# Cloud Optical and Microphysical Properties Product some "collection 5" efforts

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## Topics

- L2 collection 5 efforts & examples
  - Quantitative pixel-level uncertainty
  - Multi-layer/phase cloud detection
  - Sun glint, heavy aerosol detection
- L3 research effort

# Future Processing Effort (collection 5)

- Collection: refers to a processing/reprocessing stream
  - Terra atmosphere algorithm deliveries in fall '04, Aqua in Dec '04
  - Aqua forward processing (L1B, L2) to begin in Jan '05
- MOD06 cloud retrieval algorithm expected improvements/additions:
- Pixel-level uncertainty analysis
- Multi-layer/phase cloud detection (non-opaque cirrus over water clouds)
- Sun glint, heavy aerosol detection
  - Improved spectral surface albedo maps
  - Improved ice cloud libraries, atmospheric transmittance libraries
  - Implementation of 1.6-2.1 µm band combination retrieval for snow/ice surfaces and heavy aerosol over clouds
  - Code improvements, etc.

## Pixel-level Retrieval Uncertainty Analysis S. Platnick, B. Wind

Currently incorporating the effect of the following sources on inferred cloud-top reflectance:

- 1. Instrument calibration
- 2. Atmospheric correction uncertainty
- 3. Spectral surface albedo uncertainty

#### <u>Note</u>:

Uses sensitivity derivatives calculated from reflectance libraries,
 e.g.:

$$\frac{\partial \tau}{\partial R_1}\Big|_{R_2} = f\left(\frac{\partial R_1}{\partial \tau}\Big|_{r_e}, \frac{\partial R_1}{\partial r_e}\Big|_{\tau}, \frac{\partial R_2}{\partial \tau}\Big|_{r_e}, \frac{\partial R_2}{\partial r_e}\Big|_{\tau}\right)$$

- A likely minimum uncertainty, i.e., other missing components: ( ice cloud models, vertical cloud structure including multi-layer clouds, ...)
- Random L2 uncertainties may be reduced/eliminated during L3 aggregations

## Retrieval Example Terra granule, coastal Chile/Peru, 18 July 2001, 1530 UTC

[Platnick et al., IEEE Trans. Geosci. Remote Sens., 41]



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001)



Phase (white = ice)

**Cloud Optical Thickness** 

Cloud Optical Thickness Uncertainty

## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) $\tau$ : water clouds over ocean



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) $\tau$ : water clouds over land



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) *r*<sub>e</sub>: water clouds over ocean



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) *r*<sub>e</sub>: water clouds over land



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) $\tau$ : ice clouds over ocean



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) $\tau$ : ice clouds over land



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) $\tau$ : ice clouds over ocean



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) *IWP*: ice clouds over ocean



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) *IWP*: ice clouds over land



## Pixel-level Uncertainty Analysis - Terra MODIS orbit (20 Nov 2002)



## Pixel-level Uncertainty Analysis - Terra MODIS orbit (20 Nov 2002)



## Approximate/Qualitative Solution Space vs. Method ~ 3dB (100% relative error)



Example Validation Efforts

## Cirrus Validation - SGP ARM site

Mace, Zhang, Platnick, King, Minnis, Yang (J. Appl. Meteor., accepted)

MODIS Terra, 6 March 2001, 1735 UTC



## Cirrus Validation - SGP ARM site, cont.



## Cirrus Validation - SGP ARM site, cont.

Case study 6 March 2001



## Cirrus Validation - SGP ARM site, cont.

15 overpasses, single layer cirrus (MOD06 vs. Z-Radiance algorithm case study)



## Cloud multilevel/phase detection G. Wind, S. Platnick

- Utilizes differences between:
  - 1. Inferred above-cloud PW between CO<sub>2</sub> slicing (+ NCEP moisture fields) and 0.94 μm solar reflectance retrieval

2. IR and SWIR phase retrieval

## Cloud multilevel/phase detection G. Wind, S. Platnick

- Utilizes differences between:
  - Inferred above-cloud PW between CO2 slicing (+ NCEP moisture fields) and 0.94 µm solar reflectance retrieval ⇒identify ice retrieval contaminated by water cloud
  - 2. IR and SWIR phase retrieval ➡water cloud retrieval contaminated by ice cloud

### multilayer/phase detection

MAS, CRYSTAL-FACE 7-26-2002, track 5









Multi-layer map

## multilevel/phase detection MODIS Terra, Antarctic Ocean 11-20-2002



RGB composite (2.1, 1.6, 0.55  $\mu$ m)







