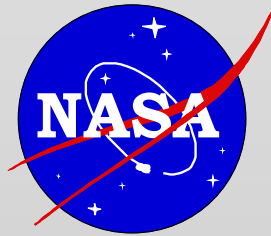


Incorporating Satellite Time-Series data into Modeling



Watson Gregg

NASA/GSFC/Global Modeling and Assimilation Office

Topics:

Models, Satellite, and In situ representations of

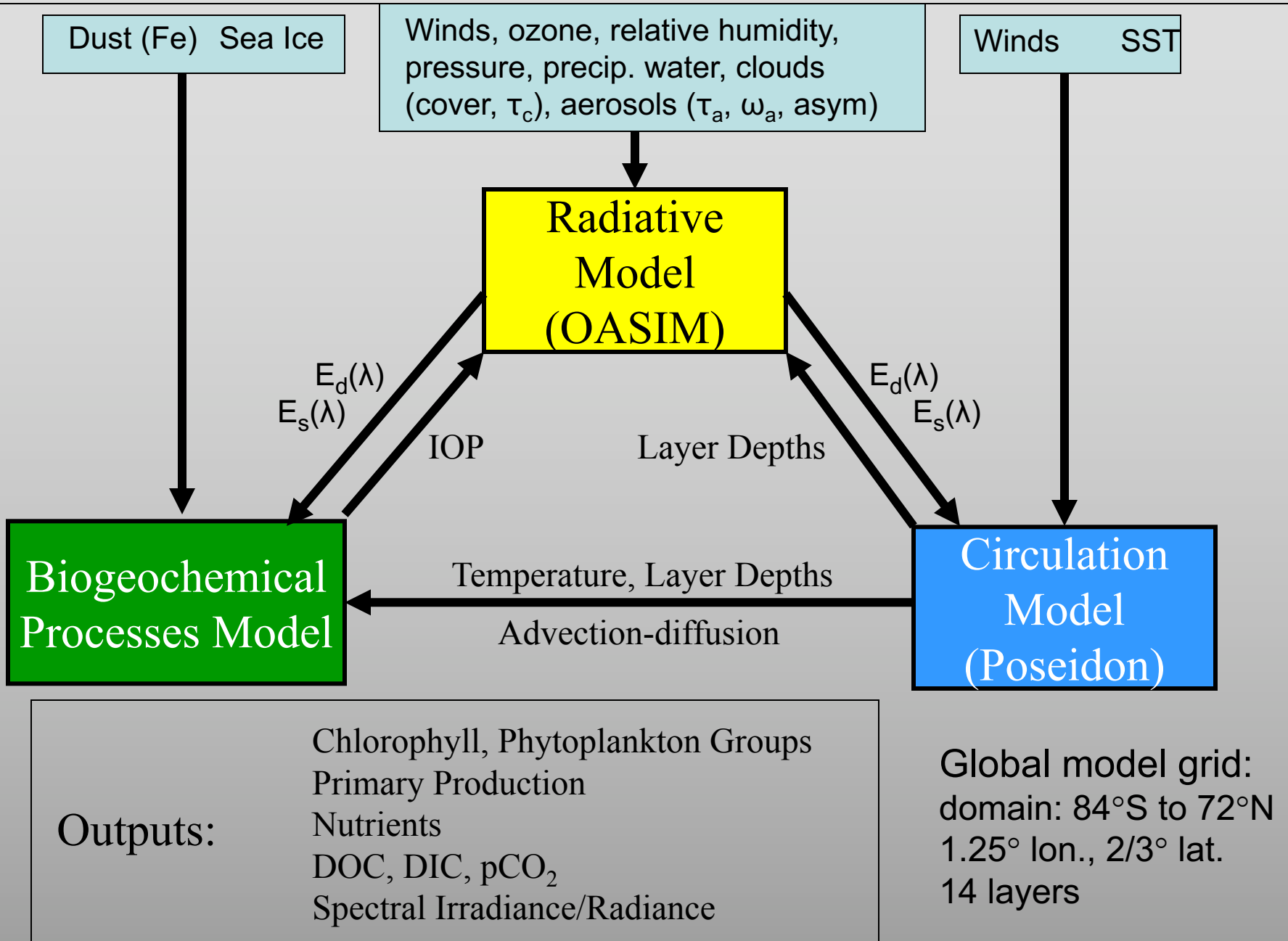
Seasonal Variability

Interannual Variability

Decadal and Longer Trends

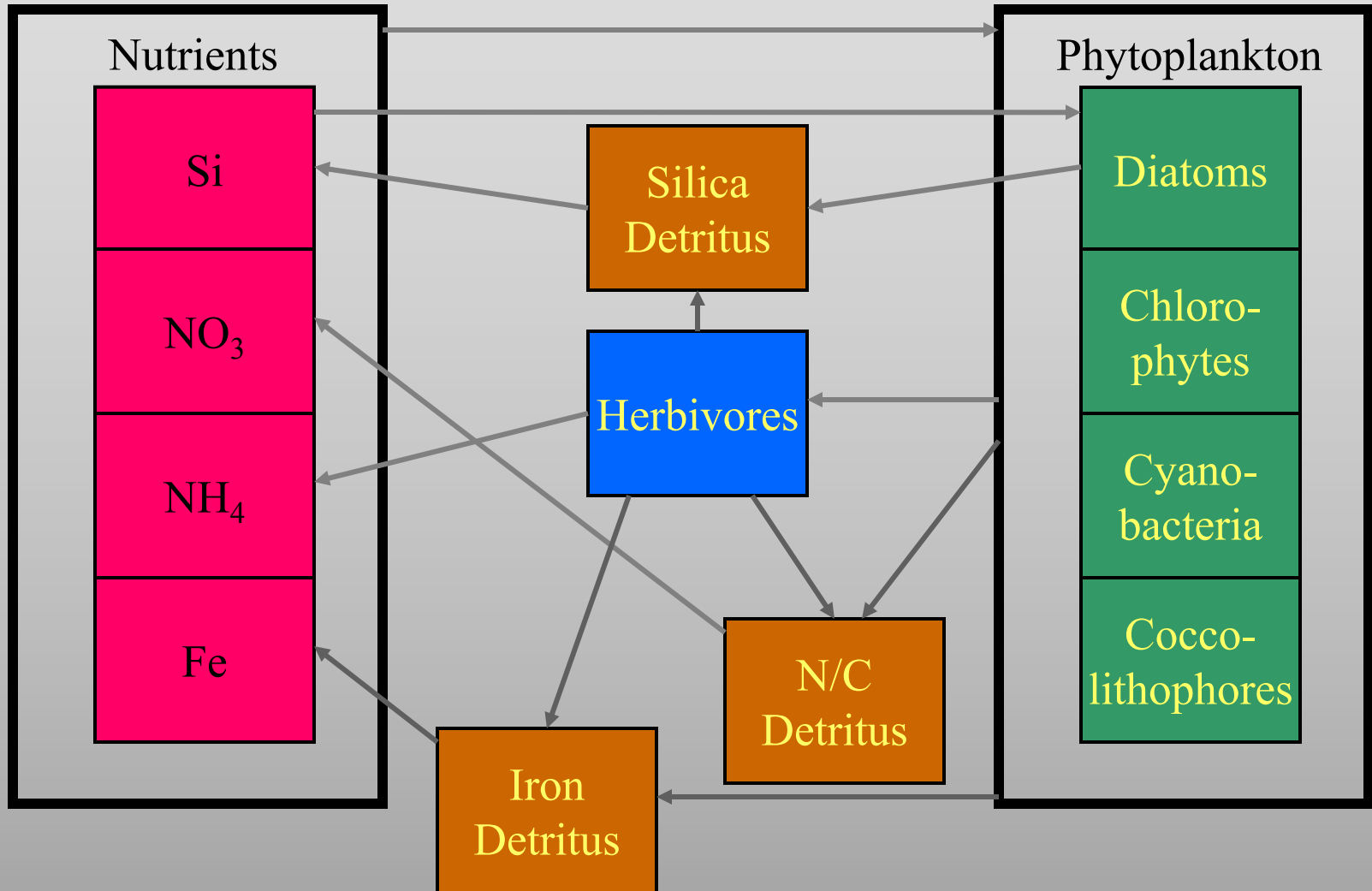
Supporting data and publications: Google gmao, click Research, then Ocean Biology Modeling (<http://gmao.gsfc.nasa.gov/research/oceanbiology>)

NASA Ocean Biogeochemical Model (NOBM)

