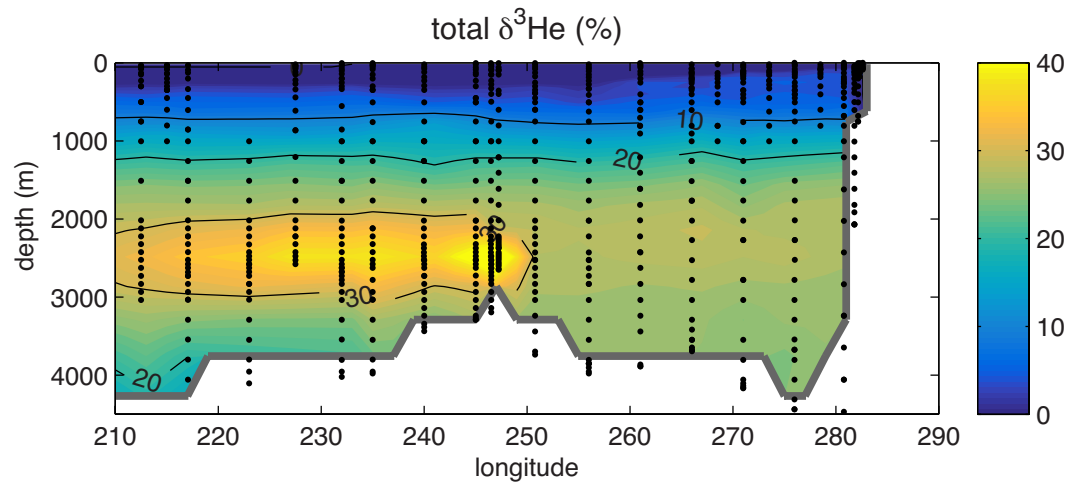
- A model that maximizes data use, *without over-relying on it*
 - Doesn't allow unrealistic parameter choices (we might just have the wrong model!)
- A model whose fluxes/rates are most consistent with data
 - Back out fluxes from section data

Rates: Lifetime of hydrothermal Fe

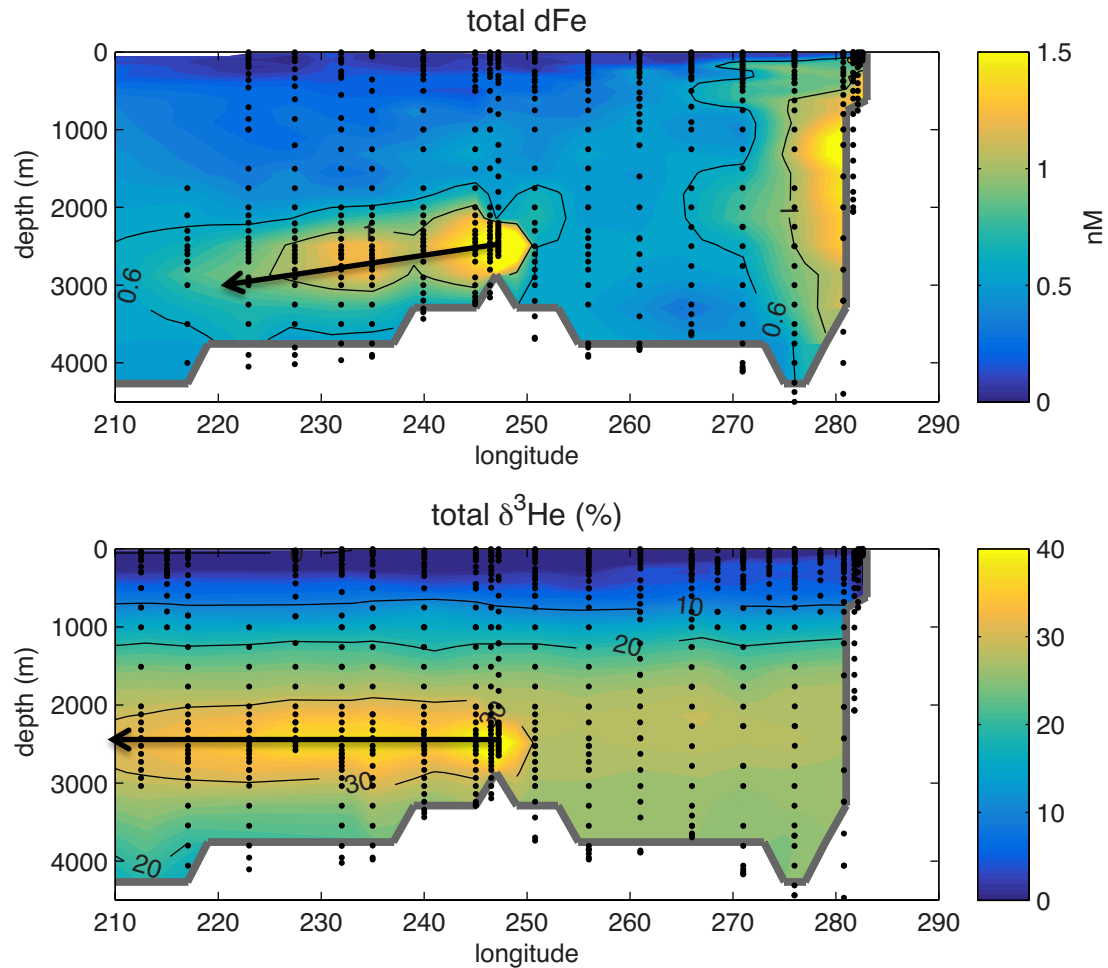


EPZT hydrothermal vent

- Fe plume transported long distances across the basin (slow scavenging)



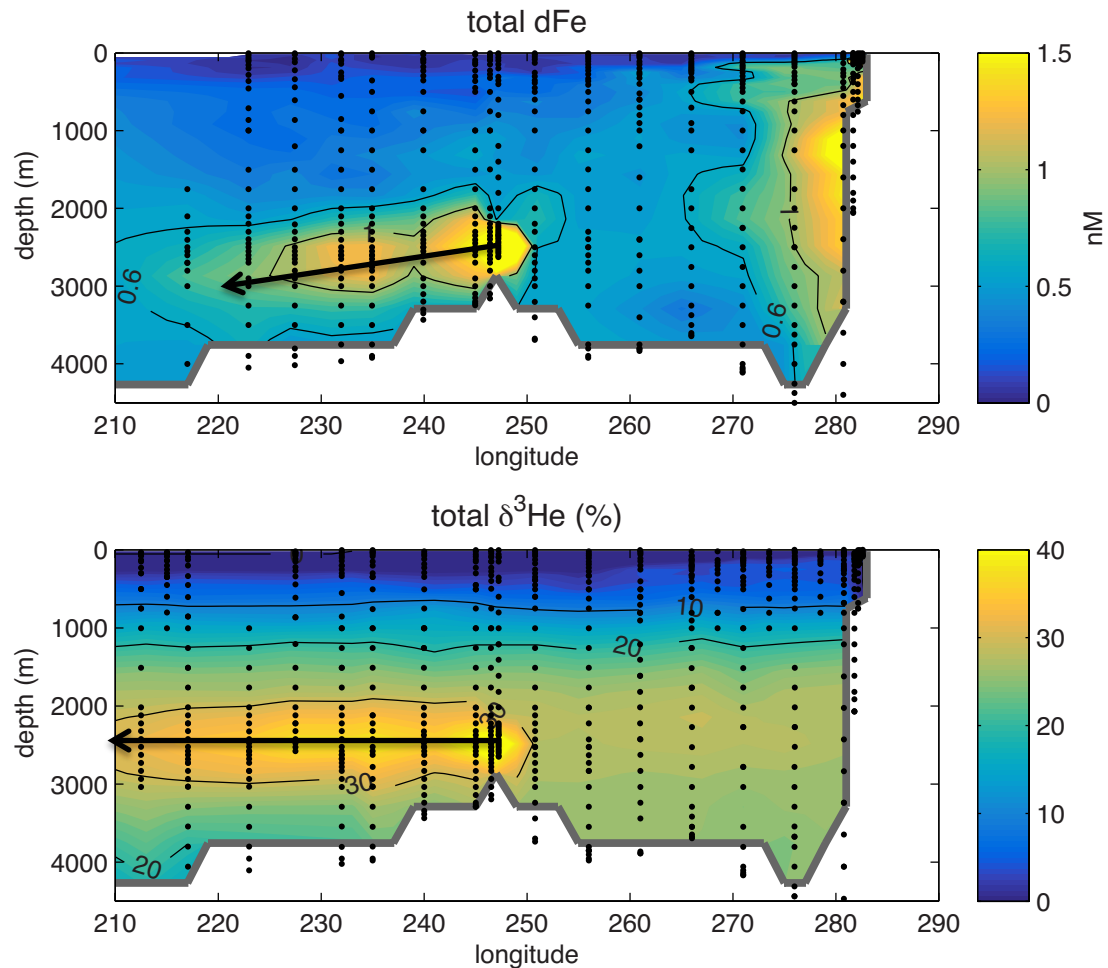
Rates: Lifetime of hydrothermal Fe



EPZT hydrothermal vent

- Fe plume transported long distances across the basin (slow scavenging)
- Slumping of plume relative to passive tracer

