

Ocean color satellite atmospheric correction

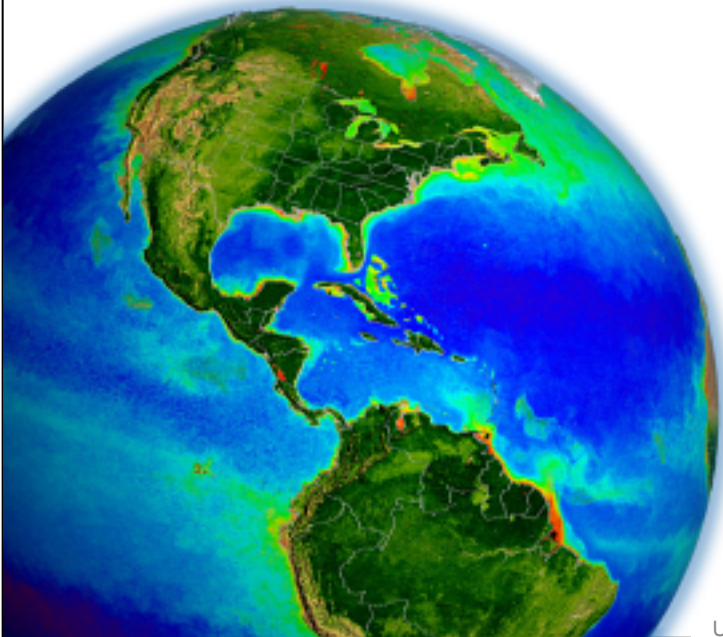
Jeremy Werdell

NASA Goddard Space Flight Center

UMaine Ocean Optics Summer Course

Jul 7 – Aug 3, 2013

Acknowledgements: Zia Ahmad, Sean Bailey,
Bryan Franz, & Wayne Robinson



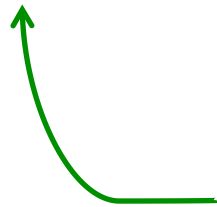
satellite ocean color

we desire measurements of marine biogeochemical stocks (e.g., []'s of phytoplankton, carbon) to further our understanding of marine ecosystems

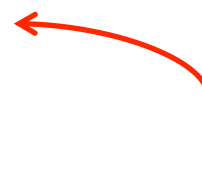
satellites provide routine, synoptic views of the marine biosphere that cannot be achieved using conventional *in situ* & aircraft platforms

ocean color satellite instruments measure light (AOPs) – not []'s

in-water constituents & their []'s → IOPs → AOPs



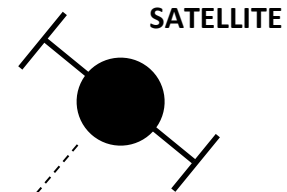
in satellite ocean color, we start here
& work from right to left



(as discussed in lectures 19 & 20 last Friday)

satellite ocean color

the satellite views the **spectral light field** at the top-of-the-atmosphere



TOP-OF-THE-ATMOSPHERE

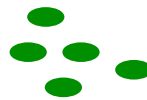
3. spatially / temporally bin and remap satellite C_a observations

1. remove atmosphere from total signal to derive estimate of light field emanating from sea surface (remote sensing reflectance, R_{rs})



SEA SURFACE

PHYTOPLANKTON



2. relate spectral R_{rs} to C_a (or geophysical product of interest)

satellite ocean color

ocean color satellites measure top-of-atmosphere radiances



$$L_t = \left(L_r + [L_a + L_{ra}] + t_{dv}L_f + t_{dv}L_w \right) t_{gv} t_{gs} f_p$$

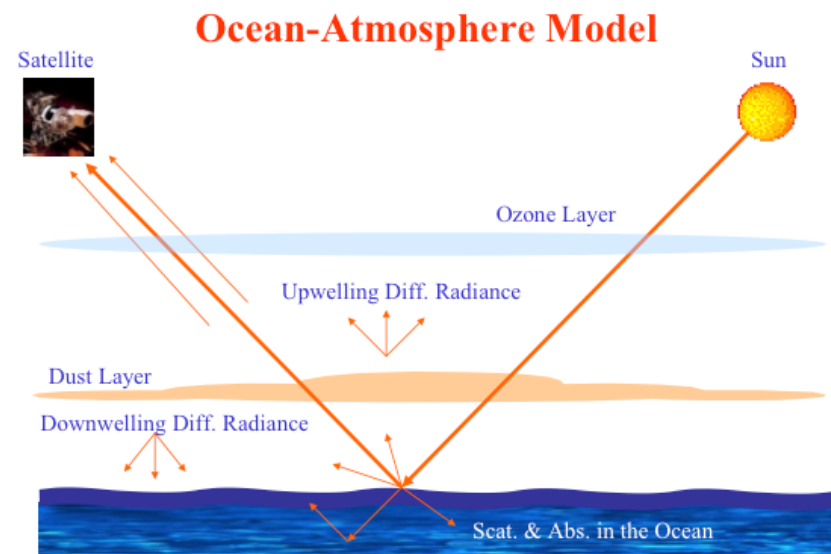
terminology:

L = radiance ($\mu\text{W cm}^{-2} \text{ nm}^{-1} \text{ sr}^{-1}$)

t = transmittance (unitless)

f = correction factor (unitless)

all terms are spectrally dependent



satellite ocean color

ocean color satellites measure top-of-atmosphere radiances



$$L_t = (L_r + [L_a + L_{ra}] + t_{dv}L_f + t_{dv}L_w) t_{gv} t_{gs} f_p$$

$$R_{rs} = \frac{L_w}{F_0 \cos(\theta_s) t_{ds} f_s f_b f_\lambda}$$

we desire (normalized) remote sensing reflectances



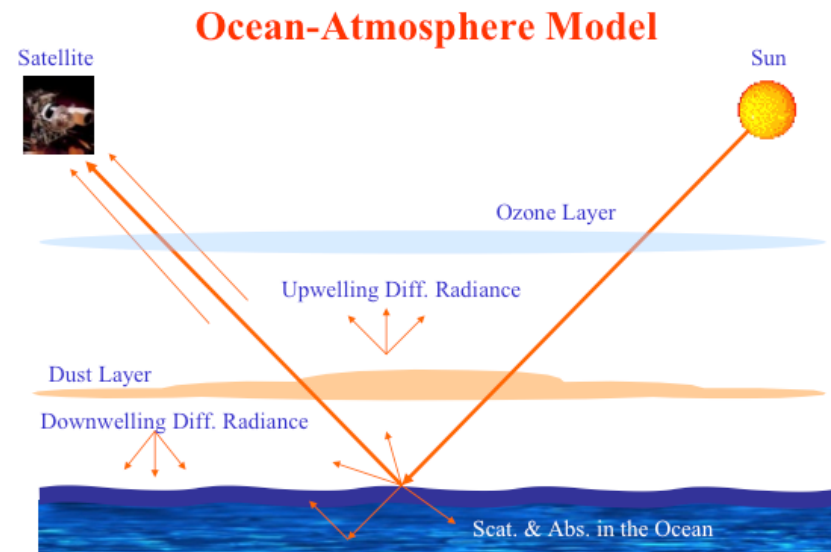
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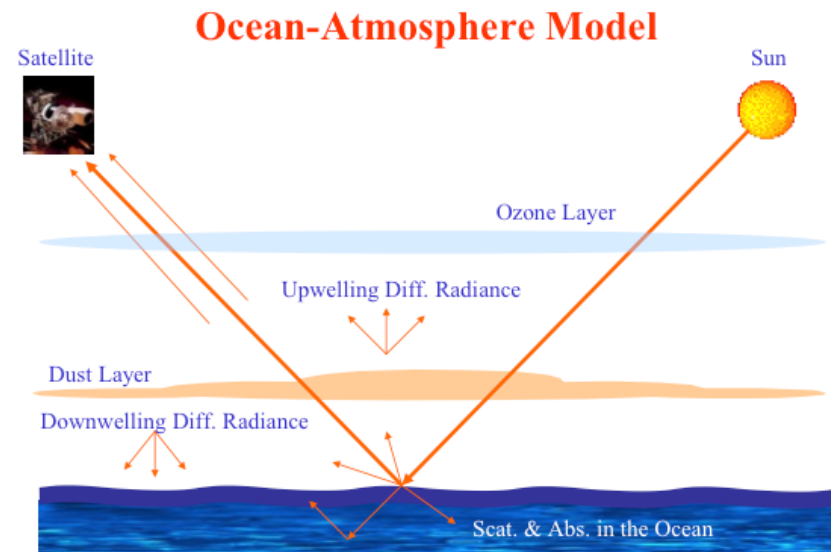
outline

atmospheric correction is the process of estimating R_{rs} from L_t

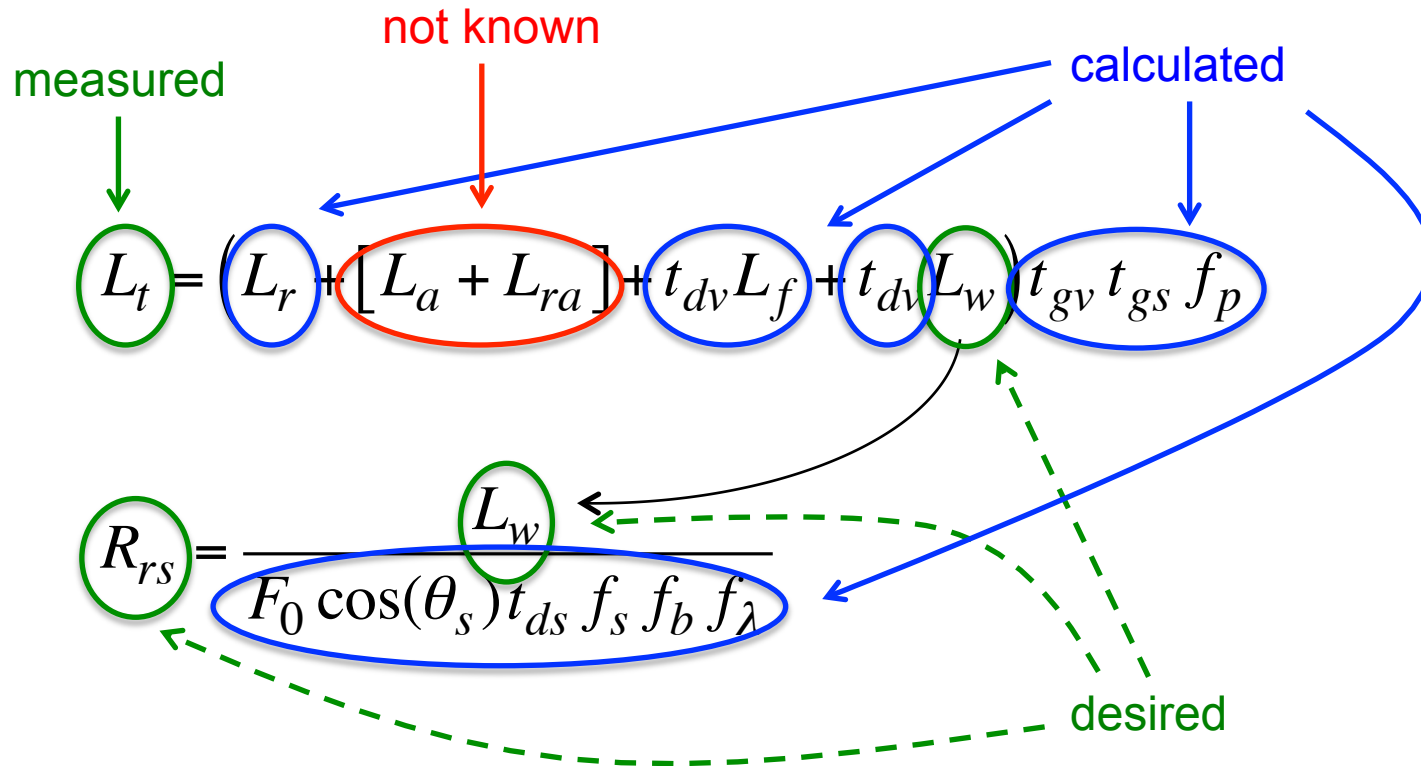
$$L_t = (L_r + [L_a + L_{ra}] + t_{dv}L_f + t_{dv}L_w) t_{gv} t_{gs} f_p$$

$$R_{rs} = \frac{L_w}{F_0 \cos(\theta_s) t_{ds} f_s f_b f_\lambda}$$

we will sequentially step through the meaning & derivation of each term in these equations




preview

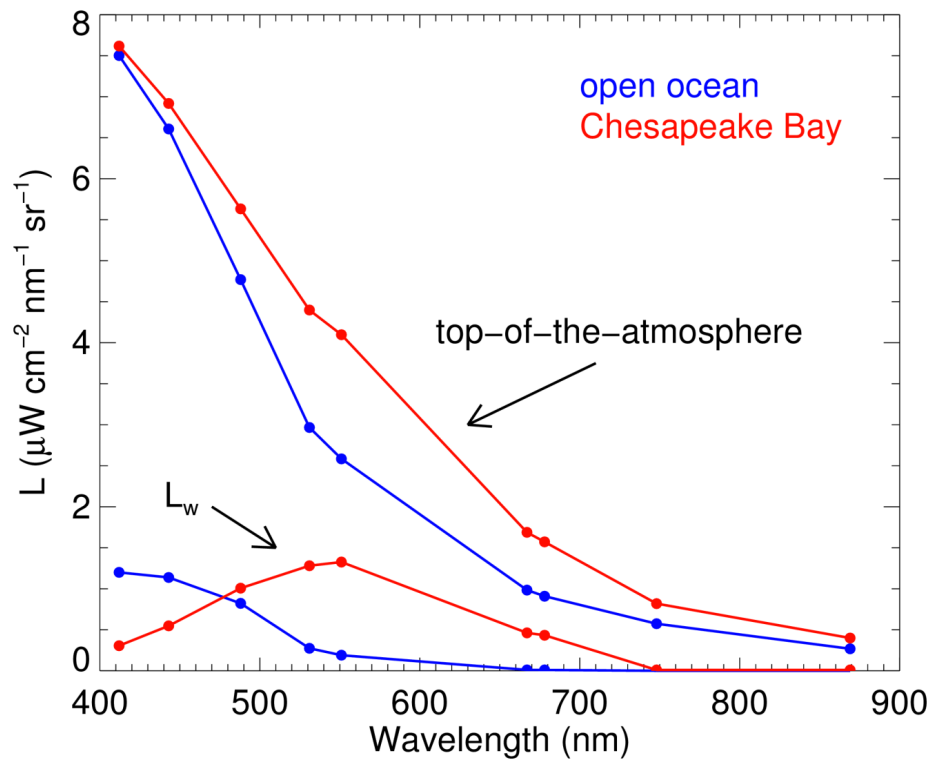


top-of-atmosphere radiance

$$L_t = (L_r + [L_a + L_{ra}] + t_{dv}L_f + t_{dv}L_w) t_{gv} t_{gs} f_p$$

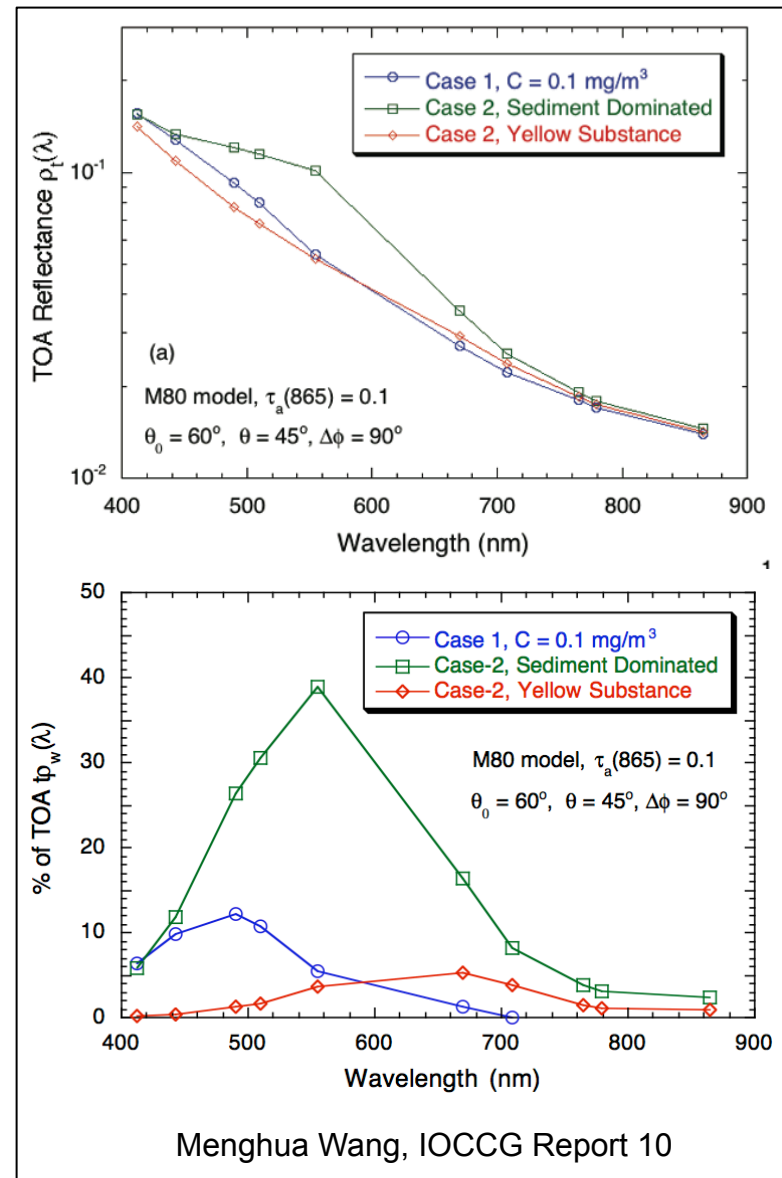
$$R_{rs} = \frac{L_w}{F_0 \cos(\theta_s) t_{ds} f_s f_b f_\lambda}$$


top-of-atmosphere radiance




L_w is often <10% of L_t !

