

# CERES Angular Distribution Model Working Group Report



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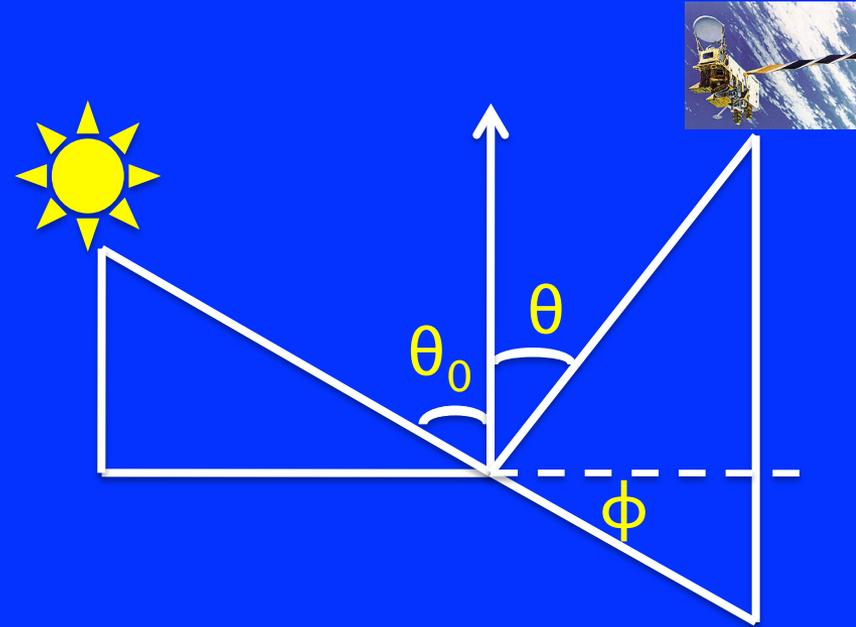


# Publications

- Wenyong Su, Norman Loeb, Lusheng Liang, Nana Liu, Chuntao Liu, The El Nino-Southern Oscillation Effect on Tropical Outgoing Longwave Radiation: A Daytime Versus Nighttime Perspective, JGR, 2017, DOI: 10.1002/2017JD027002.
- Zachary Eitzen, Wenyong Su, Kuan-Man Xu, Norman Loeb, Moguo Sun, David Doelling, Fred Rose, and Alejandro Bodas-Salcedo, Evaluation of a general circulation model by the CERES Flux-by-Cloud Type simulator, JGR. 2017, accepted.
- Wenyong Su, Lusheng Liang, Walter Miller, and Victor Sothcott, The effects of different footprint sizes and cloud algorithms on the top-of-atmosphere radiative flux calculation from CERES instrument on Suomi-NPP, Atmos. meas. tech., amt-2017-75, 2017, accepted.

# From radiance to flux: angular distribution models

- Sort observed radiances into angular bins over different scene types;
- Integrate radiance over all  $\theta$  and  $\phi$  to estimate the anisotropic factor for each scene type;
- Apply anisotropic factor to observed radiance to derive TOA flux;



$$R(\theta_0, \theta, \phi) = \frac{\pi \hat{I}(\theta_0, \theta, \phi)}{\int_0^{2\pi} \int_0^{\frac{\pi}{2}} \hat{I}(\theta_0, \theta, \phi) \cos\theta \sin\theta d\theta d\phi} = \frac{\pi \hat{I}(\theta_0, \theta, \phi)}{\hat{F}(\theta_0)}$$

$$F(\theta_0) = \frac{\pi I_o(\theta_0, \theta, \phi)}{R(\theta_0, \theta, \phi)}$$

## Predicted radiance vs. observed radiance

$$R(\theta_0, \theta, \phi) = \frac{\pi \hat{I}(\theta_0, \theta, \phi)}{\hat{F}(\theta_0)} \quad F(\theta_0) = \frac{\pi I_o(\theta_0, \theta, \phi)}{R(\theta_0, \theta, \phi)}$$

$$F(\theta_0) = \frac{I_o(\theta_0, \theta, \phi)}{\hat{I}(\theta_0, \theta, \phi)} \hat{F}(\theta_0)$$

- Predicted radiances can be used to verify the accuracy of ADM;

# Normalize predicted and observed radiance

Observed radiance:

$$I_j^o, \quad j = 1, \dots, n$$

Predicted radiance:

$$\hat{I}_j, \quad j = 1, \dots, n$$

$$\overline{I^o} = \frac{1}{n} \sum_{j=1}^n I_j^o \quad \overline{\hat{I}} = \frac{1}{n} \sum_{j=1}^n \hat{I}_j$$

$$RMS = \sqrt{\frac{1}{n} \sum_{j=1}^n \left( \frac{\hat{I}_j}{\overline{\hat{I}}} - \frac{I_j^o}{\overline{I^o}} \right)^2}$$

1°

1°

- RMS error between normalized predicted radiance and normalized observed radiance is closely related to the ADM error

## Is normalized radiance RMS error a good metric

- Simulate a radiance and flux database over different scene types, for a set of Sun-viewing geometries;
- Assuming the simulated radiances ( $I^s$ ) and fluxes ( $F^s$ ) are the truth;
- For each simulated radiance, there is a corresponding predicted radiance from CERES ADMs; and the CERES ADMs are used to convert the simulated radiances to fluxes;
- The RMS error between normalized predicted radiances and normalized simulated radiances can be calculated as follows:

$$RMS = \sqrt{\frac{1}{n} \sum_{j=1}^n \left( \frac{\hat{I}_j}{\hat{I}} - \frac{I_j^s}{I^s} \right)^2}$$

- The relative flux RMS error between simulated flux and ADMs inverted flux are calculated as:

$$\phi = \frac{\sqrt{\frac{1}{n} \sum_{j=1}^n \left( F_j - F_j^s \right)^2}}{\frac{1}{n} \sum_{j=1}^n F_j^s}$$