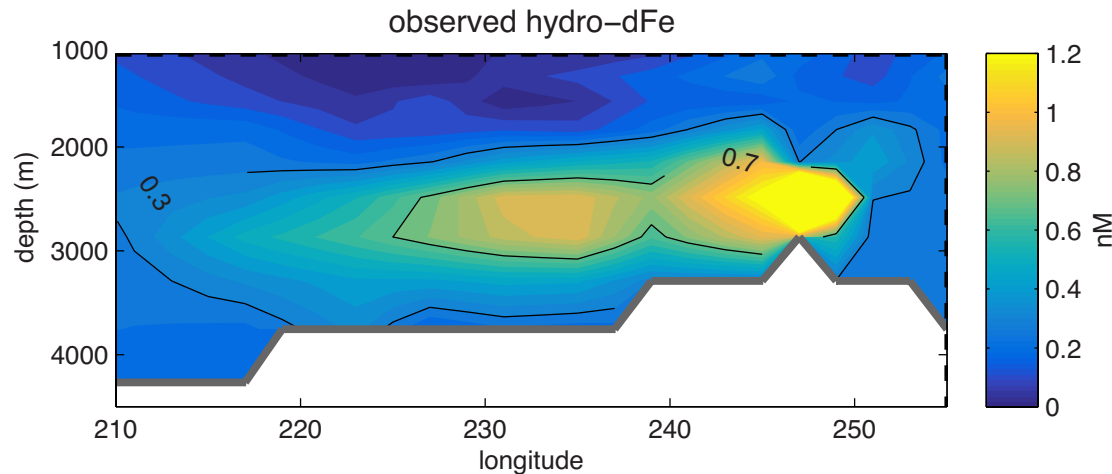
Rates: Lifetime of hydrothermal Fe



EPZT hydrothermal vent

- Fe plume transported long distances across the basin (slow scavenging)
- Slumping of plume relative to passive tracer
- Rates of Fe sinking and scavenging?
- Lifetime of hydrothermal Fe, and supply to surface?

A simple “hydro-dFe” model



hydro-dFe
 dFe in plume region, minus
 “background” concentration
 (advected, ligand-bound Fe)

$$\frac{d(\text{Fe})}{dt} = \text{circ.}(\text{Fe}) + R_{\text{Fe:He}} J_{\text{hydro}}(^3\text{He}) - k_{\text{sc}} \text{Fe} - \frac{\partial}{\partial z} (w_{\text{sink}} \text{Fe})$$

Circulation
 2°x2° Ocean
 model (OCIM)

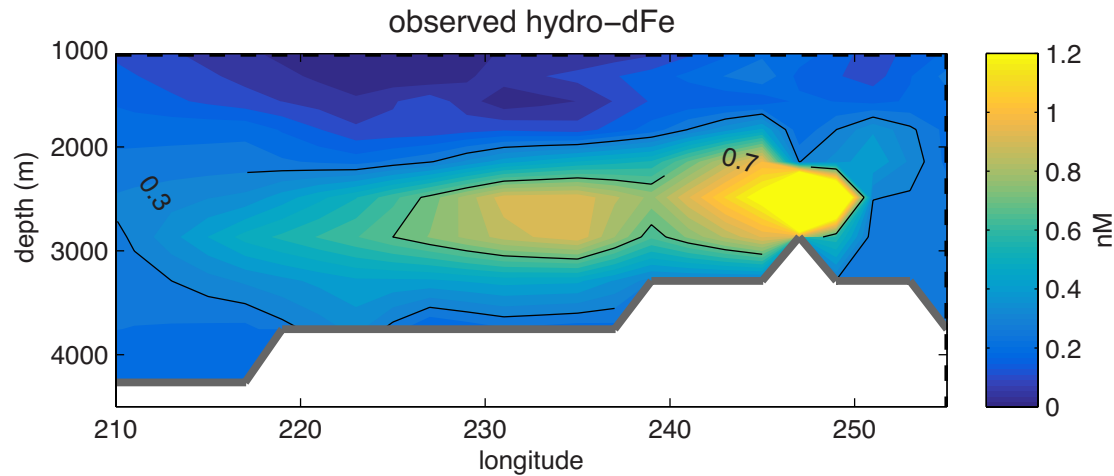
Vent source
 Coupled to
³He input

Scavenging
 First-order
 removal

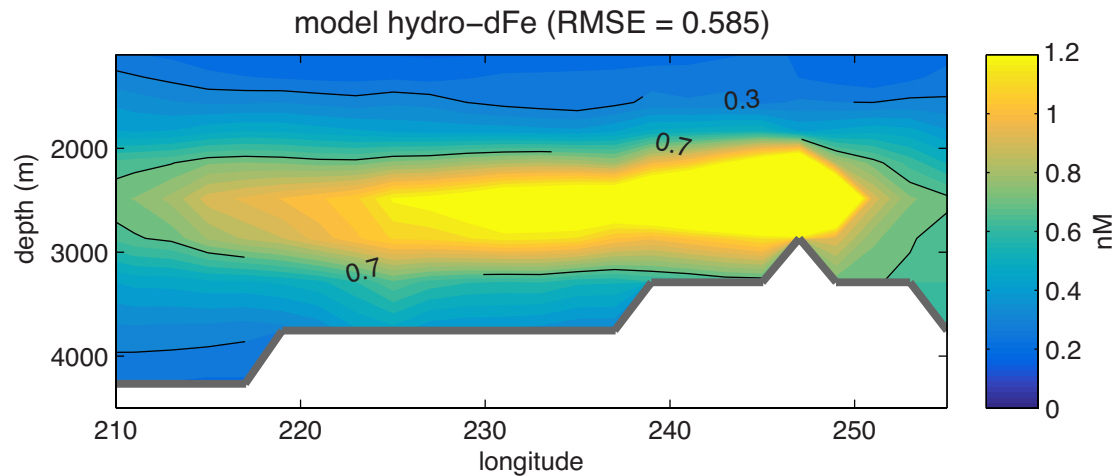
Sinking
 At constant
 sinking rate

DeVries et al 2011

A simple “hydro-dFe” model



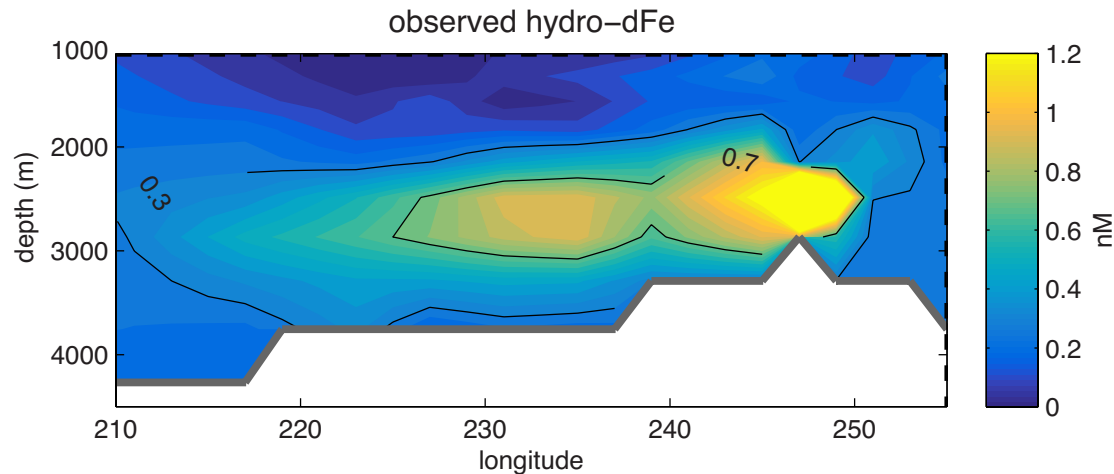
hydro-dFe
dFe in plume region, minus
“background” concentration
(advected, ligand-bound Fe)



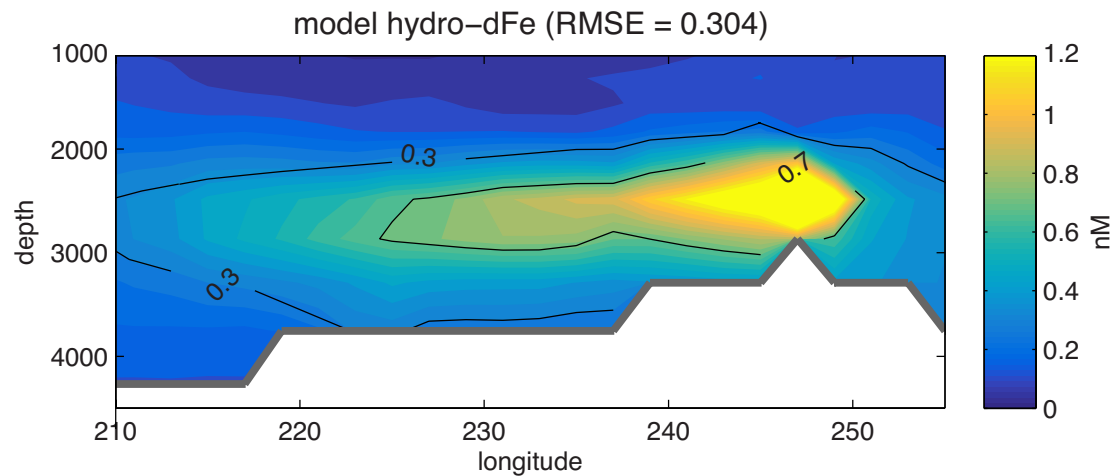
Initial guess:

