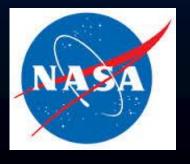
# An Accurate Absorption-Based Net Primary Production Model for the Global Ocean

Greg Silsbe
Horn Point Laboratory, UMCES

Mike Behrenfeld, Kim Halsey, Allen Milligan & Toby Westberry
Oregon State University

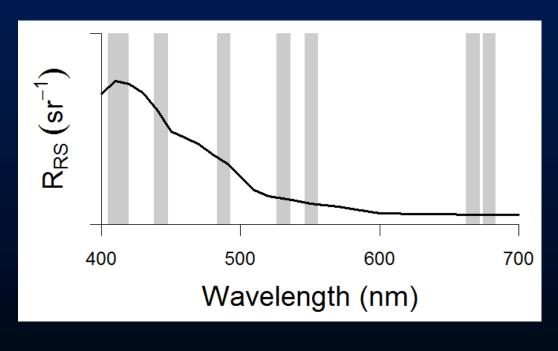


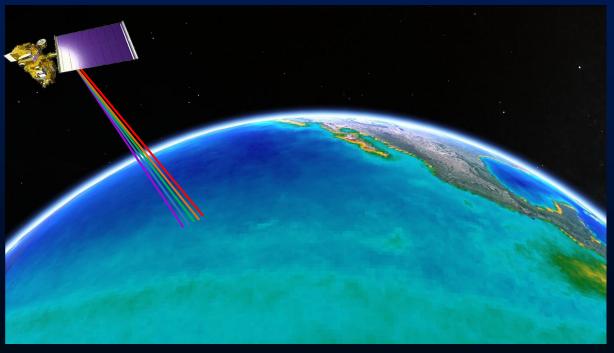




# Ocean Color Remote Sensing: Science & Challenges

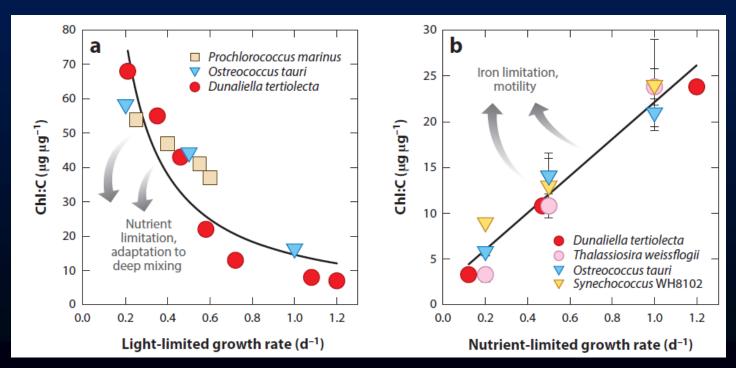
Ocean Color  $(R_{RS}(\lambda))$   $\longrightarrow$  Net Phytoplantkon Production (NPP) Growth Rates ( $\mu$ )



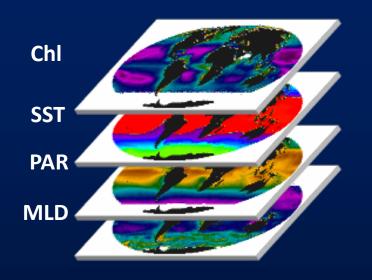


#### NPP Models

- Most published NPP models use Chl a as their central metric of phytoplankton biomass
- Disparate changes in cellular Chl:C in response to light and nutrients confound a direct relationship between Chl a and NPP



#### Traditional Products

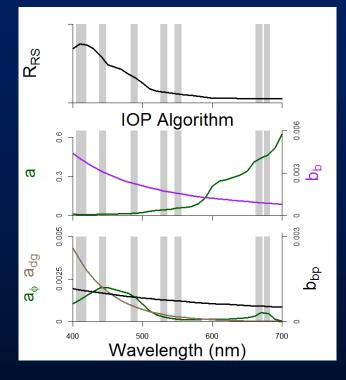


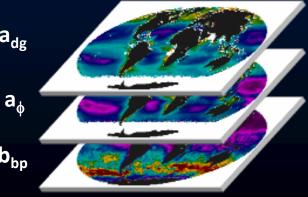
Halsey and Jones 2015. Annu. Rev. Mar. Sci. 4:260-280. Also: Laws and Bannister 1980. Marra et al. 2007.