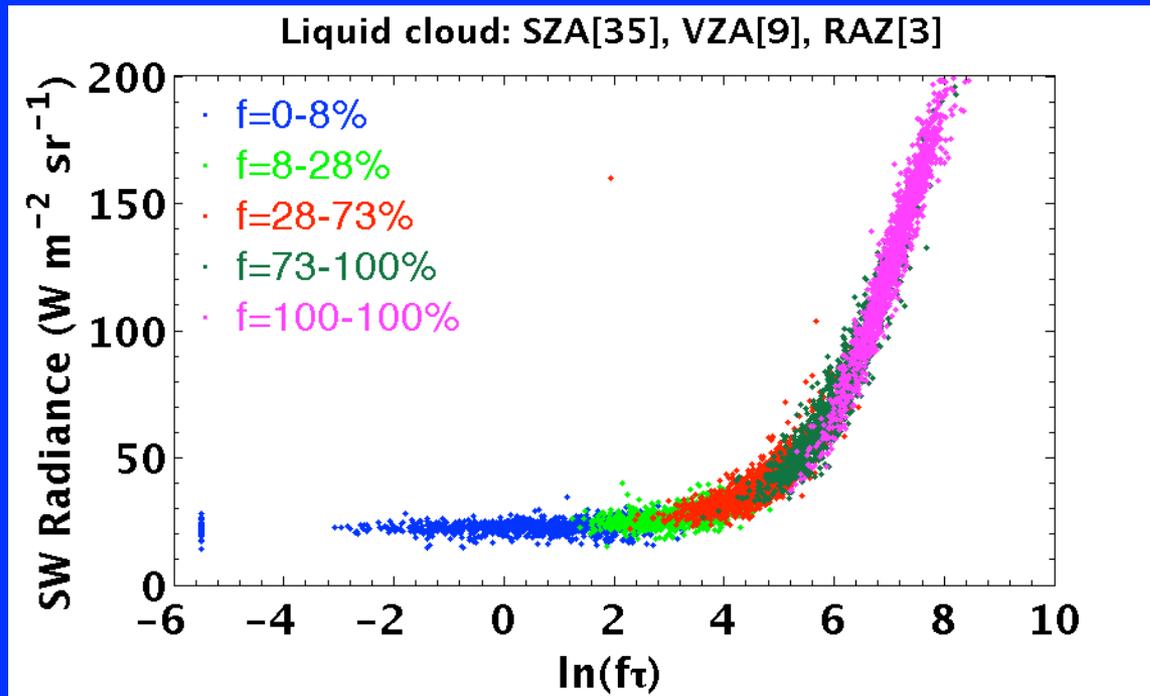
## Simulation over cloudy ocean

- Radiances and fluxes are calculated for solar zenith angles of 41.4 and 60.0 degrees over cloudy ocean for the following conditions:
  - Liquid cloud optical properties provided by Ping Yang's group, effective radius of 4, 10, 16, and 25  $\mu\text{m}$  are included;
  - Single-habit ice cloud optical properties provided by Ping Yang's group, effective diameter of 21, 46, and 115  $\mu\text{m}$  are included;
  - Cloud optical depths : 1, 2, 4, 12, 14, 20, 217;
  - Viewing zenith angles: 0 to 88 with a bin width of 4
  - Relative azimuth angles: 0 to 180 with a bin width of 10
  - Ocean wind speed: 9m/s
  - US standard atmospheric profiles

## Angular distribution model over cloudy ocean

- For glint angle  $> 20^\circ$ , or glint angle  $< 20^\circ$  and  $\ln(f\tau) > 6$ :
  - For a given  $[\theta_0, \theta, \varphi, ]$  bin, average instantaneous radiances into 775 intervals of  $\ln(f\tau)$ , separately for liquid, mixed, and ice clouds;
  - Apply a five-parameter sigmoidal fit to mean radiance and  $\ln(f\tau)$ ;

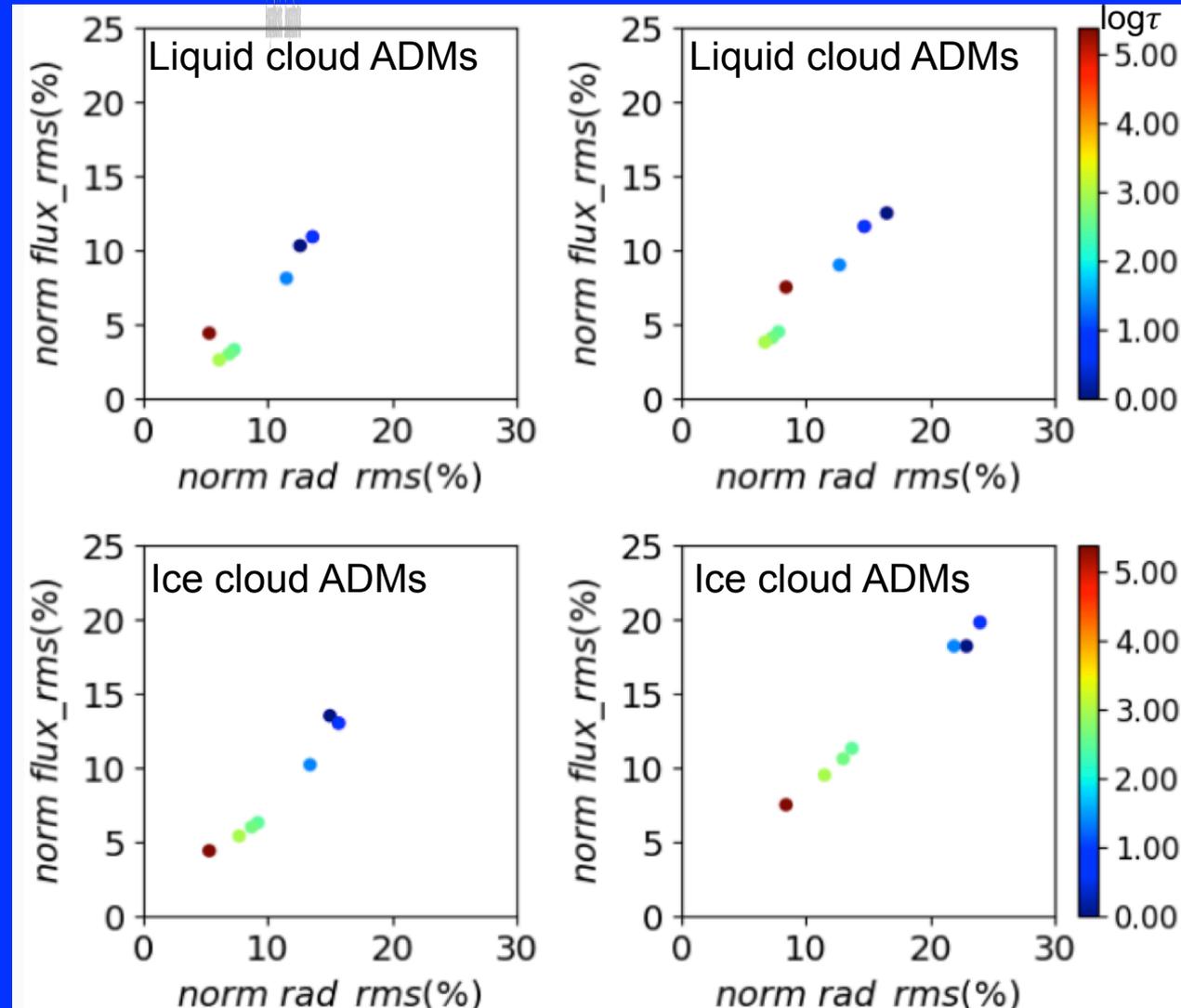
$$I = I_0 + \frac{a}{[1 + e^{-(x-x_0)/b}]^c}$$