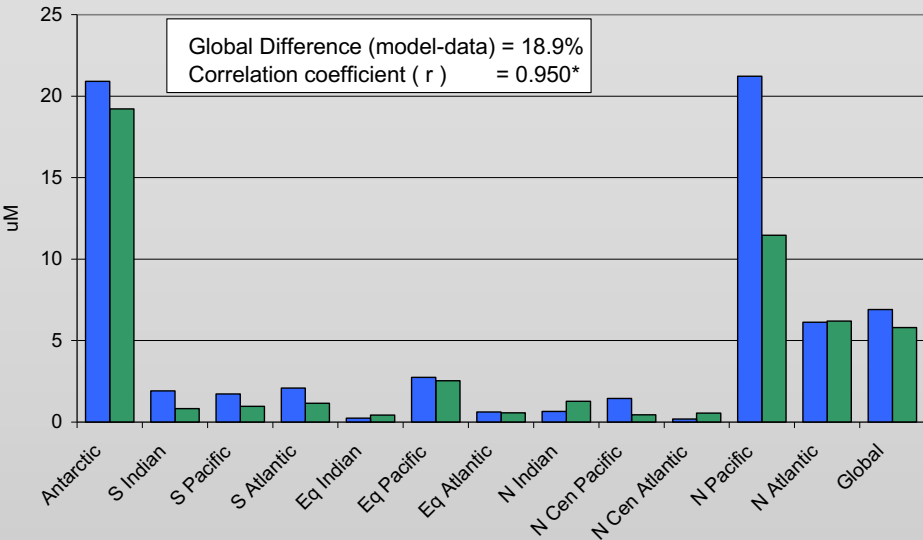


Biogeochemical Processes Model

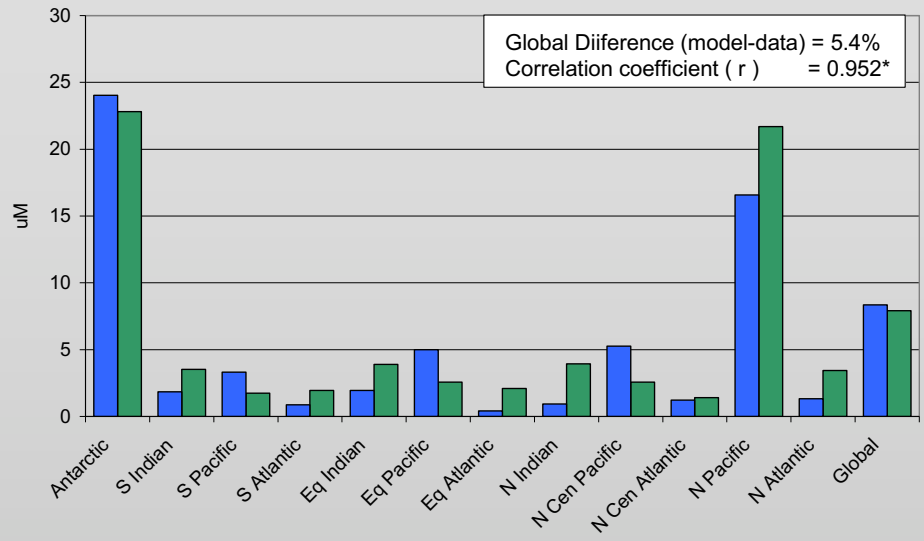
Ecosystem Component



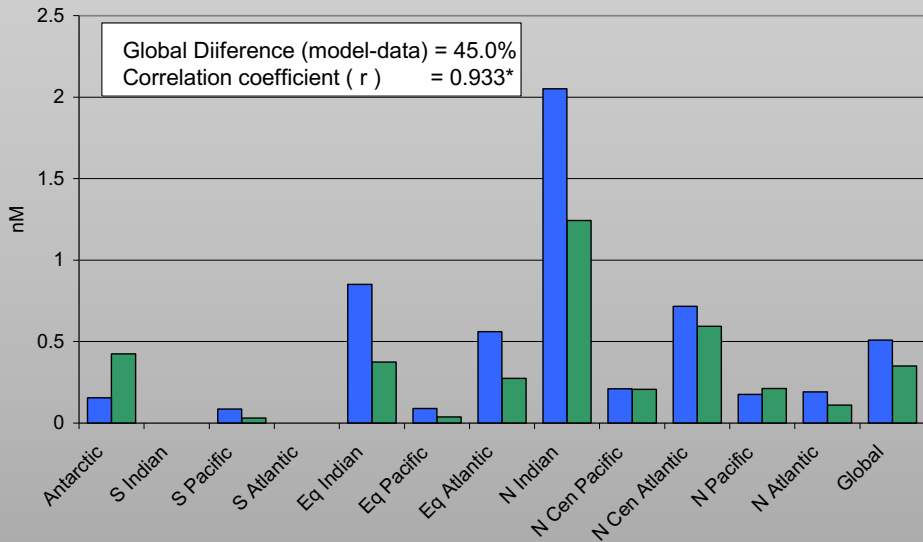
Nitrate



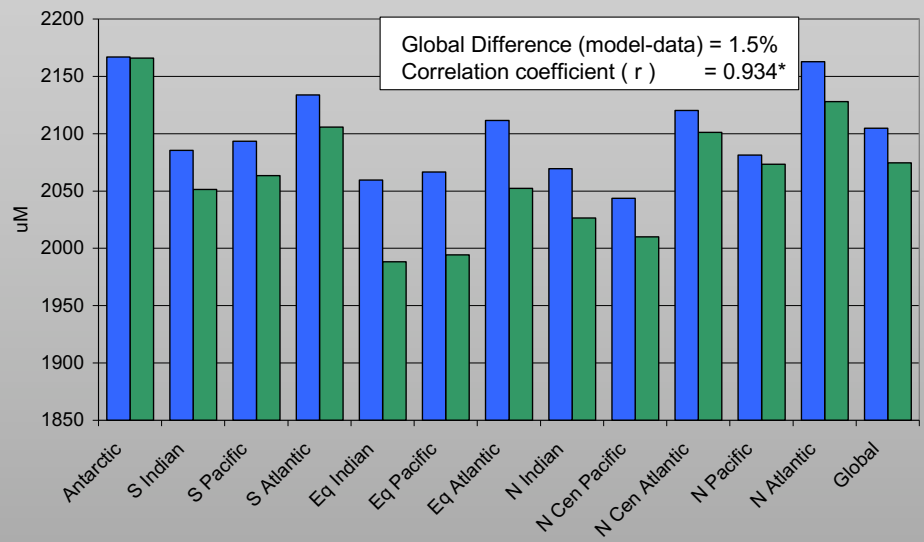
Silica



Dissolved Iron

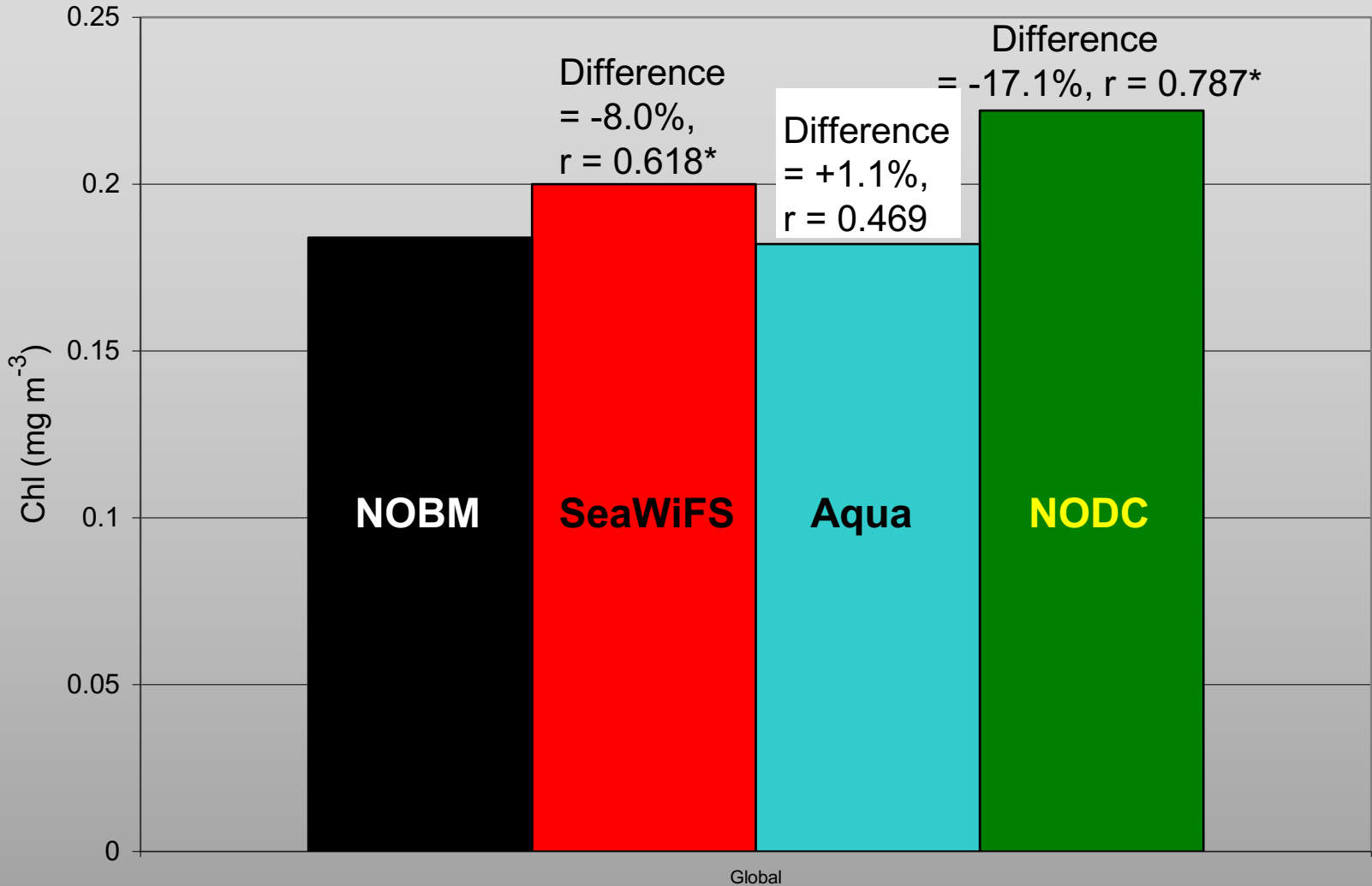


Dissolved Inorganic Carbon



Blue = NOBM; Green = Data

Global Chlorophyll



Data Assimilation

Time has finally come

>50 papers using data, 12 using satellite data
(Gregg et al., Journal of Marine Systems, in press)

In ocean biology, Two Classes:

Variational (e.g., adjoint)

Sequential (e.g., Kalman Filter)

Here we used Sequential Methodologies,
Conditional Relaxation Analysis Method
Ensemble Kalman Filter

Application to Ocean Color

Daily assimilation of gridded
data into surface layer

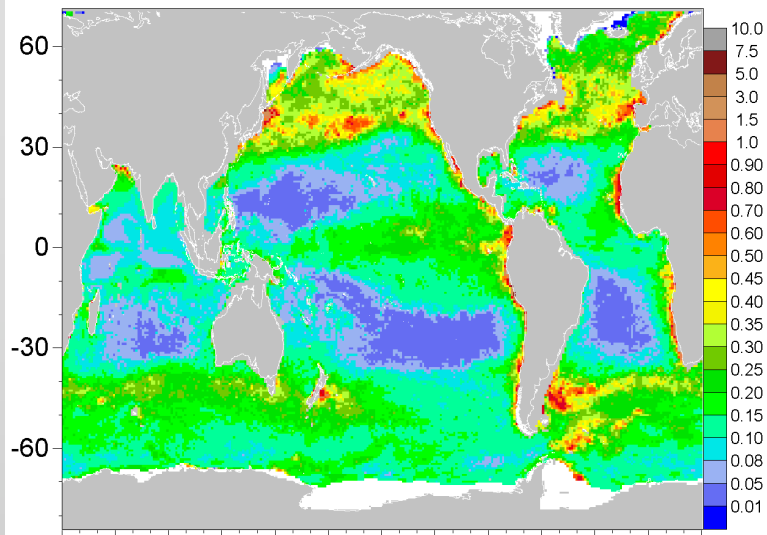
Chlorophyll distribution log-normal

assimilate logarithmic quantities

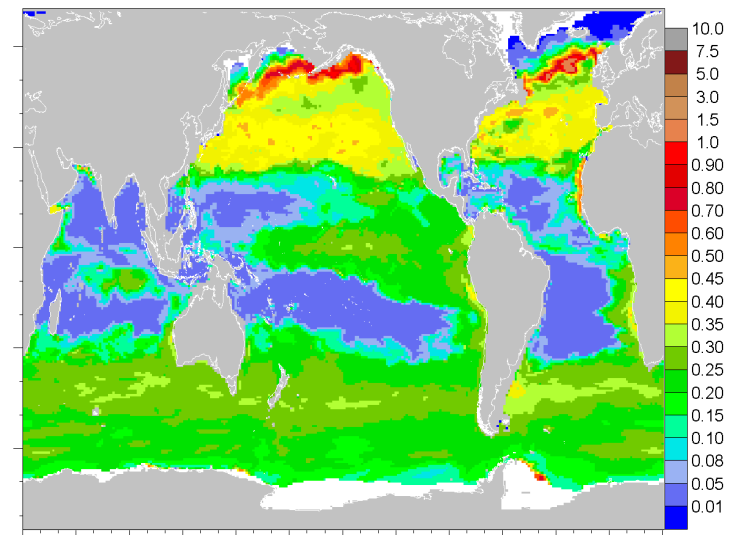
Satellite errors can affect results

explicitly define regional satellite errors estimated
from global analysis of in situ data