- $\text{Fe}/^3\text{He} = 10^8 \text{ mol/mol}$
- $w_{\text{sink}} = 10 \text{ m/yr}$
- $k_{\text{sc}} = 0.01\text{yr}^{-1}$

Optimized “hydro-dFe” model



hydro-dFe
dFe in plume region, minus
“background” concentration
(advected, ligand-bound Fe)



Optimum (~200 iterations):

- $\text{Fe}/^3\text{He} = 5.5 \times 10^7 \text{ mol/mol}$
- $w_{\text{sink}} = 33 \text{ m/yr}$
- $k_{\text{sc}} = 0.02 \text{ yr}^{-1}$

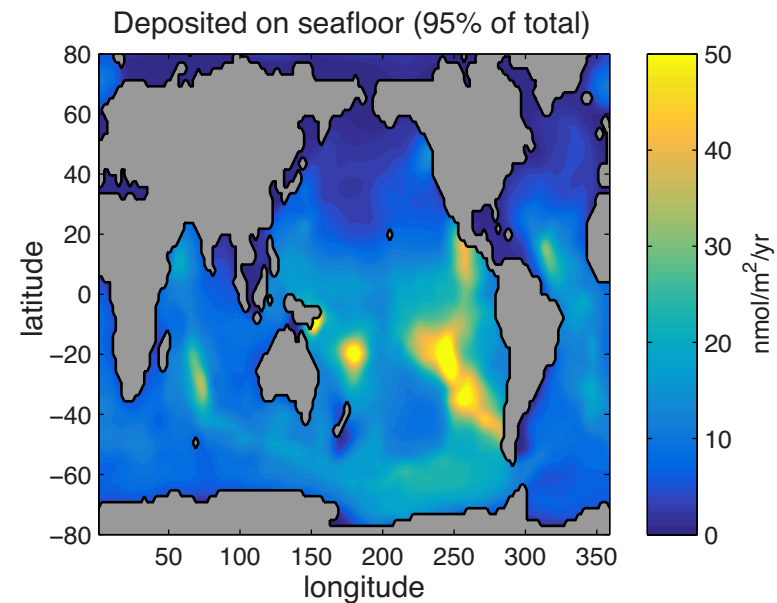
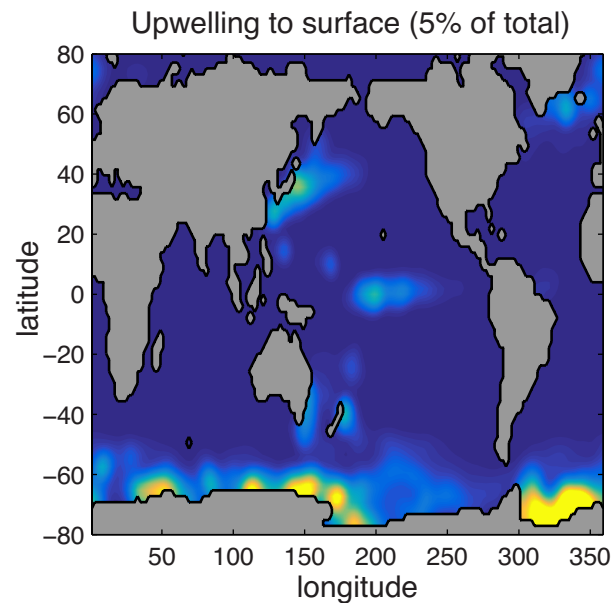
Limitations and implications

Limitations:

- Unresolved process (*data-assimilating models are often necessarily simplified*)

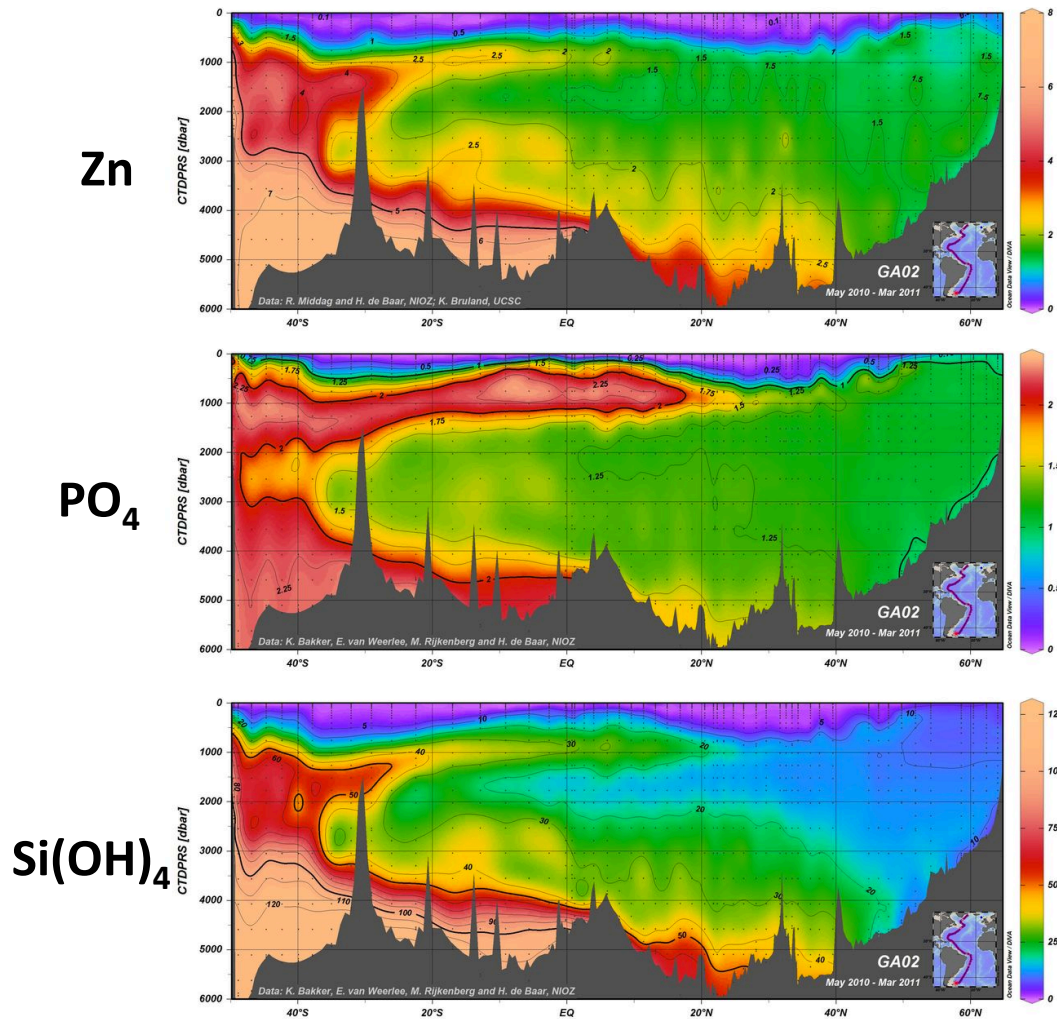
Implications (extrapolation):

- Total hydrothermal source of 4.5Gmol/yr, lifetime of ~30yrs
- Sinking and scavenging remove 95% of hydrothermal before it reaches the surface

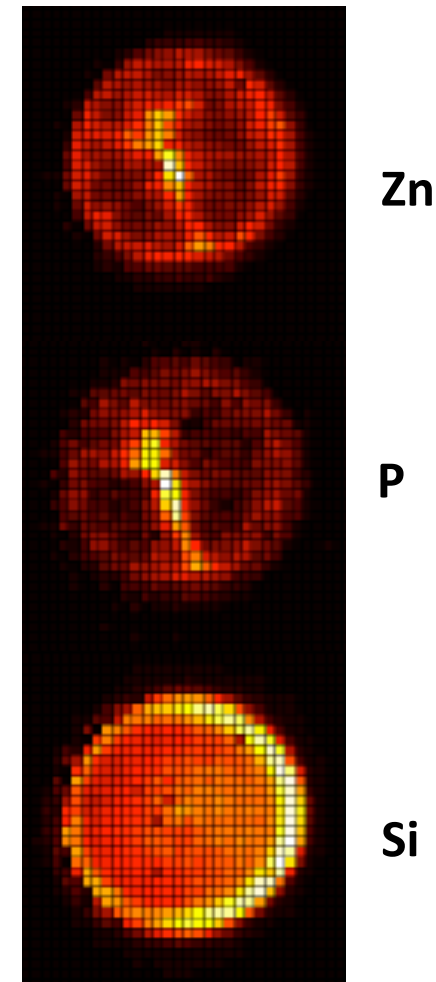


Processes: Global Zn/Si covariation

Oceanic distribution (GA02)