# Normalized radiance RMS errors and relative flux RMS errors for simulated liquid clouds

SZA=41

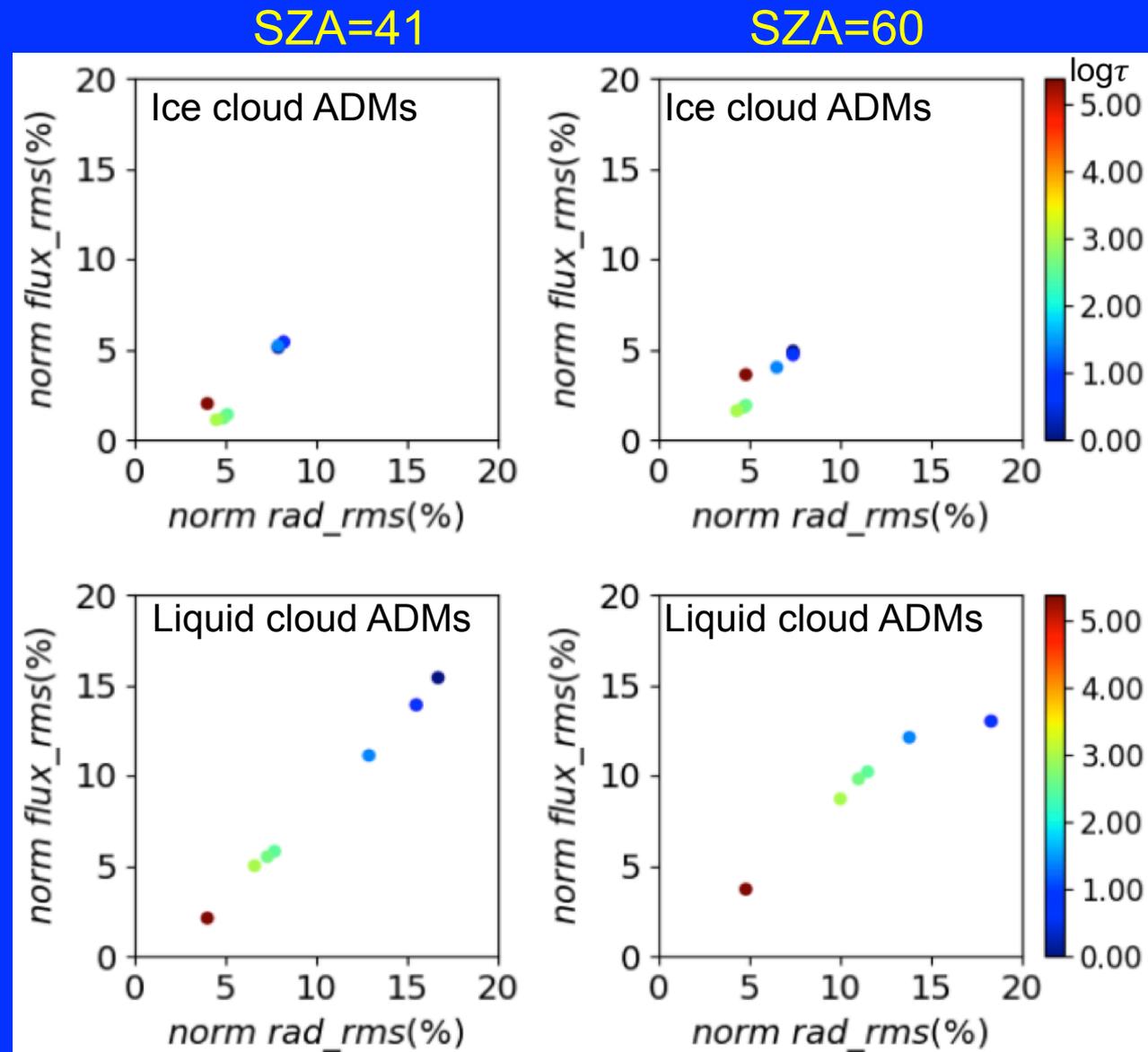
SZA=60



- There is a strong relationship between normalized radiance RMS error and the relative flux RMS error;
- Applying ice cloud ADMs for liquid clouds increases both the normalized radiance RMS error and the relative flux RMS error.

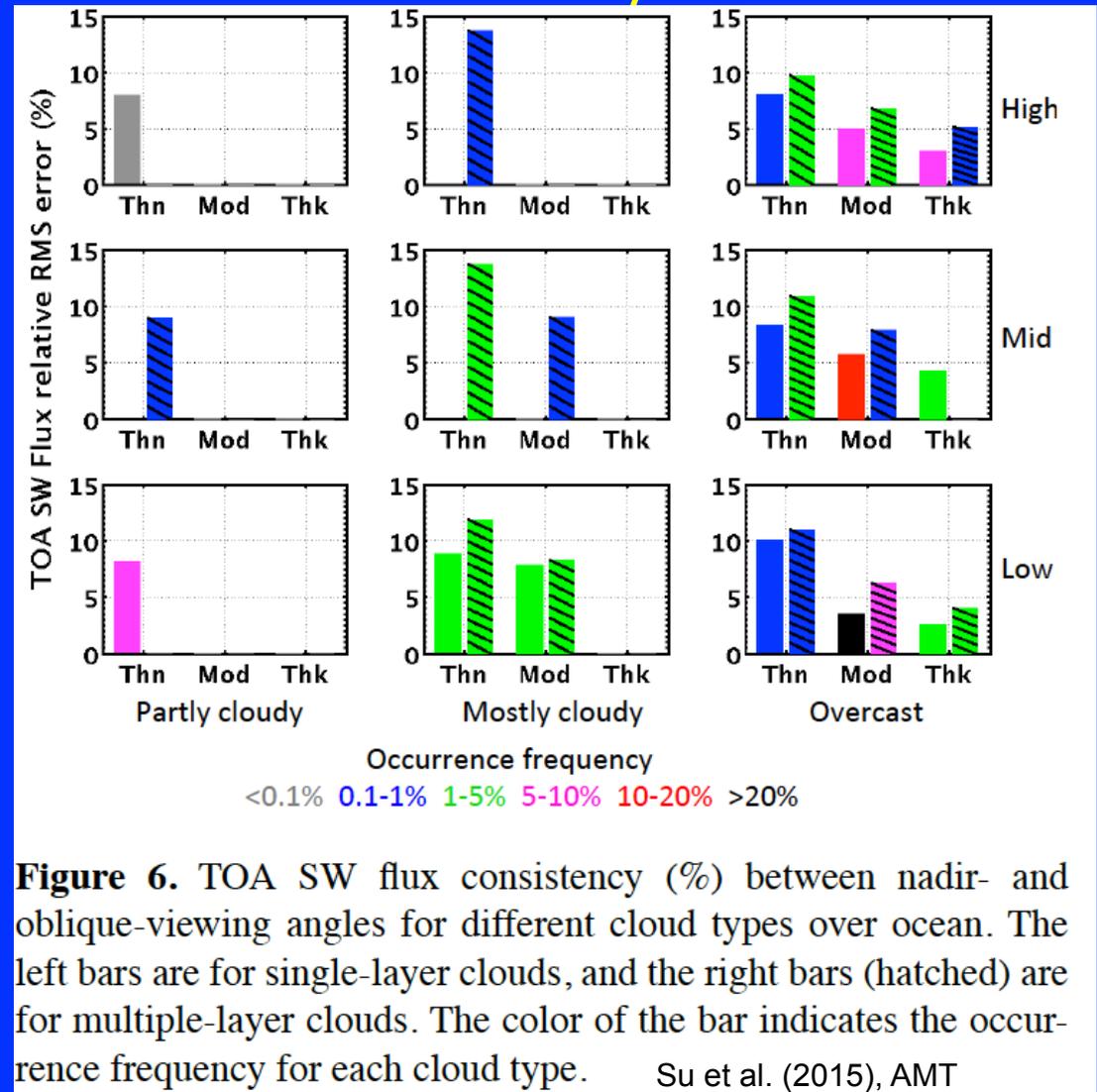
# Normalized radiance RMS errors and relative flux RMS errors for simulated ice clouds

- There is a strong relationship between normalized radiance RMS error and the relative flux RMS error;
- Applying liquid cloud ADMs for ice clouds increases both the normalized radiance RMS error and the relative flux RMS error.