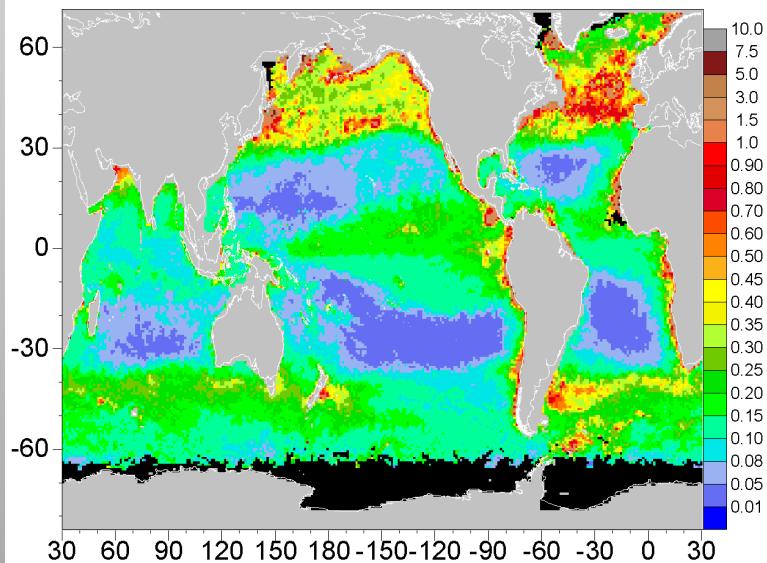
Assimilated Chlorophyll Apr 1 2001



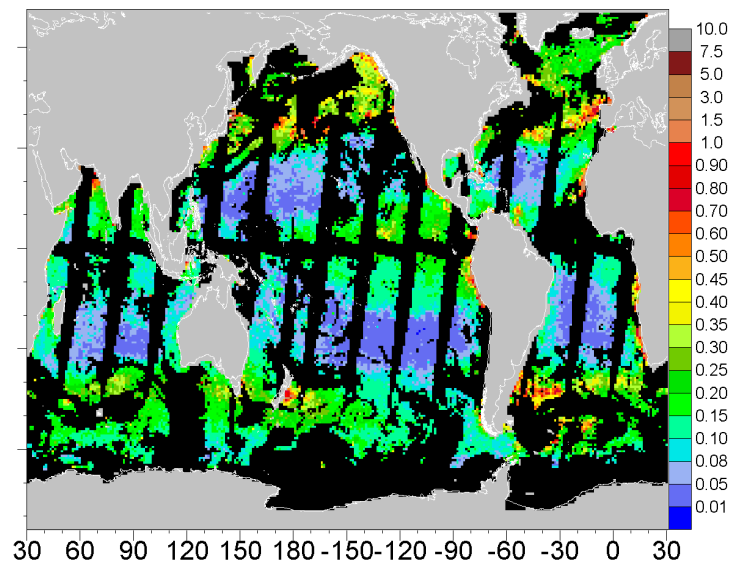
Free Run Model Chlorophyll Apr 1 2001



Monthly SeaWiFS Chlorophyll Apr 2001



Daily SeaWiFS Chlorophyll Apr 1 2001

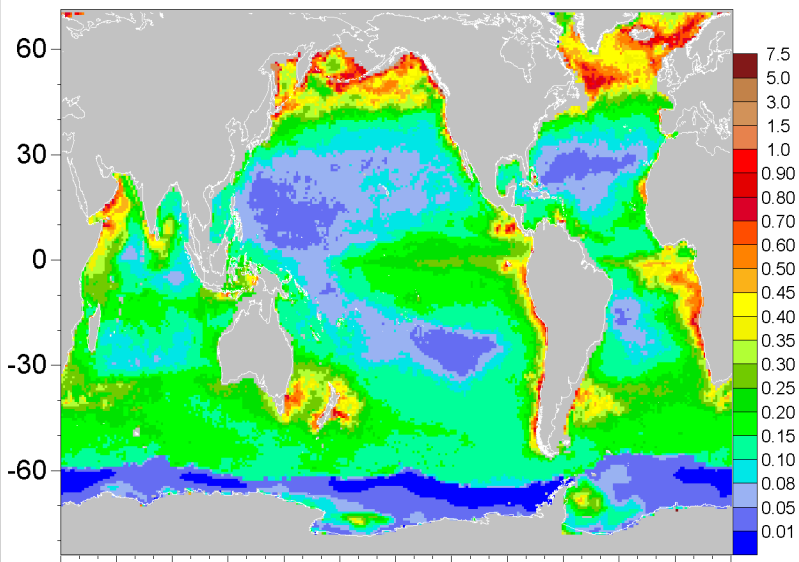


Gregg, 2008. *Journal of Marine Systems* 69: 205-225.

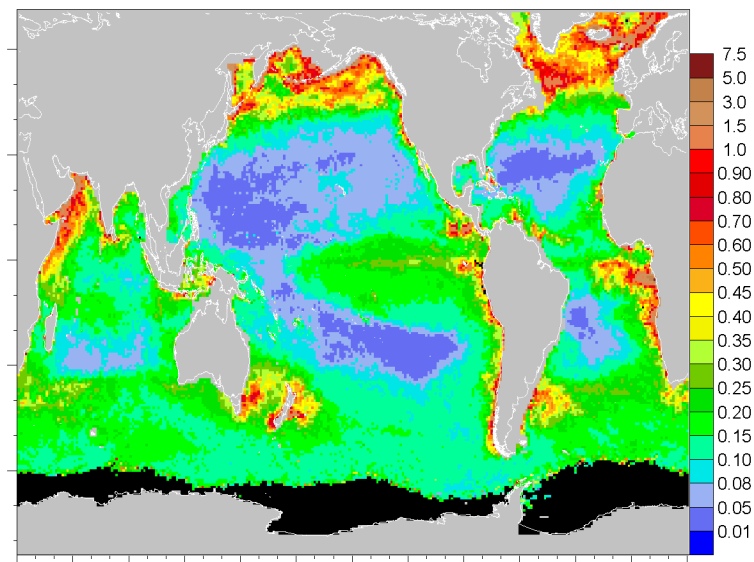
Nerger and Gregg, 2008. *Journal of Marine Systems*, in press.

Nerger and Gregg, 2007. *Journal of Marine Systems* 68: 237-254.

Assimilated Chlorophyll Sep 2001

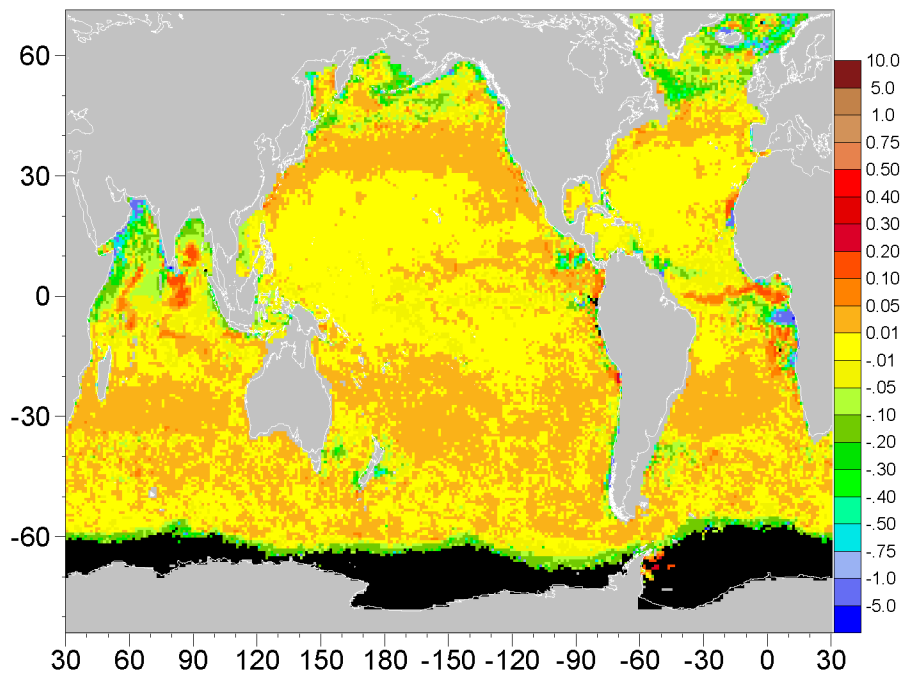


SeaWiFS Chlorophyll Sep 2001



NASA
Ocean
Biogeochemical
Model
(NOBM)

Difference (Assim-SeaWiFS) Sep 2001



Assim Model vs.
SeaWiFS:

Bias = +5.5%

Uncertainty = 10.1%

Compared to In situ Data

	Bias	Uncertainty	N
SeaWiFS	-1.3%	32.7%	2086
Free-run Model	-1.4%	61.8%	4465
Assimilation Model	0.1%	33.4%	4465