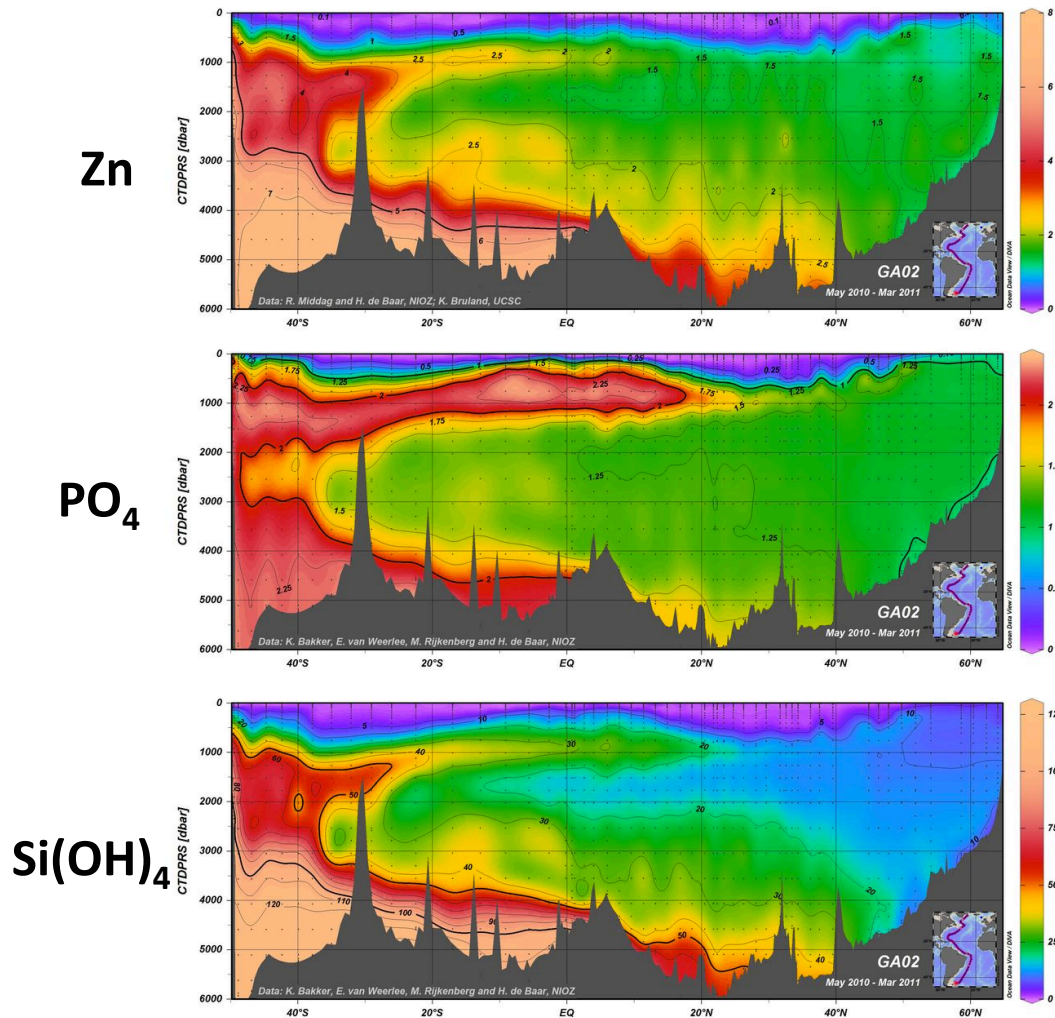
Cellular distribution (SXRF)



Twining et al 2003

Processes: Global Zn/Si covariation

Oceanic distribution (GA02)

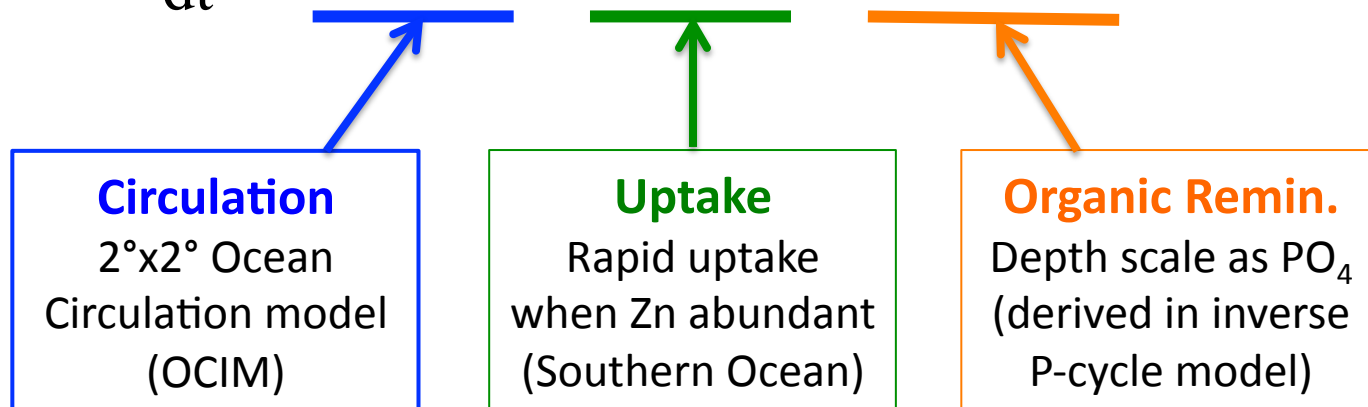


Three hypotheses

- 1. Preformed:** Rapid Southern Ocean uptake leaves surface distribution similar to Si(OH)_4
- 2. Frustules:** Incorporation of some Zn into diatom frustules results in deeper remineralization than other soft-tissue nutrients
- 3. Scavenging:** Zn adsorbed onto organic particles, hitchhikes a ride to the deep ocean (*John & Conway 2014*)

Model 1: Preformed Zn

$$\frac{dZn}{dt} = circ.(Zn) - J_{up}(Zn) - J_{rem}(POZn)$$

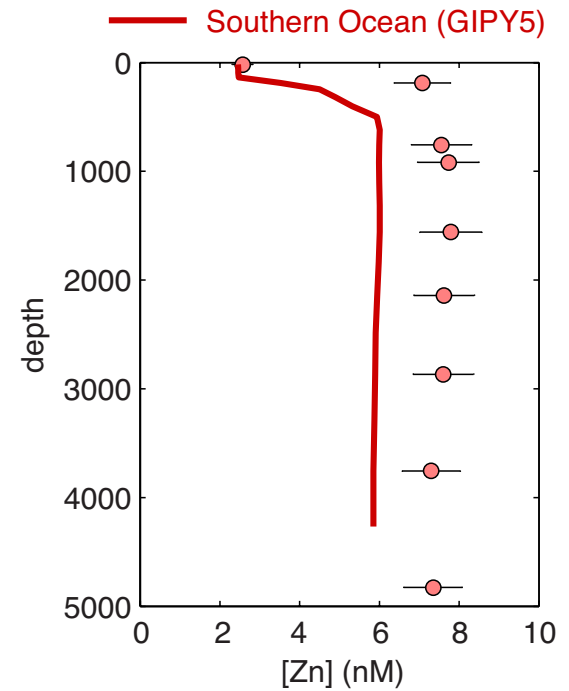
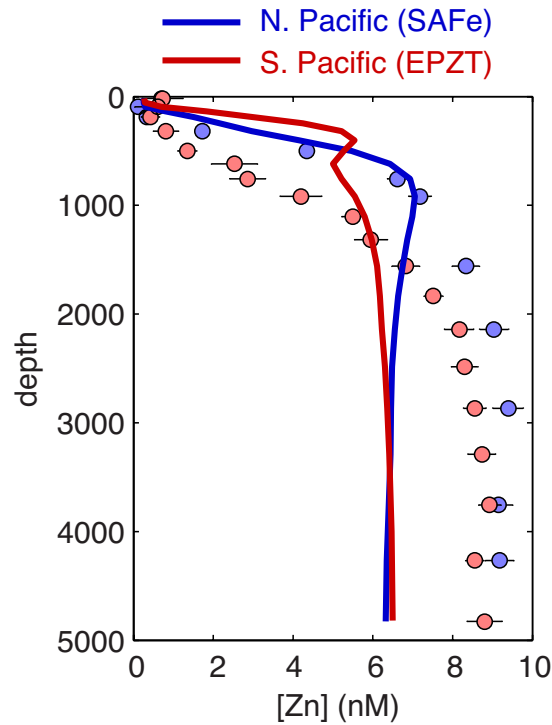
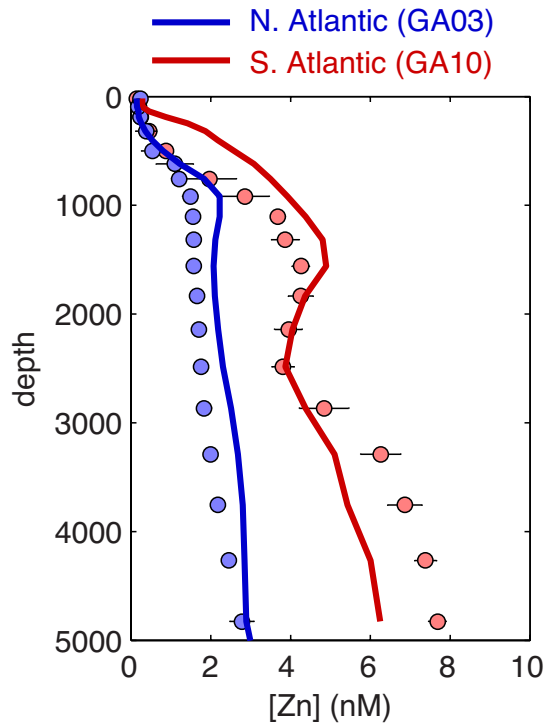


Why data-assimilating model?:

- Rule out model “failure” due to bad parameters.
- Fairest way to eliminate hypotheses.