#### NPP Models

- Spectral inversion algorithms now permit retrievals of Inherent Optical Properties (IOPs) from space (Lee et al. 2002; Maritorena et al. 2002; Werdell et al. 2013).
- The Carbon, Absorption, Fluorescence and Euphotic-Resolved (CAFE) model framework seeks to incorporate these products into a mechanistic model of NPP and μ.

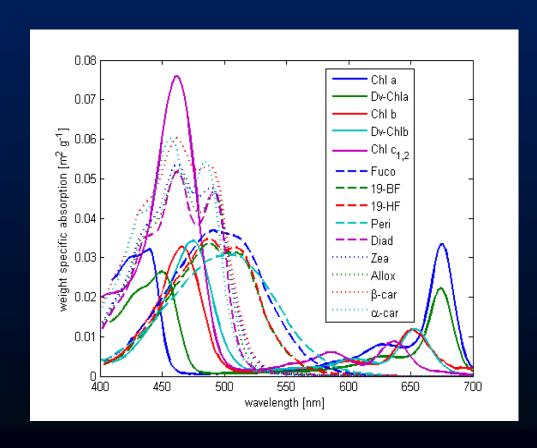
$$R_{RS}(\lambda) \sim \frac{b_b(\lambda)}{a(\lambda) + b_b(\lambda)}$$





# Phytoplankton Absorption Coefficient ( $a_{\phi}$ ): The New Chlorophyll

• The phytoplankton absorption coefficient  $(a_\phi)$  represents the sum of the product of all photosynthetic and non-photosynthetic pigments and the specific absorbance invivo







#### **Model Parameterization**

Absorption Model: 
$$NPP = E(\lambda) \times a_{\phi}(\lambda) \times \phi_{\mu}$$

Carbon Model: 
$$NPP = C_{Phyto} \times \mu$$

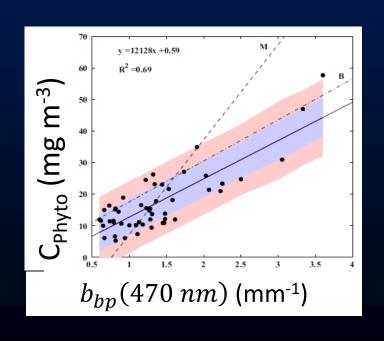
Combined Eqs: 
$$\mu = E(\lambda) \times a_{\phi}(\lambda) \times \phi_{\mu} / C_{Phyto}$$

Where:  $E(\lambda)$  is spectral extrapolation of PAR

 $C_{Phyto}$  is derived from Graff et al. (2015)

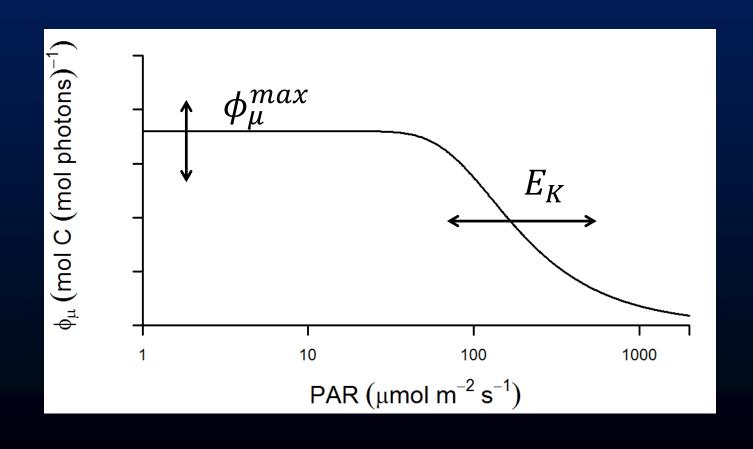
 $a_{\phi}(\lambda)$ ,  $b_{bp}(\lambda)$  are from the GIOP-DC

 $\phi_u$  is the quantum efficiency of growth



## **Model Parameterization**

$$\phi_{\mu} = \phi_{\mu}^{max} \times tanh(E_K/E)$$



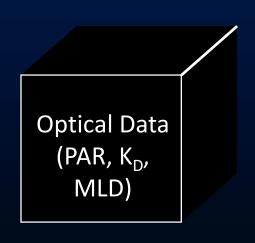
# Model Parameterization: E<sub>K</sub>

#### Other absorption-based models:

- $E_K$  is globally constant at 116 mmol m<sup>-2</sup> s<sup>-1</sup> (Marra et al. (2007)
- $E_K$  varies with sea-surface temperature (SST) (Antione and Morel 1996; Smyth et al. 2005)