Estimate of in situ data uncertainty: 22%

Seasonal Variability

Chlorophyll mg m^{-3}

BATS
Shown in Blue

SeaWiFS

MODIS-Aqua

Free-run Model

Assimilation Model

Apr

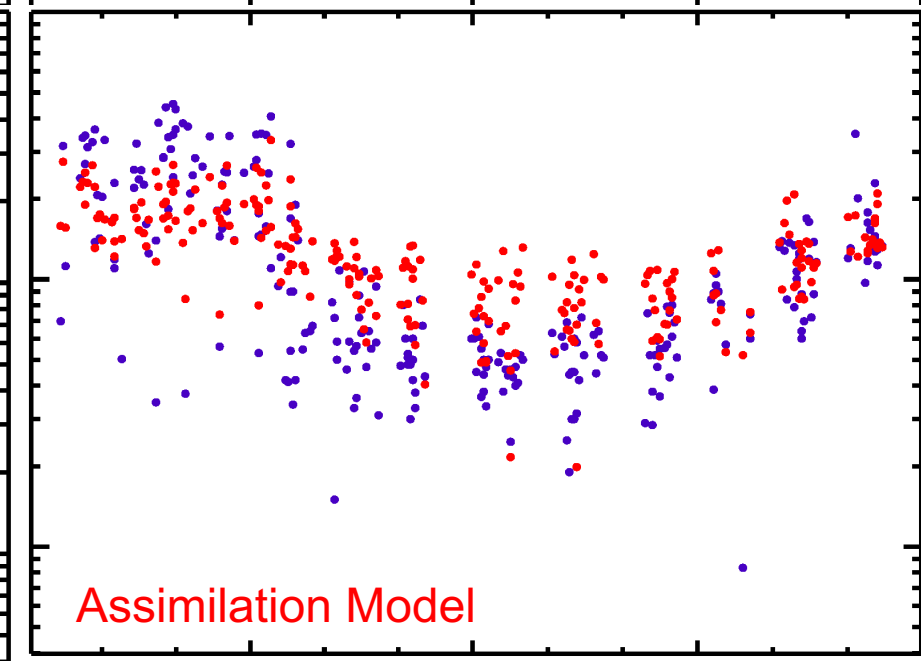
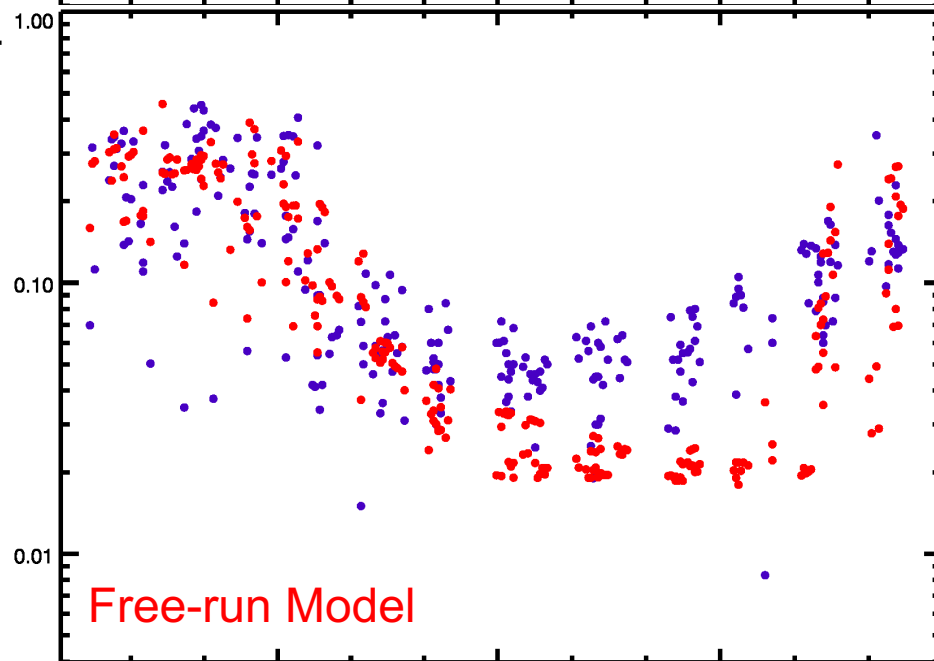
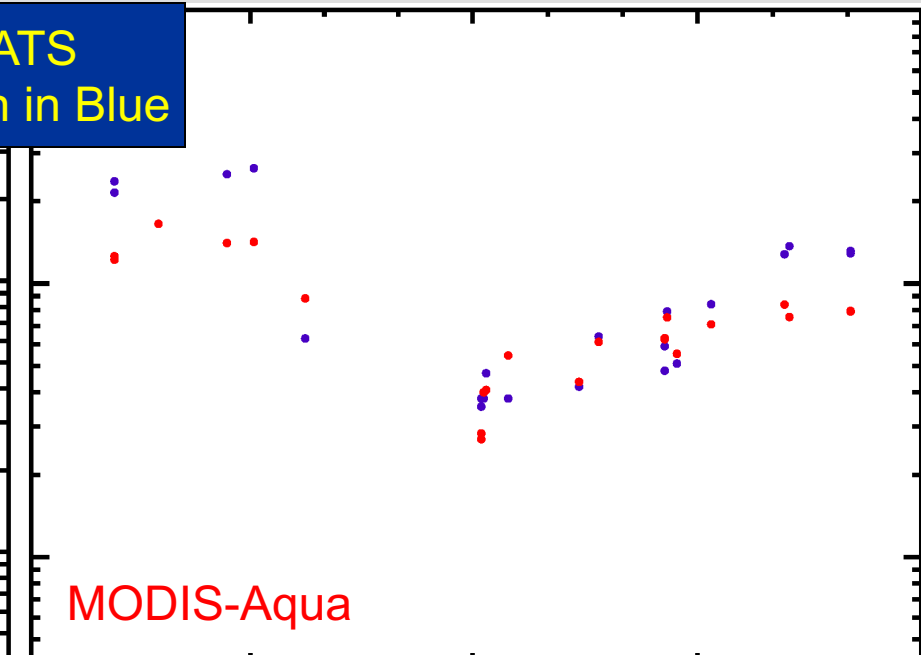
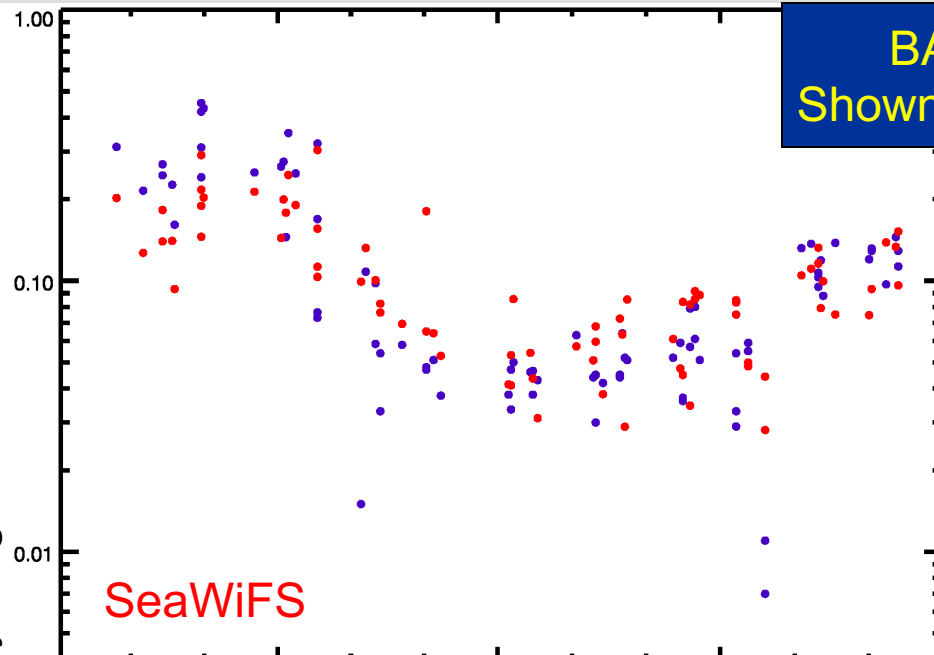
Jul