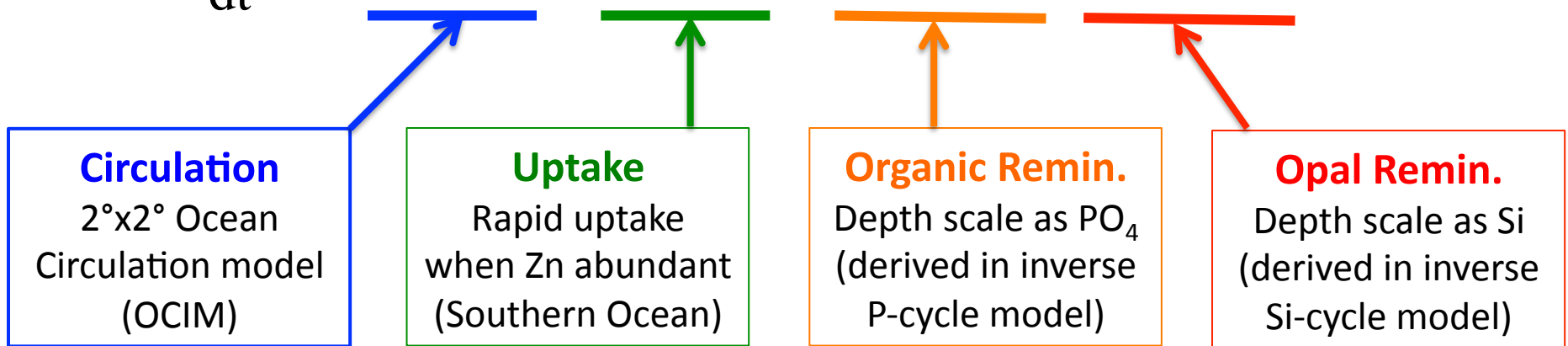
Model 1: Preformed Zn

$$\frac{dZn}{dt} = circ.(Zn) - J_{up}(Zn) - J_{rem}(POZn)$$



Model 2: Frustule-associated Zn

$$\frac{dZn}{dt} = circ.(Zn) - J_{up}(Zn) - J_{rem}(POZn) - J_{rem}(Zn_{frust})$$

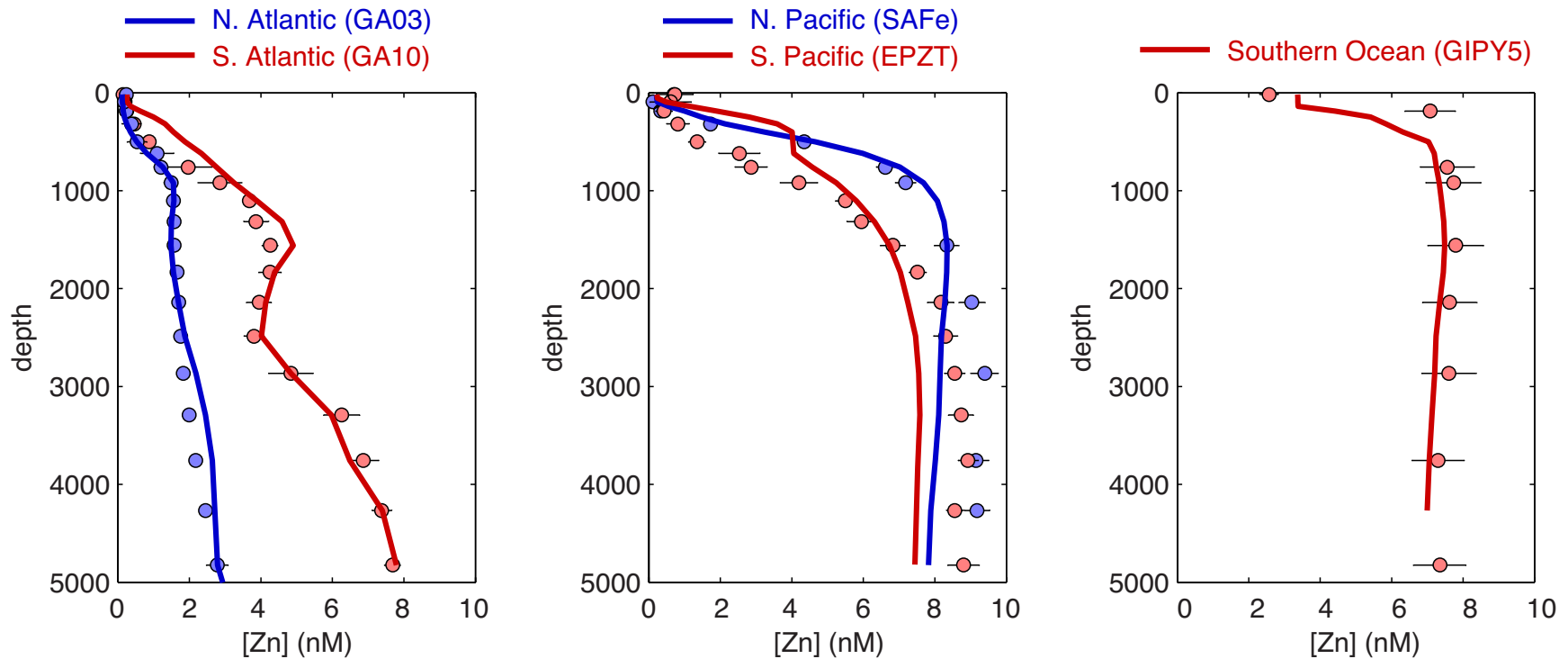


Additional constraints:

- 1-3% directly incorporated into frustules (*Ellwood & Hunter 2000*)
- <40% co-located with frustule (*SXRF*)

Model 2: Frustule-associated Zn

$$\frac{dZn}{dt} = circ.(Zn) - J_{up}(Zn) - J_{rem}(POZn) - J_{rem}(Zn_{frust})$$



Wants >40% Zn in diatom frustules

Model 3: Scavenging

$$\frac{dZn}{dt} = \text{circ.}(Zn) - J_{up}(Zn) - J_{rem}(POZn) - \frac{\partial}{\partial z} wZn_{ads}$$

Circulation
2°x2° Ocean
Circulation model
(OCIM)

Uptake
Rapid uptake
when Zn abundant
(Southern Ocean)

Organic Remin.
Depth scale as PO₄
(derived in inverse
P-cycle model)

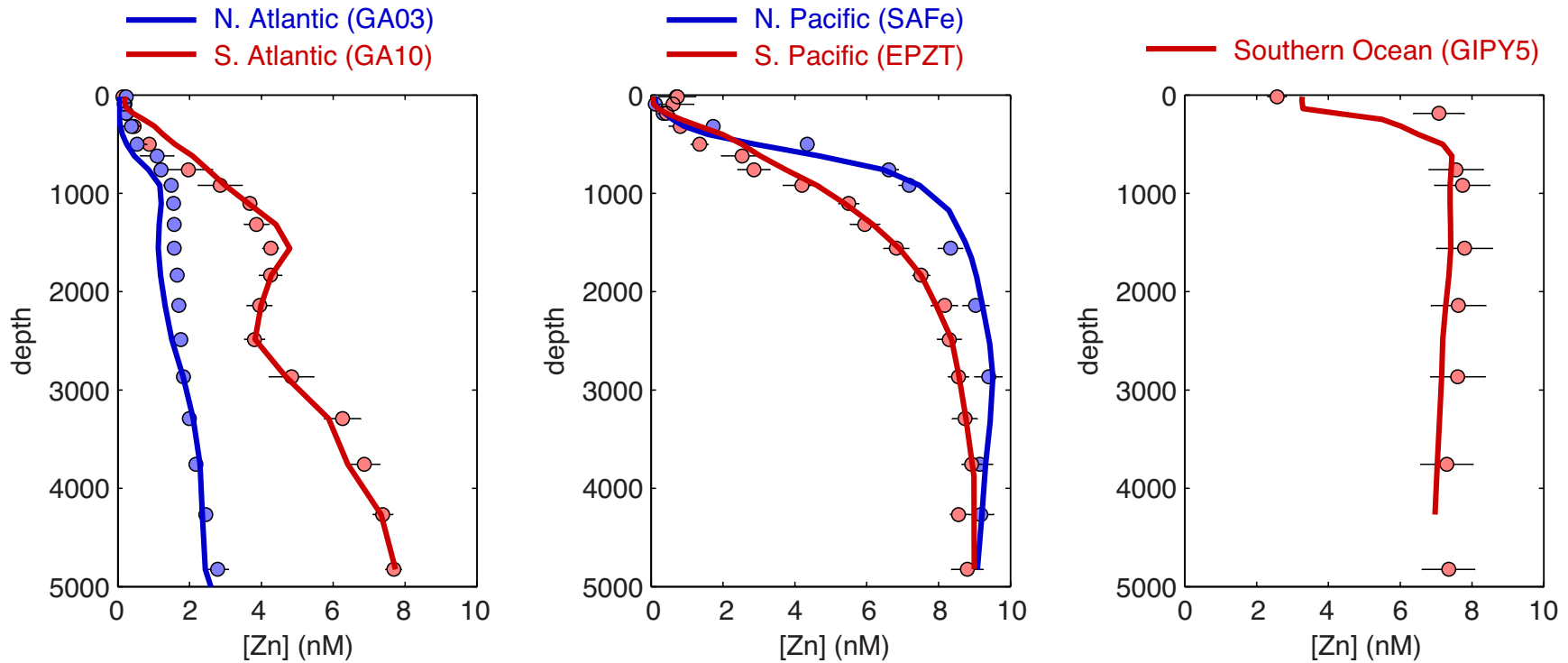
Scavenging
Depth scale as Si
(derived in inverse
Si-cycle model)

