

Multi-angle/multi-spectral/multi-pixel remote sensing of 3D convectively-driven clouds with MISR and MODIS

A Progress Report for ROSES/TASNPP

Anthony B. Davis (PI), Linda Forster, Marcin J. Kurowski, and David J. Diner (JPL);

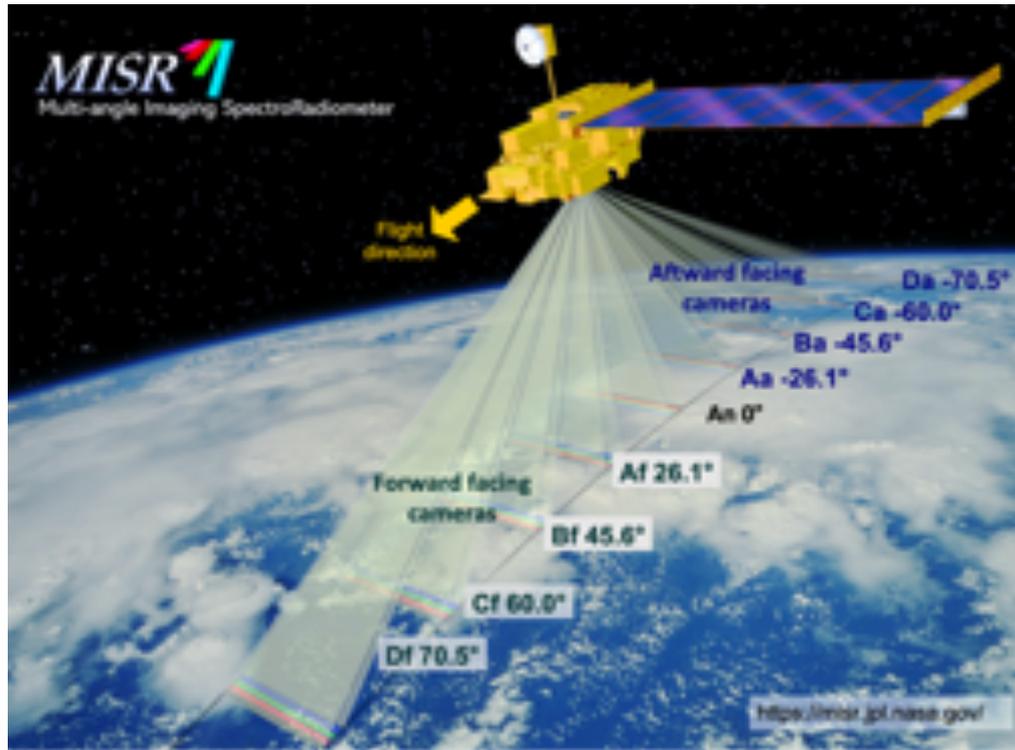
Aviad Levis (Caltech); Jesse Loveridge (UI-UC); Yoav Y. Schechner (Technion);

Bernhard Mayer (LMU)



Jet Propulsion Laboratory
California Institute of Technology

Multi-angle Imaging Spectro-Radiometer (MISR) with MODIS, on Terra



L2 aerosol products:

R.A. Kahn

(now at NASA-GSFC),

D.J. Diner,

J.V. Martonchik,

O.V. Kalashnikova,

M.J. Garay,

M.L. Witek,

et al.

L2 cloud products:

R. Davies

(now at U of Auckland),

D.J. Diner,

V.M. Jovanovic,

C.M. Moroney,

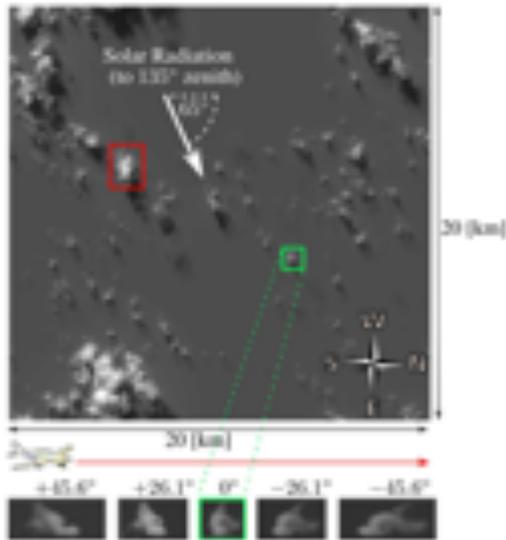
M.J. Garay,

K.J. Mueller,

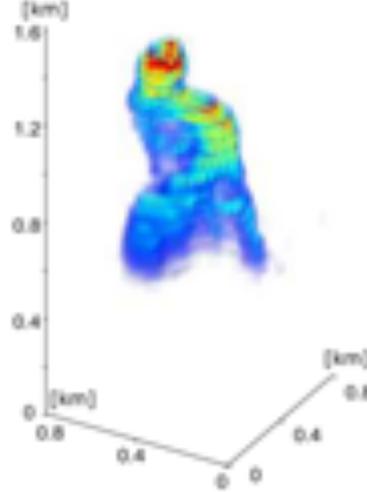
et al.

- push-broom acquisition, ~400 km swath
- 4 spectral channels, all VNIR
- 9 views, 275 m pixels (always in red-channel used here)
- ≈7 minutes from most forward to most aft-word