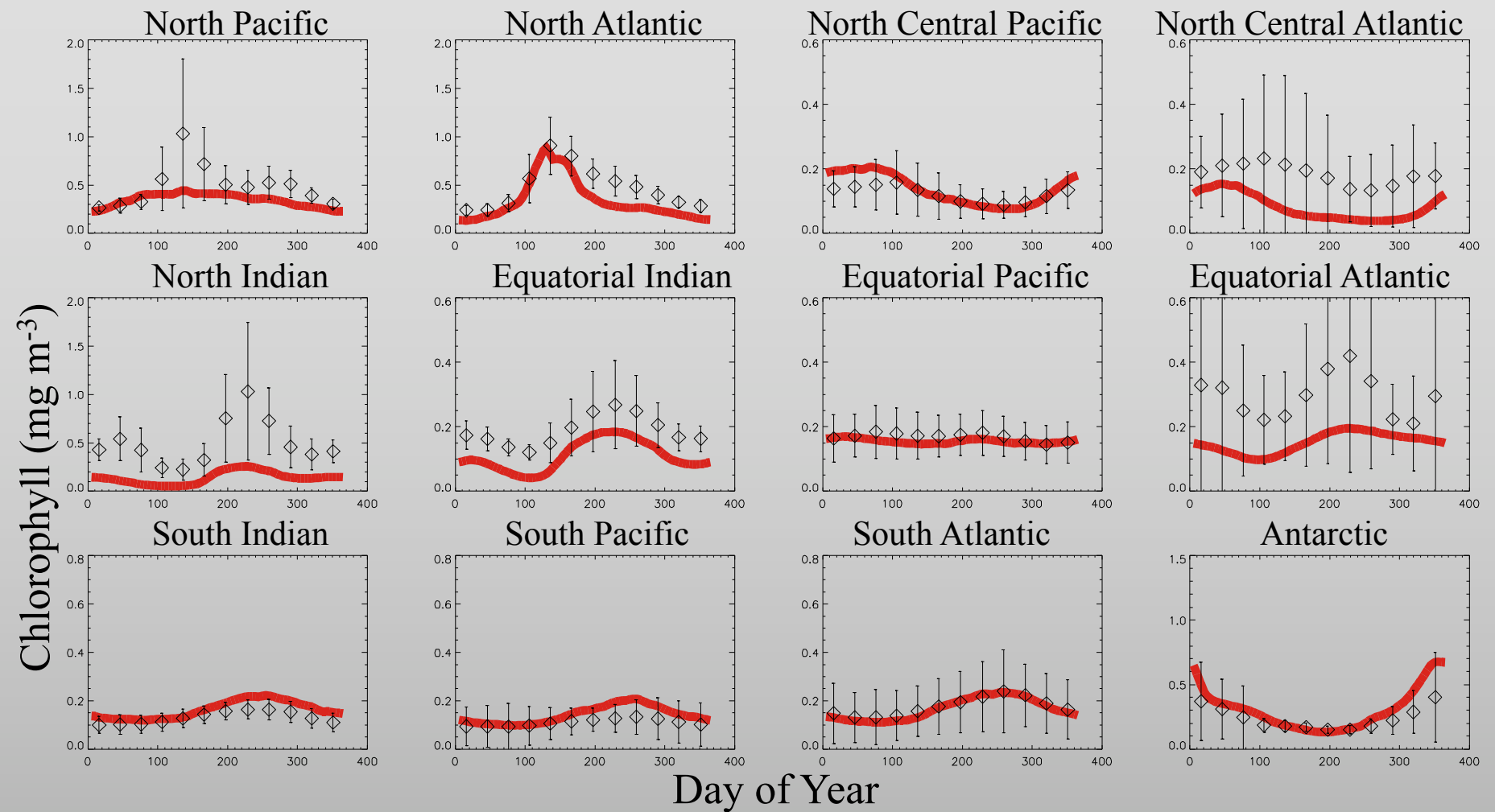
Oct

Apr

Jul

Oct





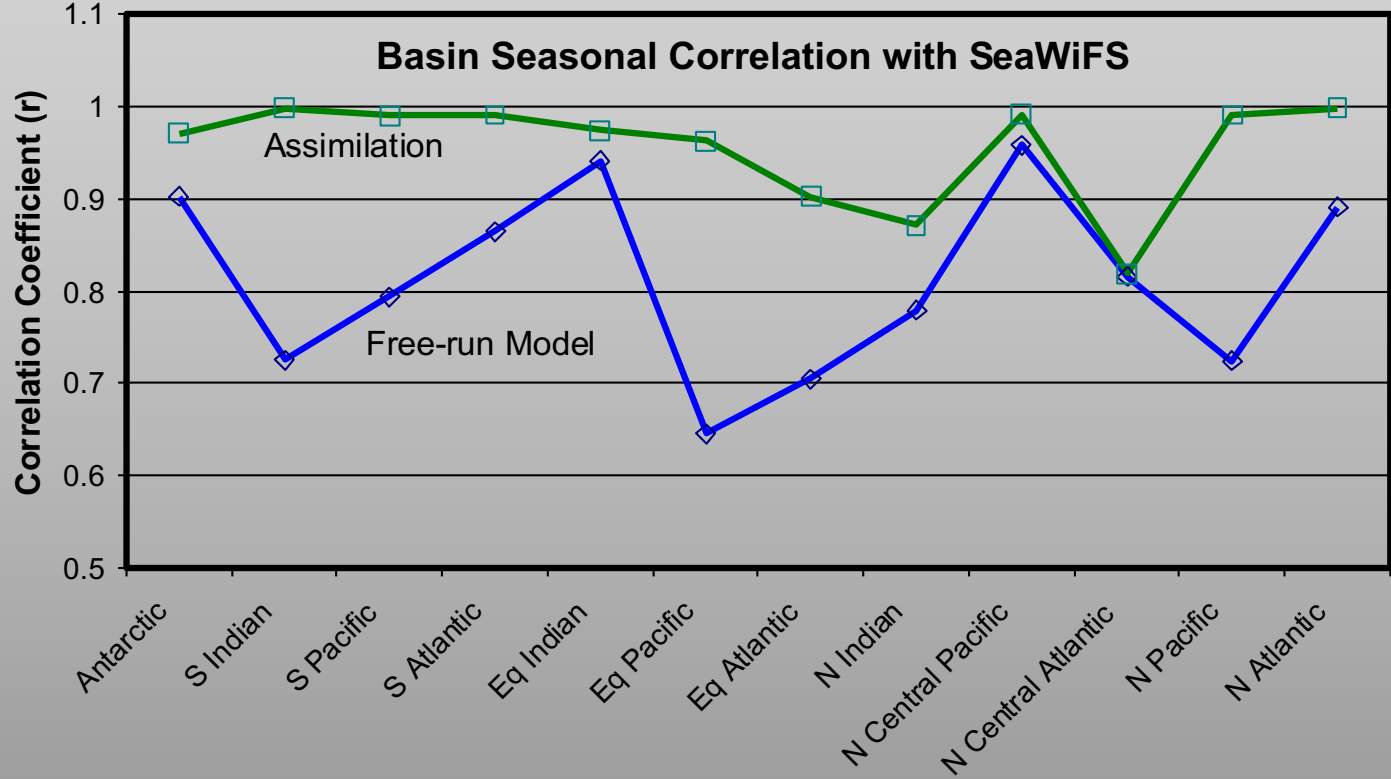
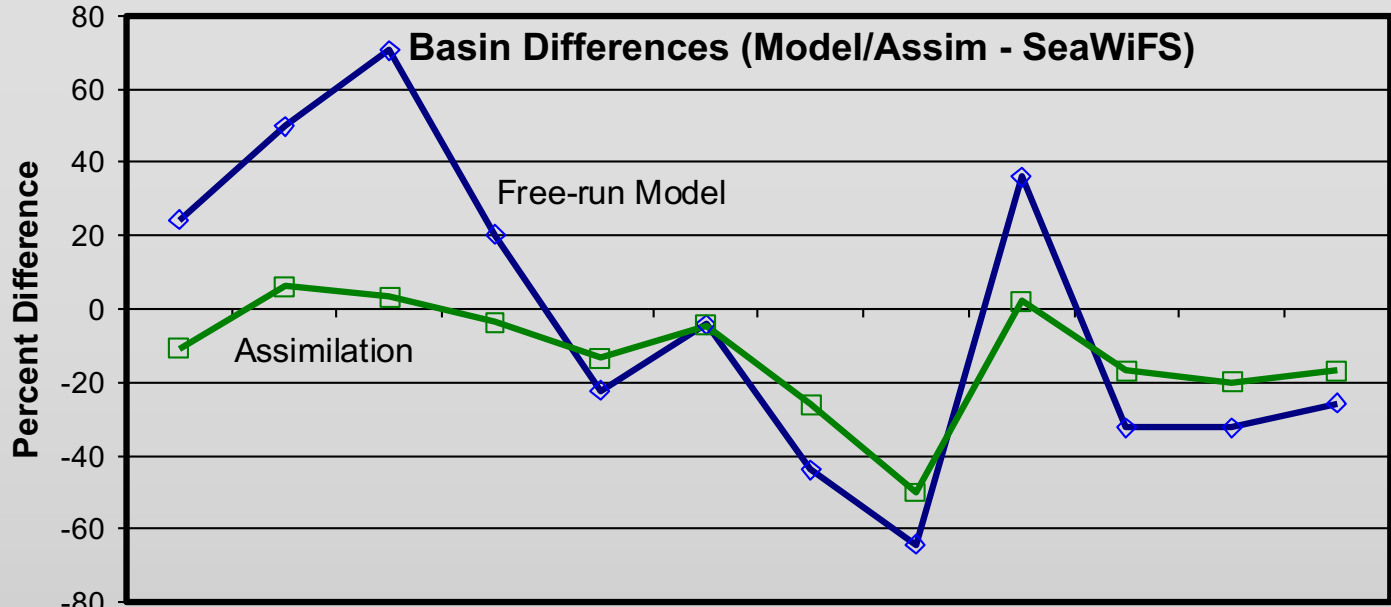
Statistically positively correlated ($P < 0.05$) all 12 basins