# Do we need to consider other variables in cloudy ocean ADMs?

- Current ADMs consider cloud optical depth, cloud fraction, and cloud phase;
- Can we further improve the ADMs by accounting for cloud inhomogeneity using standard deviation of cloud optical depth;
- For each solar zenith angular bin, determine the terciles of standard deviation of cloud optical depth, and develop ADMs for each of them.



**Figure 6.** TOA SW flux consistency (%) between nadir- and oblique-viewing angles for different cloud types over ocean. The left bars are for single-layer clouds, and the right bars (hatched) are for multiple-layer clouds. The color of the bar indicates the occurrence frequency for each cloud type. Su et al. (2015), AMT

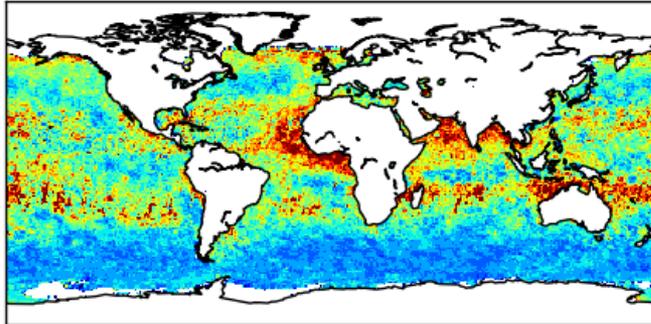
PCL: CF = 0.1-40%	High: EP < 440 hPa	Thin: $\tau < 3.35$
MCL: CF = 40-99%	Mid: EP = 440-680 hPa	Mod: $\tau = 3.35 - 22.63$
OVC: CF = 99-100%	Low: EP > 680 hPa	Thick: $\tau > 22.63$

# Normalized radiance RMS error for January 2002

1<sup>st</sup> tercile

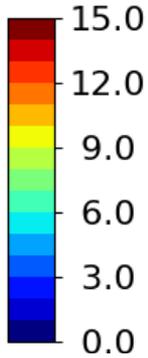
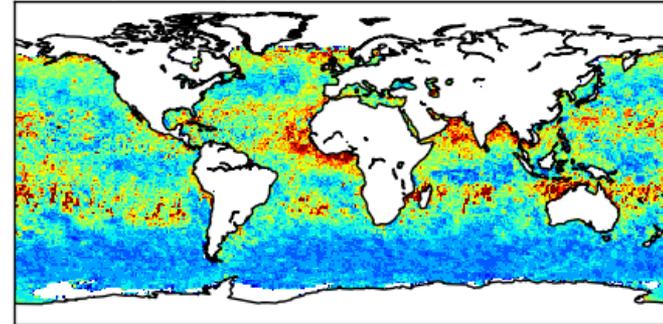
Ed4ADMs

RMS= 7.16%



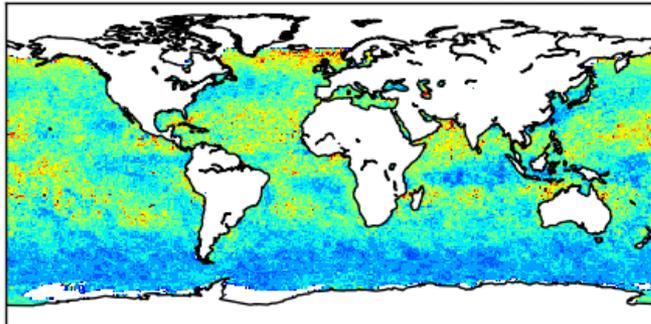
STD(COD) ADMs

RMS= 6.76%

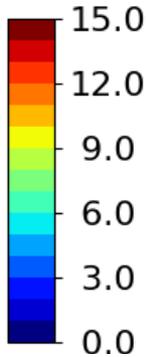
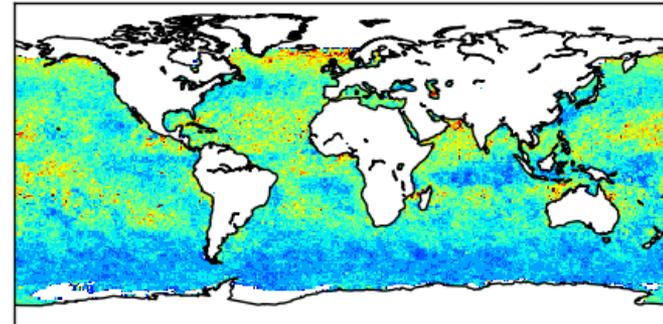


2<sup>nd</sup> tercile

RMS= 6.28%

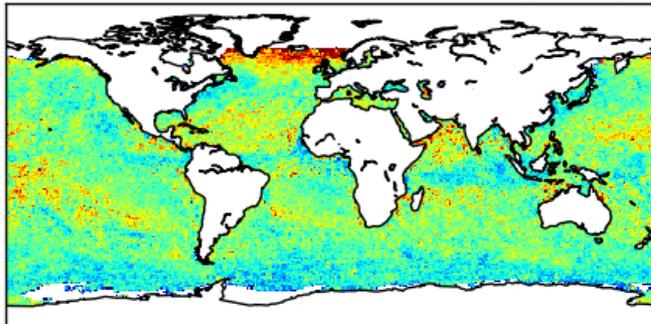


RMS= 6.20%

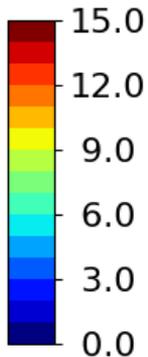
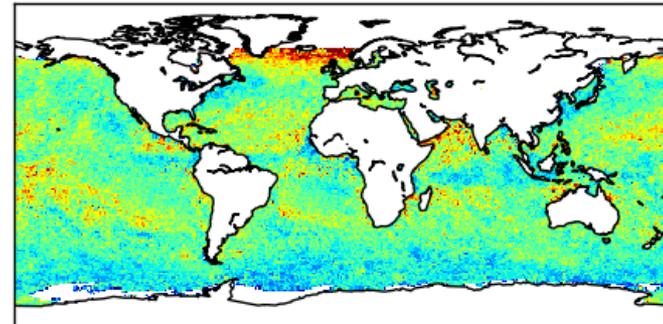