3D cloud tomography: Demonstrated

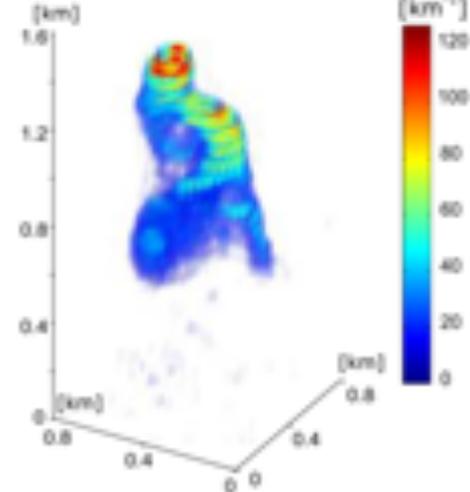


Ground-truth extinction



Ground truth

Recovered extinction

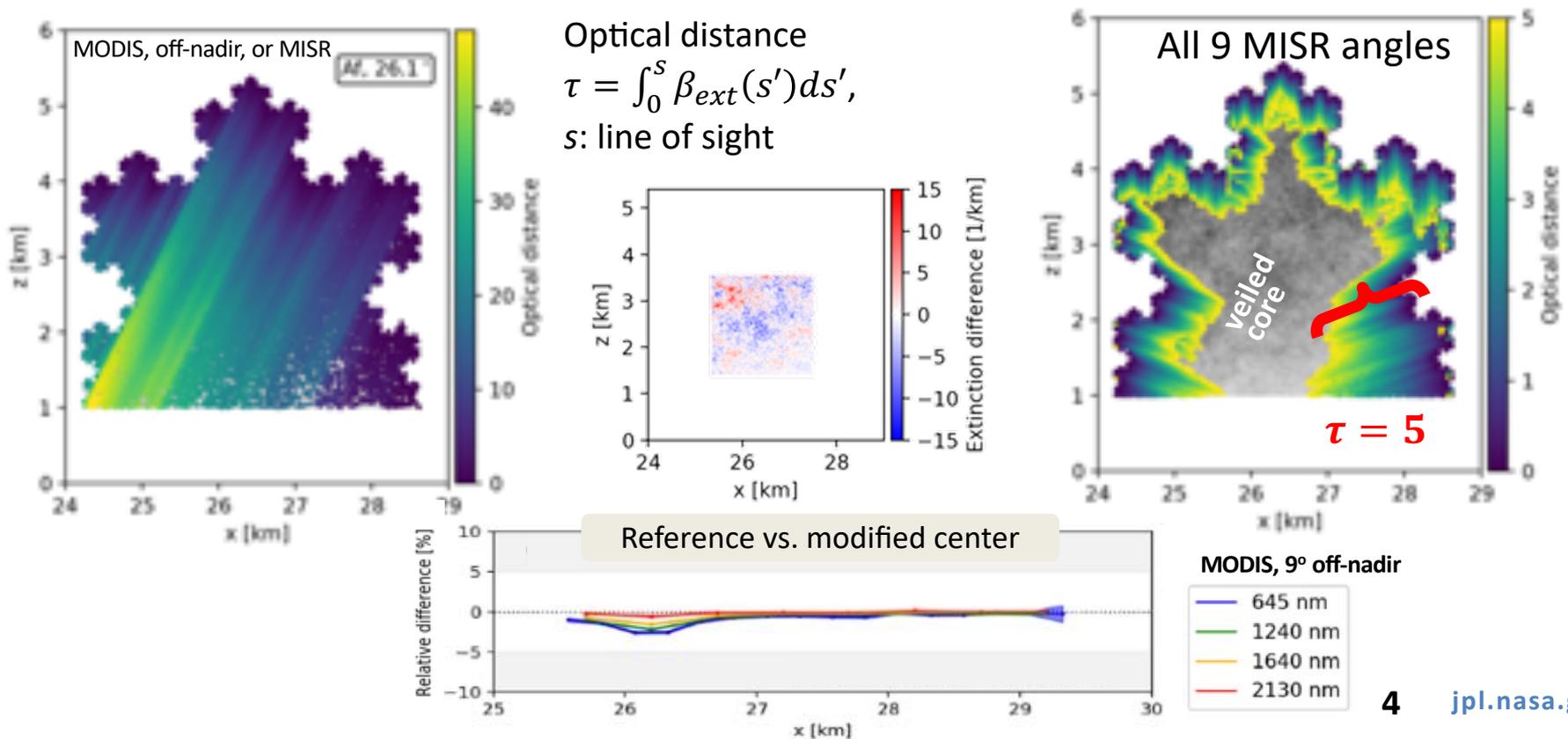


Reconstruction

- Levis et al. (2015): red channel, 9 views, 20-m (\approx AirMSPI) resolution
 - 46,656 unknowns & 315,018 unknowns, 2-step iteration scheme (1st linearized) using SHDOM
- Levis et al. (2017): VNIR multi-spectral
 - basic (profile-only) microphysics (r_e, v_e) **w/o SWIR (à la MODIS) nor polarization (à la MSPI)**
- Levis et al. (2020): VNIR multi-spectral/multi-polarimetric
 - potential for a full 3D microphysics (N_e, r_e, v_e) retrieval using polarization [I, Q, U]

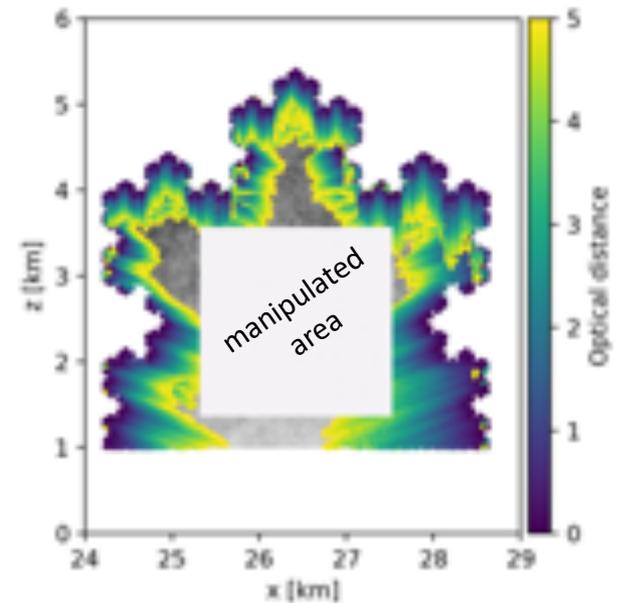
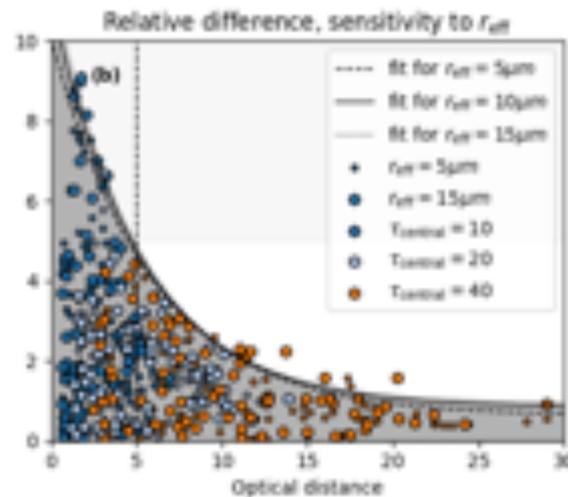
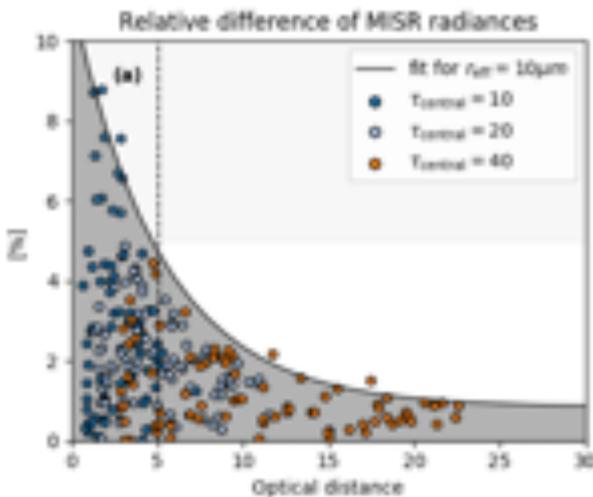
The “veiled core” of opaque clouds