0.5% error in atmospheric correction or calibration corresponds to possible 5% error in L_w



known terms

$$\checkmark L_t = (L_r + [L_a + L_{ra}] + t_{dv}L_f + t_{dv}L_w) t_{gv} t_{gs} f_p$$

$$R_{rs} = \frac{L_w}{F_0 \cos(\theta_s) t_{ds} f_s f_b f_\lambda}$$


known terms

instrument polarization correction factor (pre-launch measurement)

$$\checkmark \quad \dot{L}_t = \left(L_r + [L_a + L_{ra}] + t_{dv} L_f + t_{dv} L_w \right) t_{gv} t_{gs} f_p$$

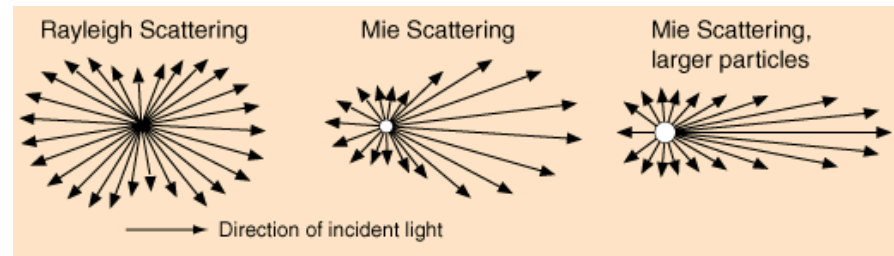
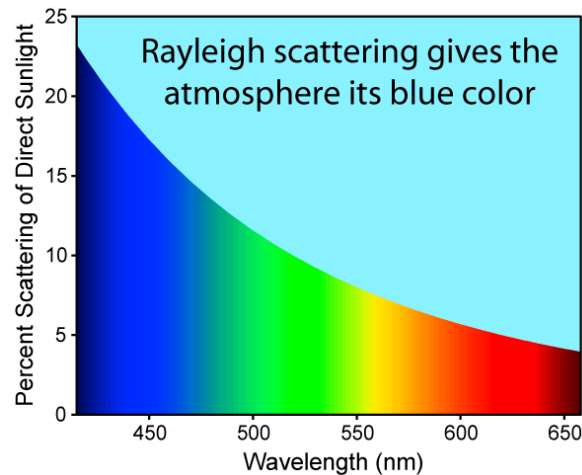
$$R_{rs} = \frac{L_w}{F_0 \cos(\theta_s) t_{ds} f_s f_b f_\lambda}$$

cosine of the instrument view angle

solar constant (irradiance) & an adjustment for the Earth-Sun distance

molecular (Rayleigh) scattering

- elastic scattering of electromagnetic radiation by **particles much smaller than the wavelength of light** (atoms or molecules)
- Rayleigh scattering of sunlight in atmosphere causes diffuse sky radiation – why the sky is blue and the Sun is yellow



<http://hyperphysics.phy-astr.gsu.edu/hbase/atmos/blusky.html>

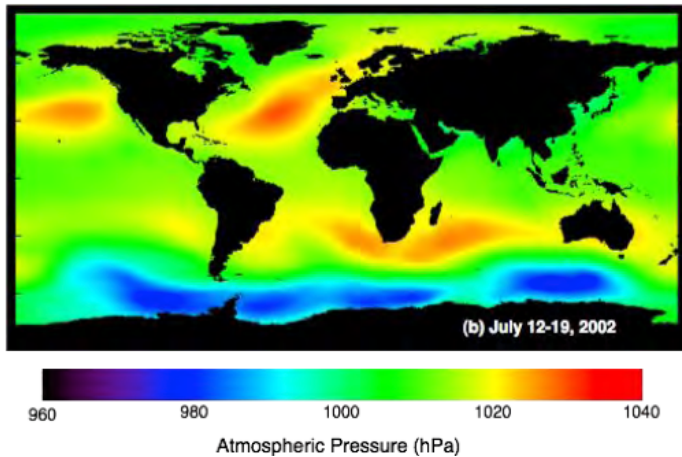
- results from electric polarizability of the particles
 - the oscillating electric field of a light wave acts on the charges within a particle, causing them to move at the same frequency
 - particle becomes a dipole whose radiation we see as scattered light
- scattering phase function is symmetrical – equal forward & backward

molecular (Rayleigh) scattering

Rayleigh optical properties are calculable (to ~0.2%) – made challenging by a rough, reflective ocean (in lieu of a flat, black ocean)

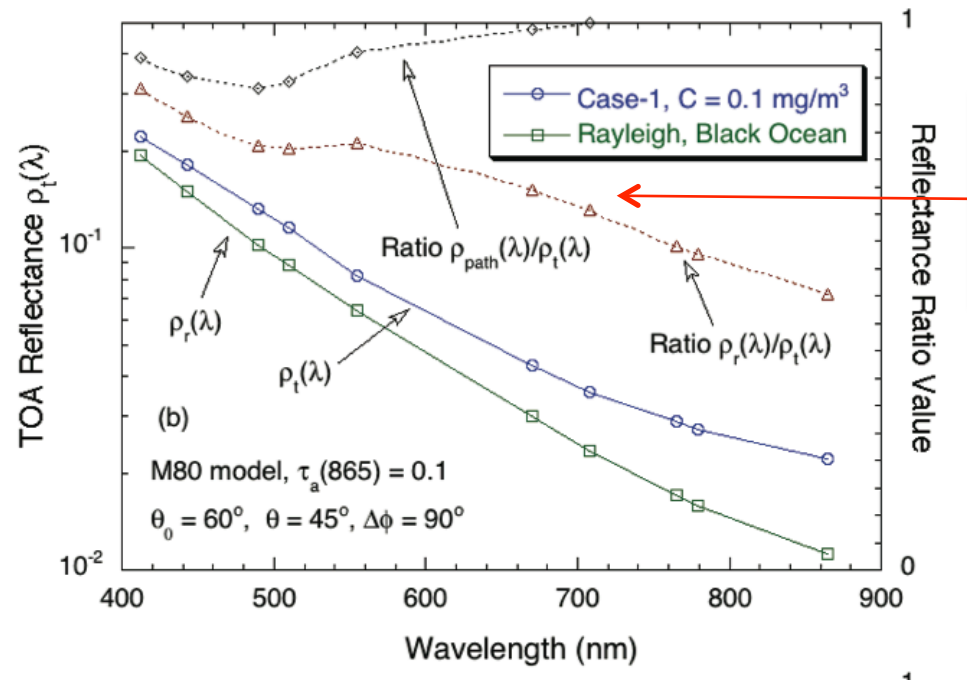
Rayleigh radiances (with polarization) are retrieved from look up tables given:

- solar & satellite viewing geometries
- wind speed (a proxy for surface roughness)
- atmospheric pressure (to adjust Rayleigh optical thickness, τ_r)



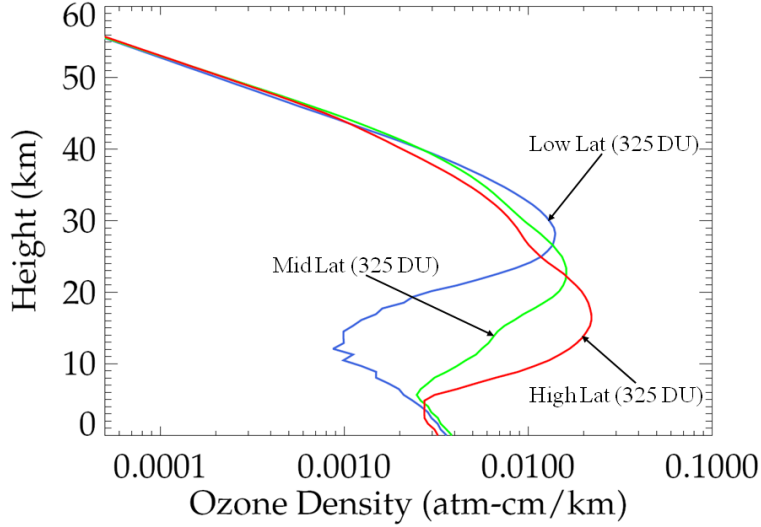
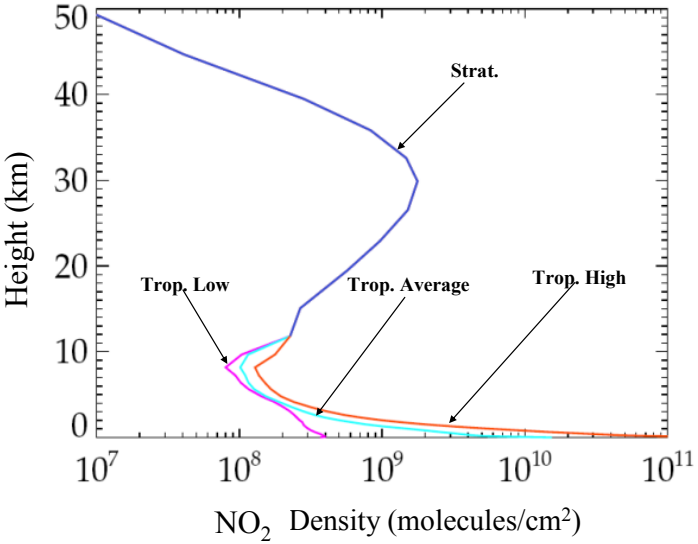
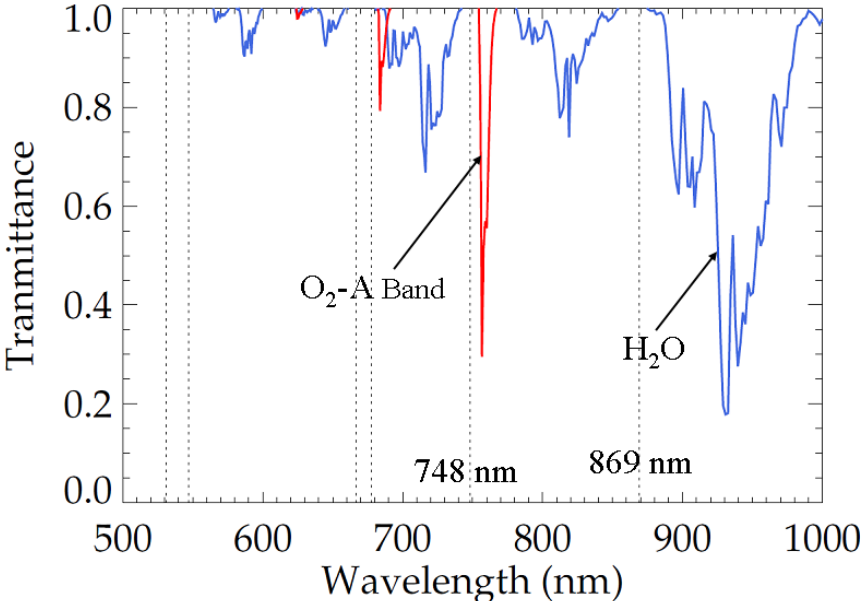
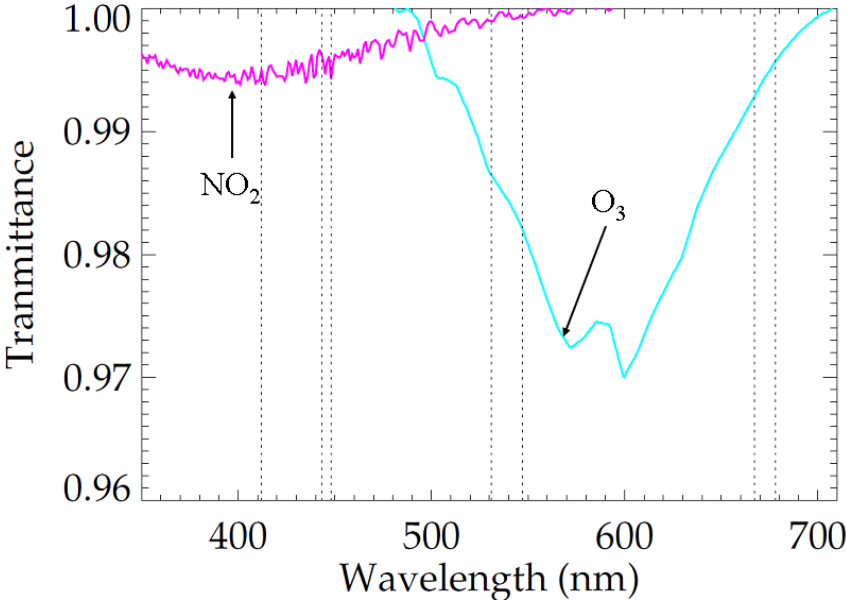
Menghua Wang, IOCCG Report 10

L_r can be 50-90% of L_t



transmittances

nitrogen dioxide, ozone, oxygen, & water vapor all attenuate sunlight

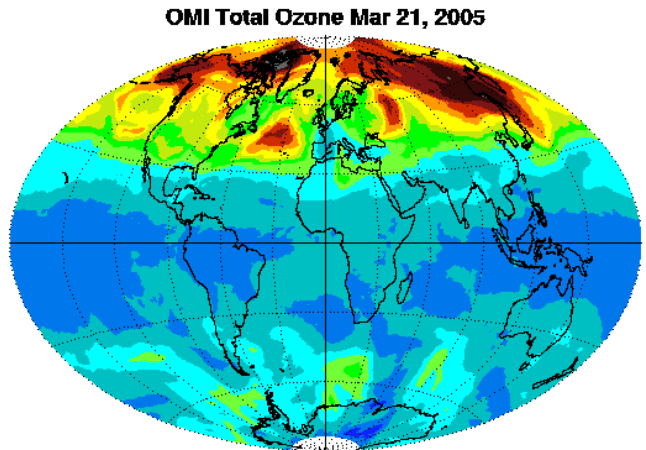
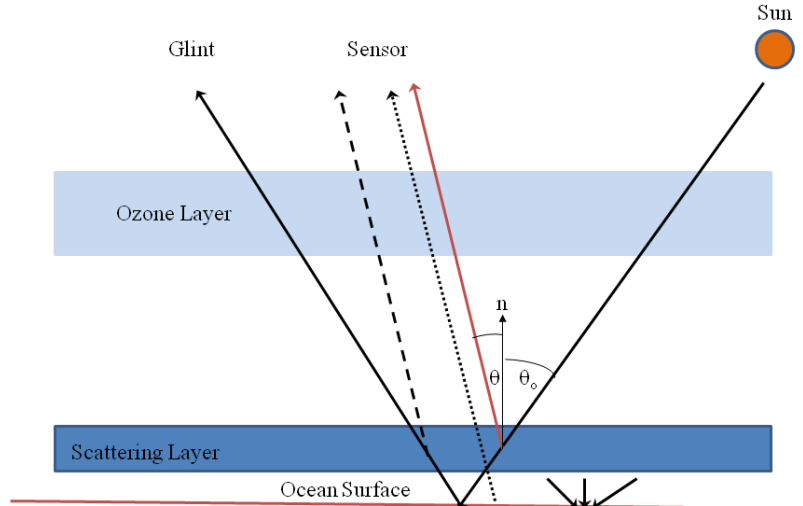


transmittances

requires ancillary data, e.g.:

- NO₂ from SCIAMACHY/GOME/OMI
- O₃ from OMI/TOMS
- water vapor from NCEP

ancillary data from varied sources for a given product often differ



example for ozone:

$$\tau_{O_3} = O_3 k_{O_3} \leftarrow \text{from LUT}$$

$$t_{O_3} = \exp \left[-\tau_{O_3} \left(\frac{1}{\cos(\theta_0)} + \frac{1}{\cos(\theta)} \right) \right]$$

transmittances

$$E_d(z) = E_d(0^-) \exp(-K_d z)$$

$$E_d(z) = E_d(0^-) \exp(-\tau)$$

$$\frac{E_d(z)}{E_d(0^-)} = \exp(-\tau)$$

$$t = \exp(-\tau)$$

I often need to mentally transfer
atmospheric terminology to
oceanic terminology

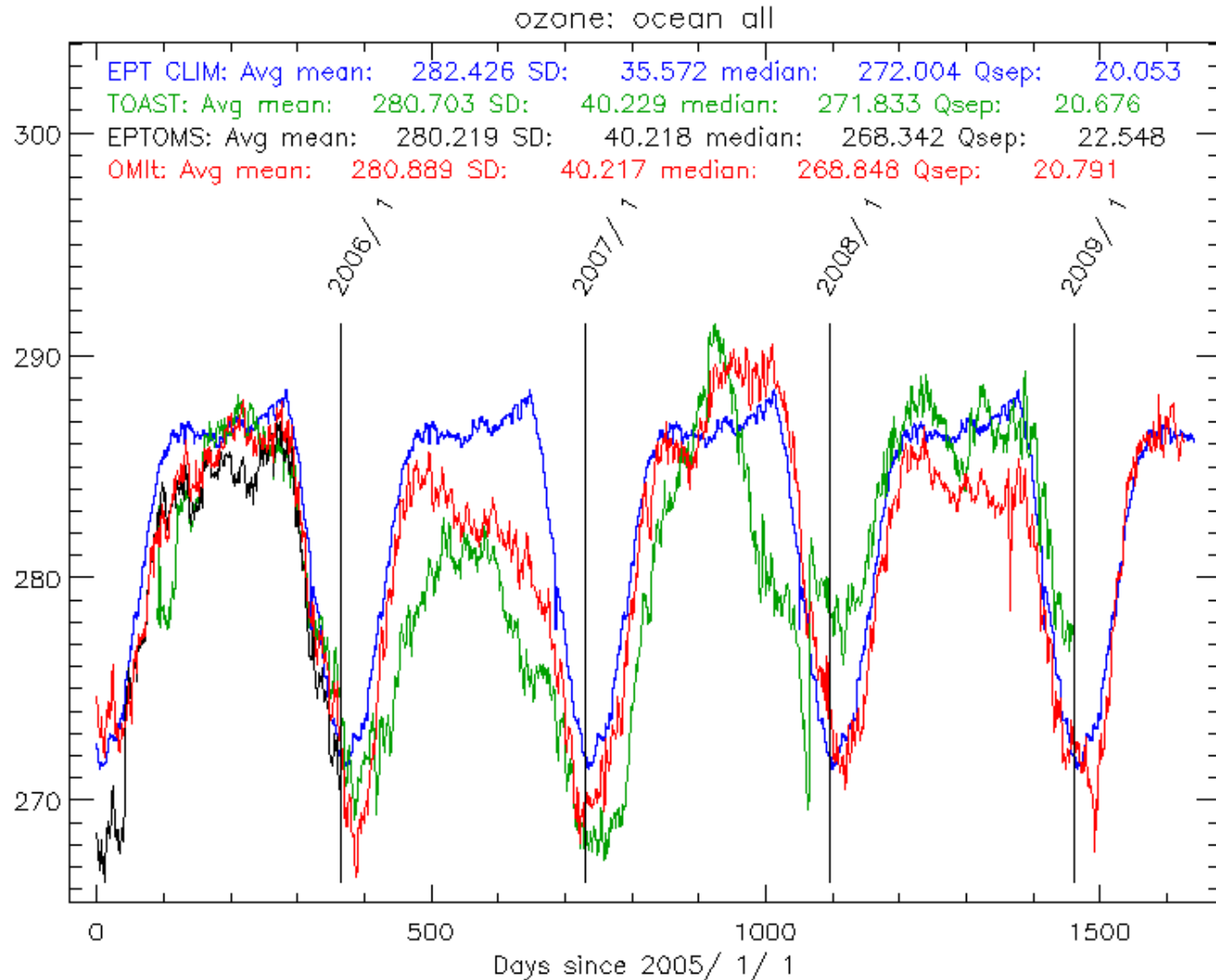


$$\tau_{O_3} = O_3 k_{O_3}$$

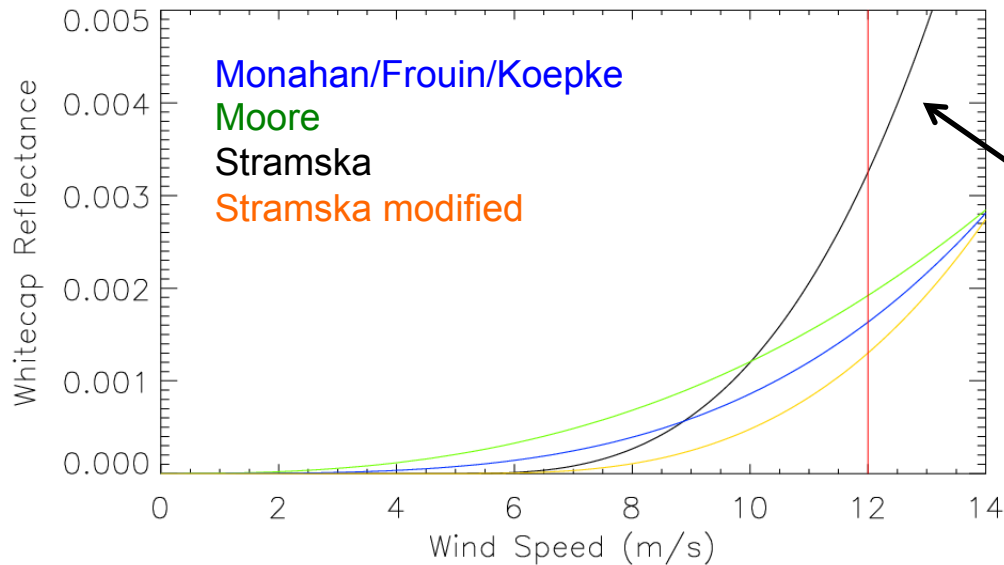
$$t_{O_3} = \exp \left[-\tau_{O_3} \left(\frac{1}{\cos(\theta_0)} + \frac{1}{\cos(\theta)} \right) \right]$$

a word about ancillary data

compare three
ancillary
sources of O₃:
TOAST
OMI
EPT climatology



foam & whitecaps



Stramska & Petelski, JGR, 2003

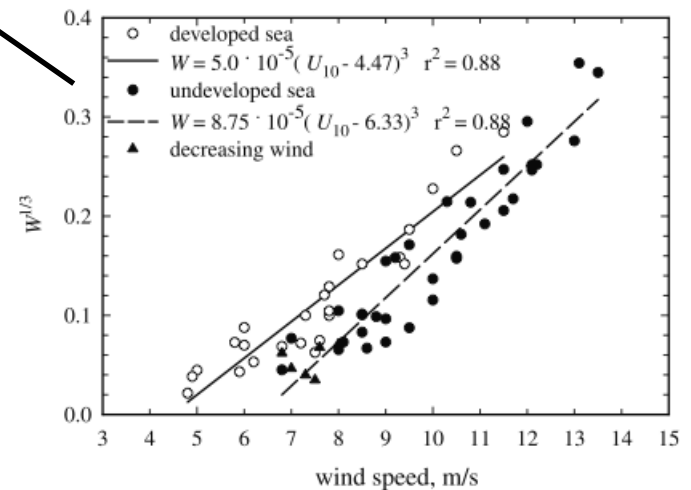
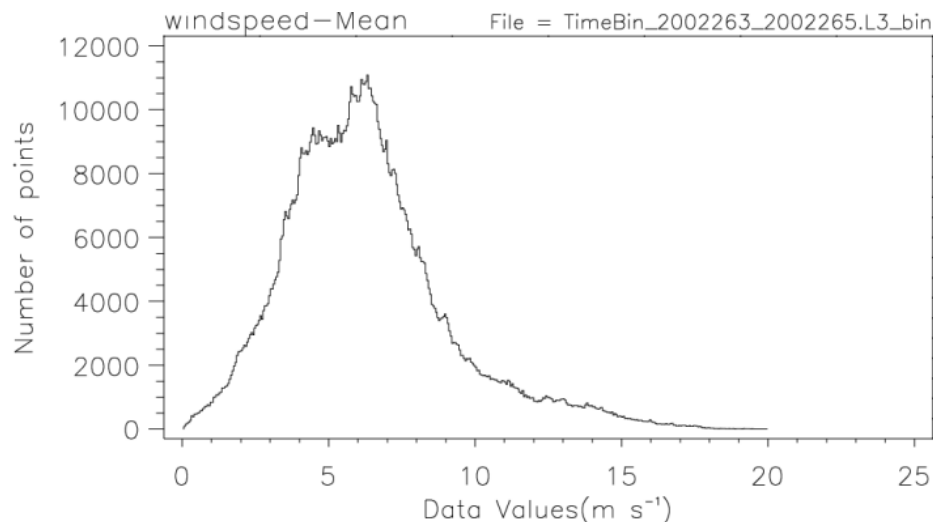


Figure 8. Oceanic whitecap coverage as a function of wind speed. Different symbols are used for the developed wave field, the undeveloped wave field, and the decreasing wind speed. See text for details.



Min :	0.0155	Mean:	6.2878
Max :	19.9611	STD :	2.7837
Median:	5.9863	Mode :	6.3260

Binsize : 0.0500000, No. of bins : 400
No. pts selected/No. pts in area : 1187244/8388608 (14.1531%)
Area : Full data

$$L_f \sim A \pi^{-1} [a (U_{10} + b)^3]$$

$A = 22\%$ (11-33%) from Koepke 1984

estimation of contribution of whitecaps & foam requires ancillary wind data (NCEP)

a quick aside about Sun glint

ideally, satellite ocean color instruments tilt away from Sun glint (e.g., SeaWiFS)

equation for top-of-atmosphere radiance can more accurately be described as:

$$L_t - F_0 T_0 T L_{GN} = \underbrace{\left(L_r + [L_a + L_{ra}] \right)}_{\text{contribution of Sun glint}} + t_{dv} L_f + t_{dv} L_w \Big) t_{gv} t_{gs} f_p$$

contribution
of Sun glint

Sun glint

Correction of sun glint contamination on the SeaWiFS ocean and atmosphere products

Menghua Wang and Sean W. Bailey

4790 APPLIED OPTICS / Vol. 40, No. 27 / 20 September 2001

$$L_g = F_0 T_0 T L_{GN}$$

L_{GN} is glint radiance normalized to no atmosphere & $F_0 = 1$

$$T_0 T = \exp \left[-(\tau_r + \tau_a) \left(\frac{1}{\cos(\theta_0)} + \frac{1}{\cos(\theta)} \right) \right]$$

two step iteration since we don't know τ_a :

$$(1) [L_t, \tau_a', W] \rightarrow L_t^{(1)} = L_t - L_g \rightarrow \tau_a^{(1)}$$

$$(2) [L_t^{(1)}, \tau_a^{(1)}, W] \rightarrow L_t^{(2)} = L_t^{(1)} - L_g \rightarrow \tau_a^{(2)}$$

with initial guess of $\tau_a' \sim 0.1$ (additional logic included to prevent overcorrection)

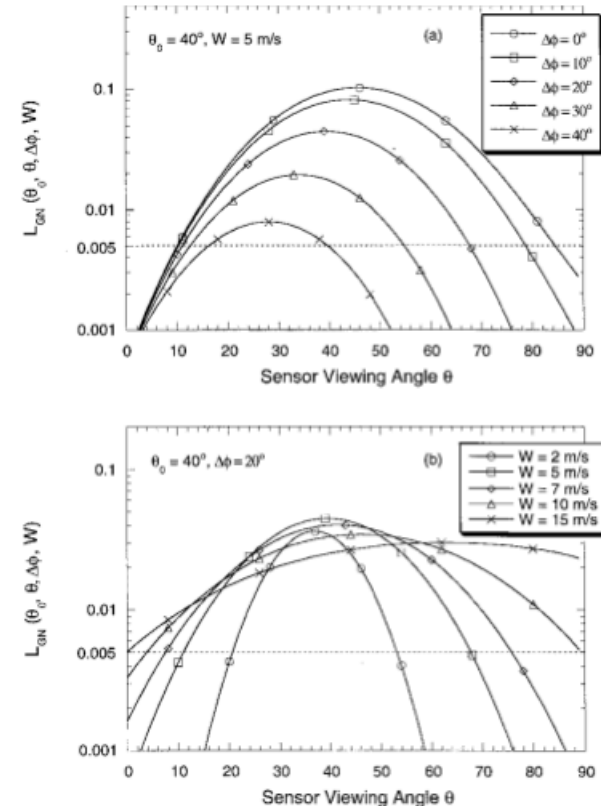


Fig. 1. Normalized sun glint radiance L_{GN} as a function of the sensor-viewing angle (solar zenith angle, 40°) and for (a) various relative azimuthal angles with surface wind speed of 5 m/s and (b) various surface wind speeds with a relative azimuthal angle of 20° .

L_{GN} from Cox and Munk (1954) requires ancillary wind speed & geometries of Sun & sensor

aerosols

$$\checkmark L_t = \left(\checkmark L_r + \checkmark [L_a + L_{ra}] + \checkmark t_{dv} \checkmark L_f + \checkmark t_{dv} \checkmark L_w \right) \checkmark t_{gv} \checkmark t_{gs} \checkmark f_p$$

$$R_{rs} = \frac{L_w}{F_0 \cos(\theta_s) t_{ds} f_s f_b f_\lambda}$$

✓ ✓ ✓ ✓

aerosols

final unknowns in top expression

$$\checkmark \quad \checkmark \quad \checkmark \quad \checkmark \quad \checkmark \quad \checkmark \quad \checkmark \quad \checkmark \quad \checkmark$$

$$\checkmark L_t = \left(\checkmark L_r + \checkmark [L_a + L_{ra}] + \checkmark t_{dv} \checkmark L_f + \checkmark t_{dv} \checkmark L_w \right) \checkmark t_{gv} \checkmark t_{gs} \checkmark f_p$$

$$R_{rs} = \frac{L_w}{F_0 \cos(\theta_s) t_{ds} f_s f_b f_\lambda}$$

$\checkmark \quad \checkmark \quad \checkmark \quad \checkmark$

additional concepts:

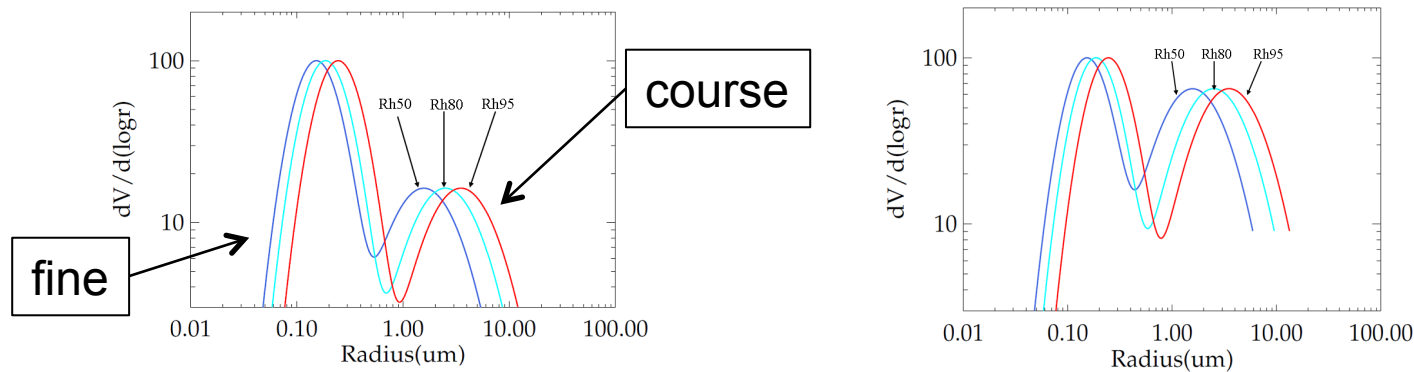
- aerosol tables
- single- vs. multi-scattering
- aerosol selection
- the “black pixel” assumption
- absorbing aerosols

aerosol tables

- aerosol properties can be characterized by their particle size distribution (PSD) & their complex index of refraction (m)
- given a PSD & m (& assuming sphericity), aerosol optical properties can be computed using Mie theory:
 - scattering phase function ($\tilde{\beta}$)
 - single scattering albedo ($\omega = b / c$)
 - extinction coefficient ($c = a + b$)
- aerosol optical thickness relates to extinction coefficient
 - $$\tau_a = \int_0^z c(z) dz$$
- aerosol tables are generated for various PSDs (& m 's) & are
 - defined by $\tilde{\beta}$, ω , τ_a (& other variables)
 - navigated using solar & satellite viewing geometries

aerosol tables

- we assume each PSD to be represented by 2 lognormal distributions
 - fine particles (continental & sometimes absorbing)
 - coarse particles (oceanic / sea salt & non-absorbing)