3<sup>rd</sup> tercile

RMS= 7.34%

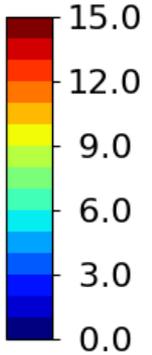
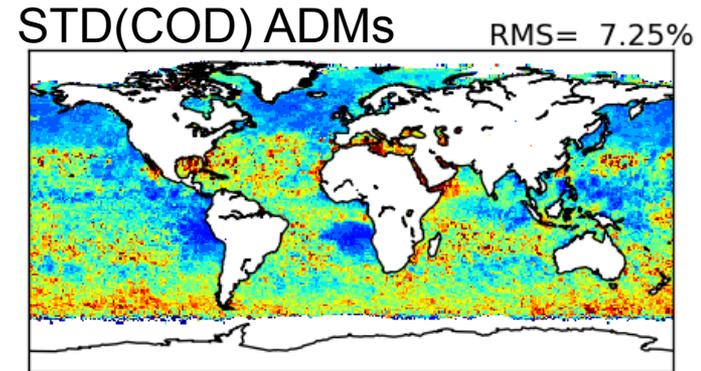
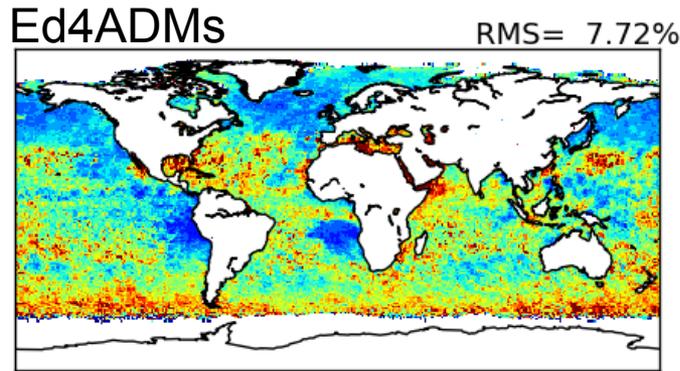


RMS= 7.15%

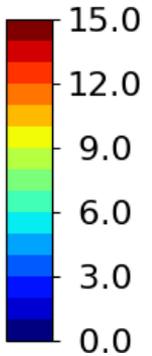
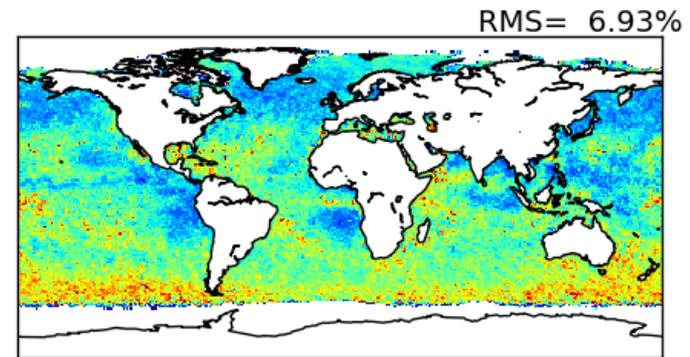
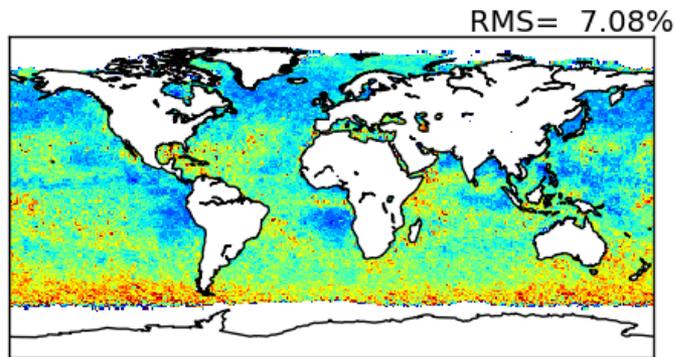


# Normalized radiance RMS error for July 2002

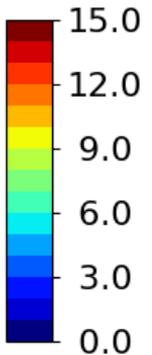
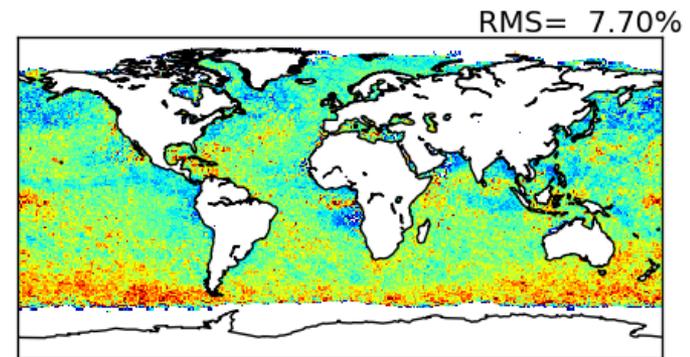
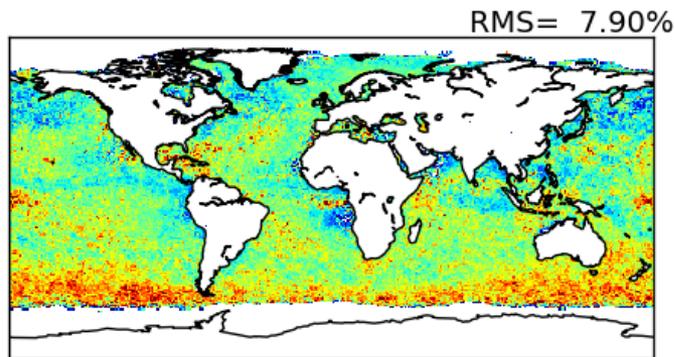
1<sup>st</sup> tercile



2<sup>nd</sup> tercile



3<sup>rd</sup> tercile



## Cloudy-sky LW/WN angular distribution models

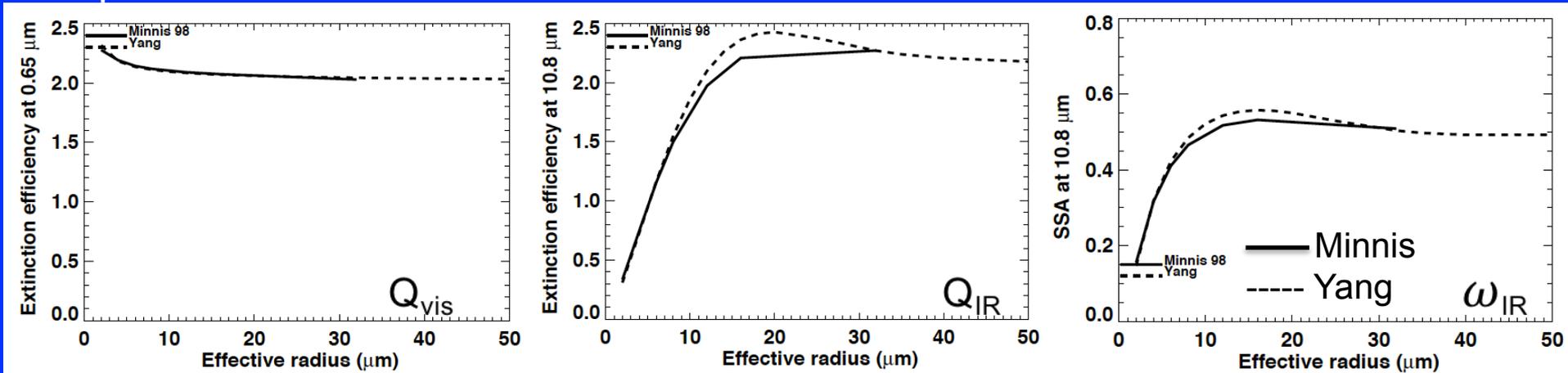
- Cloudy-sky LW ADMs are constructed based upon the relationship between radiance and 'pseudoradiance' ( $\Psi$ ):