- **Problem:** ≈ 20 m pixels from airborne sensors
 - ≈ 250 & 500 m pixels from space (MISR + MODIS)
 - SHDOM issues: opaque, internally variable pixels



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L. Forster, A. B. Davis, B. Mayer, and D. J. Diner (2020), Toward Convective Cloud Tomography from Space using MISR and MODIS: Locating the “Veiled Core” in Opaque Convective Clouds, *J. Atmos. Sci.* (in press). <http://arxiv.org/abs/1910.00077>

Cloud image formation in VNIR+SWIR: A story of two diffusions

Diffusion process #1:

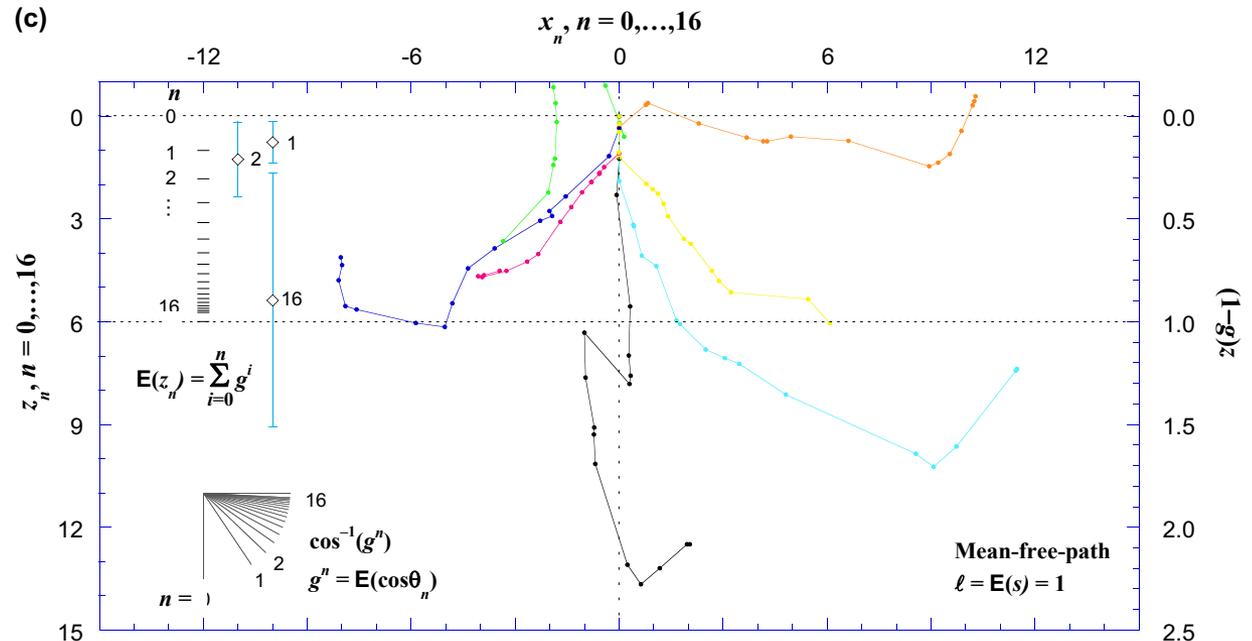
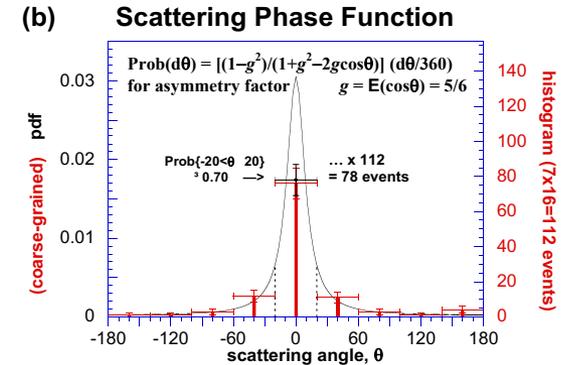
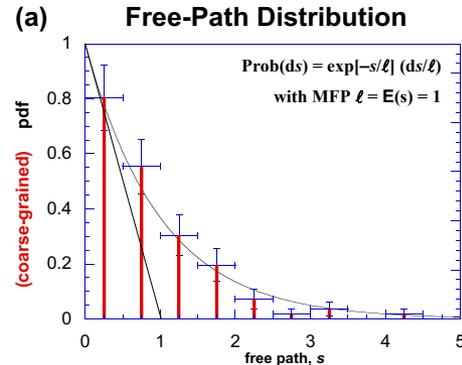
- random walk is on the sphere (*direction space*)
- in the *outer* shell
- gradual loss of *directional* memory

Characteristic time:

$$N^* = 1/\ln(1/g) \approx (1/g - 1)^{-1}$$

$$N^* \approx 6 \text{ for } g \approx 0.85$$

... explains empirical (≈ 5)
threshold for definition of
veiled core in optical distance



Cloud image formation in VNIR+SWIR: A story of two diffusions

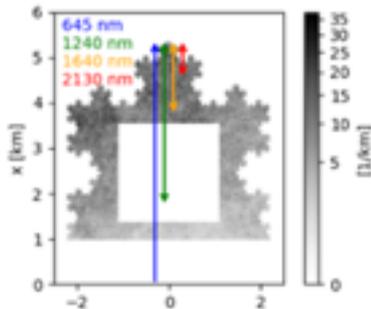
Diffusion process #2:

- random walk is in 3D *physical* space
- in the *veiled* core
- gradual loss of *positional* memory

Characteristic scale:

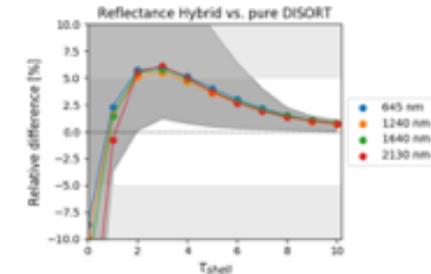
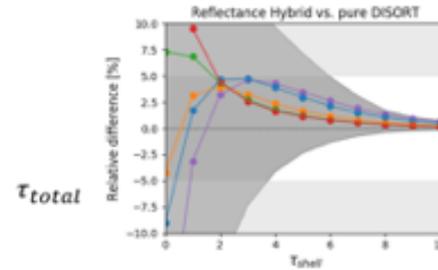
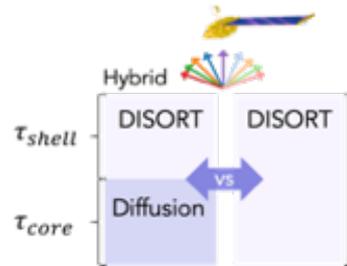
diffusion scale L_d