- each PSD modulated by varying relative humidity
 - humidity changes particle size
 - requires ancillary data from NCEP
- 80 aerosol tables total, built from AERONET measurements
 - 10 PSDs
 - 8 relative humidities

see Ahmad et al.,
Applied Optics, 2010

aerosol tables

- the Angstrom exponent (α) provides an estimator of particle size

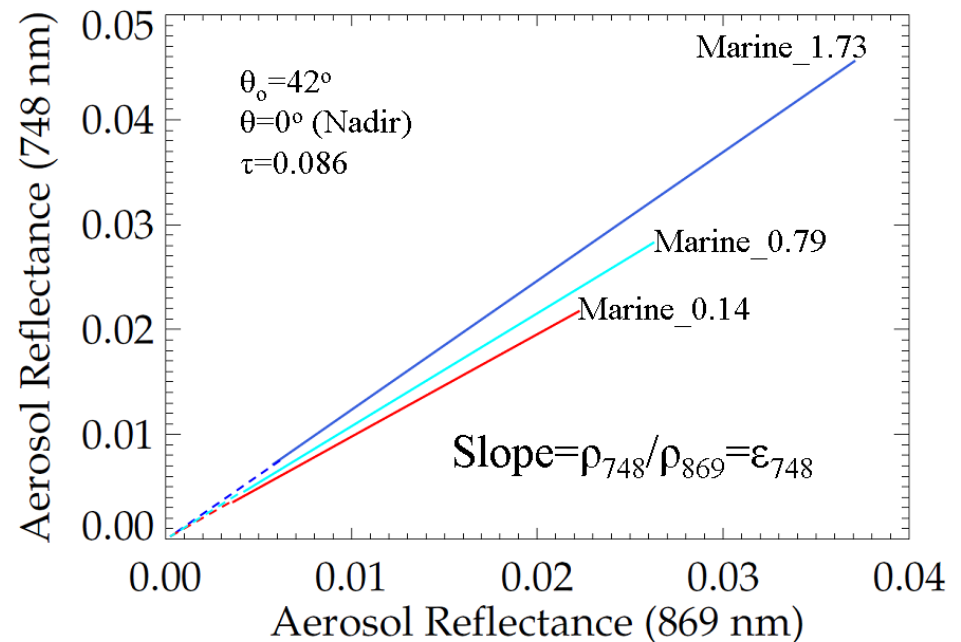
- high α = small particles

- low α = large particles

- defined via
$$\frac{\tau_a(\lambda)}{\tau_a(\lambda_0)} = \left(\frac{\lambda_0}{\lambda} \right)^\alpha$$

- aerosol models often defined by epsilon (ϵ)

- $$\epsilon(748, 869) = \frac{L_a(748)}{L_a(869)}$$



black pixel assumption

final unknowns in top expression

$$\checkmark L_t = \left(\checkmark L_r + \checkmark [L_a + L_{ra}] + \checkmark t_{dv} L_f + \checkmark t_{dv} L_w \right) \checkmark t_{gv} \checkmark t_{gs} \checkmark f_p$$

in the open ocean, we can assume (???) that L_w in the near-infrared (NIR) is = 0 (rather, is *black*)

thus, in the NIR (e.g., 748 and 869 nm):

$L_a(\text{NIR}) + L_{ra}(\text{NIR}) = L_t(\text{NIR}) -$ the terms we computed

$$L_a + L_{ra} = L_t t_{gv} t_{gs} f_p - L_r - t_{dv} L_f$$

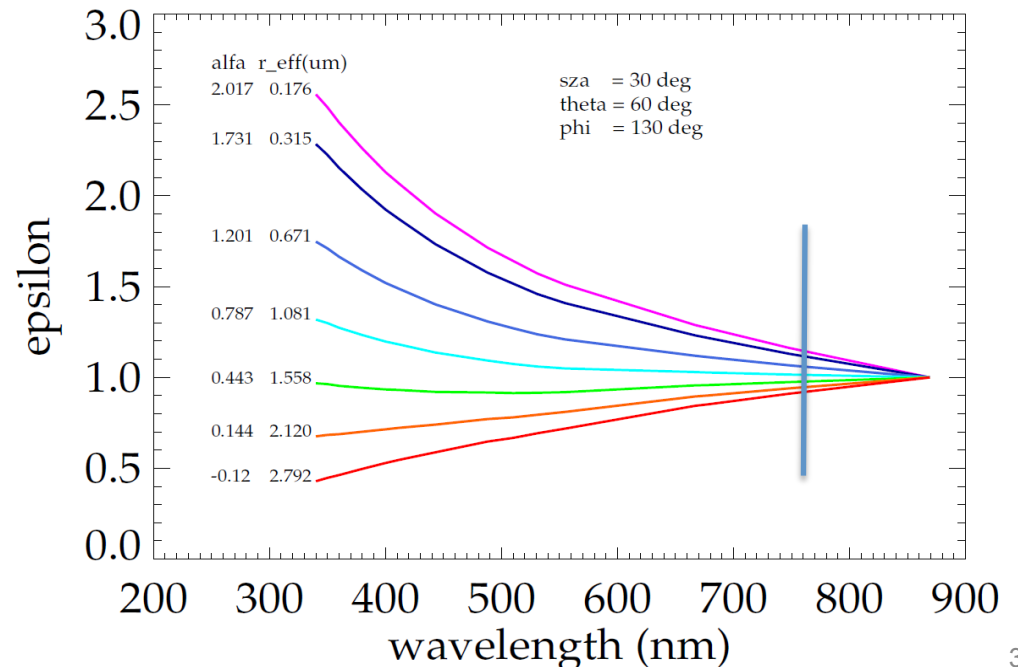
aerosol selection

$L_w(\text{NIR}) = 0$, so $L_a(\text{NIR}) + L_{ra}(\text{NIR}) = L_t(\text{NIR}) -$ (everything previously computed)
how do we estimate $L_a(\text{visible}) + L_{ra}(\text{visible})$?

- let's refer to $[L_a + L_{ra}]$ simply as L_a & ignore single- vs. multi-scattering issues
- select the 10 aerosol tables that match the observed NCEP relative humidity
- compute epsilon values for the 10 tables $[\varepsilon(748,869) = L_a(748) / L_a(869)]$
- perform an iterative determination of the mean $\varepsilon(748,869)$ value (can be describe offline) & select a final bounding 2 aerosol models
- using 2 bounding models, calculate $\varepsilon(\lambda,869)$ from $\varepsilon(748,869)$
- calculate $L_a(\lambda) = \varepsilon(\lambda,869) L_a(869)$

see Gordon & Wang,
Applied Optics, 1994

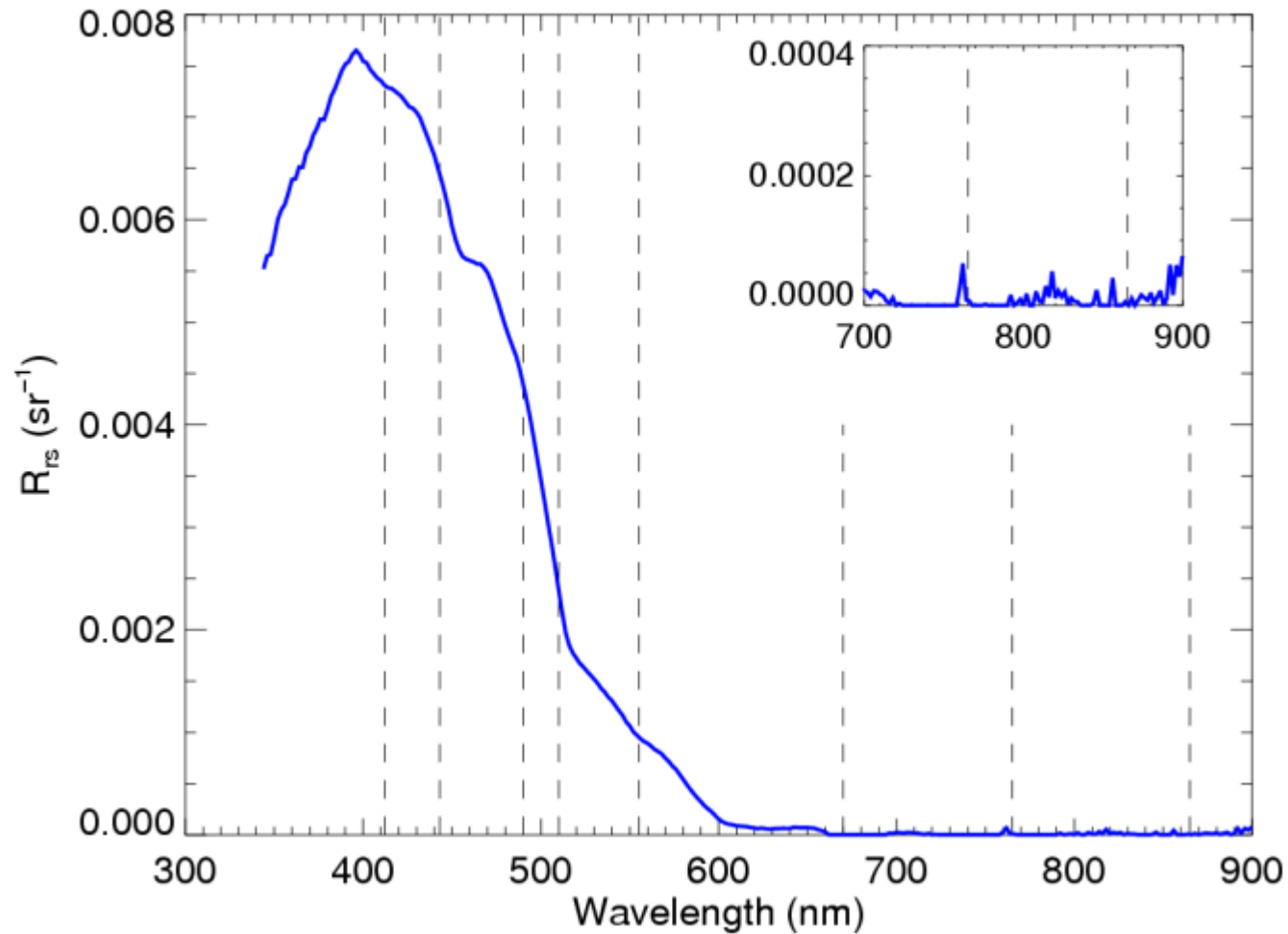
final retrieval of $L_a(\lambda)$ is more accurate than that of τ_a and α ;
not unlike retrievals of $a(\lambda)$ being more accurate than $a_{dg}(\lambda)$ & $a_{ph}(\lambda)$ in inversion models (lectures 19 & 20)



black pixel assumption

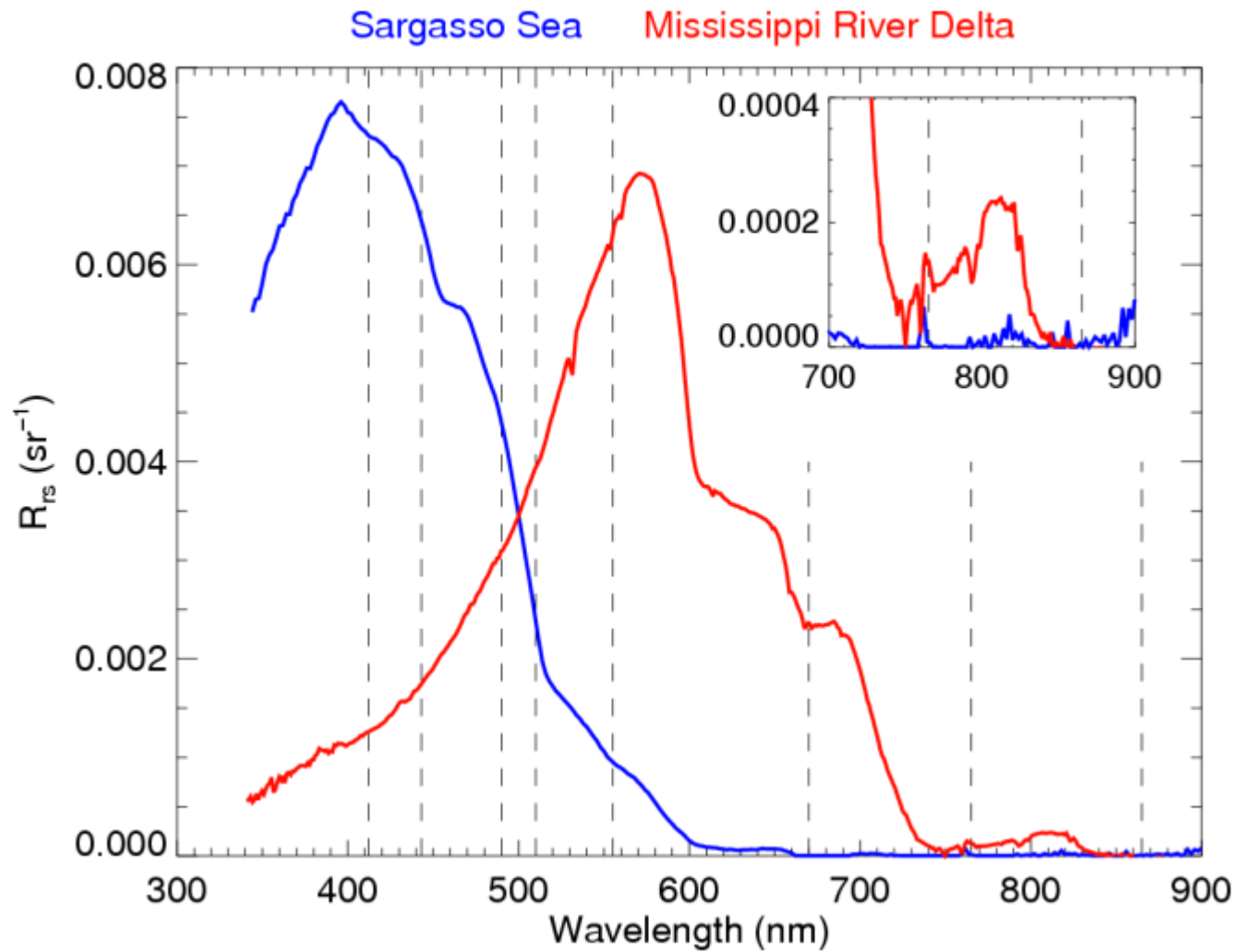
are R_{rs} (NIR) really black?

Sargasso Sea



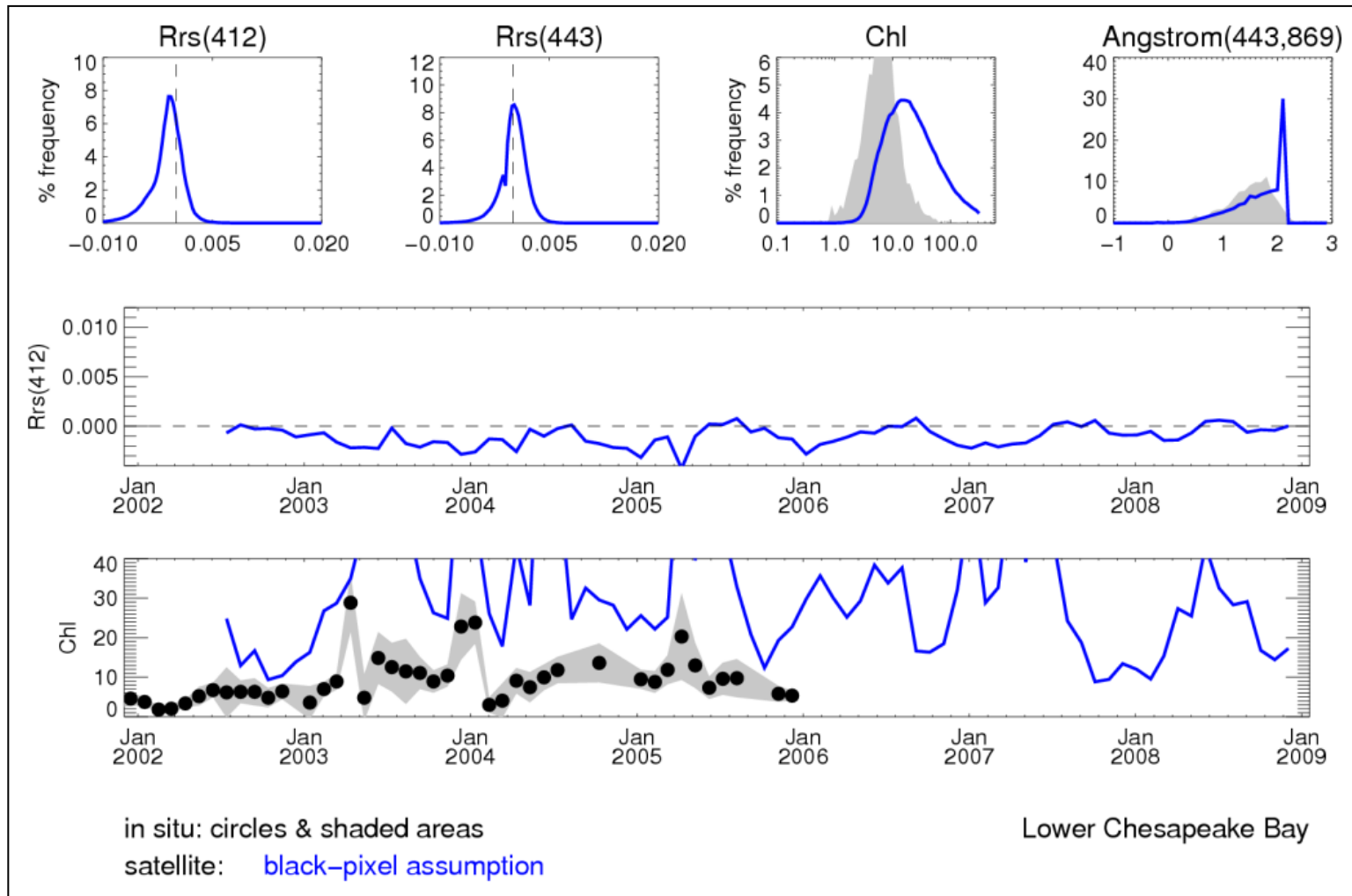
black pixel assumption

are $R_{rs}(\text{NIR})$ really black?