## Sun Glint & Heavy Aerosol Detection J. Riédi, G. Wind, et al.

- Problem
  - Difficulty in discriminating heavy aerosol (e.g., dust outbreak) and sun glint from cloud in current version
  - Dust aerosol —> water cloud; Pollution aerosol —> ice cloud
- Approach
  - Combination of spatial variance tests and possibly spectral dependence tests (TBD)

## sunglint detection

MAS, CRYSTAL-FACE 7-26-2002, track 3



Clear sky

Water cloud

Ice cloud

Sunglint, no cloud

Cloud edge detected

R(1.61) G(0.66) B(1.87)

Sunglint/phase map

Gicklime™ and a TIFF (12W) elecompressor are needed to see this picture.

## Sun Glint & Heavy Aerosol Detection Example Terra, 8 May 2001, 1200 UTC, Saharan Dust



## Sun Glint & Heavy Aerosol Detection Example Terra, 8 May 2001, 1200 UTC, Saharan Dust



Global Analysis of MODIS Level-3 Cloud Properties and their Sensitivity to Aggregation Strategies (data & analysis grant)

#### Investigators

**PI:** Steve Platnick

**Co-l's:** Steve Ackerman (U. Wisconsin), Robert Pincus (NOAA-CIRES), Michael King, Bryan Baum (LaRC, U. Wisconsin CIMSS)

**Collaborators:** Lazaros Oreopoulos (JCET, UMBC), Jean-Jacques Morcrette (ECMWF)

#### Consequences of pixel-level errors?

 MODIS L3 aggregations provide statistics relevant to large-scale GCM domains. Therefore ...

Overarching science question:

To what extent do systematic pixel-level retrieval errors bias spatial/temporal aggregations?

An approach:

Since difficult to determine error as well as separate into random and bias components, what is the aggregation sensitivity to parameters expected to influence retrieval error (solar/viewing geometry w/segregation by cloud type, phase, surface,  $\tau_c$ ,  $r_e$ , ...)?

#### **Research Approach**

- Investigate global L3 distribution and correlations of various cloud products. Initial emphasis on hemispheric, land/ocean, tropical/midlatitude convective, marine stratocumulus regimes.
- Design/create research-level aggregation code (i.e., exist outside of production facility) w/capability of answering science questions.
- Analyze aggregation sensitivities by excluding various parts of geometry/retrieval space.
- Explore use of theoretical retrieval sensitivity calculations in weighting L2 data.
- Make a variety of L3 daily and monthly data sets available for use by researchers interested in MODIS cloud aggregations, including ECMWF (non-angular grid, reduced volume), UMBC (L. Oreopoulos), GMAO.

# Extra Slides

## MODIS Solar Reflectance Retrieval MOD06 - Cloud Optical & Microphysical Properties

Pixel-level cloud product for daytime observations at 1 km

- Cloud optical thickness ( $\tau$ ), effective particle radius ( $r_{\rm e}$ ), water path, thermodynamic phase
- liquid water and ice clouds, global retrievals (land, water, snow/ice)
- Algorithm overview
  - -Use single water non-absorbing band (0.65, 0.86, 1.2  $\mu$ m) w/three absorbing bands (1.6, 2.1, 3.7  $\mu$ m) => 1  $\tau$ , 3  $r_e$  (2.1  $\mu$ m derived  $r_e$  is primary).
  - -Short-wavelength band choice: 0.65  $\mu m$  (land), 0.86  $\mu m$  (ocean), 1.2  $\mu m$  (snow/ice)
  - -Surface spectral albedo from MODIS ecosystem and albedo products
  - -Retrieval gives homogeneous-equivalent cloud properties

## Solar Reflectance Method retrieval space example - ice cloud over ocean surface



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001)



Phase (white=ice)

Effective radius (µm)

Effective radius Uncertainty (µm)

## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) *r*<sub>e</sub>: water clouds over ocean



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) *r*<sub>e</sub>: ice clouds over ocean



## Pixel-level Uncertainty Analysis Peru granule (18 July 2001) *r*<sub>e</sub>: ice clouds over land



## Pixel-level Uncertainty Analysis Cyclone granule (20 Nov. 2002) *IWP*: ice clouds



## Cloud optical/microphysical properties from reflectance measurements - **Spherical Particles**

In general:

$$R_{\lambda} = R(\tau_{\lambda}, \varpi_{0,\lambda}, g_{\lambda})$$

For Mie scattering (spheres, **water droplets**), 3 optical variables can be reduced to 1 optical & 1 microphysical:

$$R_{\lambda} \approx R(\tau_{\lambda_0}, r_e)$$

$$r_{e} \equiv \frac{\int_{0}^{\infty} r^{3}n(r)dr}{\int_{0}^{\infty} r^{2}n(r)dr} = \frac{3}{4} \frac{\langle V \rangle}{\langle A_{cs} \rangle}$$