$$L_d \times (\text{mean extinction}) = [3(1-\omega)(1-\omega_g)]^{-1/2}$$



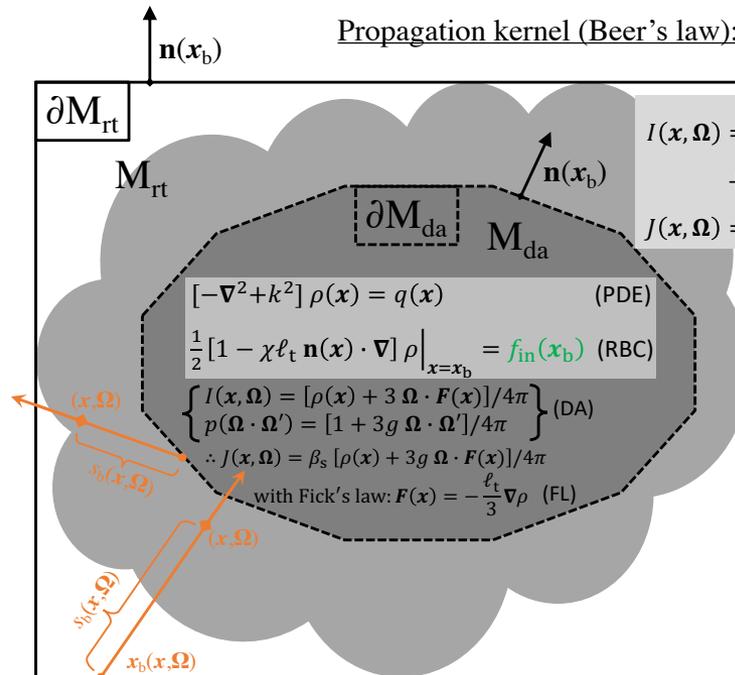
→ veiled core grows with absorption in MODIS-SWIR

Hybrid RT: Implementation in 1D



Hybrid RT: Possible Implementation in 3D

Propagation kernel (Beer's law): $T(\mathbf{x}, \boldsymbol{\Omega}; s) = \exp[-\int_0^s \beta(\mathbf{x} - \boldsymbol{\Omega}s') ds']$ (BL)



$$I(\mathbf{x}, \boldsymbol{\Omega}) = \int_0^{s_b(\mathbf{x}, \boldsymbol{\Omega})} [J(\mathbf{x}_s, \boldsymbol{\Omega}) + Q(\mathbf{x}_s, \boldsymbol{\Omega})] T(\mathbf{x}, \boldsymbol{\Omega}; s) ds + I_b(\mathbf{x}_b, \boldsymbol{\Omega}) T(\mathbf{x}, \boldsymbol{\Omega}; s_b(\mathbf{x}, \boldsymbol{\Omega})) \quad (\text{RTE})$$

$$J(\mathbf{x}, \boldsymbol{\Omega}) = \beta_s(\mathbf{x}) \int_{4\pi} p(\boldsymbol{\Omega} \cdot \boldsymbol{\Omega}') I(\mathbf{x}, \boldsymbol{\Omega}') d\boldsymbol{\Omega}' \quad (\text{SF})$$

Boundary condition at $\mathbf{x}_b(\mathbf{x}, \boldsymbol{\Omega}) \in \partial M_{rt}$
 $I_b(\mathbf{x}_b, \boldsymbol{\Omega}) = 0$, for $\mathbf{n}(\mathbf{x}_b) \cdot \boldsymbol{\Omega} < 0$ (IBC)

Two-way couplings at $\mathbf{x}_b(\mathbf{x}, \boldsymbol{\Omega}) \in \partial M_{da}$

RT → DA:
 $f_{in}(\mathbf{x}_b) = \int_{\mathbf{n}(\mathbf{x}_b) \cdot \boldsymbol{\Omega} < 0} |\mathbf{n}(\mathbf{x}_b) \cdot \boldsymbol{\Omega}| I(\mathbf{x}_b, \boldsymbol{\Omega}) d\boldsymbol{\Omega}$

DA → RT:

$$I_b(\mathbf{x}_b, \boldsymbol{\Omega}) = f_{out}(\mathbf{x}_b) \frac{\mathbf{n}(\mathbf{x}_b) \cdot \boldsymbol{\Omega}}{\pi}, \text{ if } \mathbf{n}(\mathbf{x}_b) \cdot \boldsymbol{\Omega} > 0,$$
 where $f_{out}(\mathbf{x}_b) = \frac{1}{2} [1 + \chi l_t \mathbf{n}(\mathbf{x}) \cdot \nabla] \rho \Big|_{x=x_b}$

Tomography initialization: Outer cloud shape ... *quickly!*

Progress in 3D Tomographic Cloud Reconstruction, Part 2*: Efficient determination of outer shapes of convective clouds from MISR data using radiosity



Anthony B. Davis, JPL / Caltech, Pasadena, Ca, USA
Guillaume Bal, University of Chicago, Chicago, Il, USA
Céline Cornet, Laboratoire d'Optique Atmosphérique, Lille, France

* Part 1 is Linda Forster's oral presentation on Wednesday PM.

Objective & Results

Tomographic reconstruction of 3D convective clouds is a large and challenging inverse problem that can benefit from a reasonably good initial guess. In this study, we use synthetic and real MISR data to demonstrate a robust and highly efficient determination of the effective outer shape of a 3D cloud, one slice at a time. Although time will tell, we believe this will constitute a viable initial guess for the full 3D cloud tomography.

In our methodology, the outer boundary of a 2D slice from a 3D cloud along any convenient direction (say N-S) is represented as an irregular n -sided off-grid polygon with the lowest facet (i.e., cloud base) being constrained to remain horizontal. Radiance emerging from each facet (except the base) is modeled with a radiosity-type angular distribution inspired by asymptotic radiative transfer theory; this function has at most two free parameters. In total, $2n-1+2(n-1) = 4n-3$ parameters need to be determined by fit to the MISR data. We found that $n = 8$ is a reasonable balance for a wide range of cloud sizes, in view of the coarse nature of the cloud boundary model. Thus, a nonlinear minimization problem is solved to estimate the 29 parameters per 2D slice, 15 of which will be used in the subsequent full 3D cloud tomographic reconstruction to construct an initial guess. This preliminary fitting problem also calls for an initial guess that we distill out of cloud masks for the nadir and most oblique views.