black pixel assumption

what happens when we don't account for $R_{rs}(\text{NIR}) > 0$?



use the “black pixel” assumption (e.g., SeaWiFS 1997-2000)

black pixel assumption

what to do when $R_{rs}(\text{NIR}) > 0$?

many approaches exist, here are a few examples:

assign aerosols (ϵ) and/or water contributions ($R_{rs}(\text{NIR})$)

e.g., Hu et al. 2000, Ruddick et al. 2000

use shortwave infrared bands

e.g., Wang & Shi 2007

correct/model the non-negligible $R_{rs}(\text{NIR})$

Siegel et al. 2000 used in SeaWiFS Reprocessing 3 (2000)

Stumpf et al. 2003 used in SeaWiFS Reprocessing 4 (2002)

Lavender et al. 2005 MERIS

Bailey et al. 2010 used in SeaWiFS Reprocessing 6 (2009)

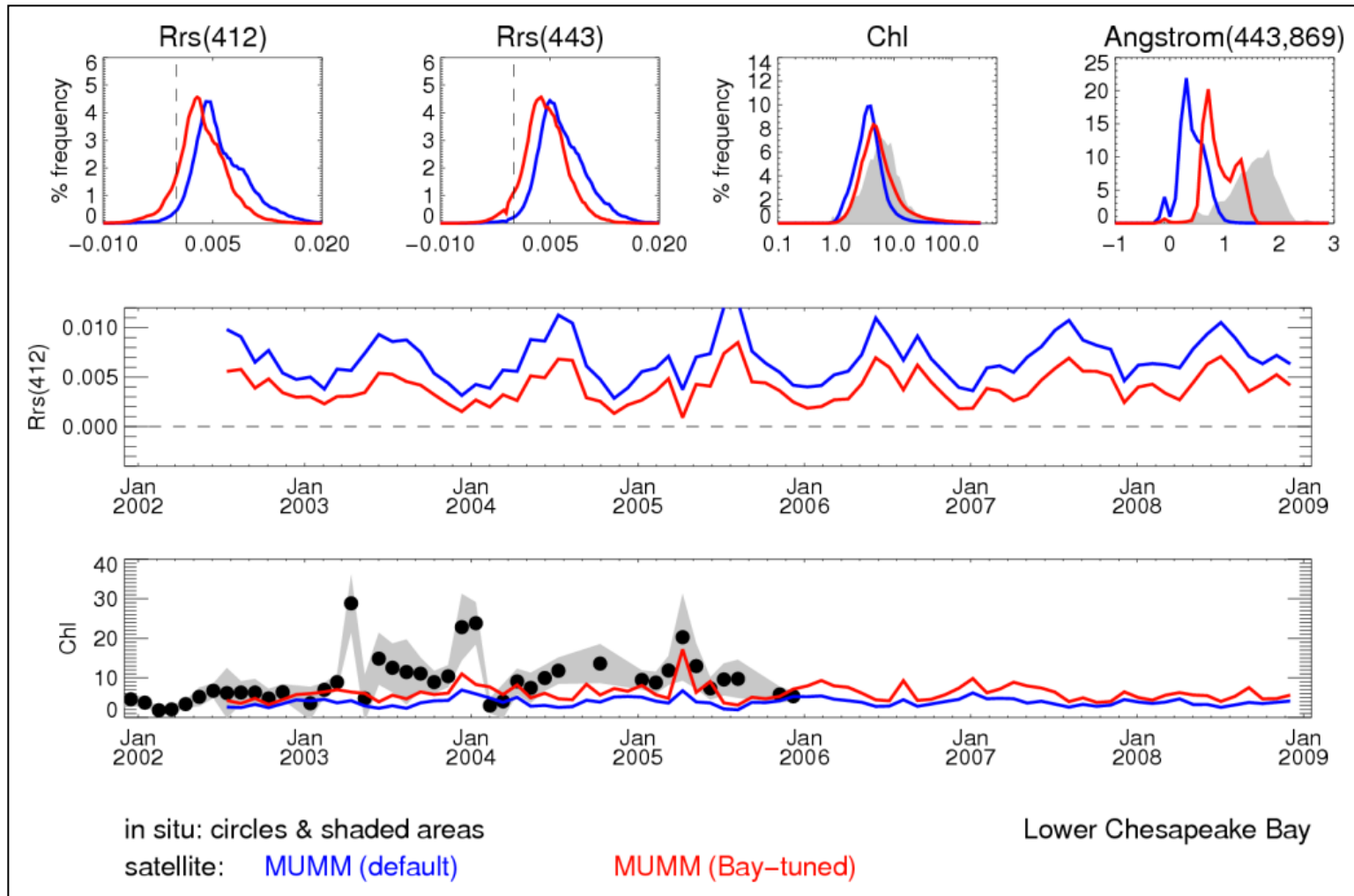
use a coupled ocean-atmosphere optimization

e.g., Chomko & Gordon 2001, Stamnes et al. 2003, Kuchinke et al. 2009

field data!

black pixel assumption

fixed aerosol & water contributions (MUMM)



assign ε & ρ_w (NIR) (via fixed values, a climatology, nearby pixels)

black pixel assumption

advantages:

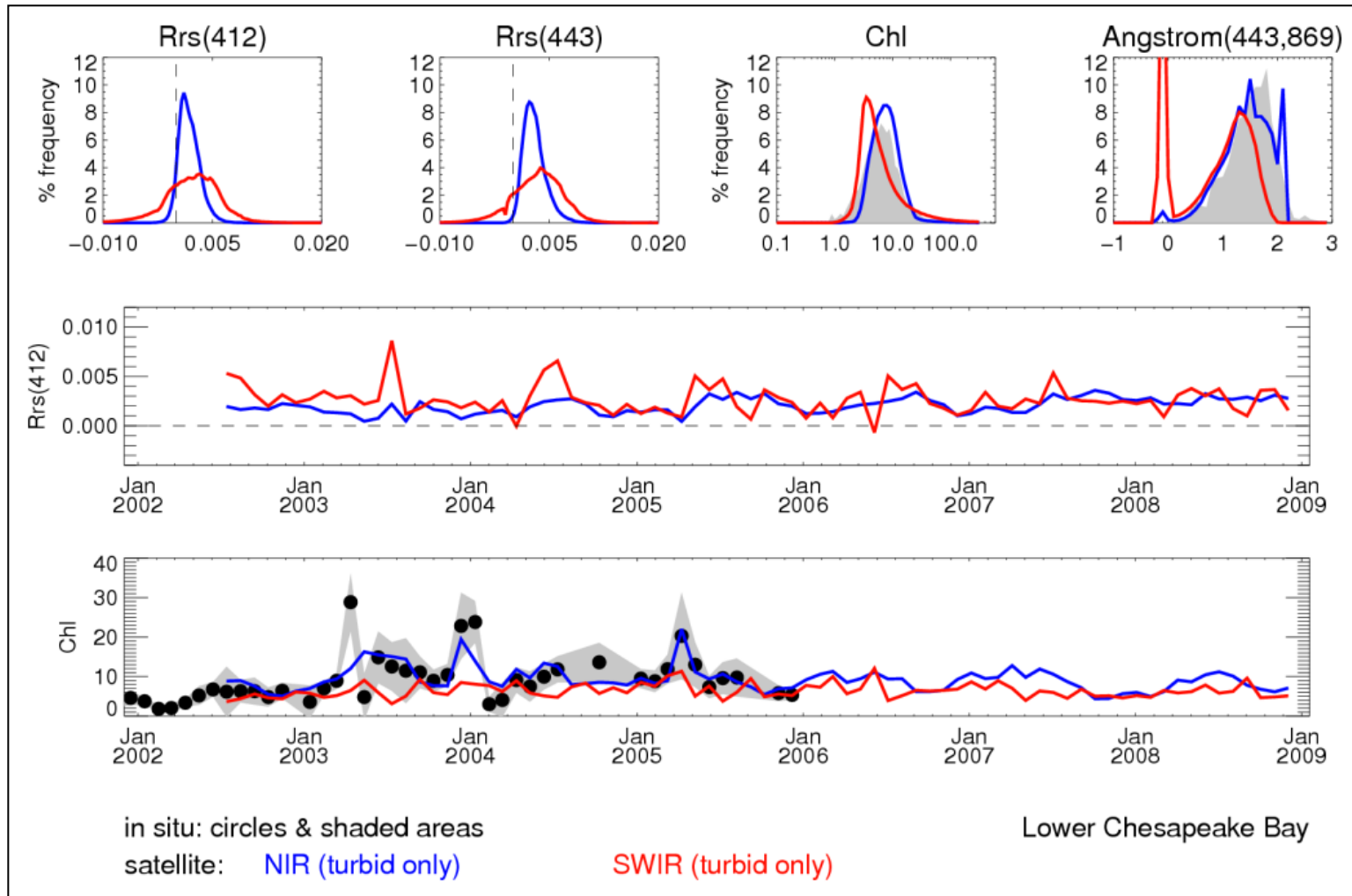
accurate configuration leads to accurate aerosol & R_{rs} (NIR) retrievals
several configuration options: fixed values, climatologies, nearby pixels
method available for all past, present, & future ocean color satellites

disadvantages:

no configuration is valid at all times for all water masses
requires local knowledge of changing aerosol & water properties
implementation can be complicated for operational processing

black pixel assumption

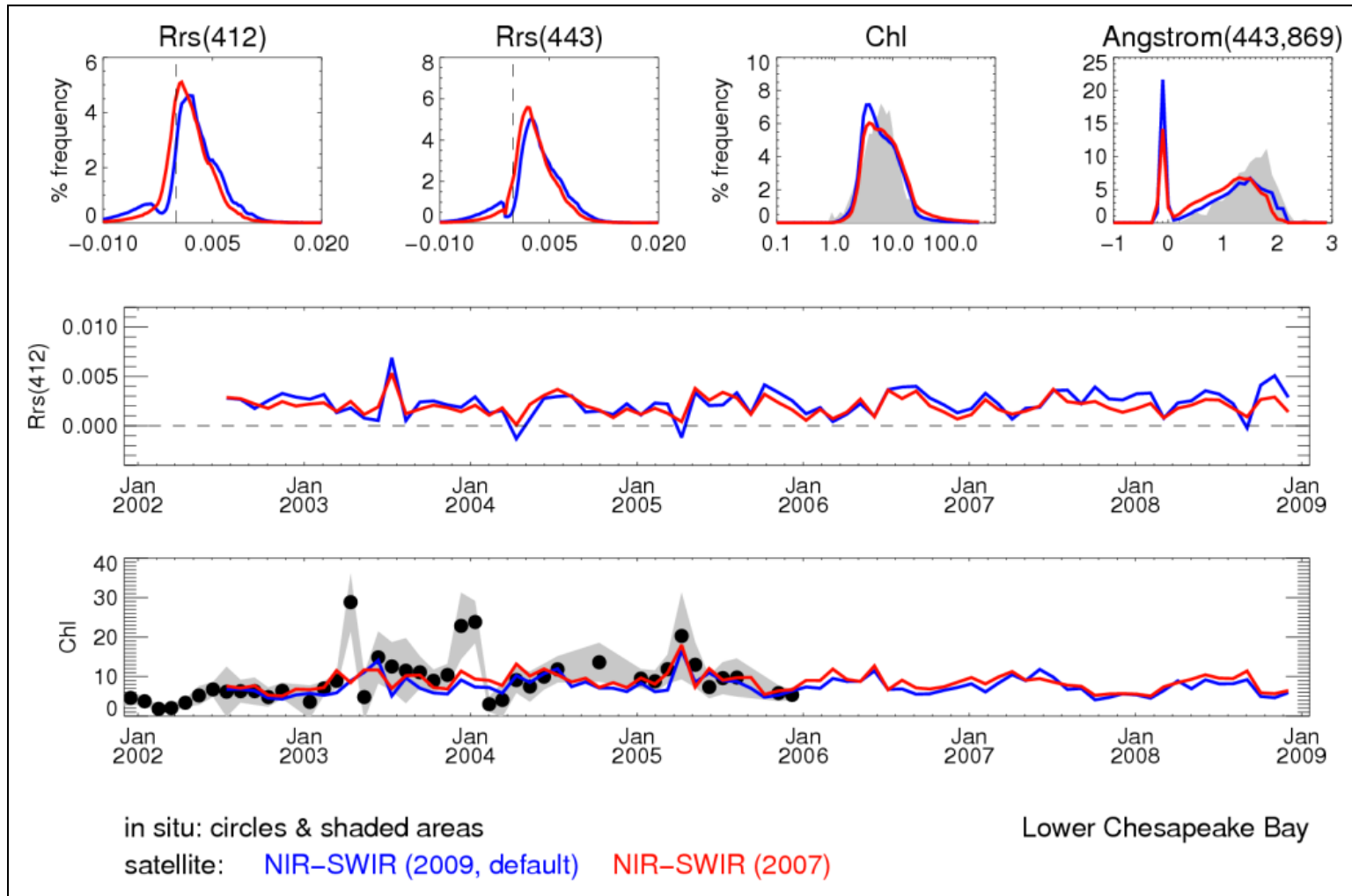
use of SWIR bands only



compare NIR & SWIR retrievals when considering only “turbid pixels”

black pixel assumption

use of NIR + SWIR bands



use SWIR bands in “turbid” water, otherwise use NIR bands

black pixel assumption

advantages:

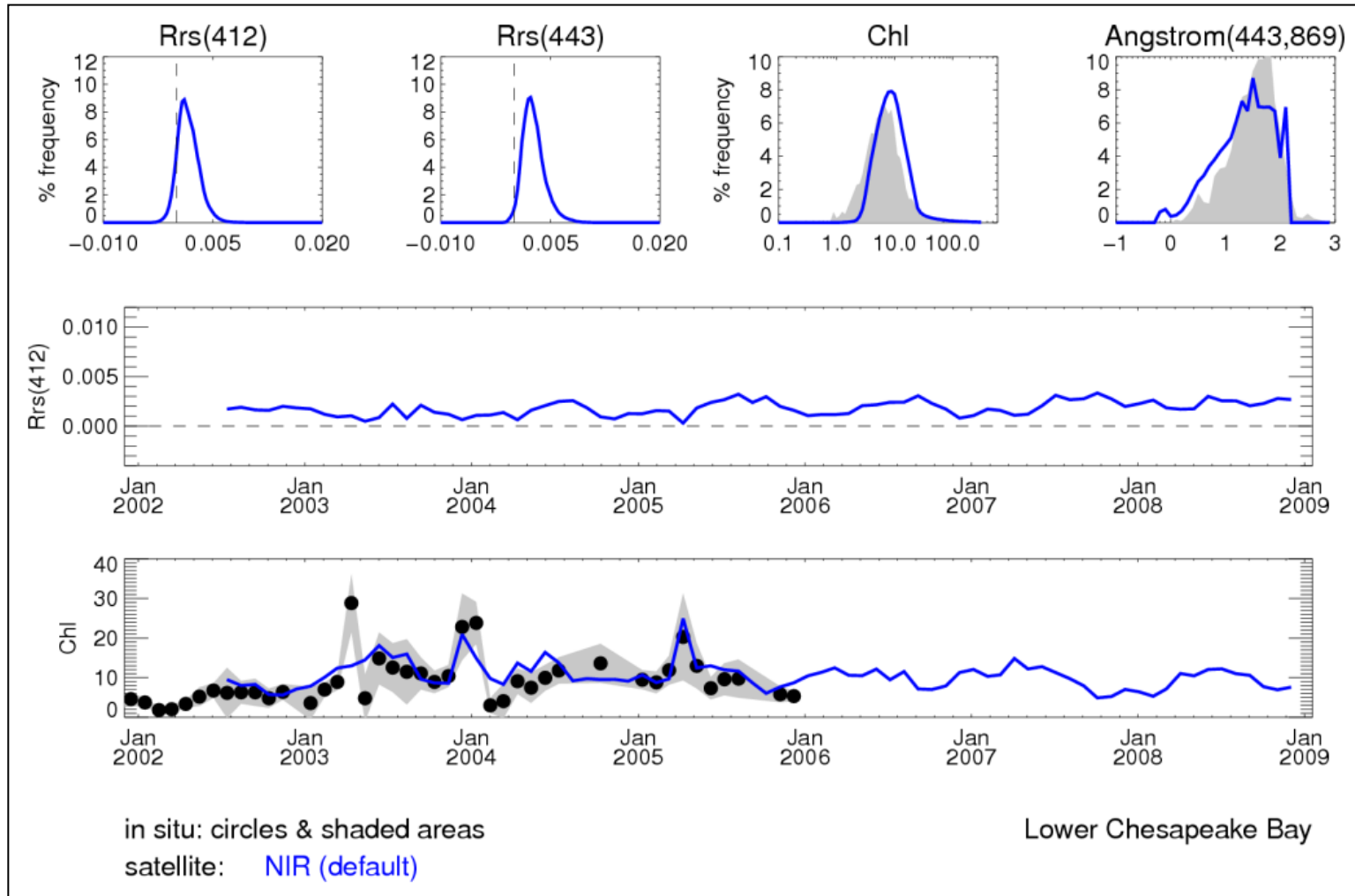
“black pixel” assumption largely satisfied in SWIR region of spectrum
straightforward implementation for operational processing