#### **CAFE Model:**

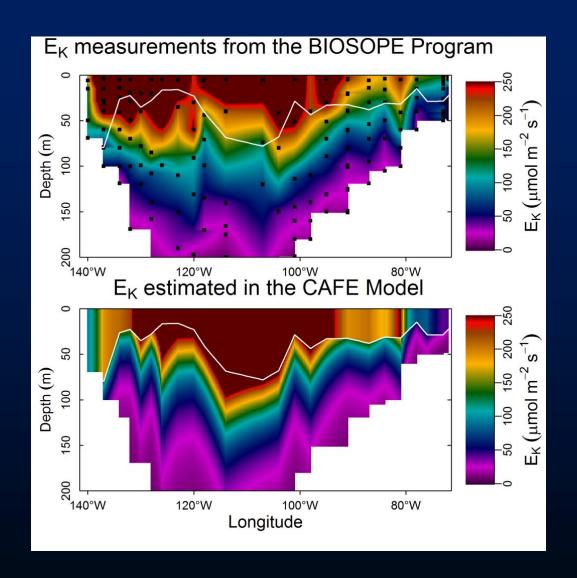
•  $E_K$  varies with Growth Irradiance (Behrenfeld et al. 2015)

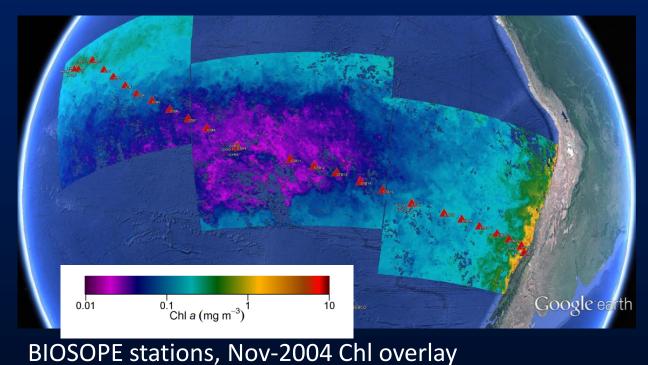




Michael J. Behrenfeld<sup>1\*</sup>, Robert T. O'Malley<sup>1</sup>, Emmanuel S. Boss<sup>2</sup>, Toby K. Westberry<sup>1</sup>, Jason R. Graff<sup>1</sup>, Kimberly H. Halsey<sup>3</sup>, Allen J. Milligan<sup>1</sup>, David A. Siegel<sup>4</sup> and Matthew B. Brown<sup>1</sup>