Reversible scavenging model:

$$Zn_{ads} = \frac{K_{sc} [\text{particles}]}{K_{sc} [\text{particles}] + 1} Zn$$

Model 3: Scavenging

$$\frac{dZn}{dt} = circ.(Zn) - J_{up}(Zn) - J_{rem}(POZn) - \frac{\partial}{\partial z} wZn_{ads}$$

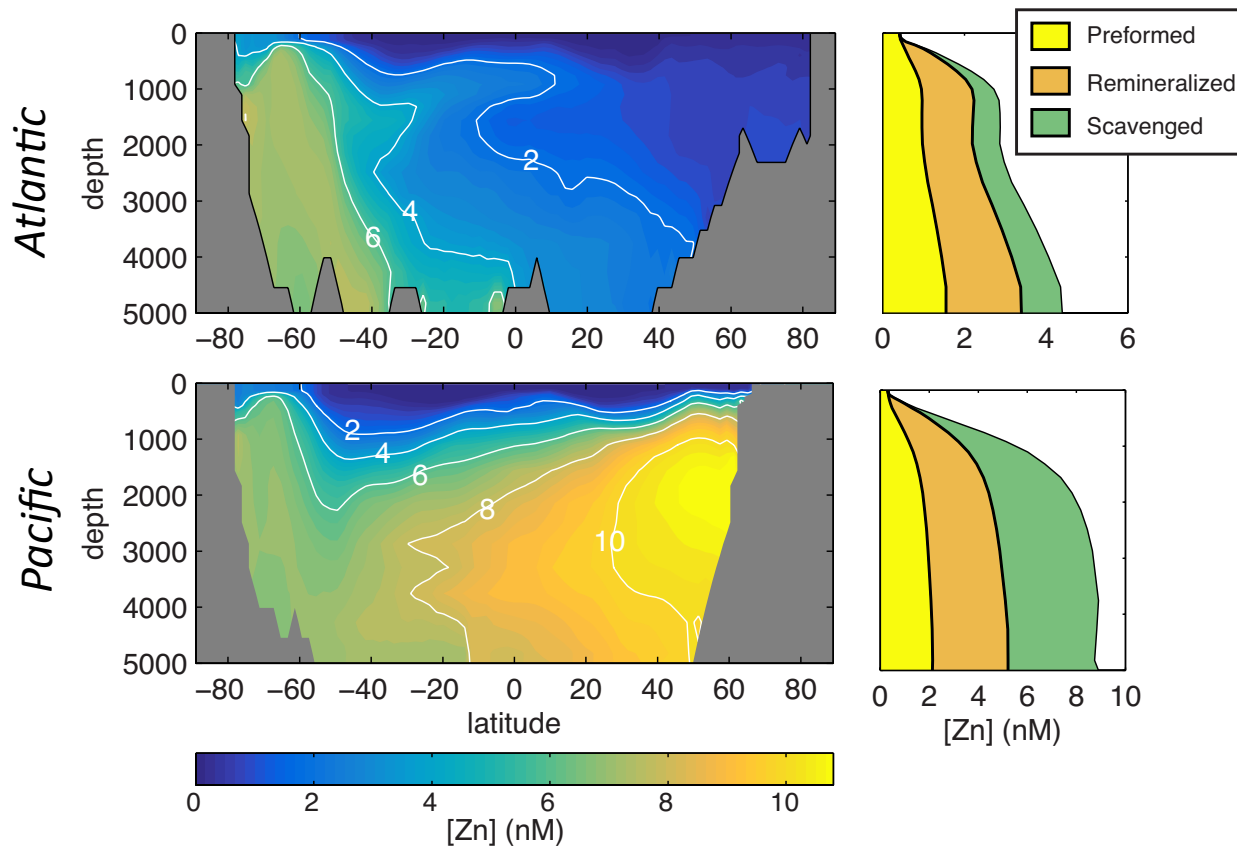


Predicts between 0.01% (deep) and 1% (surface) in adsorbed phase

Limitations and implications

Limitations:

- Lack of direct constraints on adsorbed fraction
- Cannot unravel combination of frustule/scavenging mechanisms

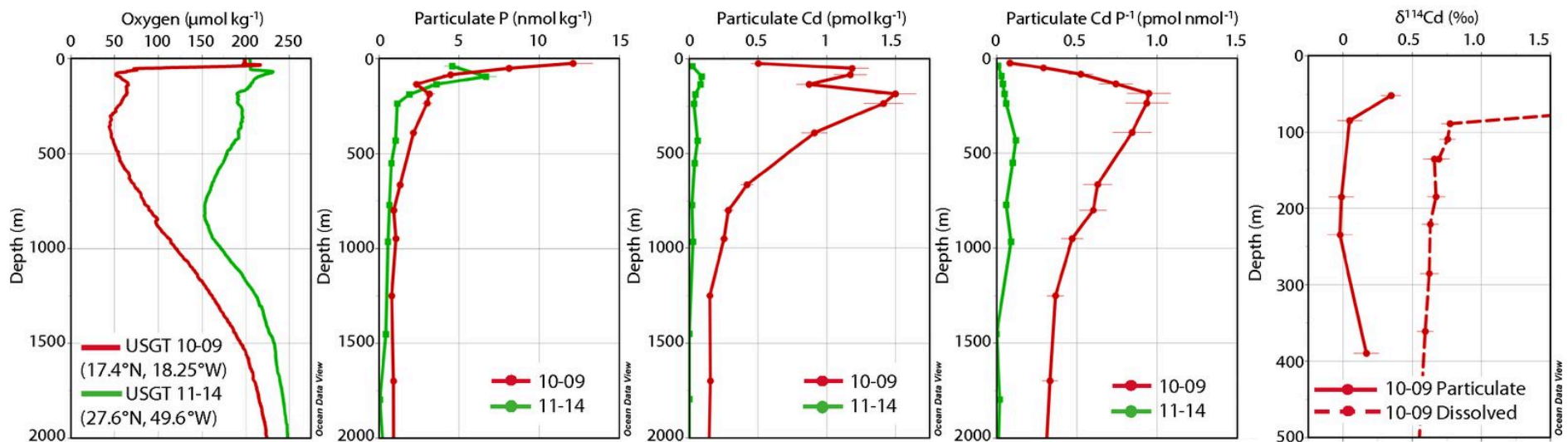


Implications:

- Power of reversible scavenging: <1% in adsorbed phase results in major transfer from intermediate to deep water
- Zn/Si covariance not caused by close mechanistic coupling of their cycles