disadvantages:

only available for instruments with SWIR bands
SWIR bands on MODIS have inadequate signal-to-noise (SNR) ratios
difficult to vicariously calibrate the SWIR bands on MODIS
must define conditions for switching from NIR to SWIR

black pixel assumption

correction of non-negligible R_{rs} (NIR)



estimate R_{rs} (NIR) using a bio-optical model

operational SeaWiFS & MODIS processing ~ 2000-present

black pixel assumption

advantages:

method available for all past, present, & future ocean color missions
straightforward implementation for operational processing

disadvantages:

bio-optical model not valid at all times for all water masses

black pixel assumption – bio-optical model

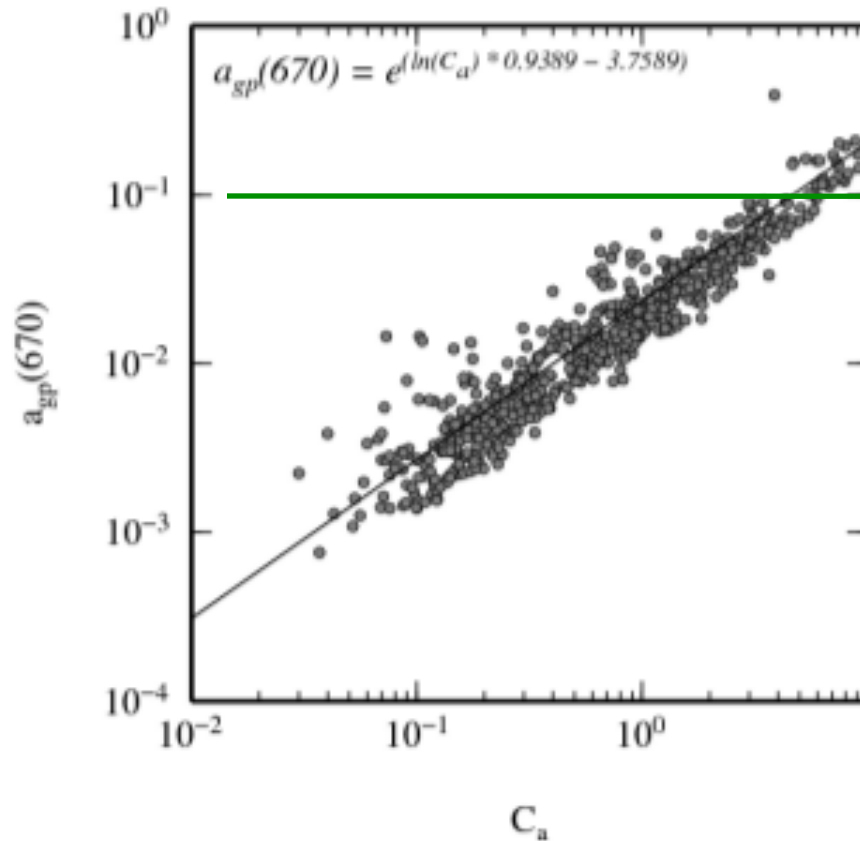
initial $R_{rs}(670)$ measured by satellite (using $R_{rs}(765) = 0$)

Bailey et al., Optics
Express, 2010

black pixel assumption – bio-optical model

initial $R_{rs}(670)$ measured by satellite (using $R_{rs}(765) = 0$)

model $a(670) = a_w(670) + a_{pg}(670)$



= 0.1 m^{-1}

$a_w(670) = 0.44 \text{ m}^{-1}$

Bailey et al., Optics Express, 2010

black pixel assumption – bio-optical model

initial $R_{rs}(670)$ measured by satellite (using $R_{rs}(765) = 0$)

model $a(670) = a_w(670) + a_{pg}(670)$

estimate $b_b(670)$ using $R_{rs}(670)$, $a(670)$, & $G(670)$ [Morel et al. 2002]

$$R_{rs}(670) = G(670) \frac{b_b(670)}{a(670) + b_b(670)}$$

Bailey et al., Optics
Express, 2010

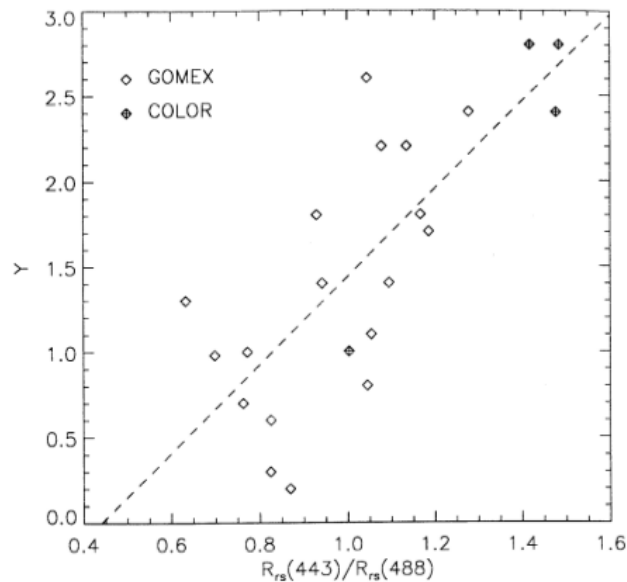
black pixel assumption – bio-optical model

initial $R_{rs}(670)$ measured by satellite (using $R_{rs}(765) = 0$)

model $a(670) = a_w(670) + a_{pg}(670)$

estimate $b_b(670)$ using $R_{rs}(670)$, $a(670)$, & $G(670)$ [Morel et al. 2002]

model η using $R_{rs}(443)$ & $R_{rs}(555)$ [Lee et al. 2002]



$$\eta = 2.0 \left[1 - 1.2 \exp\left(-0.9 \frac{R_{rs}(443)}{R_{rs}(555)}\right) \right]$$

from Carder et al. 1999

Bailey et al., Optics Express, 2010

black pixel assumption – bio-optical model

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estimate $b_b(765)$ using $b_b(670)$ & η

$$b_b(765) = b_{bw}(765) + b_{bp}(670) \left(\frac{670}{765} \right)^\eta$$

Bailey et al., Optics
Express, 2010

black pixel assumption – bio-optical model

initial $R_{rs}(670)$ measured by satellite (using $R_{rs}(765) = 0$)

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model η using $R_{rs}(443)$ & $R_{rs}(555)$ [Lee et al. 2002]

estimate $b_b(765)$ using $b_b(670)$ & η

reconstruct $R_{rs}(765)$ using $b_b(765)$, $a_w(765)$, & $G(765)$

$$R_{rs}(765) = G(765) \frac{b_b(765)}{a_w(765) + b_b(765)}$$

$$a_w(765) = 2.85 \text{ m}^{-1}$$

Bailey et al., Optics
Express, 2010

black pixel assumption – bio-optical model

initial $R_{rs}(670)$ measured by satellite (using $R_{rs}(765) = 0$)

model $a(670) = a_w(670) + a_{pg}(670)$

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estimate $b_b(765)$ using $b_b(670)$ & η

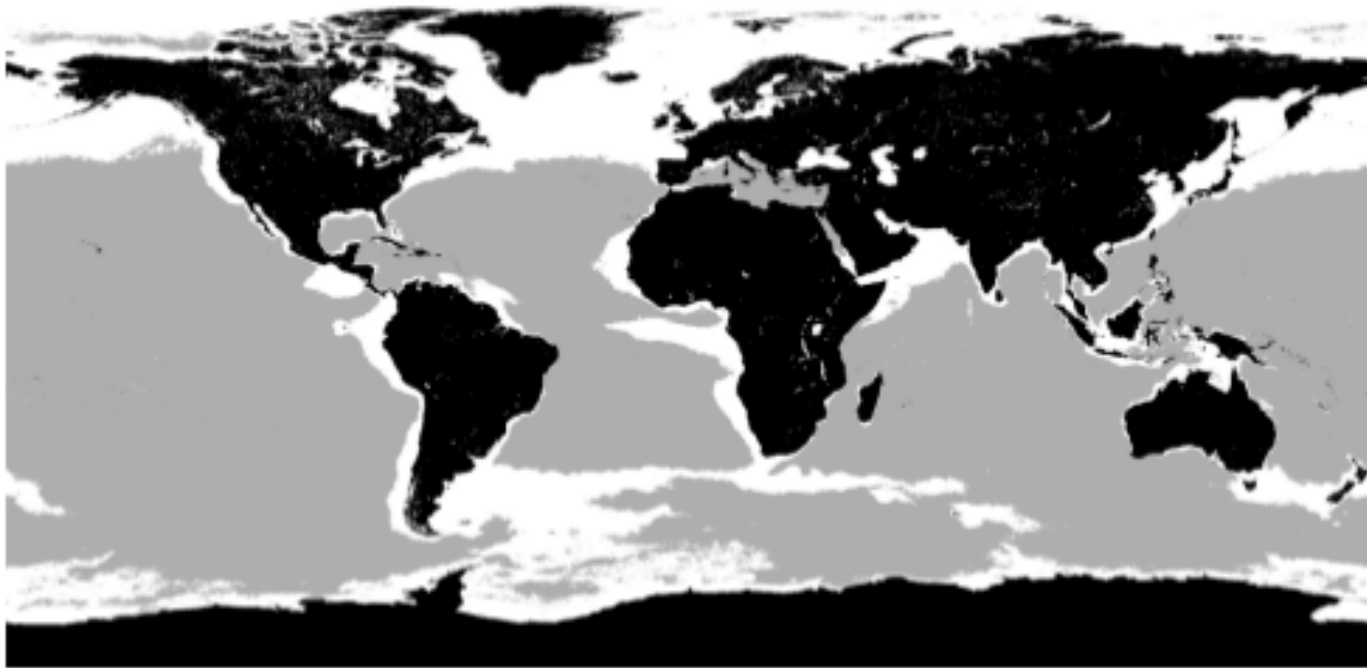
reconstruct $R_{rs}(765)$ using $b_b(765)$, $a_w(765)$, & $G(765)$

iterate until $R_{rs}(765)$ changes by $<2\%$ (typically 3-4 iterations)

Bailey et al., Optics
Express, 2010

black pixel assumption – bio-optical model

locations of application of bio-optical model



black = land; grey = $\text{Chl} < 0.3 \text{ mg m}^{-3}$; white $\text{Chl} > 0.3 \text{ mg m}^{-3}$

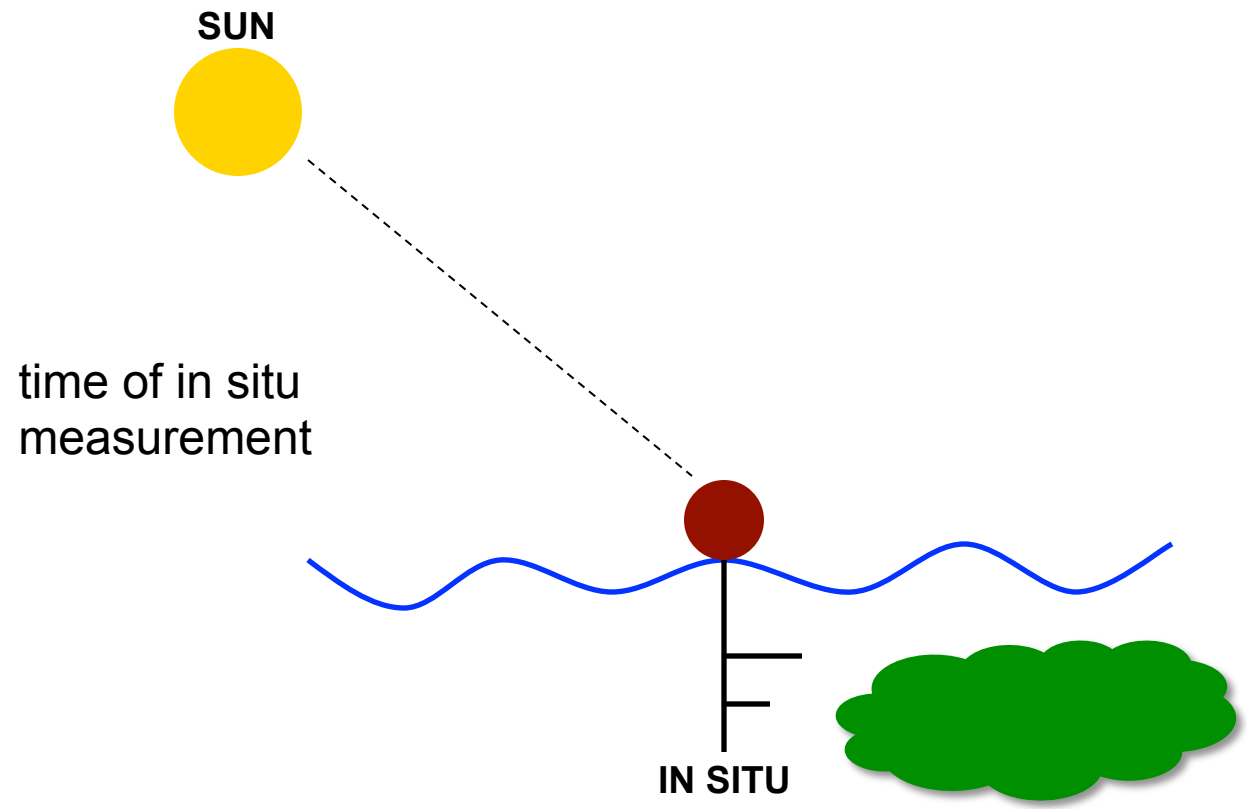
not applied when $\text{Chl} < 0.3 \text{ mg m}^{-3}$

weighted application when $0.3 < \text{Chl} < 0.7 \text{ mg m}^{-3}$

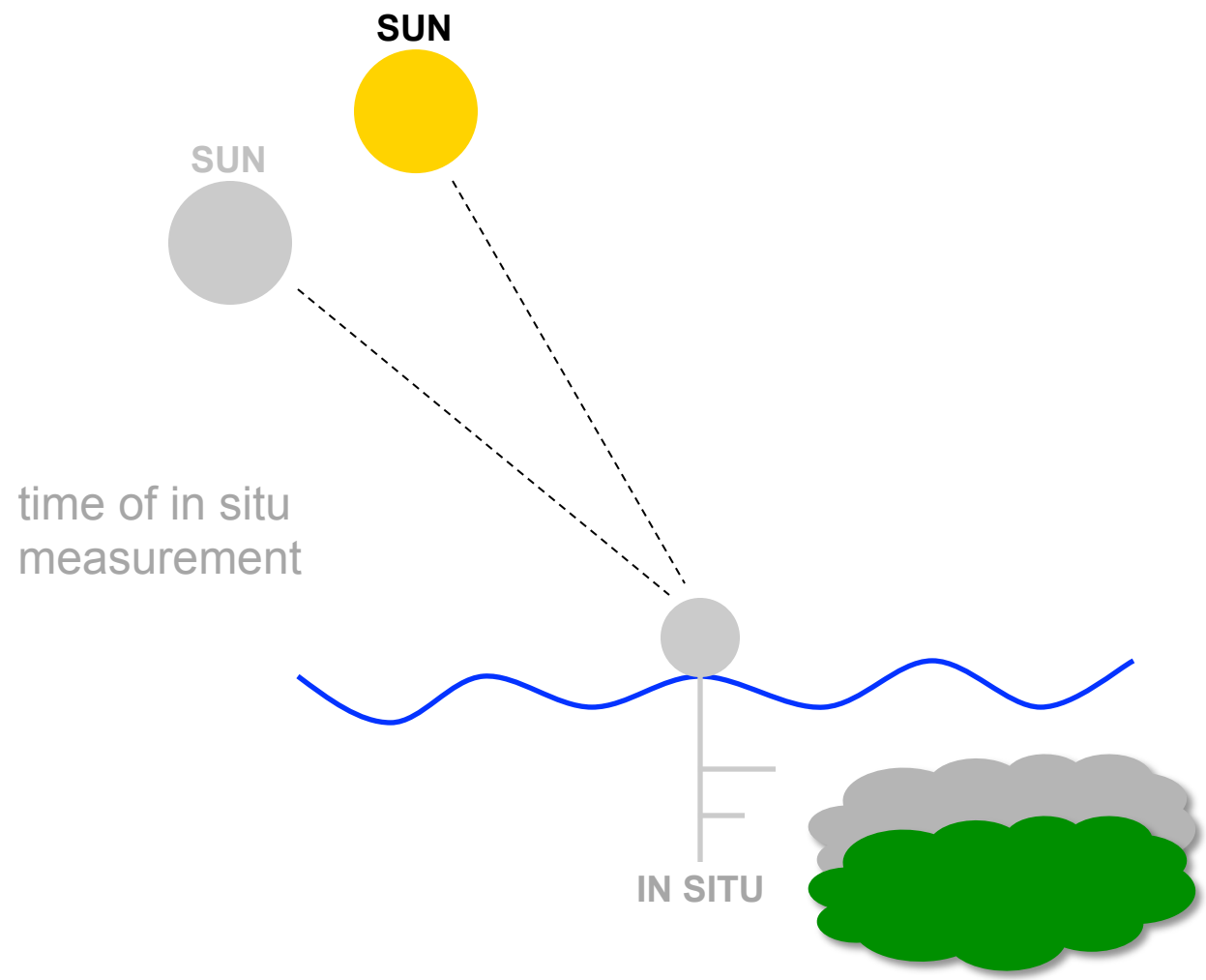
fully applied when $\text{Chl} > 0.7 \text{ mg m}^{-3}$