Two-parameter (a_i, t_i) radiosity model:

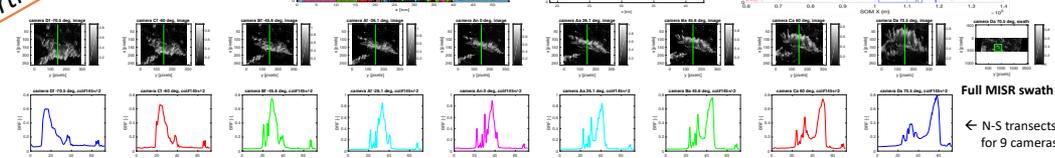
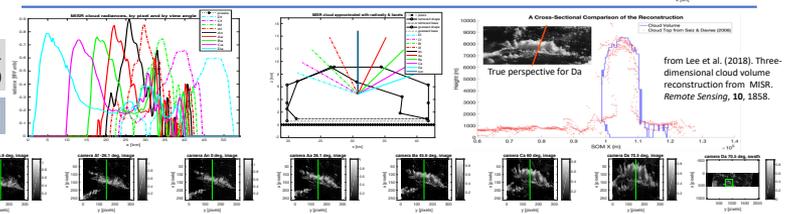
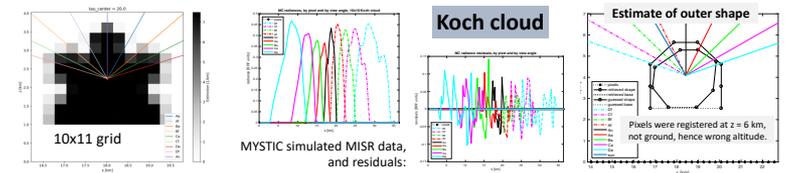
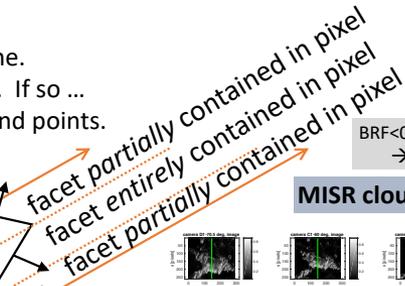
$$I(\Omega_0, \Omega_v, n_i; a_i, t_i) = a_i \times (\Omega_v \cdot n_i) \times \max(-\Omega_0 \cdot n_i, t_i)$$

where: Ω_0 is solar incidence direction;
 Ω_v is viewing direction;
 n_i is the outward normal of the i^{th} facet;
 $a_i > 0$ is an albedo-like parameter;
 $t_i > 0$ is a diffuse Lambertian background.

For a non-illuminated facet ($\Omega_0 \cdot n_i > 0$) or a grazing sun ($-\Omega_0 \cdot n_i < t_i$), a_i, t_i combines into a single free parameter.

Algorithm:

- Select one MISR view angle (Ω_v).
- Browse 7 up-looking facets, one-by-one.
- Is facet in view ($\Omega_v \cdot n > 0$)? If not, skip. If so ...
- Find what pixel or pixels contain the end points.
- Distribute radiance according to:
 - relative orientation to sun;
 - relative orientation to view;
 - projected facet area.



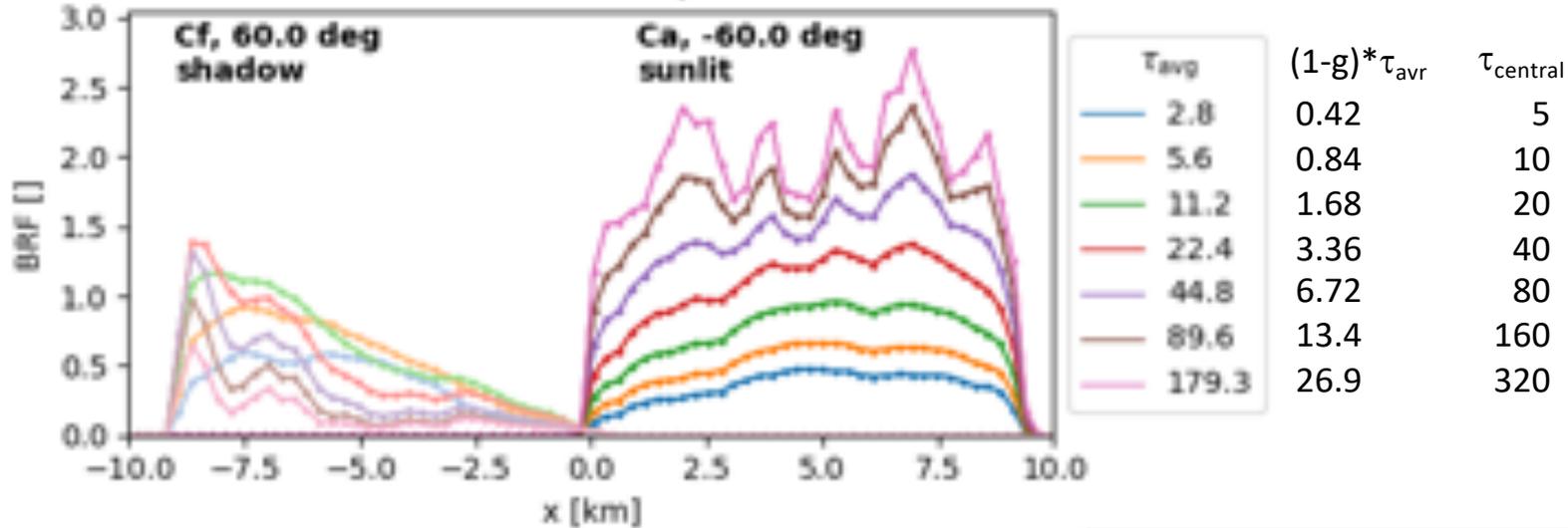
Seiz, G., and R. Davies (2006). Reconstruction of cloud geometry from multi-view satellite images. *Remote Sens. Environ.*, **100**, 143-149.

Bal, G., J. Chen, and A. B. Davis (2018). Reconstruction of cloud geometry from high-resolution multi-angle images. *Inv. Prob. Imaging*, **12**, 261-280.

a single MISR Ca ground pixel

Tomography initialization: Mean extinction ... *quickly!*

Reference Koch cloud, SZA = 60°



Exact diffusion theory:

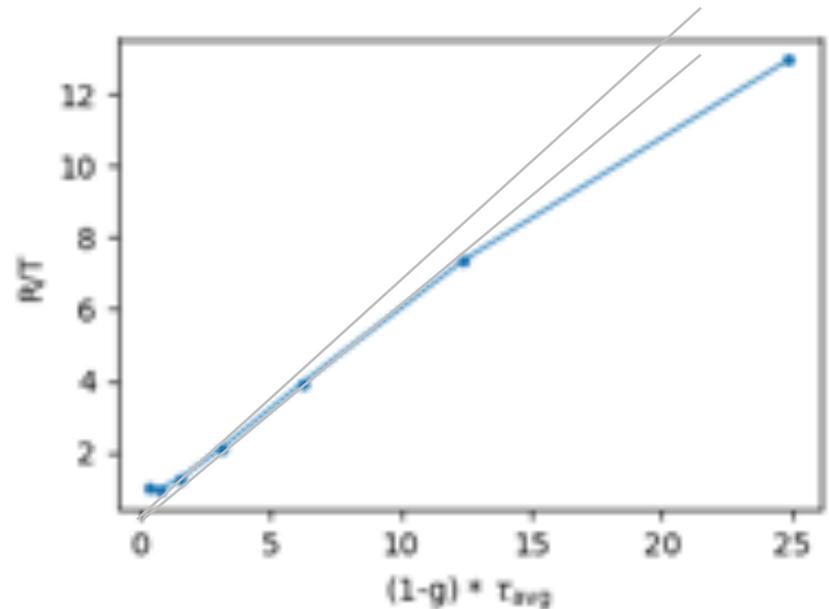
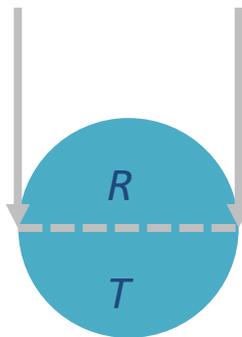
$$R/T = (1-g)\tau_{diam}/2\chi$$

1D: segments ($\chi = 1$)

2D: circles ($\chi = \pi/4$)

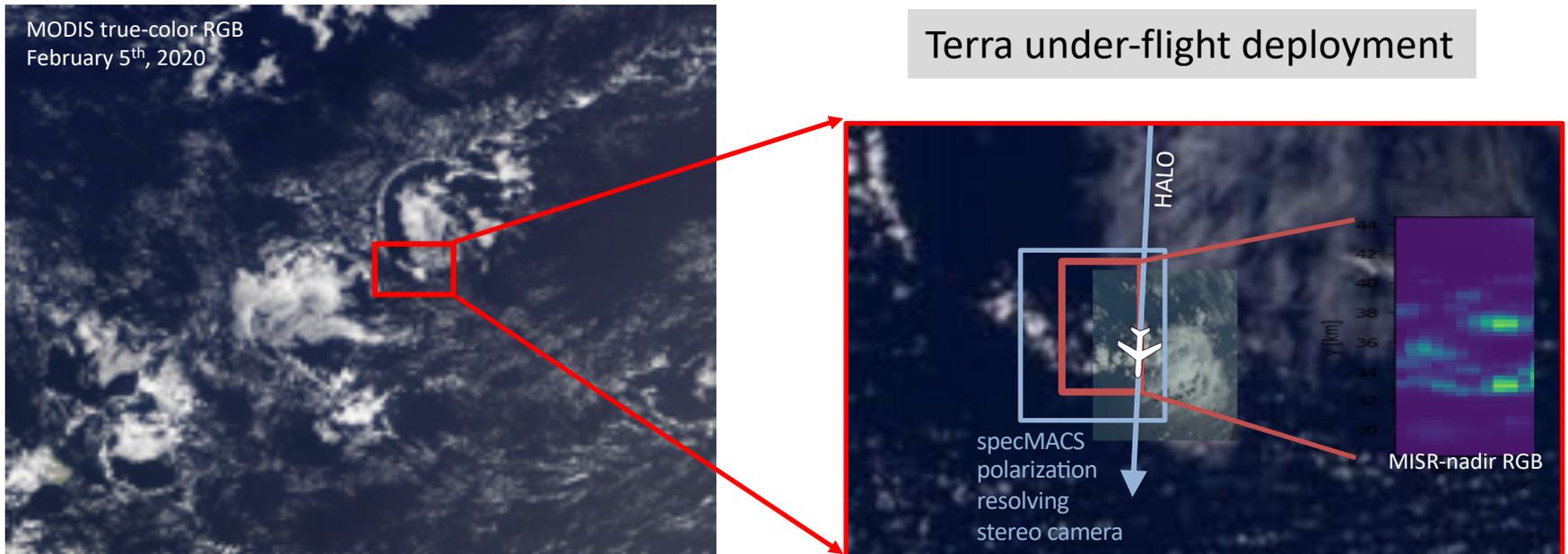
3D: slabs/cylinders/spheres ($\chi = 2/3$)

(χ is the "extrapolation" scale)



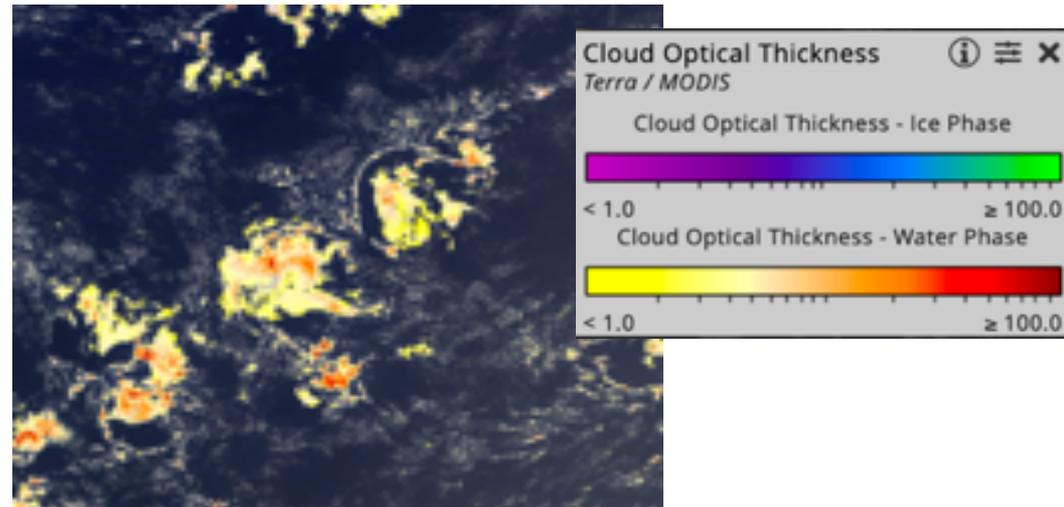
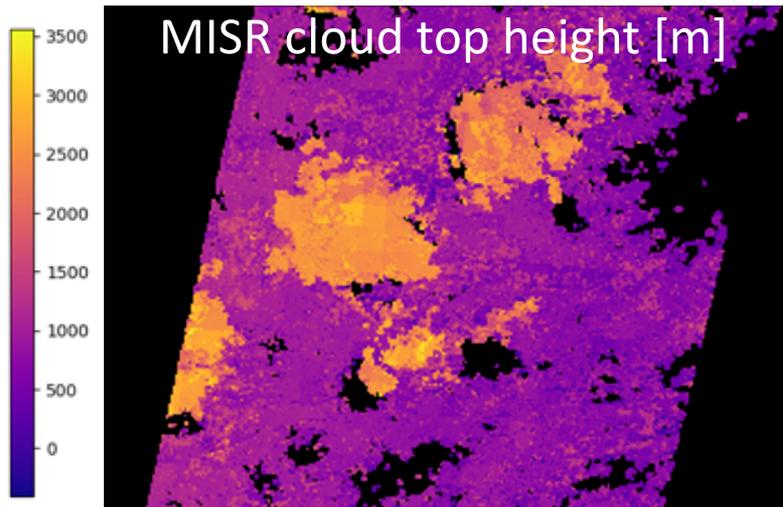
EUREC⁴A (Elucidating the Role of Clouds-Circulation Coupling in Climate) field campaign