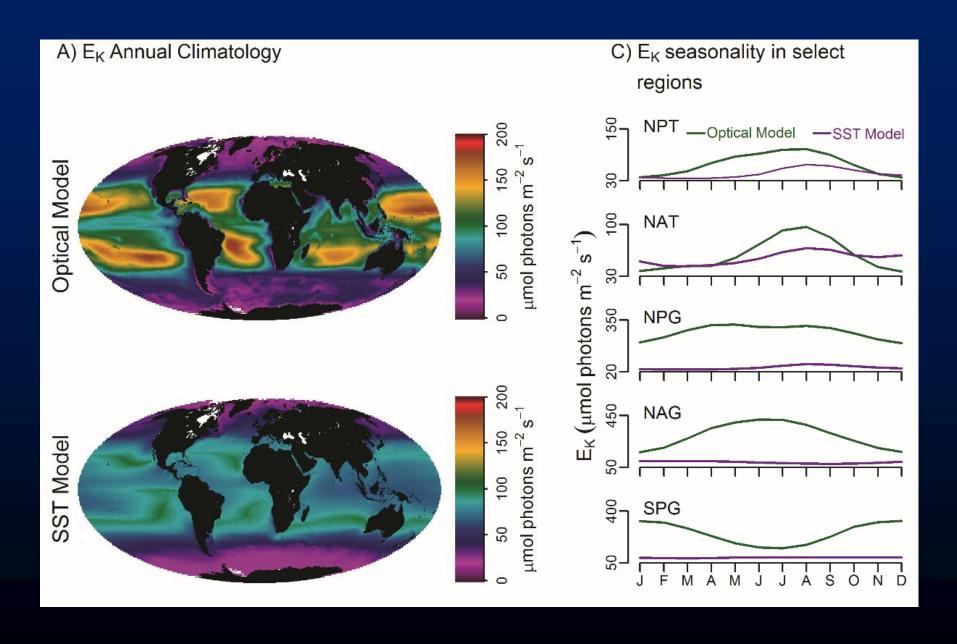
# Model Validation: E<sub>K</sub>





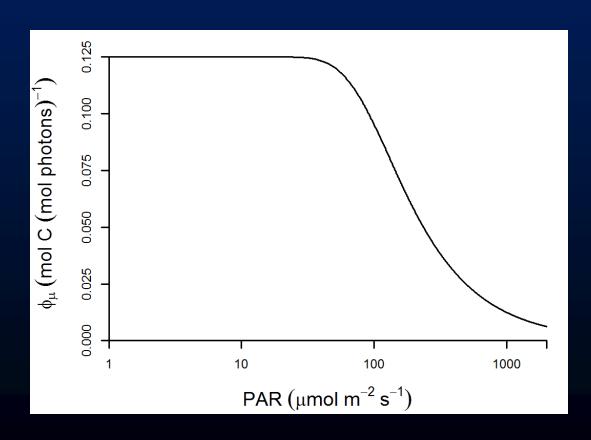
Huot et al. 2007. Relationship between photosynthetic parameters and different proxies of phytoplankton biomass in the subtropical ocean. *Biogeosciences*. **4**: 853-868.

# Model Parameterization - E<sub>K</sub>



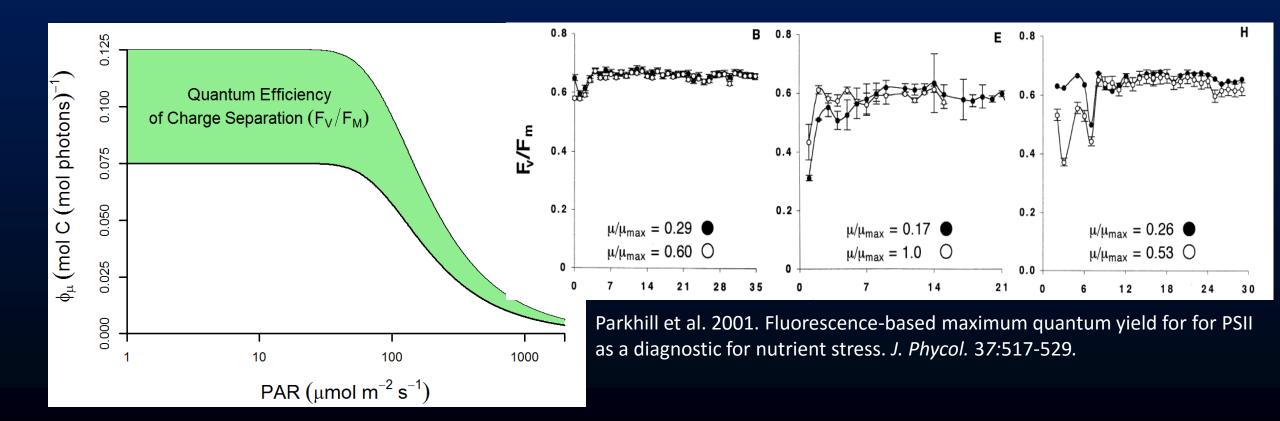
#### Other absorption-based models:

- $\phi_{\mu}^{Max}$  is globally constant: 0.060 mol C (mol photons)<sup>-1</sup> (Smyth et al. 2005; Marra et al. (2007)
- $\phi_{\mu}^{Max}$  is globally variable:  $0.058 \pm 0.038$  mol C (mol photons)<sup>-1</sup> (Antione and Morel 1996)



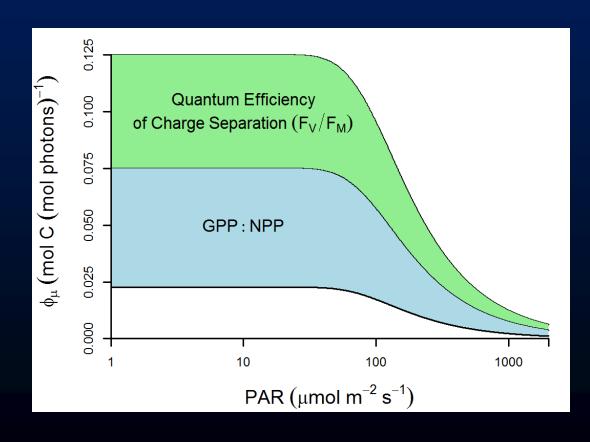
#### Other absorption-based models:

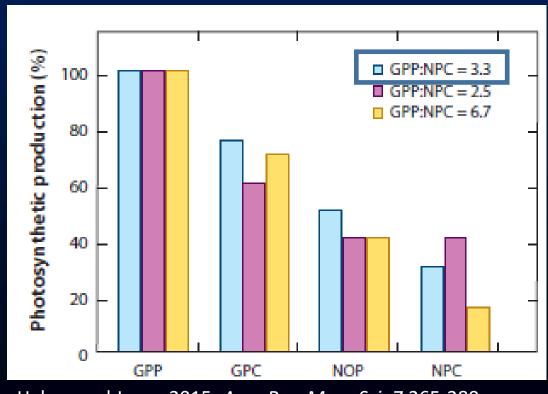
- $\phi_{\mu}^{Max}$  is globally constant: 0.060 mol C (mol photons)<sup>-1</sup> (Smyth et al. 2005; Marra et al. (2007)
- $\phi_{\mu}^{Max}$  is globally variable: 0.058 ± 0.038 mol C (mol photons)<sup>-1</sup> (Antione and Morel 1996)



#### Other absorption-based models:

- $\phi_{\mu}^{Max}$  is globally constant: 0.060 mol C (mol photons)<sup>-1</sup> (Smyth et al. 2005; Marra et al. (2007)
- $\phi_{\mu}^{Max}$  is globally variable: 0.058 ± 0.038 mol C (mol photons)<sup>-1</sup> (Antione and Morel 1996)

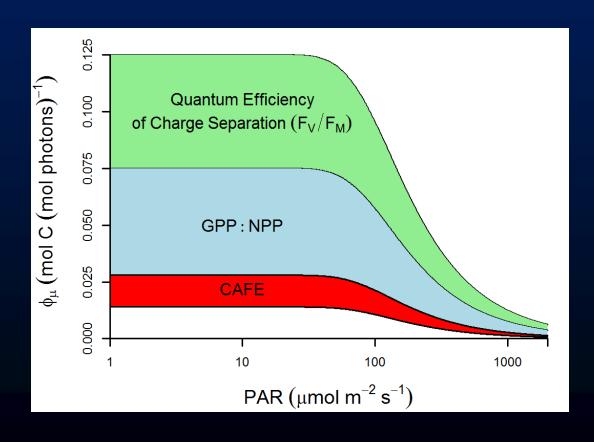




Halsey and Jones 2015. Ann. Rev. Mar. Sci. 7:265-280.

#### Other absorption-based models:

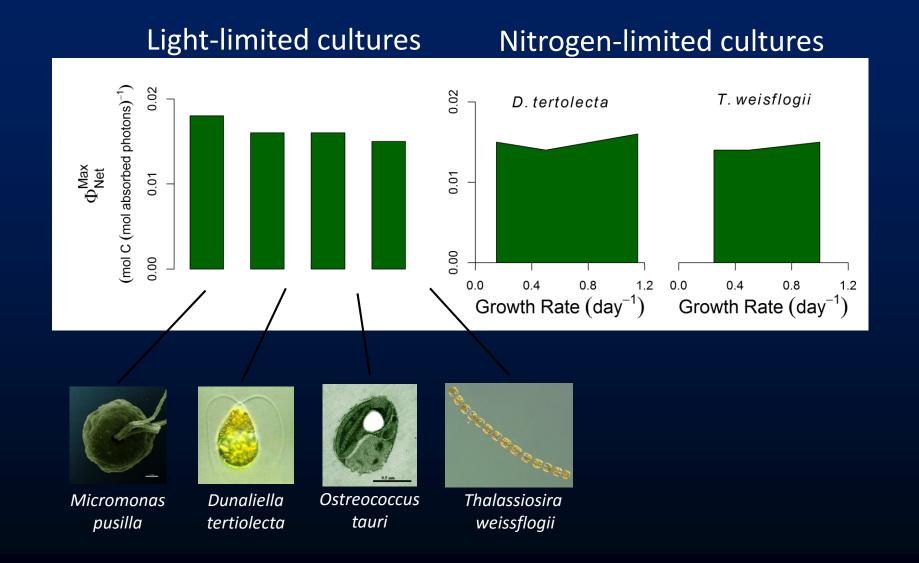
- $\phi_{\mu}^{Max}$  is globally constant: 0.060 mol C (mol photons)<sup>-1</sup> (Smyth et al. 2005; Marra et al. (2007)
- $\phi_{\mu}^{Max}$  is globally variable:  $0.058 \pm 0.038$  mol C (mol photons)<sup>-1</sup> (Antione and Morel 1996)