Bailey et al., Optics Express, 2010

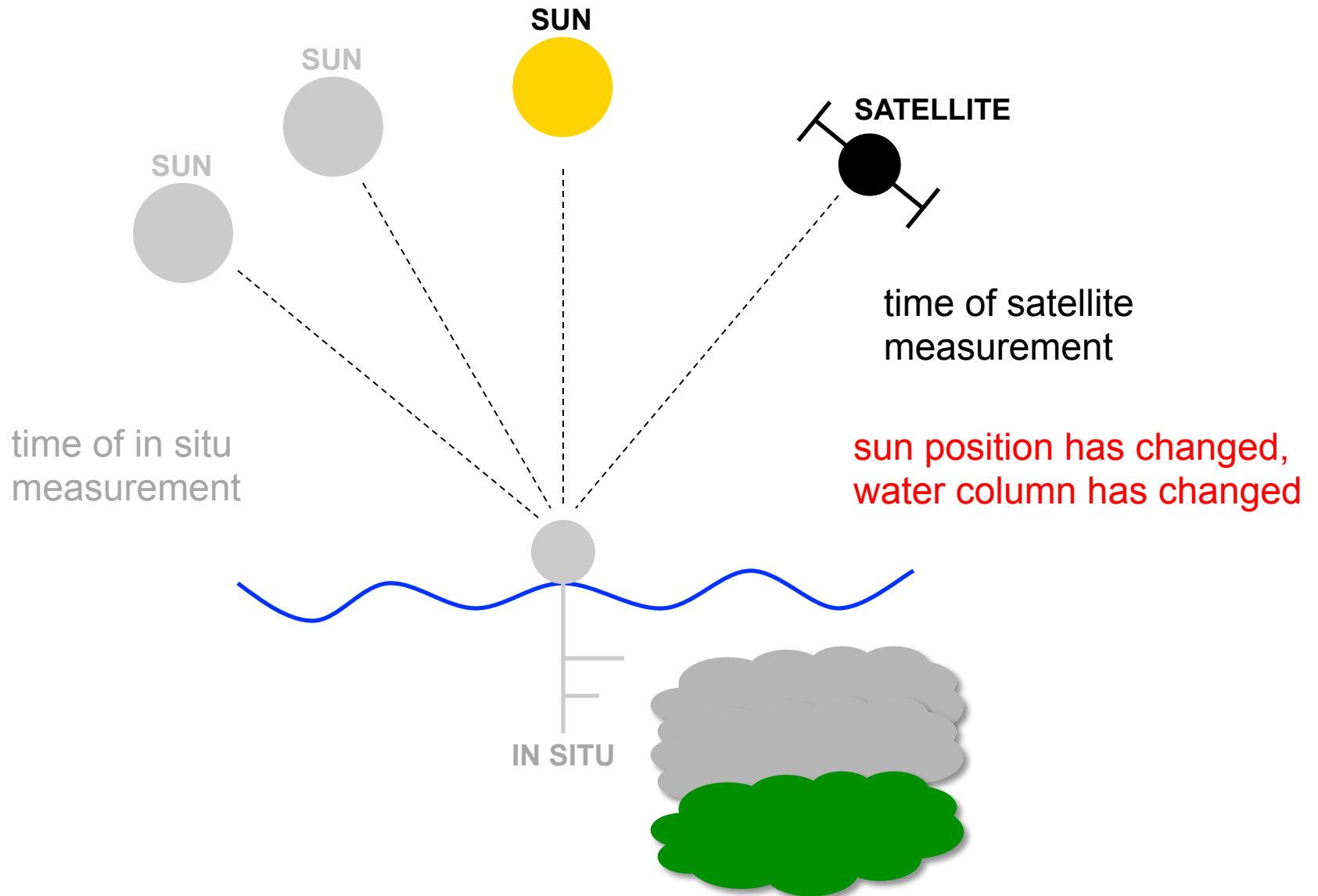
bidirectional reflectance correction



bidirectional reflectance correction



bidirectional reflectance correction



bidirectional reflectance correction

we normalize R_{rs} to account for Sun's changing position in the sky:

pathlengths through atmosphere

transmission of light through air-sea & sea-air interfaces

angular features of in-water volume scattering functions

Morel et al., Applied Optics, 2002

normalize all measurements to
condition of overhead Sun

$\mathfrak{R}, \mathfrak{R}_0, F, F_0, Q, Q_0$

from look up tables based on Chl
& geometries of Sun & sensor

$$[L_w]_N^{\text{ex}} = [L_w]_N \frac{\mathfrak{R}_0}{\mathfrak{R}(\theta', W)} \frac{f_0(\tau_a, W, \text{IOP})}{Q_0(\tau_a, W, \text{IOP})} \times \left(\frac{f(\theta_s, \tau_a, W, \text{IOP})}{Q(\theta_s, \theta', \phi, \tau_a, W, \text{IOP})} \right)^{-1},$$

bidirectional reflectance correction

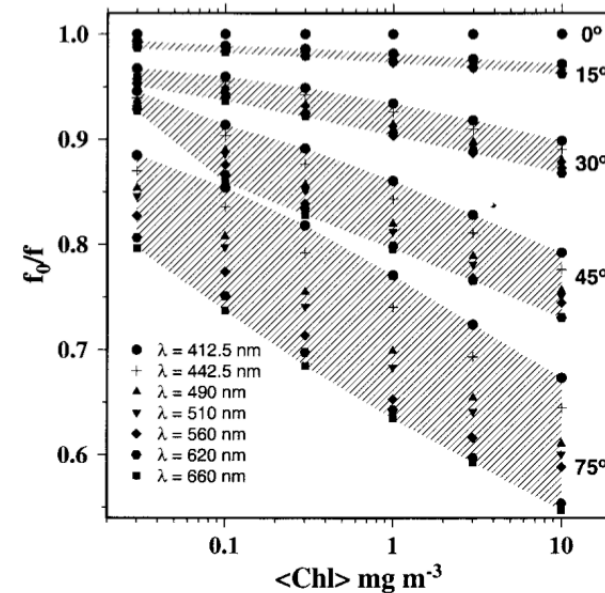


Fig. 13. Evolution of the f_0/f ratio with the Chl and for solar angles as indicated by the shaded areas (the values for $\theta_s = 60^\circ$, which overlap those for 45° and 75° , are not displayed); the symbols are for the various wavelengths.

Morel et al., Applied Optics, 2002

normalize all measurements to condition of overhead Sun

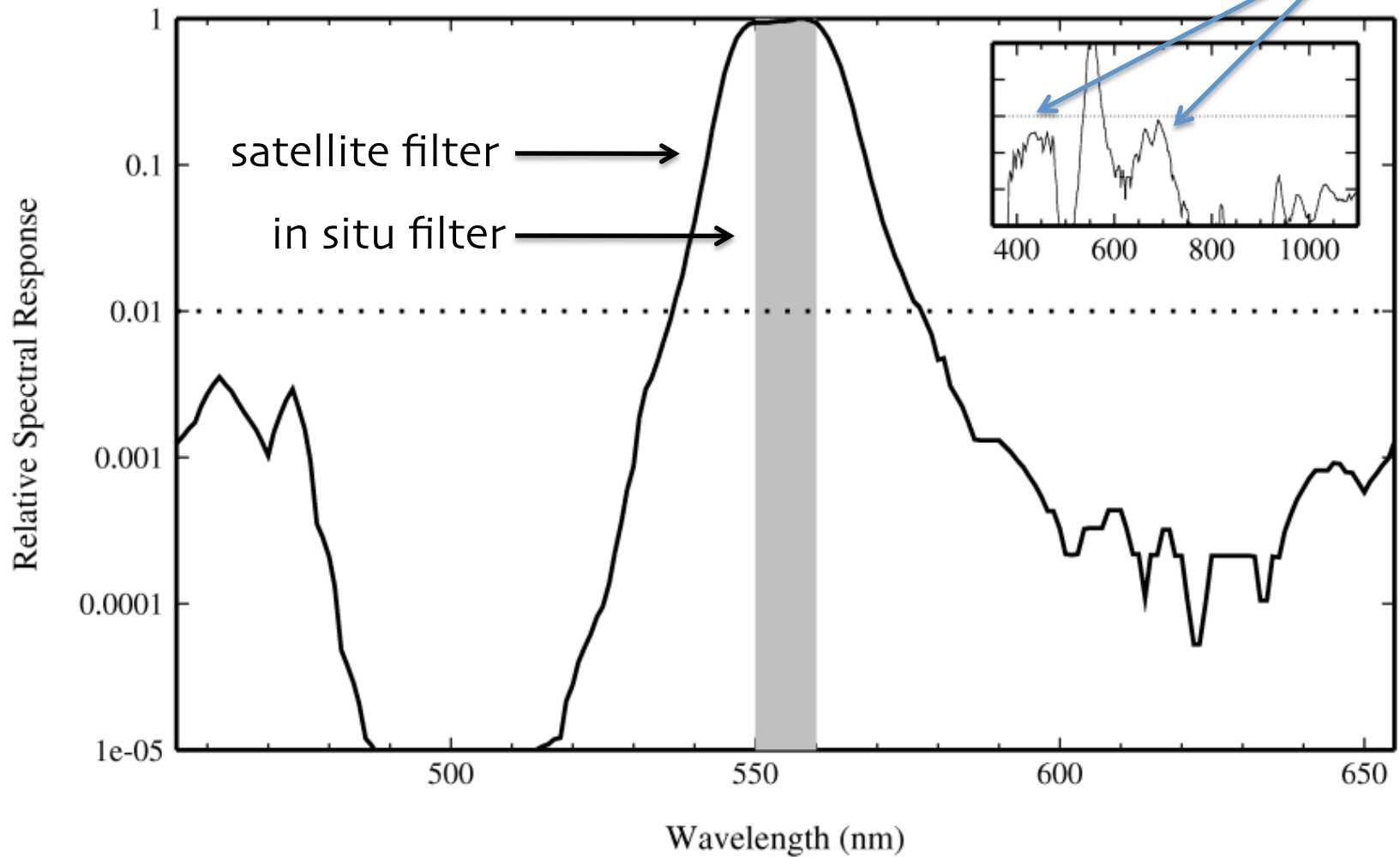
$$\mathfrak{R}, \mathfrak{R}_0, F, F_0, Q, Q_0$$

from look up tables based on Chl & geometries of Sun & sensor

$$[L_w]_N^{\text{ex}} = [L_w]_N \frac{\mathfrak{R}_0}{\mathfrak{R}(\theta', W)} \frac{f_0(\tau_a, W, \text{IOP})}{Q_0(\tau_a, W, \text{IOP})} \times \left(\frac{f(\theta_s, \tau_a, W, \text{IOP})}{Q(\theta_s, \theta', \phi, \tau_a, W, \text{IOP})} \right)^{-1},$$

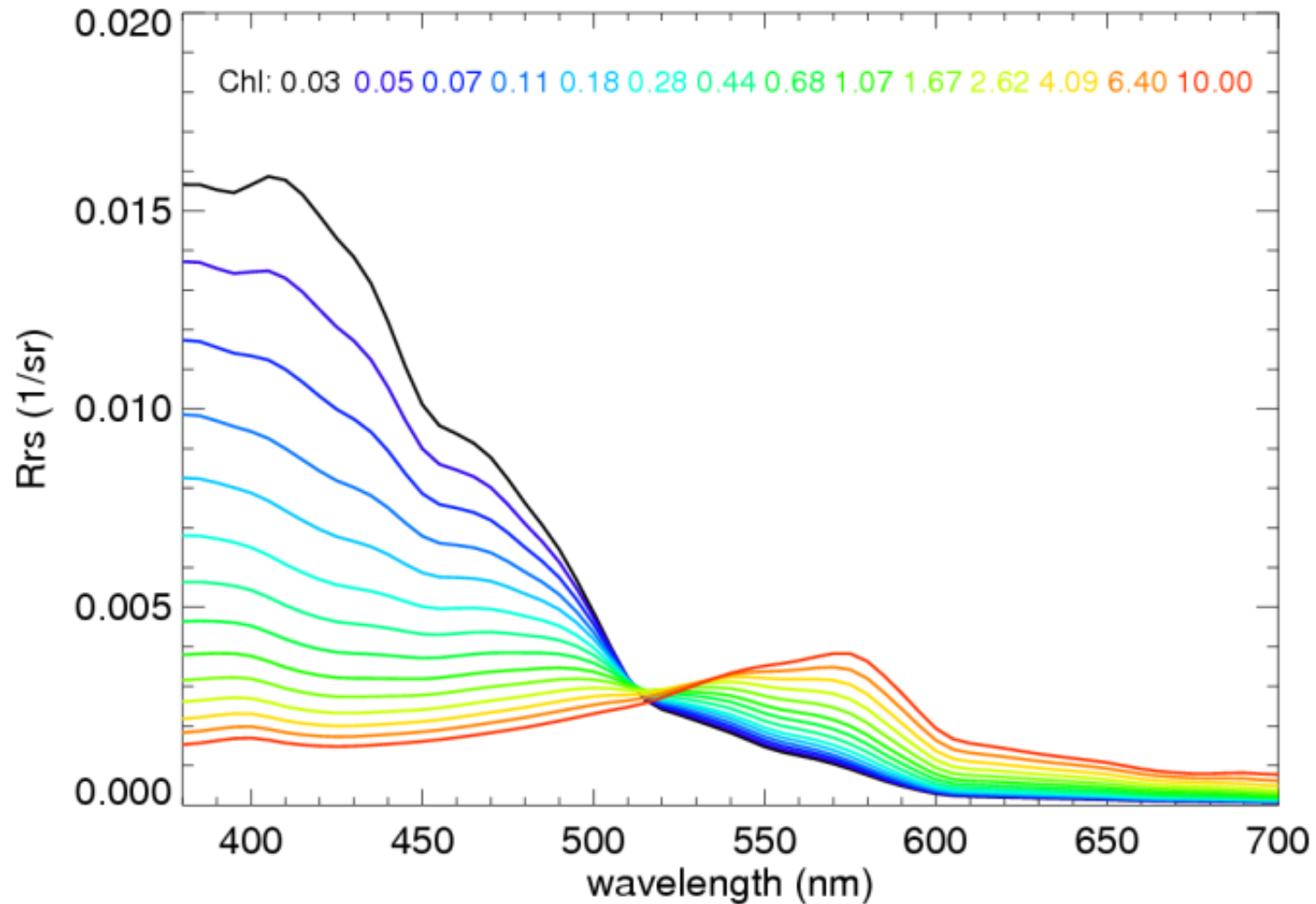
spectral bandpass correction

SeaWiFS Spectral Response for Band 5 (555 nm)



spectral bandpass correction

calculate $R_{rs}(\lambda)$ using Morel & Maritorena (2001) for $0.01 < \text{Chl} < 3 \text{ mg m}^{-3}$



spectral bandpass correction

calculate $R_{rs}(\lambda)$ using MM01 for $0.01 < \text{Chl} < 3 \text{ mg m}^{-3}$

for each satellite band λ_i :