Barbados, 1/20-2/20, 2020



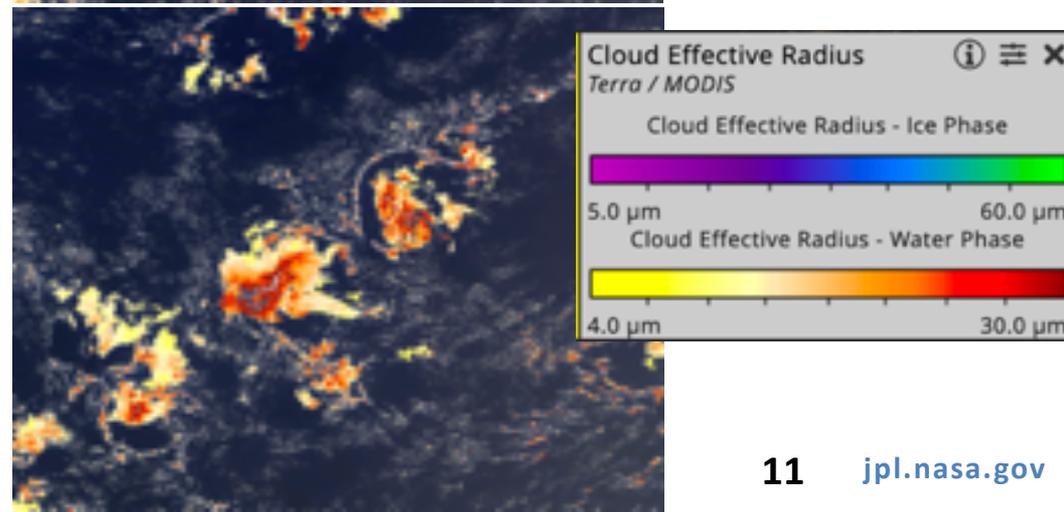
collaborator-participants at LMU: Linda Forster & Bernhard Mayer

EUREC⁴A (Elucidating the Role of Clouds-Circulation Coupling in ClimAte) field campaign

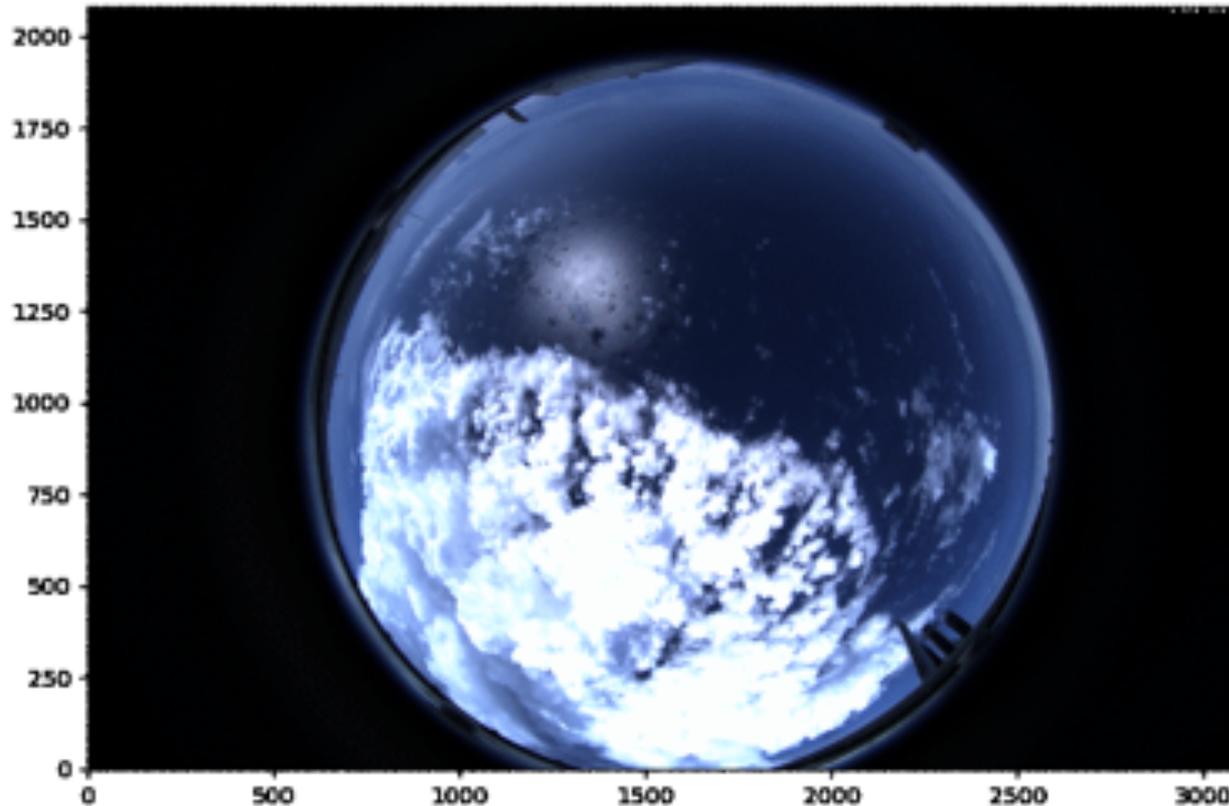


Apart from a hyperspectral imager, aircraft has lidar and radar.