$$\Psi(\omega, T_s, T_c, f, \epsilon_s, \epsilon_c) = (1 - f)\epsilon_s B(T_s) + \sum_{j=1}^2 \left[ \epsilon_s B(T_s)(1 - \epsilon_{c_j}) + \epsilon_{c_j} B(T_{c_j}) \right] f_j$$

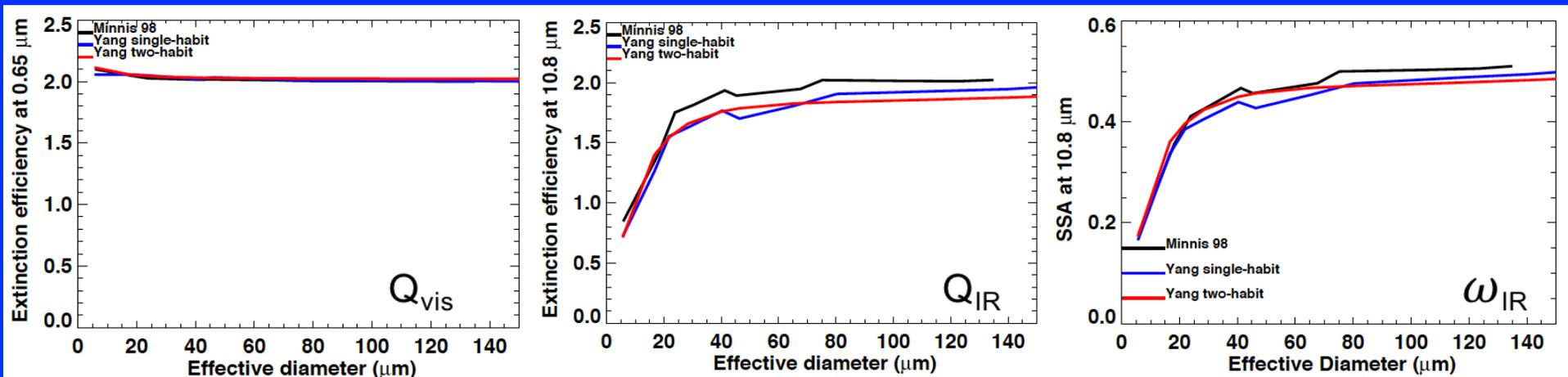
- Cloud infrared emissivity is :  $\epsilon_c = 1 - e^{-\tau_a}$
- The IR absorption optical depth is:  $\tau_a = (1 - \omega_{IR})\tau_{IR}$
- The infrared optical depth:  $\tau_{IR} = \tau_{vis} \frac{Q_{IR}}{Q_{vis}}$
- The visible and infrared extinction efficiency and the infrared scattering albedo are all based upon the values provided in Minnis et al. (1998) for various liquid and ice clouds.

# Comparison of cloud optics

## Liquid cloud



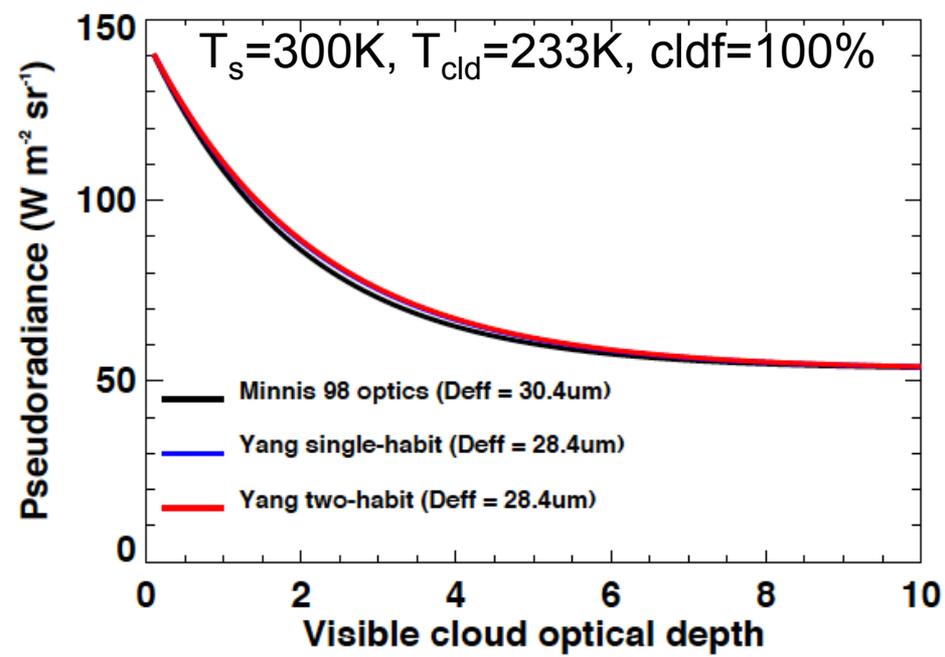
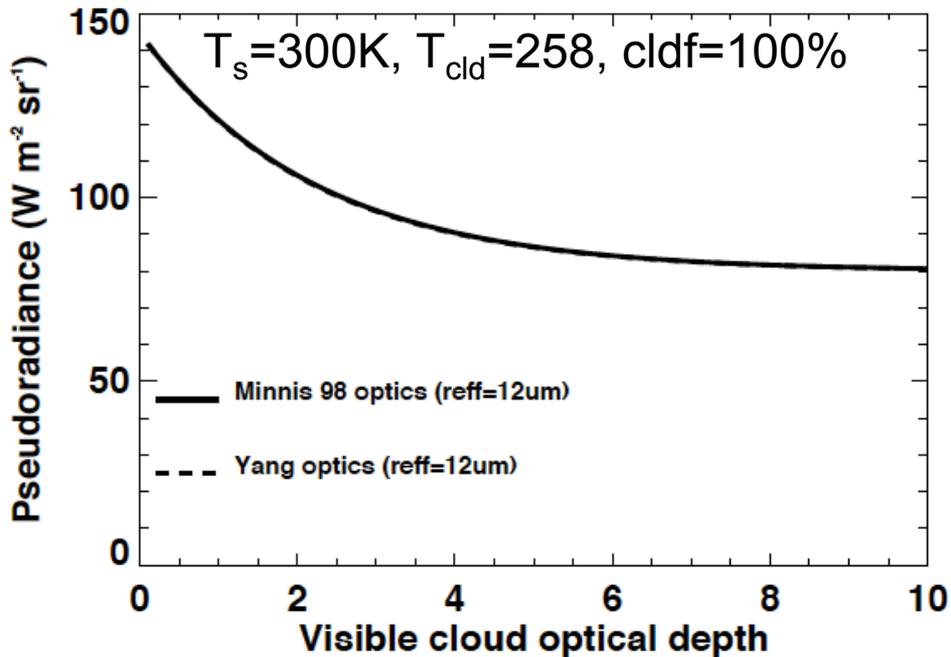
## Ice cloud