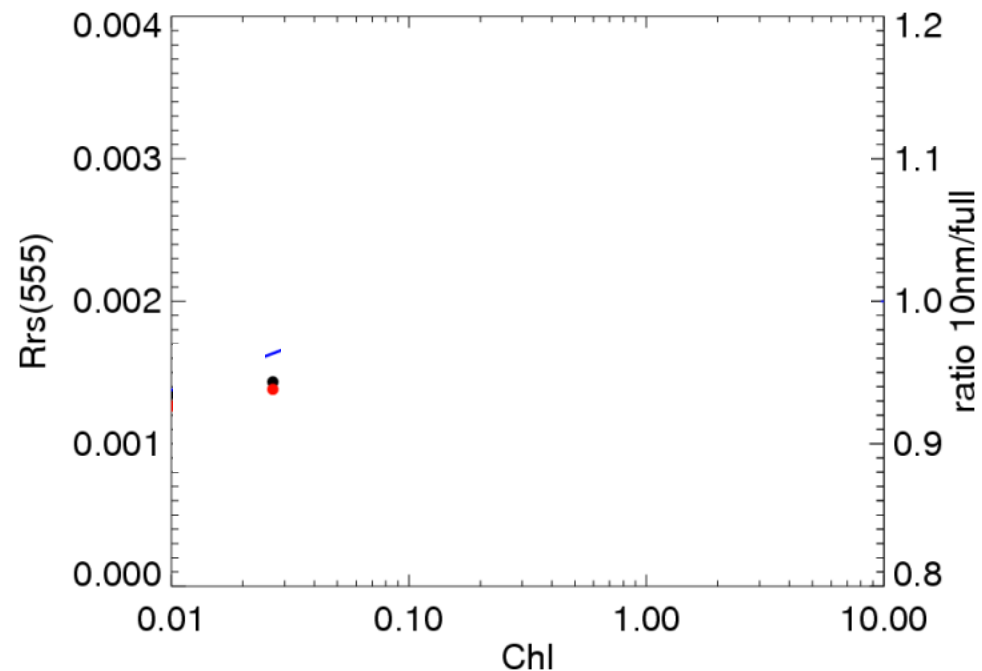
for each Chl_j :

calculate 10-nm mean $R_{rs}(\lambda_i, \text{Chl}_j)$

calculate full-spectral-response $R_{rs}(\lambda_i, \text{Chl}_j)$

ratio $r(\lambda_i, \text{Chl}_j) = \text{10-nm} / \text{full-band}$

10-nm
full-band
ratio



spectral bandpass correction

calculate $R_{rs}(\lambda)$ using MM01 for $0.01 < \text{Chl} < 3 \text{ mg m}^{-3}$

for each satellite band λ_i :

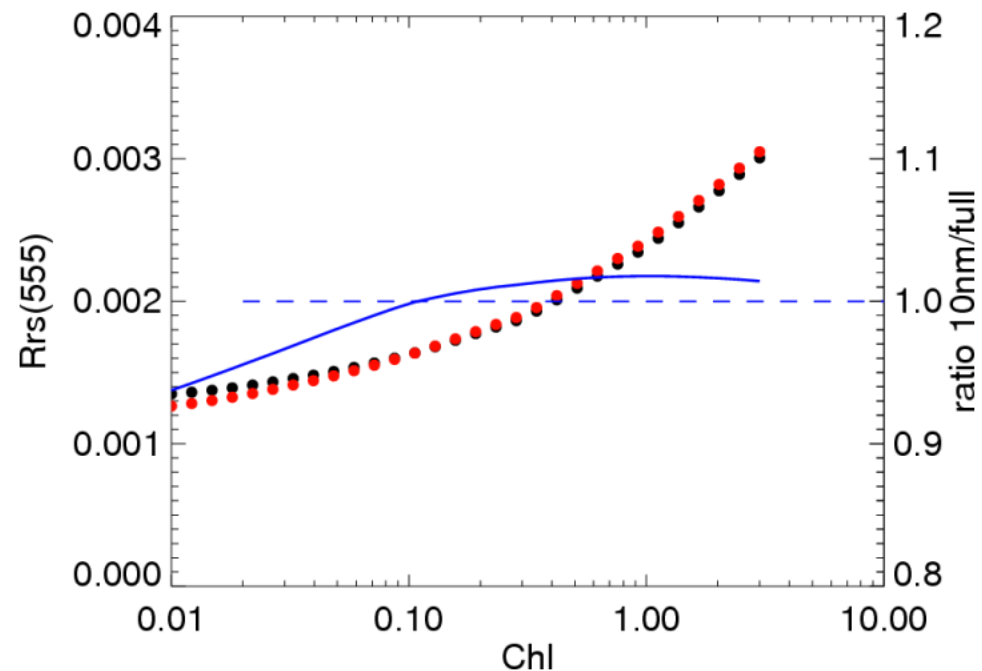
for each Chl_j :

calculate 10-nm mean $R_{rs}(\lambda_i, \text{Chl}_j)$

calculate full-spectral-response $R_{rs}(\lambda_i, \text{Chl}_j)$

ratio $r(\lambda_i, \text{Chl}_j) = 10\text{-nm} / \text{full-band}$

10-nm
full-band
ratio



spectral bandpass correction

calculate $R_{rs}(\lambda)$ using *MMo1* for $0.01 < \text{Chl} < 3 \text{ mg m}^{-3}$

for each satellite band λ_i :

for each Chl_j :

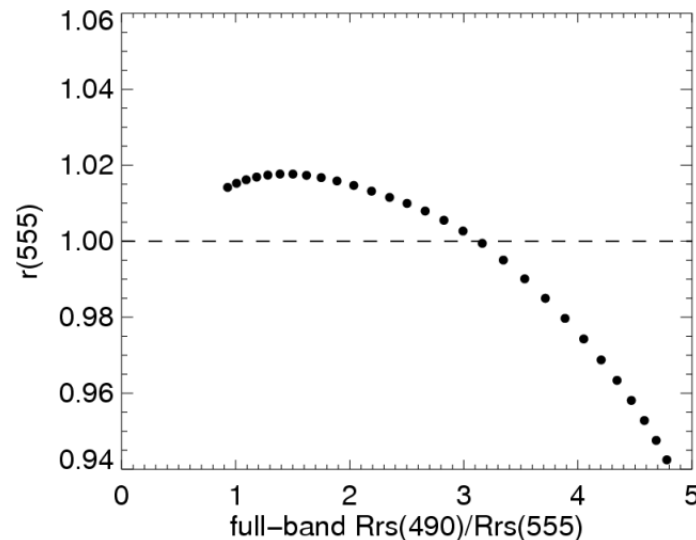
calculate 10-nm mean $R_{rs}(\lambda_i, \text{Chl}_j)$

calculate full-spectral-response $R_{rs}(\lambda_i, \text{Chl}_j)$

ratio $r(\lambda_i, \text{Chl}_j) = 10\text{-nm} / \text{full-band}$

plot / regress ratio vs. full-band $R_{rs}(490) / R_{rs}(555)$

derive polynomial expression to estimate ratio from full-band ratio



spectral bandpass correction

calculate $R_{rs}(\lambda)$ using *MM01* for $0.01 < \text{Chl} < 3 \text{ mg m}^{-3}$

for each satellite band λ_i :

for each Chl_j :

calculate 10-nm mean $R_{rs}(\lambda_i, \text{Chl}_j)$

calculate full-spectral-response $R_{rs}(\lambda_i, \text{Chl}_j)$

ratio $r(\lambda_i, \text{Chl}_j) = 10\text{-nm} / \text{full-band}$

plot / regress ratio vs. full-band $R_{rs}(490) / R_{rs}(555)$

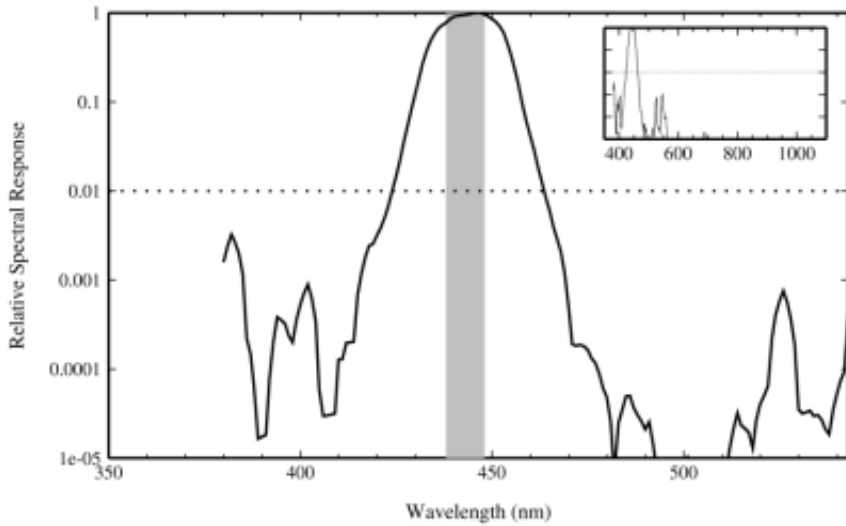
derive polynomial expression to estimate ratio from full-band ratio

to “adjust” satellite full-band to 10-nm, apply correction factors

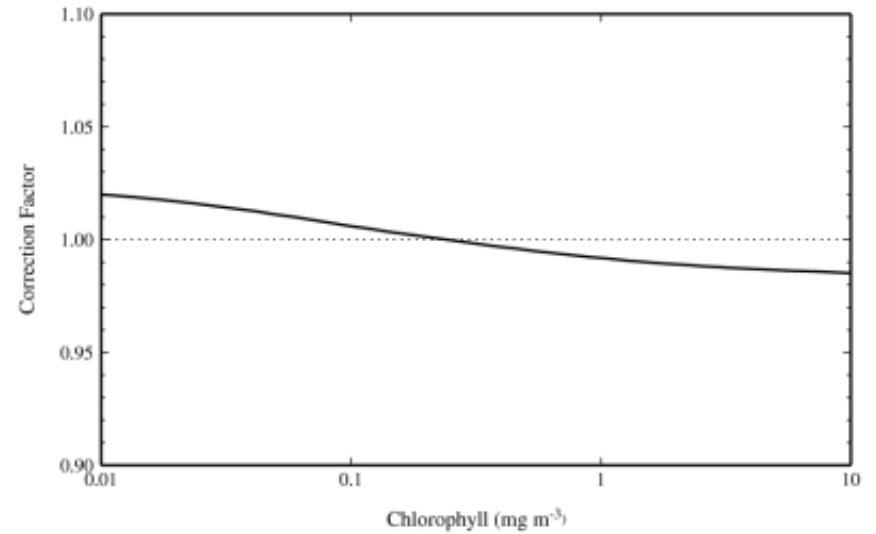
$R_{rs}(\lambda_i, 10\text{-nm}) = r(\lambda_i) * R_{rs}(\lambda_i, \text{full-band})$

spectral bandpass correction

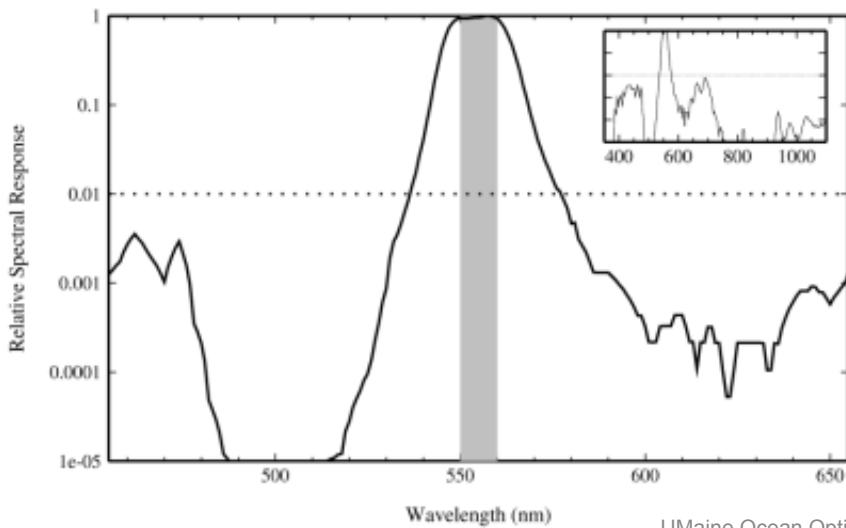
SeaWiFS Spectral Response for Band 2 (443 nm)



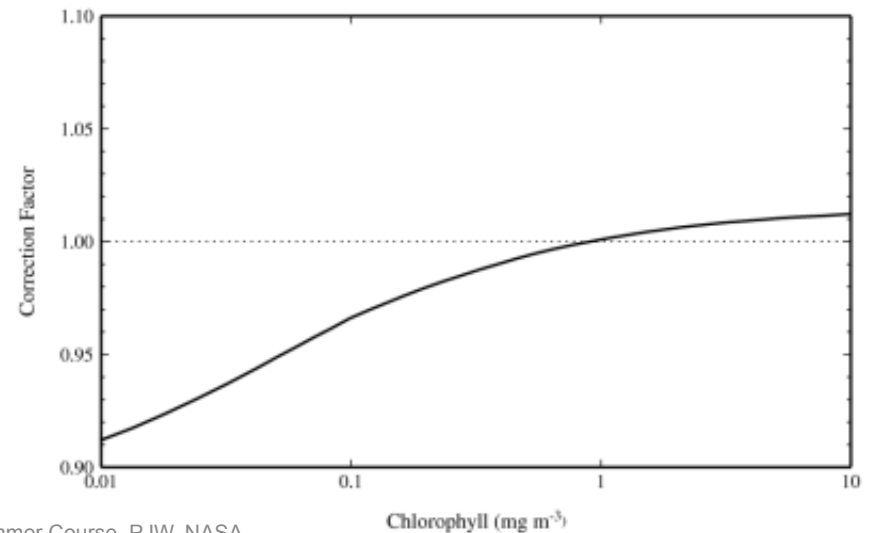
SeaWiFS Out-of-Band Correction for Band 2 (443 nm)



SeaWiFS Spectral Response for Band 5 (555 nm)



SeaWiFS Out-of-Band Correction for Band 5 (555 nm)



spectral bandpass correction

why do we care?

satellite R_{rs} adjusted using a bio-optical model

take care when executing satellite-to-in situ match-ups

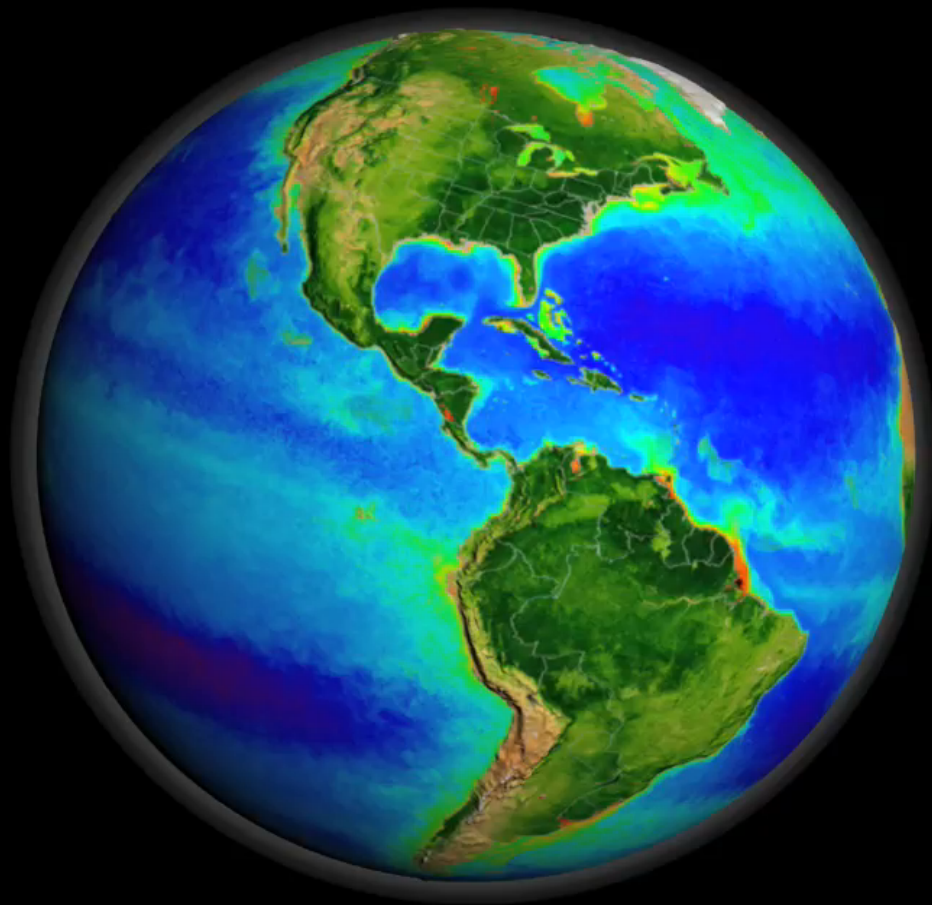
when using multispectral in situ radiometers:

enable the bandpass adjustment

when using hyperspectral in situ radiometers:

enable the adjustment when applying 10-nm filter to in situ R_{rs}

disable the adjustment when applying full-spectral-response to in situ R_{rs}



scorecard – ancillary data requirements

ancillary data

atmospheric pressure
water vapor
relative humidity
wind speed
ozone
NO₂
sea surface temperature
sea ice

ancillary source

NCEP
NCEP
NCEP
NCEP
OMI/TOMS
Sciamachy/OMI/GOME
Reynolds
NSIDC

uses

Rayleigh
transmittance
aerosol models
white caps, Sun glint, Rayleigh
transmittance
transmittance
bio-optical algorithms
masking

look-up tables, coefficients

aerosol models
Rayleigh
Rayleigh optical thickness
ozone absorption
NO₂ absorption
pure seawater absorption, scattering, index of refraction (temp/sal dependent)
f/Q (bidirectional reflectance distributions)
others ...