	$E_{\mathbf{k}}$	P B max	αВ	ā*	$\Phi_{ m cmax}$
$E_{\mathbf{k}}$	1.000				
$P_{\text{max}}^{\text{B}}$	0.508	1.000			1 1
$\alpha^{\mathbf{B}}$	-0.500	0.206	1.000		1 1
ā*	0.177	0.193		1.000	1 1
$\Phi_{ m cmax}$	-0.451	0.109	0.796	-0.364	1.000
[Chl <i>a</i> ]		0.290		-0.301	0.214
$f_{micro}$		0.258		-0.214	0.106
$f_{nano}$	-0.234		0.229		0.165
$f_{pico}$	0.116	-0.231	-0.176	0.261	-0.247
NPP	0.604	0.138	-0.468	0.283	-0.486
T	0.378		-0.369	-0.139	-0.150
[Nut]	-0.201		0.116	-0.123	0.158
$z/Z_{eu}$	-0.465	-0.320	0.254	-0.220	0.317

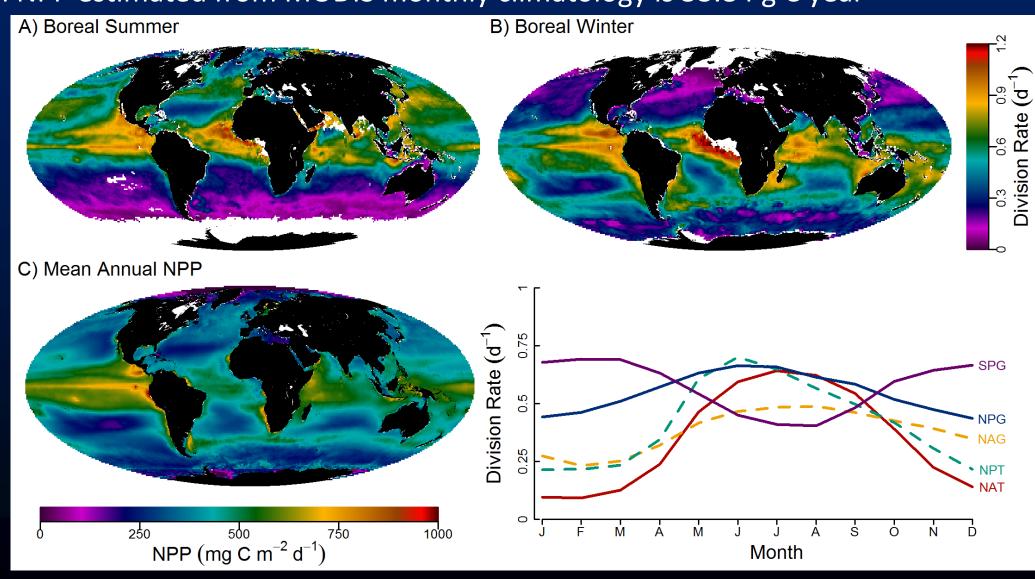
Uitz et al. 2008. Relating phytoplankton photophysiological properties to community structure. *Limnol. Oceanogr.* 53: 614-630

# Model Validation: $\phi_{\mu}^{max}$



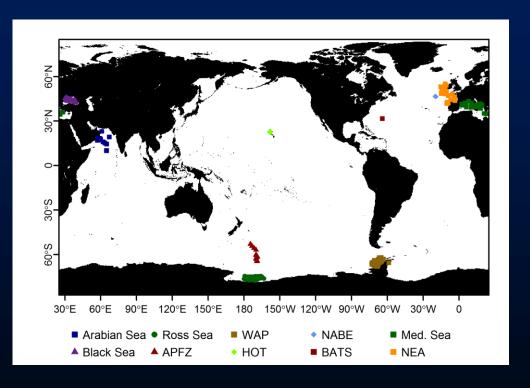
## **Model Climatology**

Global NPP estimated from MODIS monthly climatology is 53.8 Pg C year-1



## Model Validation – PPARR Approach

- CAFE NPP model results were tested against in-situ NPP measurements at 10 sites (n=1048)
- Data and methods follow PPARR4 (Saba et al. 2011)