# Impact on pseudoradiance calculation

Liquid cloud with  $r_{\text{eff}}=12 \mu\text{m}$

Ice cloud with  $D_{\text{eff}}=30 \mu\text{m}$



# Quantify Suomi-NPP flux error caused by using Aqua ADMs

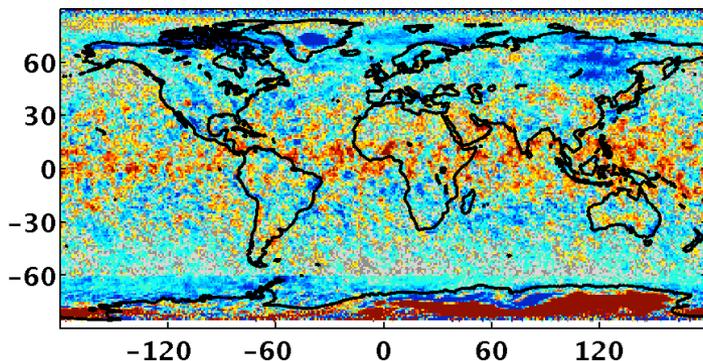
- Aqua ADMs are used to invert fluxes for CERES observations on NPP
- Footprint size for S-NPP is larger than that for Aqua.
- Cloud properties retrieved from VIIRS can also be different from those retrieved from MODIS.

Daytime

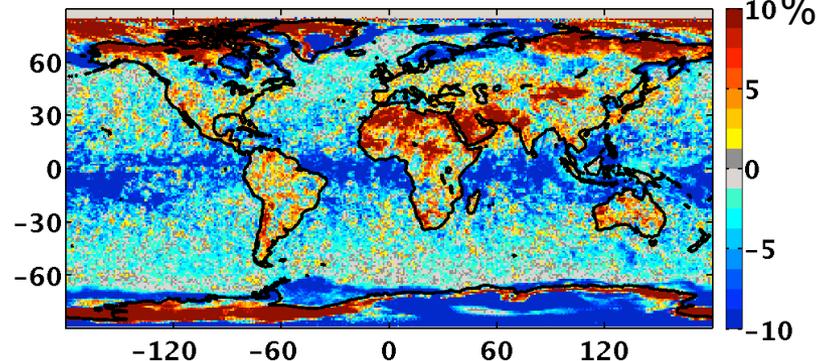
Nighttime

$\Delta f$

201304:NPP-Aqua Cloud fraction mean  $\Delta f = -0.4\%$

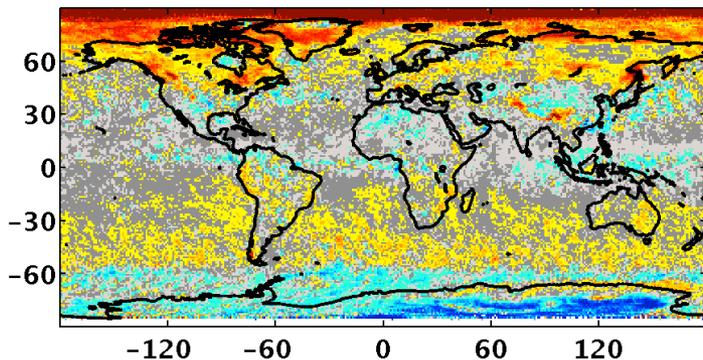


201304:NPP-Aqua Cloud fraction mean  $\Delta f = -2.0\%$

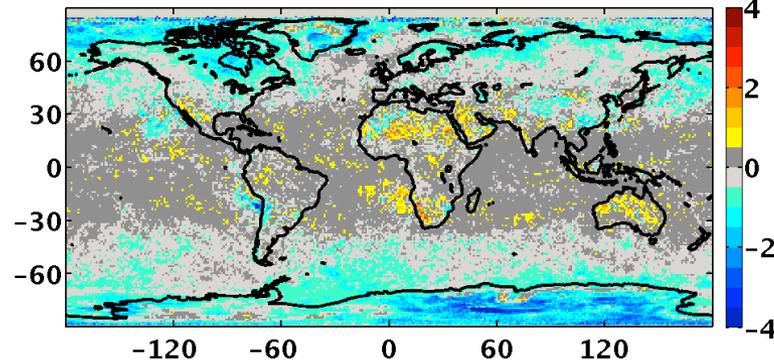


$\Delta \tau$

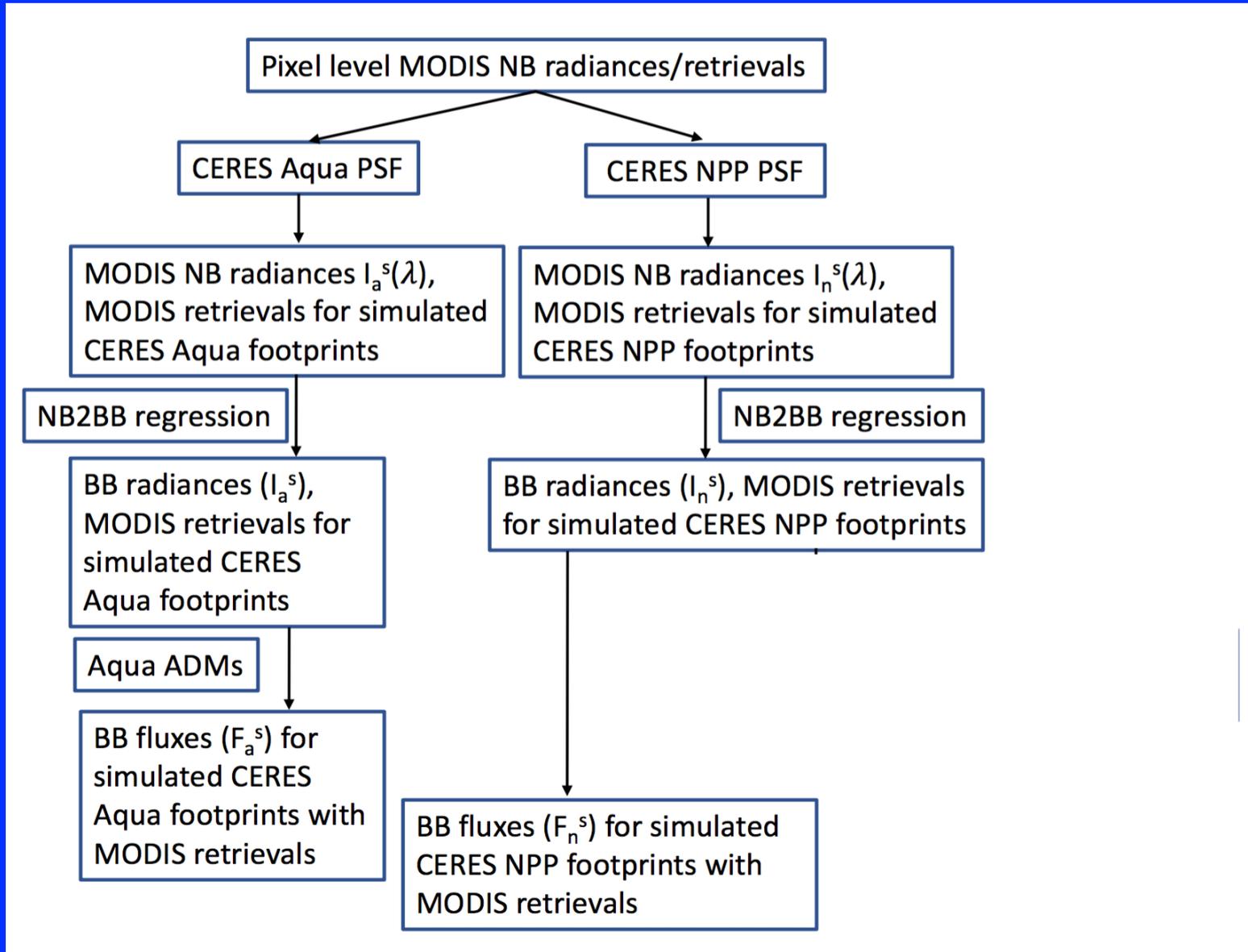
201304:NPP-Aqua Cloud optical depth mean  $\Delta \tau = 0.2$