Red = model

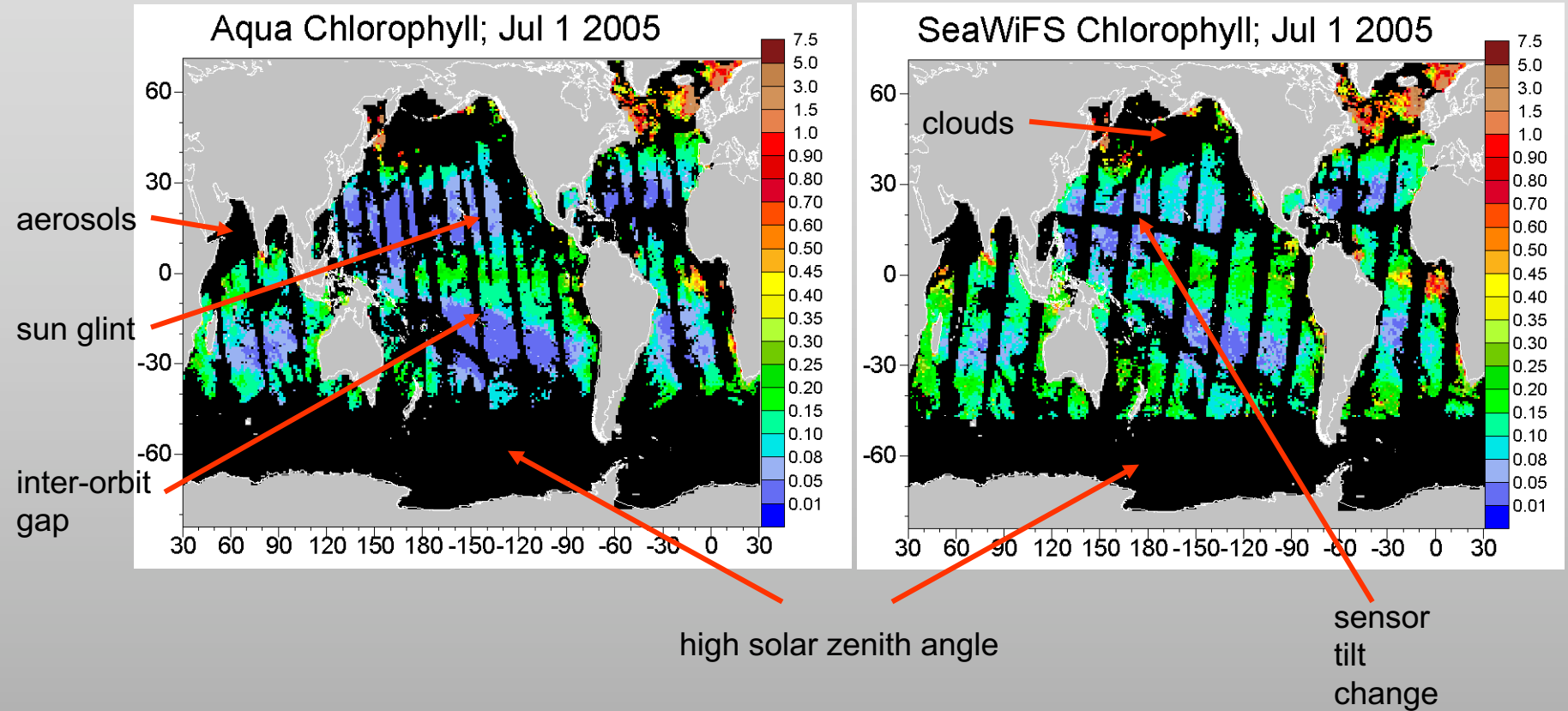
Diamonds = SeaWiFS monthly mean

Gregg, 2002, Deep-Sea Research II

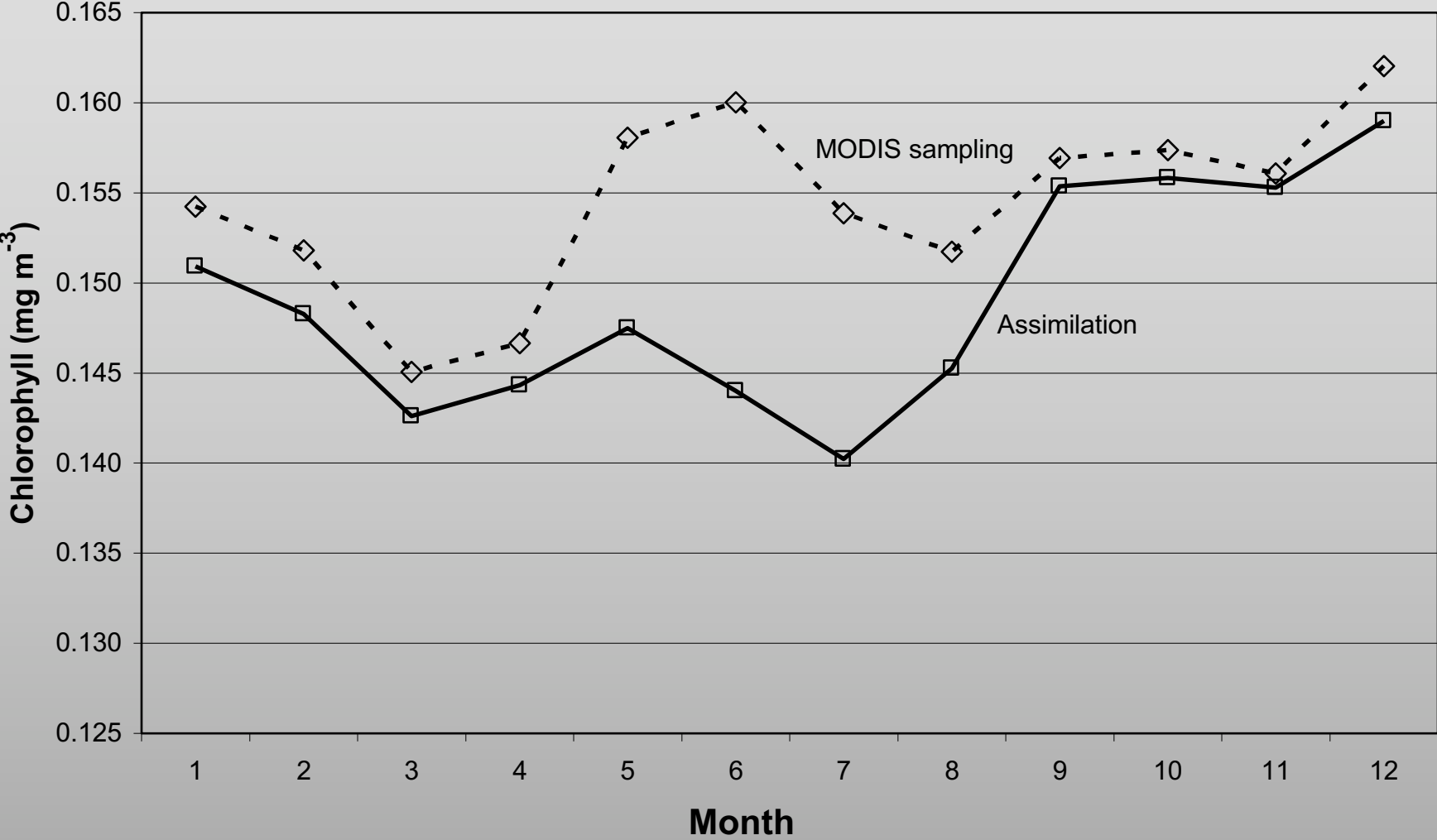
Gregg et al., 2003, Deep-Sea Research II



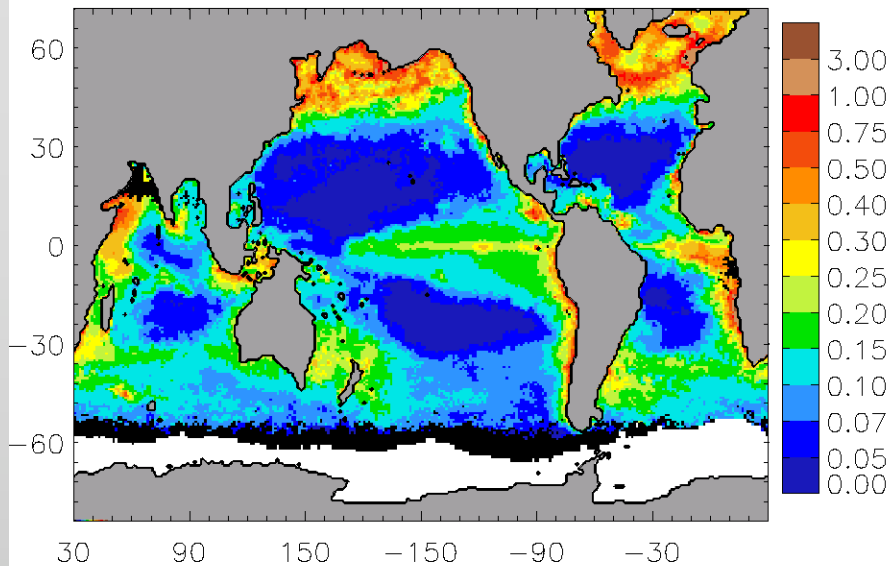
Daily ocean coverage by MODIS-Aqua and SeaWiFS.