## Cloud optical/microphysical properties from reflectance measurements - **Spherical Particles**, cont.

 $r_{\rm e}$  is a radiative parameter, but with certain assumptions, it can be used with  $\tau$  to estimate column water mass/unit area (water path):

$$\tau \approx \Delta z \int_{0}^{\infty} 2A_{cs}(r) n(r) dr = 2 \langle A_{cs} \rangle N \Delta z$$
$$WP = \Delta z \int_{0}^{\infty} \rho_{W} V(r) n(r) dr = \rho_{W} \langle V \rangle N \Delta z$$
$$WP = \frac{2\rho_{W}}{3} \tau r_{e}$$

**Assumption**: vertically homogenous cloud layer, i.e.,  $N, r_e \neq f(z)$ 

## Cloud optical/microphysical properties from reflectance measurements - Crystal/Irregular Particles

In general:

$$R_{\lambda} = R(\tau_{\lambda}, \varpi_{0,\lambda}, g_{\lambda})$$

3 optical variables can perhaps(?) be reduced to 1 optical & 2 microphysical:

$$R_{\lambda} \approx R(\tau_{\lambda_0}, r_e, Habit mixture)$$

if 
$$r_e \equiv \frac{3}{4} \frac{\langle V \rangle}{\langle A_{cs} \rangle} \implies WP = \frac{2\rho_i}{3} \tau r_e$$

# MODIS operational (collection 4) ice crystal library habits/mixtures



## Example Pseudo-Empirical $r_e - D_{me}$ Relations B. Baum, A. Heymsfield, P. Yang



Note: the tail can wag the Dme

## Sensitivity of Scattering Parameters to Habits/Mixture B. Baum



## Terra geometry (Nov. 15, 2003)



## multilevel/phase detection, example MODIS Terra, Antarctic Ocean 11-20-2002

RGB composite (2.1, 1.6, 0.55 µm)



Cloud top pressure (CO2 slicing)



## multilevel/phase detection, example MODIS Terra, Antarctic Ocean 11-20-2002

RGB composite (2.1, 1.6, 0.55 µm)



0.94 µm above-cloud PW



multilayer/phase detection MAS, CRYSTAL-FACE 7-26-2002, track 5



R(1.61) G(0.66) B(1.87)

sunglint detection MAS, CRYSTAL-FACE 7-26-2002, track 3



## Sun Glint & Heavy Aerosol Detection Example Terra, 10 April 2001, 1200 UTC, Asian Dust & Pollution



## Sun Glint & Heavy Aerosol Detection Example Terra, 10 April 2001, 1200 UTC, Asian Dust & Pollution



MODIS Atmosphere Level-3 Aggregation Summary

- 1° grid spatial; daily, 8-day, monthly temporal all atmosphere products
- Statistics (mean, sdev, min, max, QA-weighting)
- Histograms (pdf's): 1-D and 2-D

2-D cloud parameter combinations (collection 4):

parameter	$ au_{ m c}$	<i>r</i> e	$T_{\rm c}$	$\mathcal{E}_{c}$
$ au_{ m c}$		Х	Х	Х
r <sub>e</sub>			Х	Х
p <sub>c</sub>				Х

 L3 code designed to aggregate L2 data sets only (monthly file contains ~ 800 statistical data sets). For maintenance (sanity) reasons, code not capable of mathematical or logical manipulation of L2 data!

#### Science Questions

- To what extent do aggregations show significant differences and/or correlations by hemisphere, land/ocean, regionally (e.g., tropical convection vs. midlatitude ice clouds; marine stratocumulus regimes)?
- Are aggregations sensitive to the geometry/retrieval space (due to 3-D geometry, pixel-level retrieval sensitivity, etc.)? How do aggregations change by elimination of certain parts of the space (e.g, exclude view angles regions, backscatter azimuth, etc.)? To what extent can changes be equated with bias "error"?
- Can pixel-based retrieval sensitivity/error calculations (include geometry and retrieval solution dependence) be used to weight L2 retrievals to reduce bias error?
- Are other girds or statistics more useful for forecast/climate model evaluation and diagnosis?