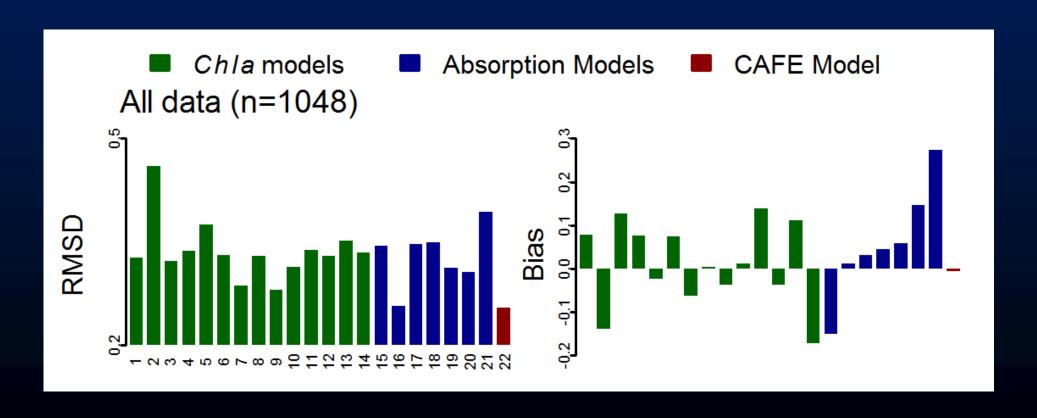


Metadata	Chl	PAR	SST	MLD	NPP
BATS	0.097	17.8	21.78	83.26	218.98
BATS	0.096	29.38	20.88	123.05	306.06
BATS	0.207	32.16	20.01	125.13	799.44

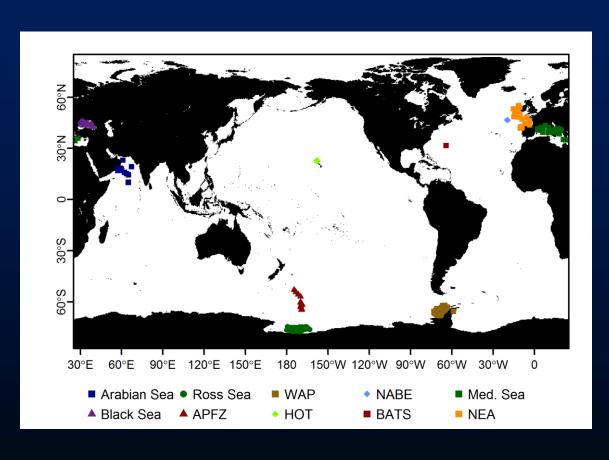
## Model Validation – PPARR Approach

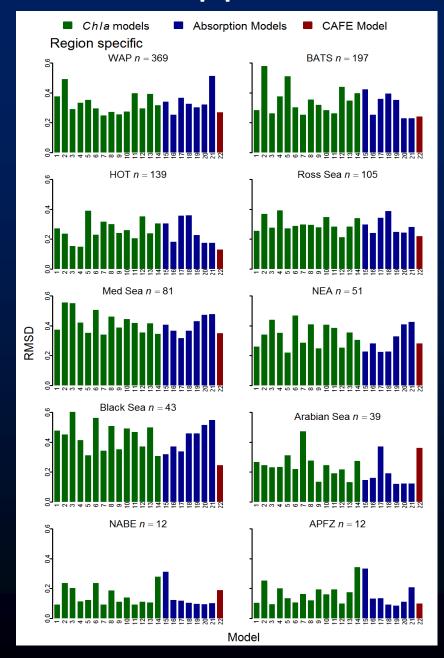
$$RMSD = \left(\frac{1}{n} \sum_{i=1}^{n} \Delta(|\log_{10} NPP_{mod} - \log_{10} NPP_{obs}|)^{2}\right)^{0.5}$$