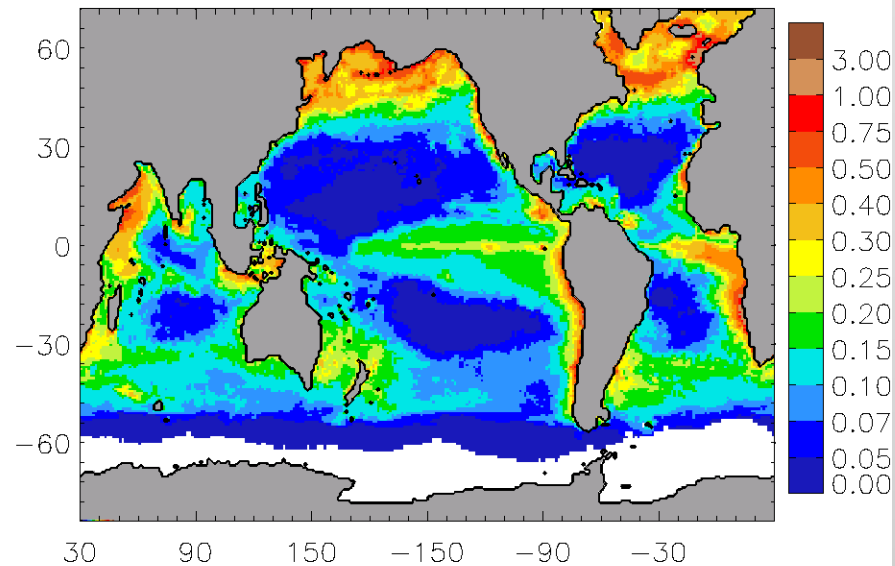
Monthly Mean Global Chlorophyll



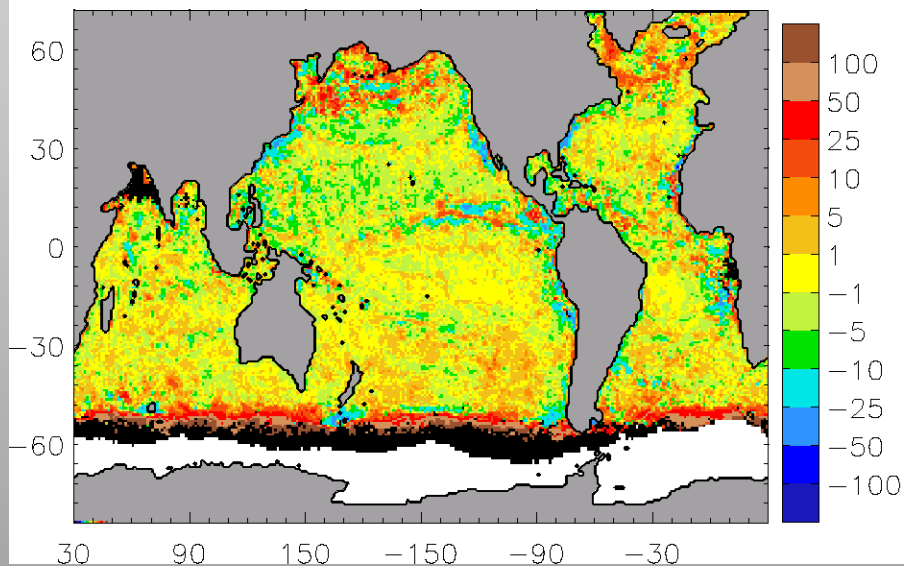
MODIS Sampling Aug 2003



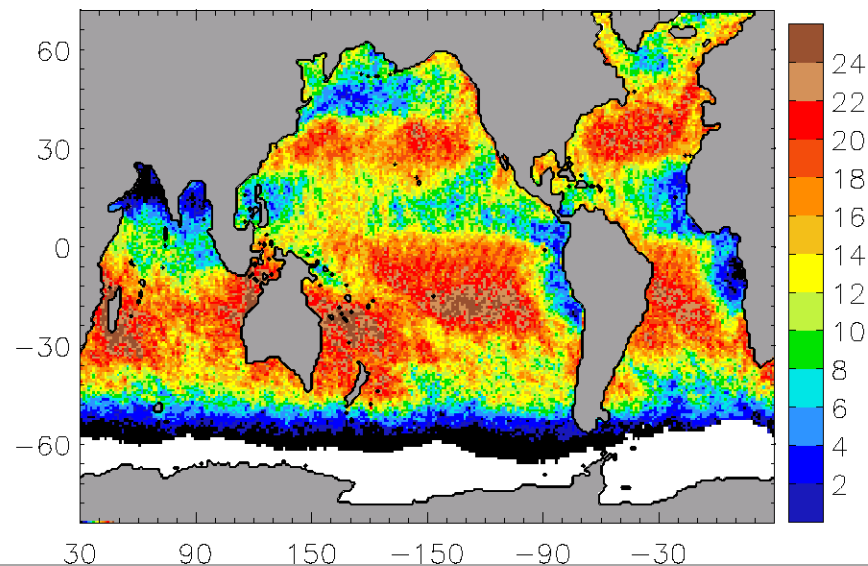
Assimilation Aug 2003



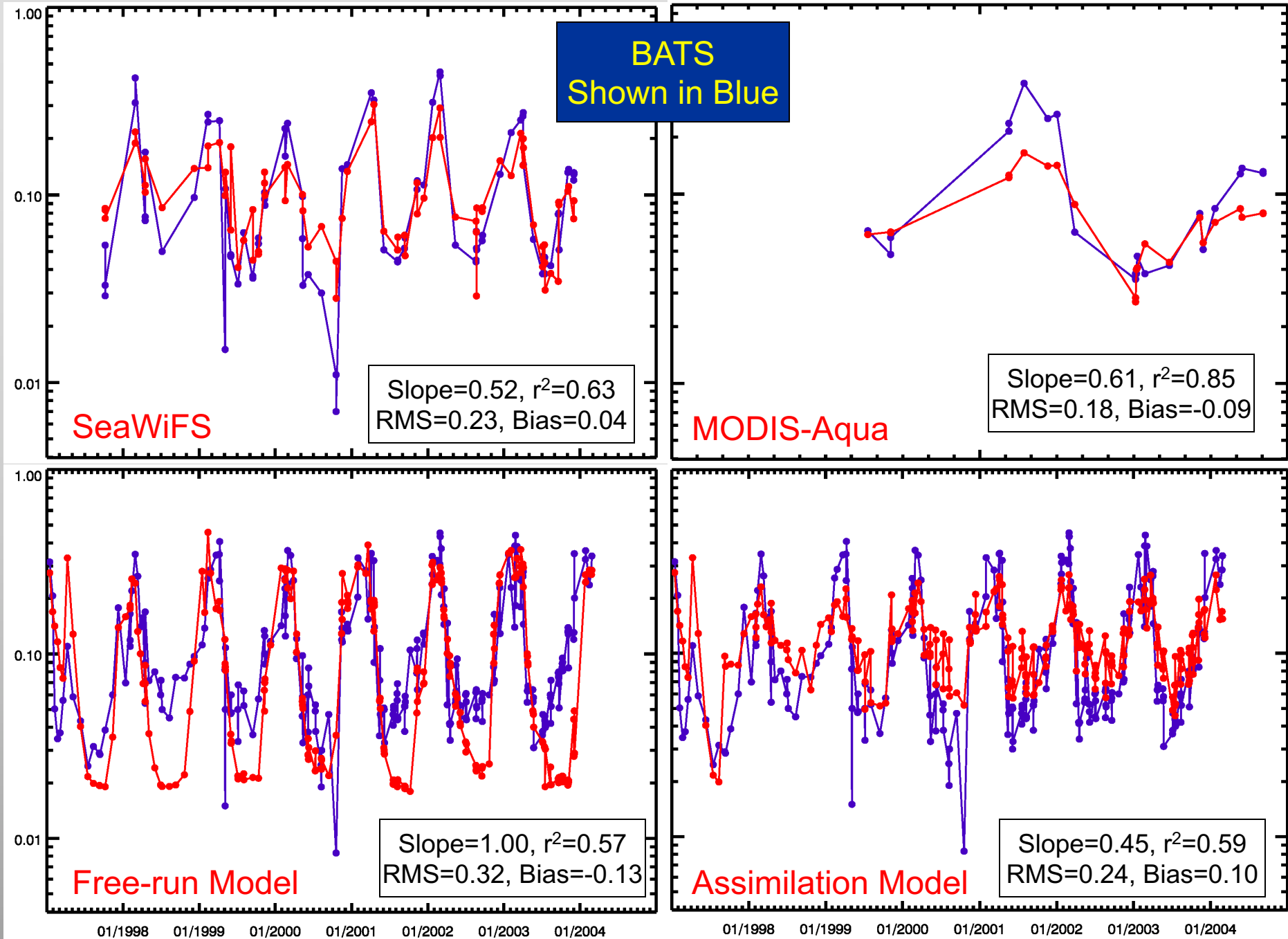
Difference (MODIS-assimilation) Aug 2003



No. Days Sampled by MODIS Aug 2003



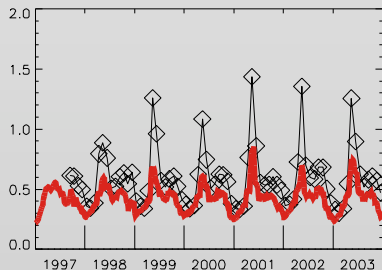
Interannual Variability



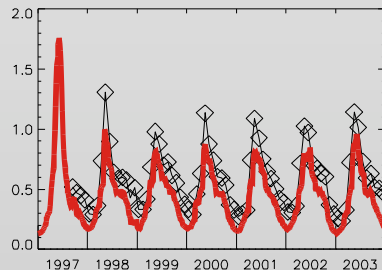
Regression statistics are for log-transformed data

Interannual Variability, SeaWiFS and Assimilation

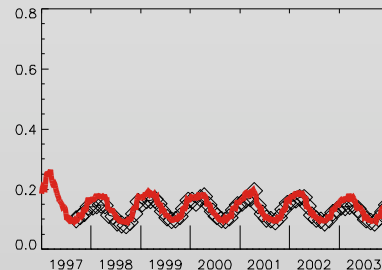
North Pacific



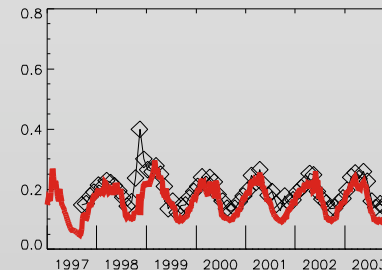
North Atlantic



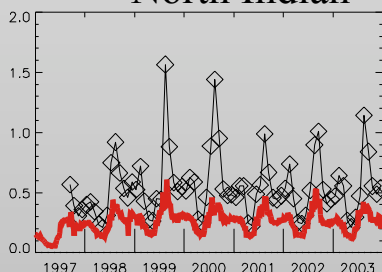
North Central Pacific



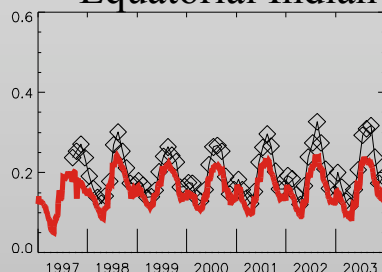
North Central Atlantic



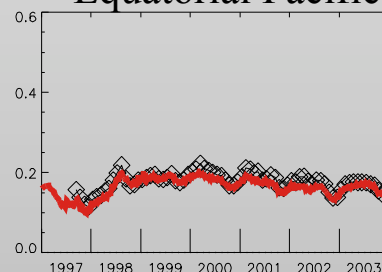
North Indian



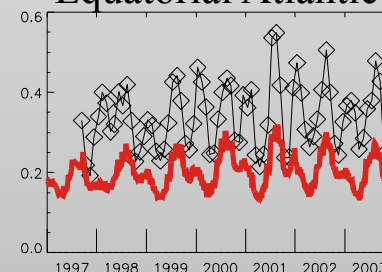
Equatorial Indian



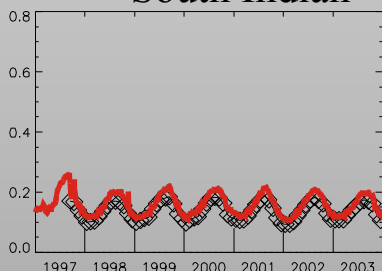
Equatorial Pacific



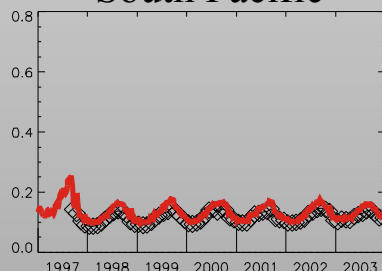
Equatorial Atlantic



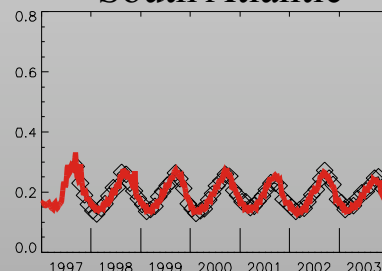
South Indian



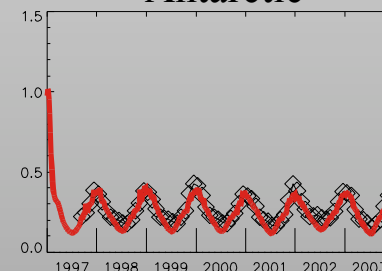
South Pacific



South Atlantic



Antarctic



Chlorophyll (mg m^{-3})