Mechanisms: Cd scavenging in ODZ

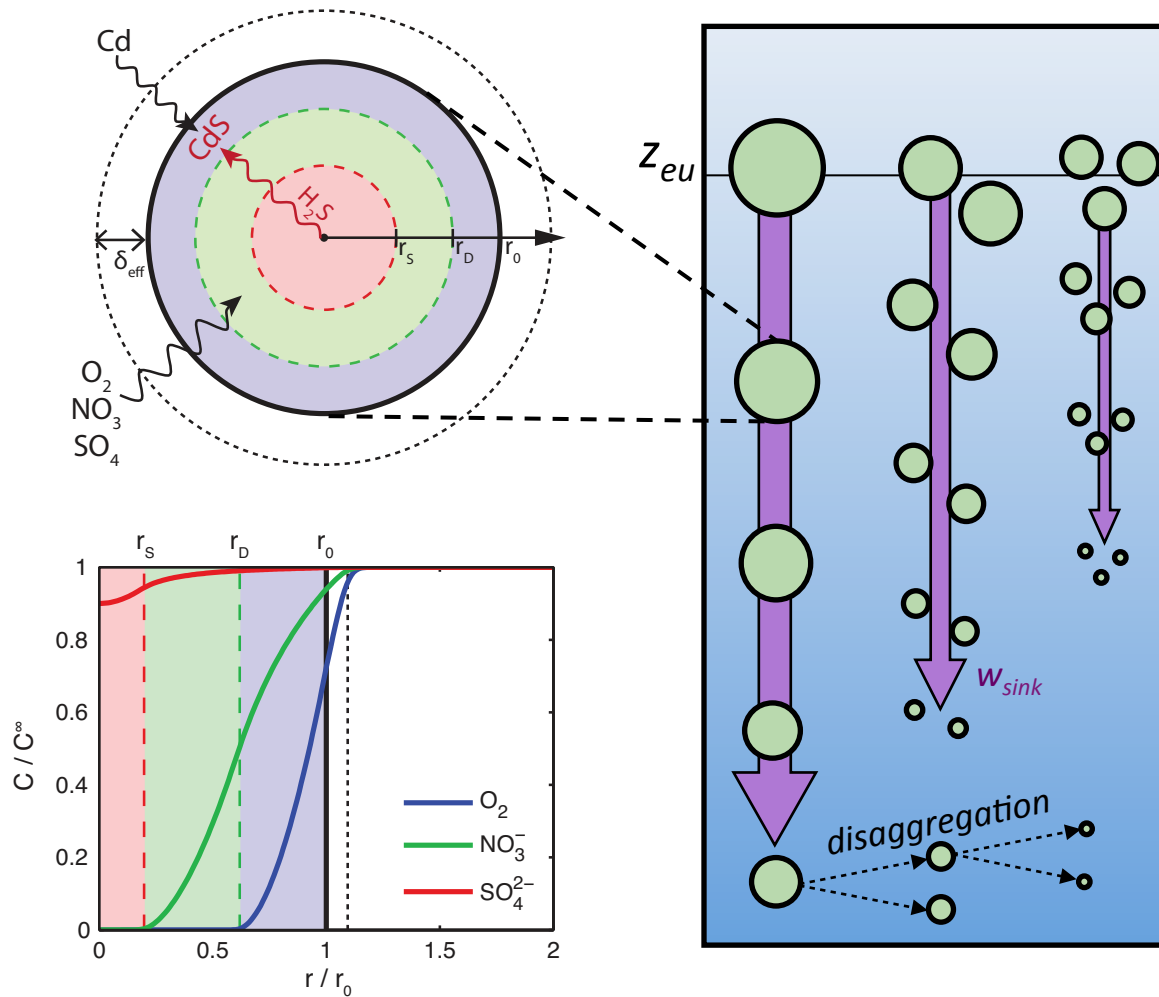
Observation: Scavenging of “light” Cd from dissolved to particulate phase in ODZ in east of GA03 transect.



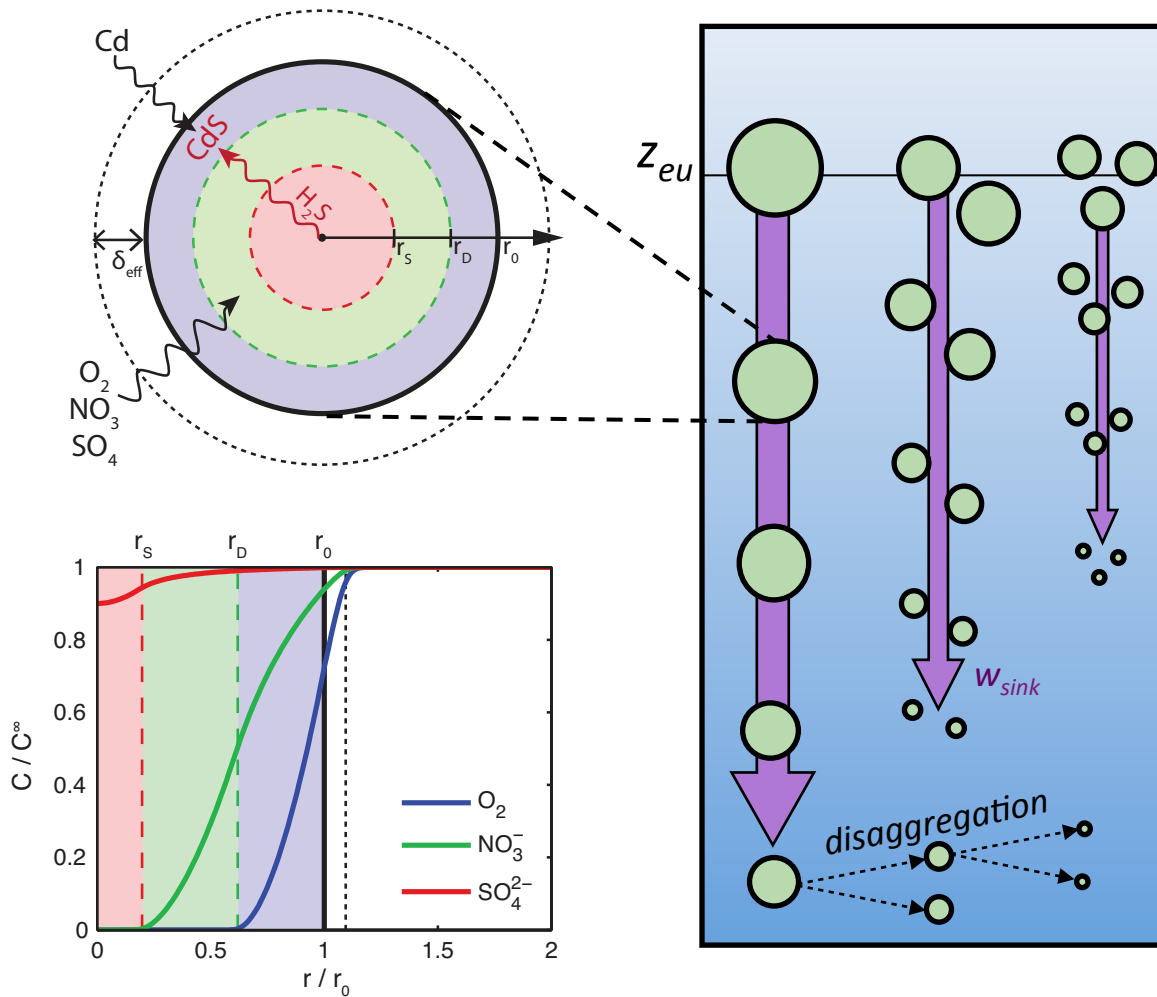
Janssen et al 2014

Hypothesized Mechanism: CdS precipitates in sulfidic “microzones” inside organic particles undergoing rapid respiration

Mechanistic model



Mechanistic model



Boundary conditions:

- Surface particle size spectrum (UVP-TARA)
- Chemical environment (GEOTRACES)

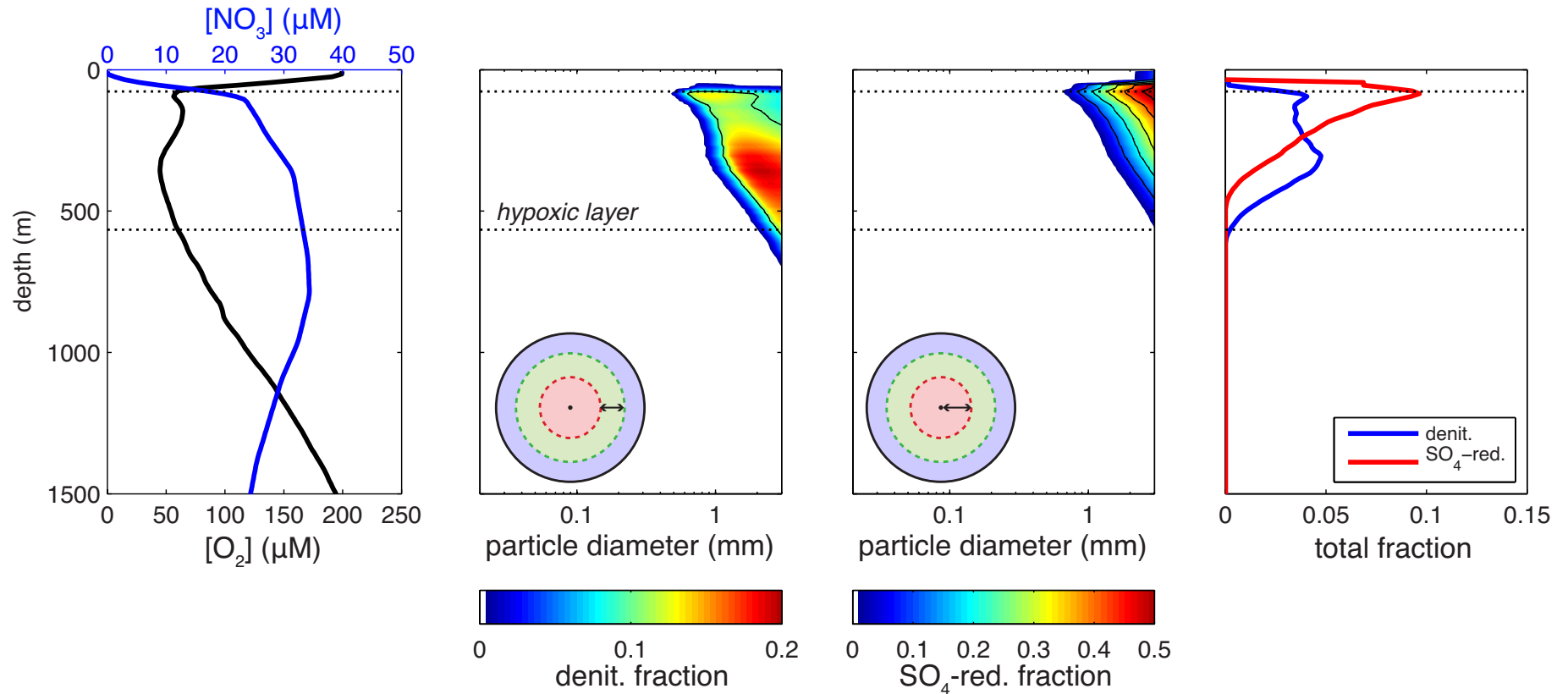
Parameters (and ranges)