 $Bias = mean(\log_{10}NPP_{mod}) - mean(\log_{10}NPP_{obs})$ 

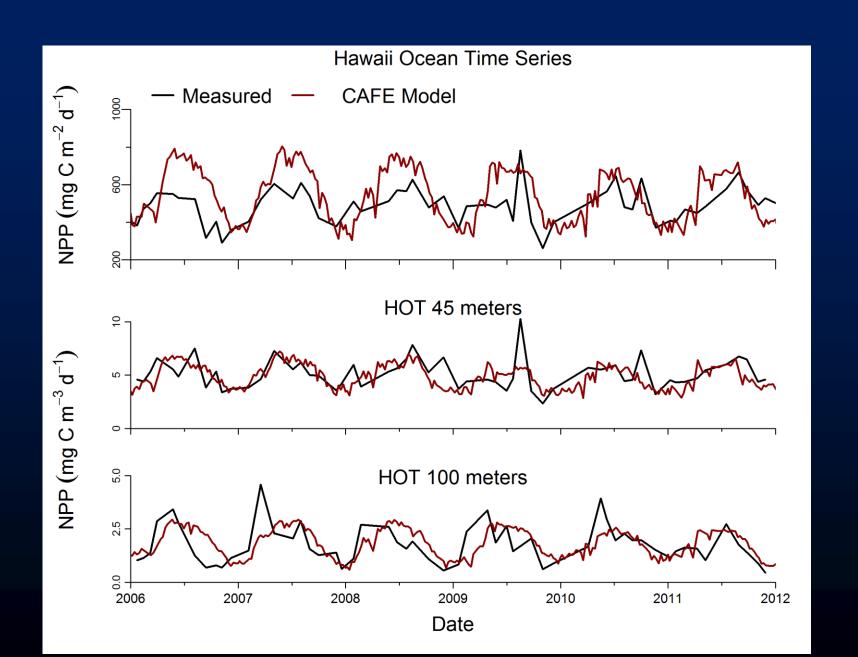


# Model Validation – PPARR Approach



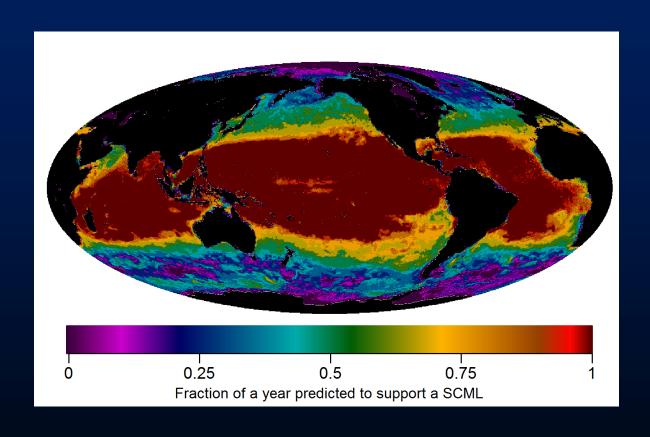


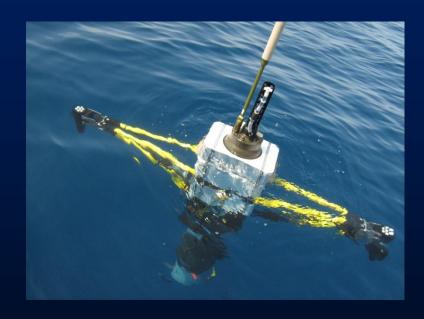
## Model Validation – Direct Satellite Measurements



## **Future Directions**

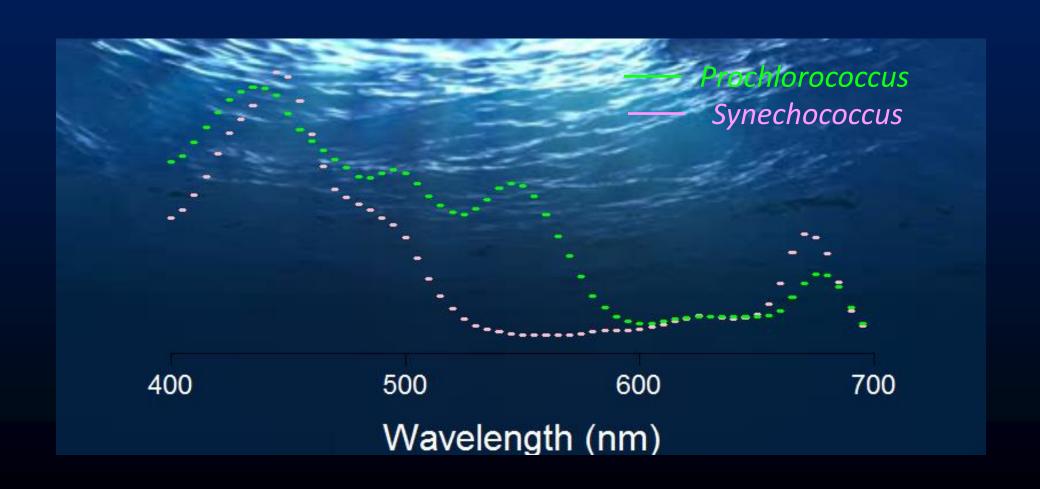
- Most phytoplankton biomass is hidden from satellite measurements of ocean color.
- BIO-Argo profiles can help fill in this missing data





## **Future Directions**

Hyperspectral ocean color data (e.g. PACE) will provide improved derivation of IOPs,
 potentially allowing for taxonomic discrimination from space



# Acknowledgements

NASA: The Science of Terra and Aqua

Questions?