→ Opportunity to compare with operational MISR & MODIS cloud products



Cloud, Aerosol and Monsoon Processes Philippines Experiment (CAMP²Ex)



collaborator-participants at UI-UC: Jesse Loveridge & Larry di Girolamo

From 3D-TRACE to CloudCT, and beyond

3D-TRACE: Three-Dimensional Tomographic Reconstruction of the Aerosol-Cloud Environment
... ESTO/AIST (~FY13), then ESD/RST (2013–2015) seed funds for JPL

CloudCT: Cloud Computed Tomography ... 14M€ ERC Horizon 2020 grant (~10 cubesat mission)

Wurzenberg: Zentrum für Telematik

Munich: Ludwig-Maximilians-Universität

University of Illinois – Urbana Champaign

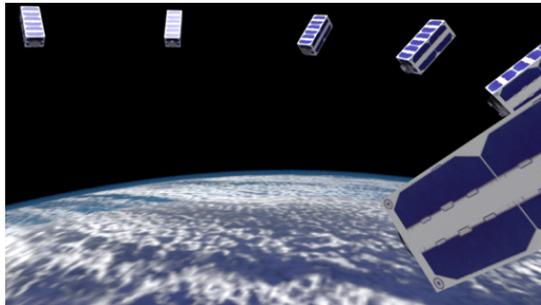
Support also acknowledged from:
JPL (RTD, CAP & JROC),
ROSES (TASNPP & FINNEST), BSF, EC's Marie Skłodowska-Curie Fellowship Program, and others ...

Pasadena:

- JPL
- Caltech

Haifa: Technion – IIT

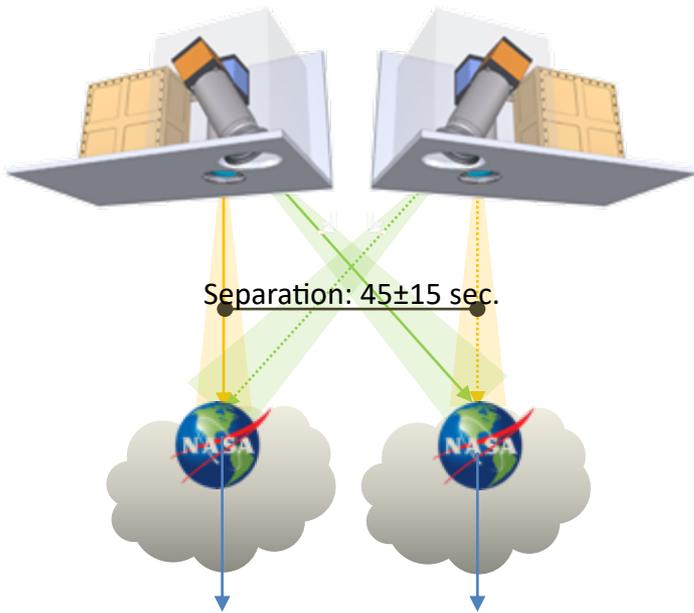
Rehovot: Weizmann Institute



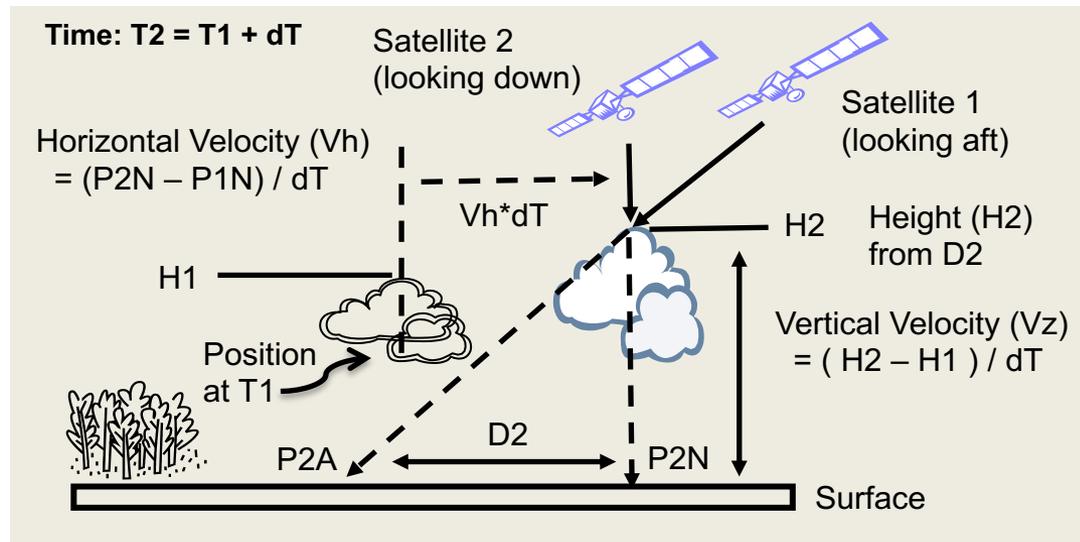
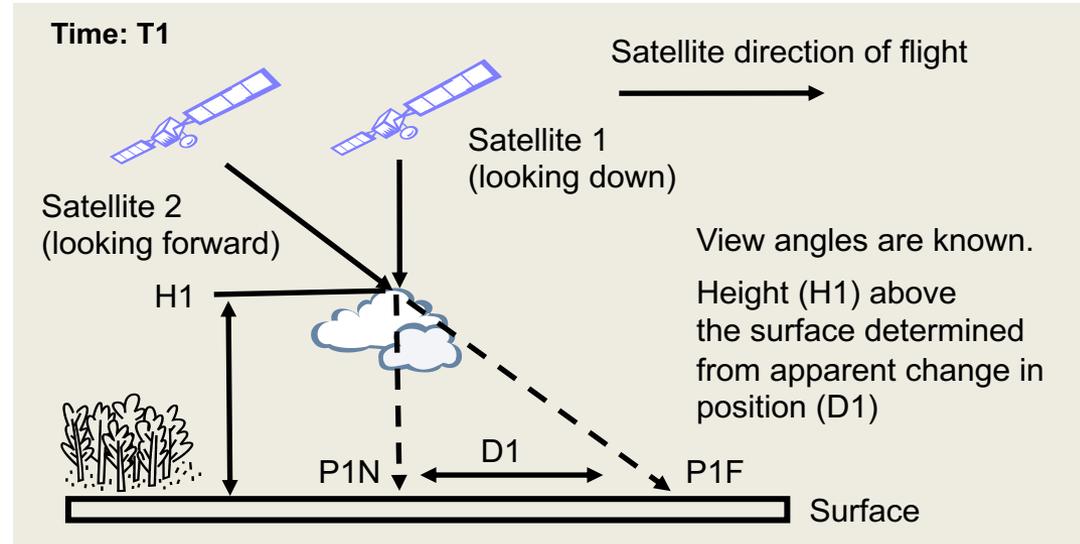
CloudCT consortium

jpl.nasa.gov

Support of A/CCP mission formulation: Tandem StereoCam “delta t” concept



Same cloud feature is observed from 2 view angles (stereo/3D image) twice (cloud motion after 45 sec.).



Summary & Outlook

- 2 papers published:
 - Levis et al. (2020)
 - Forster et al. (2020)
- 2 papers in preparation:
 - Davis et al. (two diffusion processes)
 - Davis et al. (robust tomography initialization)
 - other projects in progress, for later publications
- 2 field campaigns for validation data
 - EUREC⁴A
 - CAMP²EX