Red = Assimilation model

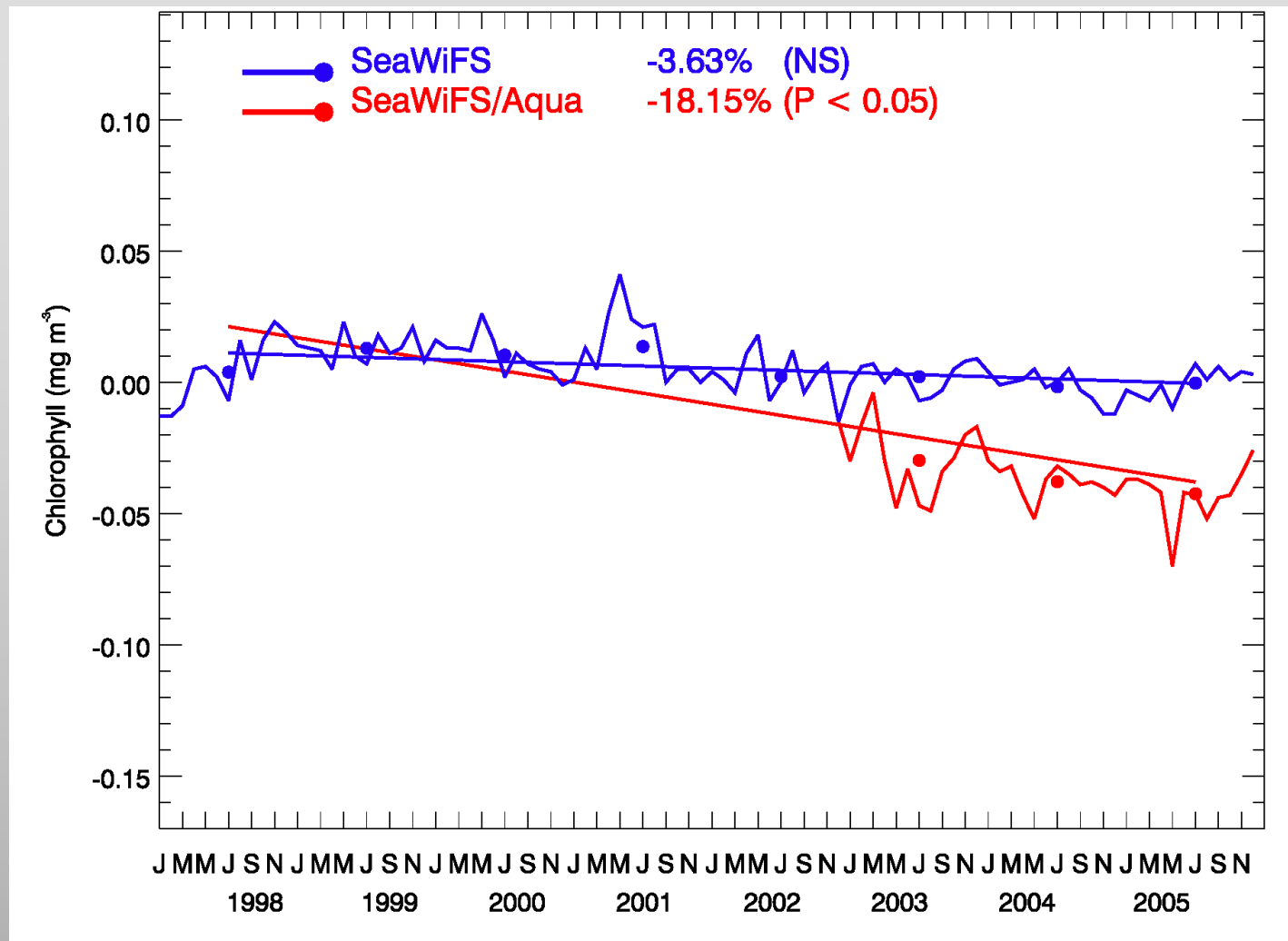
Diamonds = SeaWiFS monthly mean

Gregg, 2008, Journal of Marine Systems

Nerger and Gregg, 2007, Journal of Marine Systems

Decadal and Longer Trends

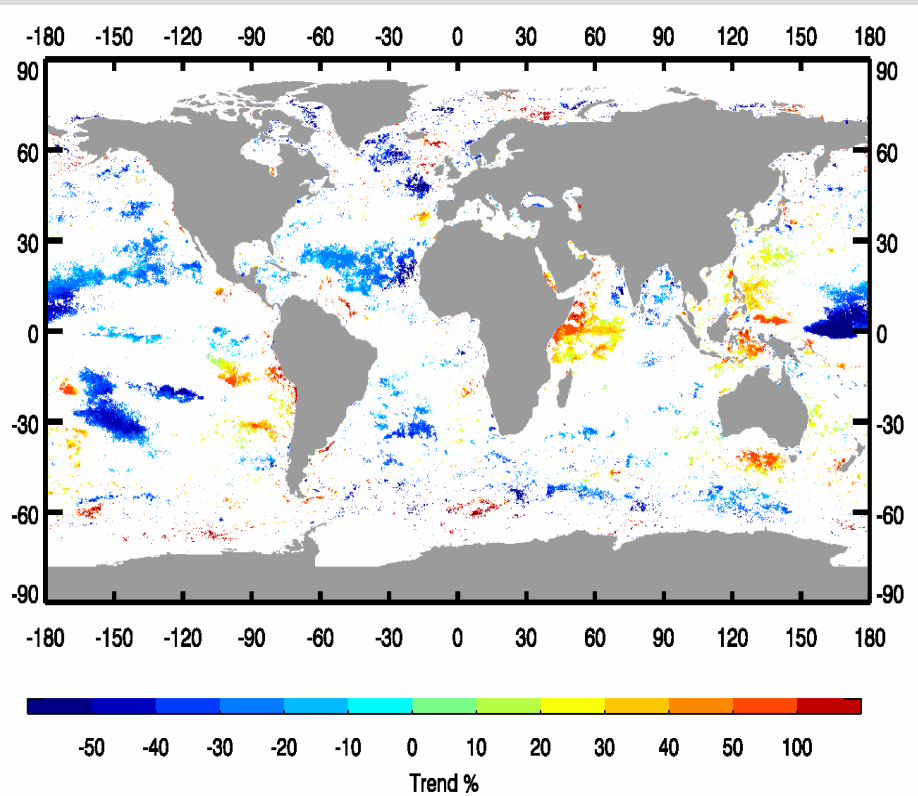
Global Annual Anomaly Trends with SeaWiFS, and SeaWiFS/Aqua



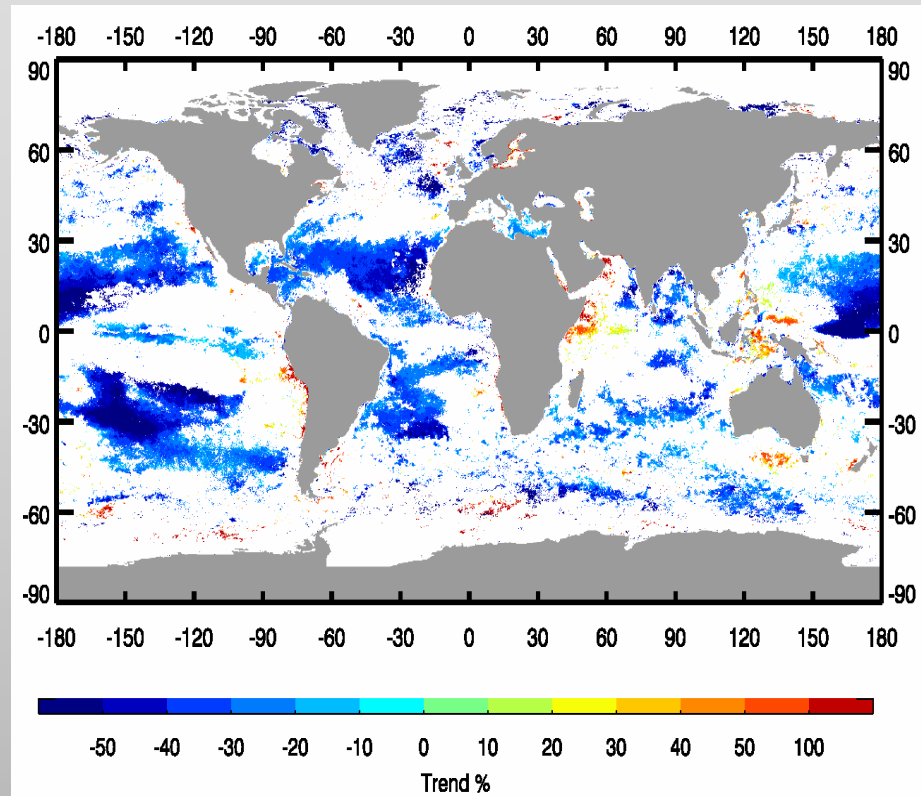
Same calibration, same algorithms, same processing

Regional Annual Trends

SeaWiFS

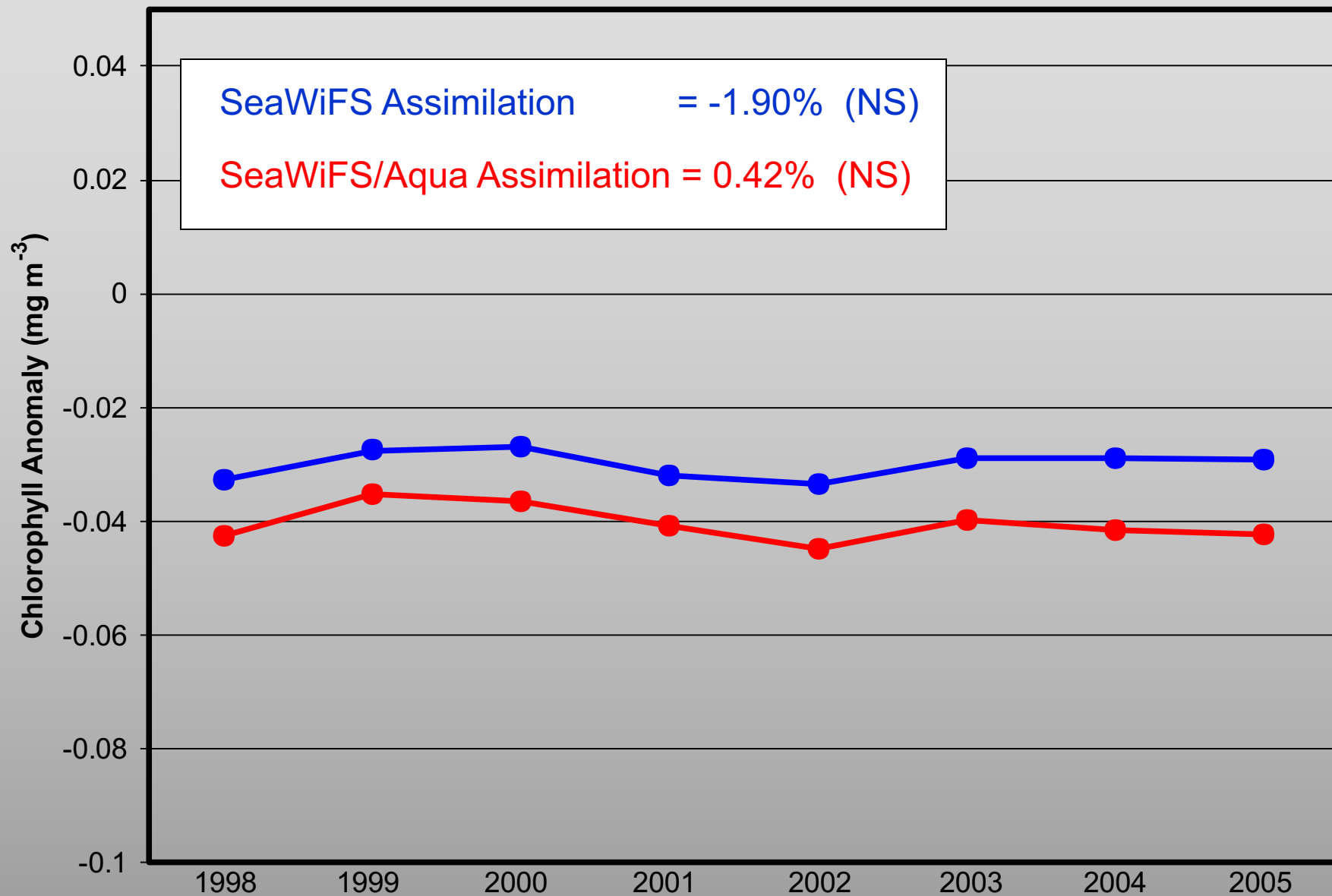


SeaWiFS/Aqua



Linear trends using 7-year average/composite images were calculated, and when significant ($P < 0.05$), shown here.

Global Annual Anomaly Trends, with Data Assimilation



Conclusions

Assimilation improves representation of seasonal, interannual, and decadal chlorophyll

Satellite data provide good representation of local and most regional seasonal variability, but global seasonal variability is poor.

Global and regional satellite data provide good representation of interannual variability.

Model and assimilation do not represent local temporal variability well, but better on regional and global scales (better than satellite for global seasonal variability)

Extending satellite time series into decadal and longer time scales is problematic due to inconsistencies between sensors/missions. Data assimilation can alleviate these problems but relies upon the model for chlorophyll abundance (not spatial and temporal variability)

Assimilation: Challenges

New work on assimilation methods needed and ongoing
multi-variate assimilation (nutrients)
dynamic state covariance matrix
multi-dimensional assimilation

Can we fill a gap in satellite data using enhanced ship observations and data assimilation?

<u>Sampling %</u>	<u>Global Difference</u>	<u>Maximum difference by basin</u>
10% sampling (about 1500 obs/day)	-2.3%	-7.6% North Pacific
1% sampling (about 150 obs/day)	1.4%	-21.9% North Indian