- Sinking (Smayda 1971)
- Respiration rates (McDonnell 2015)
- Disaggregation rates
- Fractionation factors

Constraints

- Particulate P, Cd, $\delta^{114}Cd$ (Janssen 2014)
- Particle size spectrum (UVP-TARA)

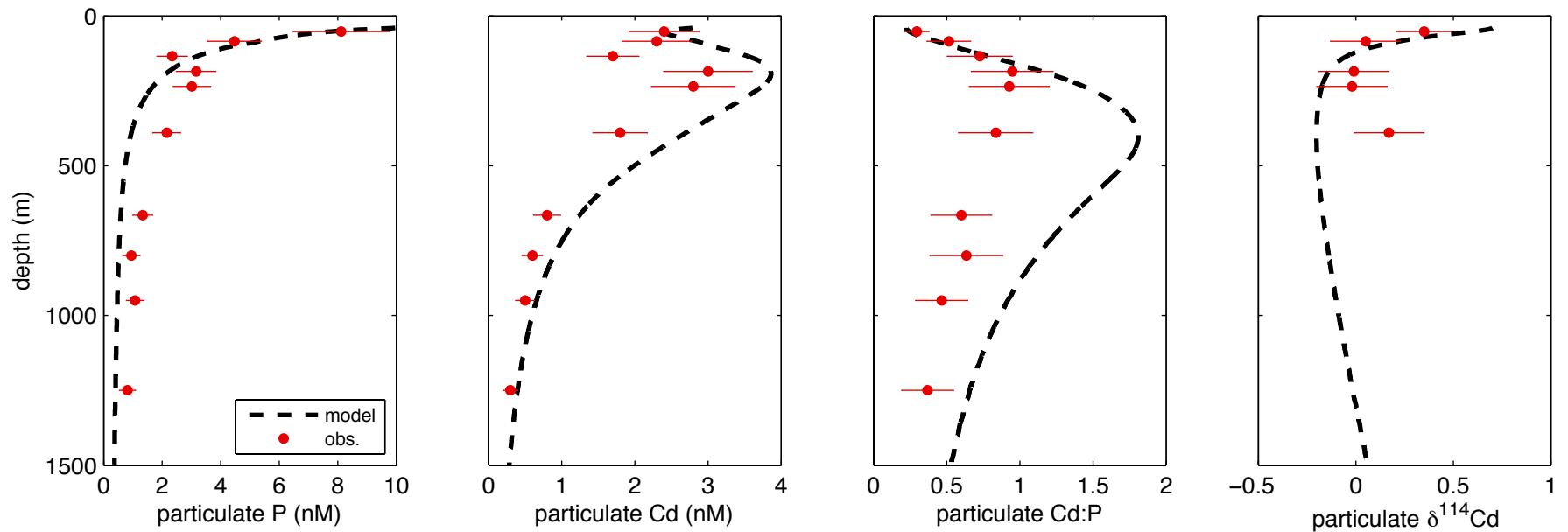
Trop. Atlantic particle microzones



Predictions from model with **prior parameters** (“best-guess” from observations)

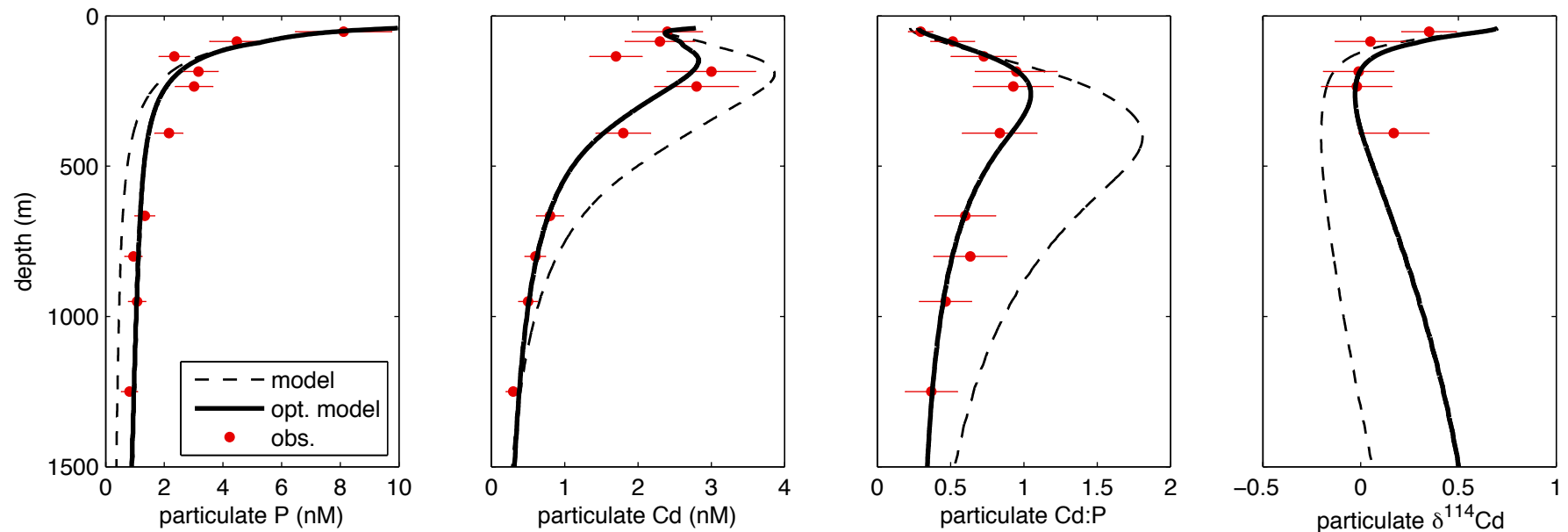
Particles $>1\text{mm}$ can precipitate CdS (and then disaggregate)

Trop. Atlantic particle chemistry



Predictions from model with **prior parameters** (“best-guess” from observations)

Trop. Atlantic particle chemistry



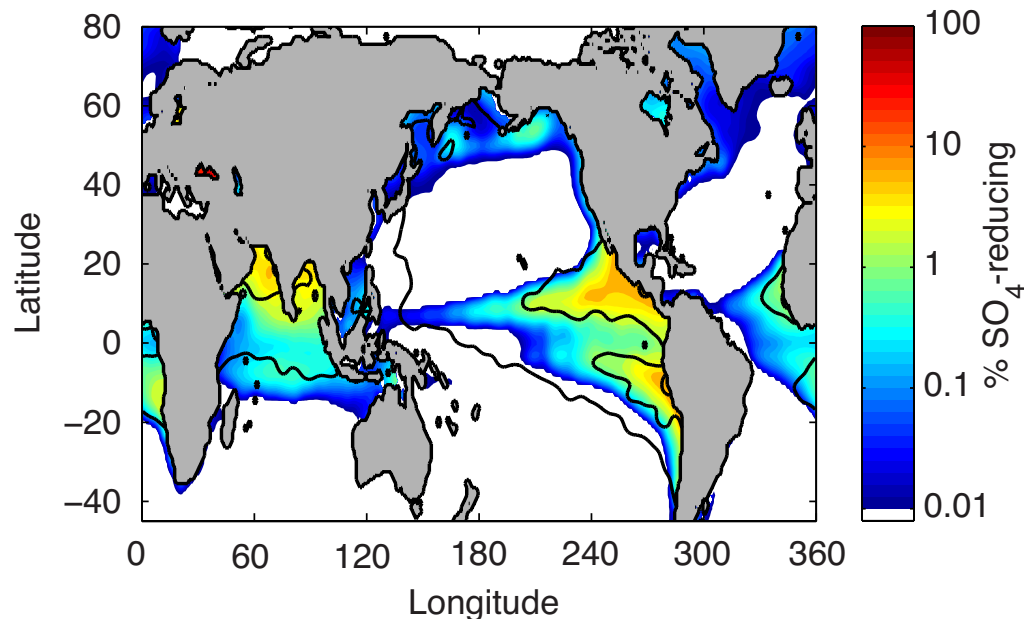
Predictions from model with **optimized parameters**

(Requires minor reduction in respiration rates, remain within observed range)

Limitations and implications

Limitations:

- Highly idealized model particles (spherical geometry, “fast” communities)
- *Cannot prove* the CdS-precipitation mechanism (need new observations), can only show that it *is consistent with observations* and other constraints.



Implications:

- Potential for CdS precipitation through tropics and N.Pacific (identifies new regions for observations)
- Major sink and redistribution process for all chalcophile trace metals?

Concluding remarks

- New scavenging insights from data-constrained models:
 - Hydrothermal Fe lifetime of ~35yrs (limited supply to surface)
 - Reversible scavenging can explain global covariation of Zn and Si
 - Particulate observations in Tropical Atlantic are consistent with CdS precipitation mechanism
- Important take-homes:
 - Inverse modeling methods can be applied to wide range of datasets to answer a wide array of questions
 - Can help *you* test *your* ideas, and interpret *your* observations
 - Best outcomes from observationalist-modeler collaboration!