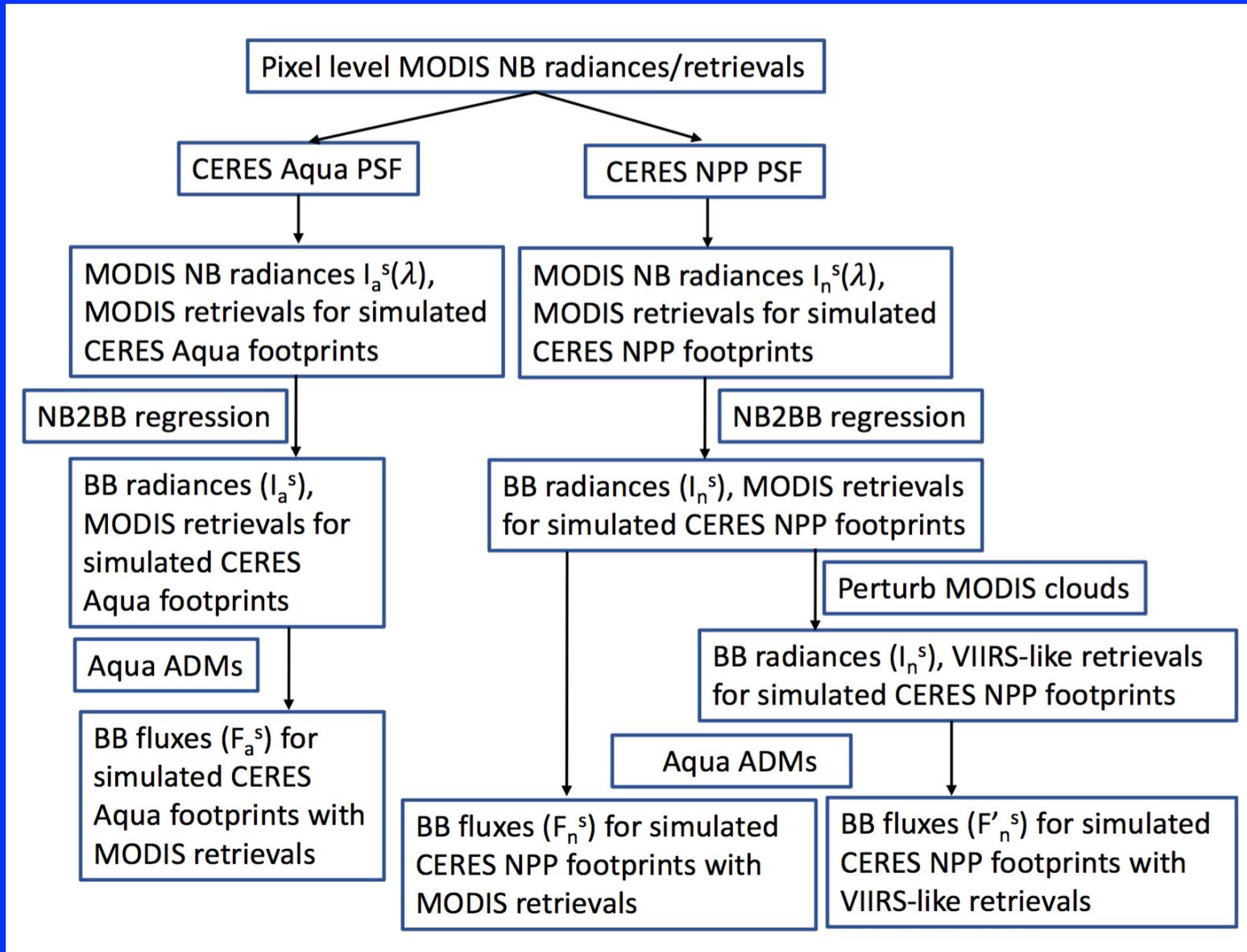
201304:NPP-Aqua Cloud optical depth mean  $\Delta \tau = -0.1$



# Simulate Aqua and NPP footprints to quantify flux error due to different footprint size and cloud property



# Simulate Aqua and NPP footprints to quantify flux error due to different footprint size and cloud property



## Flux errors due to footprint size and cloud property differences

- Footprint size difference between CERES instruments on Aqua and on Suomi-NPP leads to:
  - Underestimation of global monthly mean instantaneous SW flux by  $0.4 \text{ Wm}^{-2}$  and the RMS error is  $2.4 \text{ Wm}^{-2}$ .
  - A close to zero bias in global monthly mean LW flux and the RMS errors are  $0.8 \text{ Wm}^{-2}$  and  $0.2 \text{ Wm}^{-2}$  for daytime and nighttime.
  - Regionally, the differences are less than  $4.0$  and  $1.0 \text{ Wm}^{-2}$  for SW and LW.
- Footprint size and cloud property difference between CERES instruments on Aqua and on Suomi-NPP leads to:
  - Overestimation of global monthly mean SW flux by  $1.1 \text{ Wm}^{-2}$  and the RMS error is increased to  $5.2 \text{ Wm}^{-2}$ .
  - LW RMS errors increase slightly to  $0.9 \text{ Wm}^{-2}$  and  $0.5 \text{ Wm}^{-2}$  for daytime and nighttime.
  - Regionally, SW flux error up to  $20.0 \text{ Wm}^{-2}$  and LW error up to  $2.0 \text{ Wm}^{-2}$  are observed over polar regions.

# Plan to further evaluate the effect of scene identification on flux inversion

