

MISR Science Team Meeting (STM) & Data Users Symposium (DUS)
Pasadena (Ca), 12-14 December, 2011



MISR as Trailblazer in Atmospheric (Cloud and/or Aerosol) Tomography

Anthony B. Davis

Michael J. Garay, David J. Diner,
John V. Martonchik, Paul von Allmen (JPL/Caltech)
and Guillaume Bal (Columbia University)

Topics / Outline

- **Back to the basics:**
 - Photons & their state space*
- **Physics-based remote sensing:**
 - Is “the end” in sight?*
- **Two wide-open frontiers!**
 - Time-domain with multiple scattering
 - Multi-pixel retrievals, e.g., tomography (by adapting biomedical imaging technologies)
- **Two examples of later using MISR**
 - Aerosol extinction field reconstruction
 - 3D Clouds (work in progress)

What is a photon?

- **Wikipedia:**

- In physics, the photon (from Greek "phos," meaning light) is the quantum of the time-dependent electromagnetic field [i.e., EM waves], for instance light.
- The term "photon" was coined by G. N. Lewis in 1926.

**Gilbert Newton Lewis,
10/23/1875 – 3/23/1946,
in his UC Berkley Lab.
*N.B. He died therein.***



Photon Attributes / State Space

Quantum EM theory

↔

Classical EM

Remote sensing

Energy

$$E = \hbar\omega = h\nu$$

↔

wavelength

$$\lambda = c/\nu$$

position along spectral axis

Photon Attributes / State Space

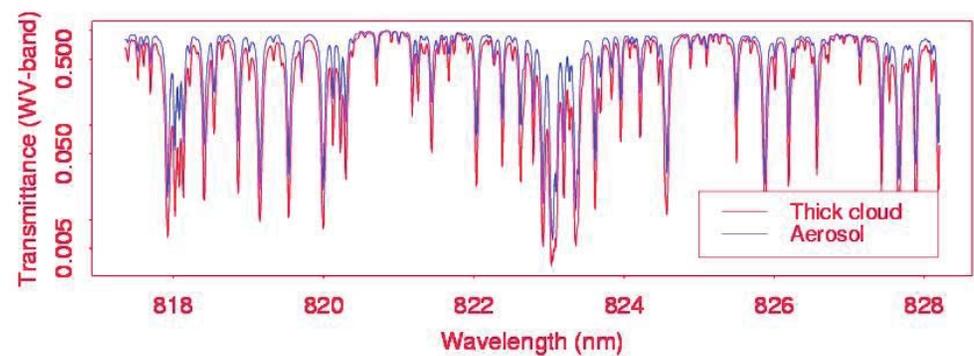
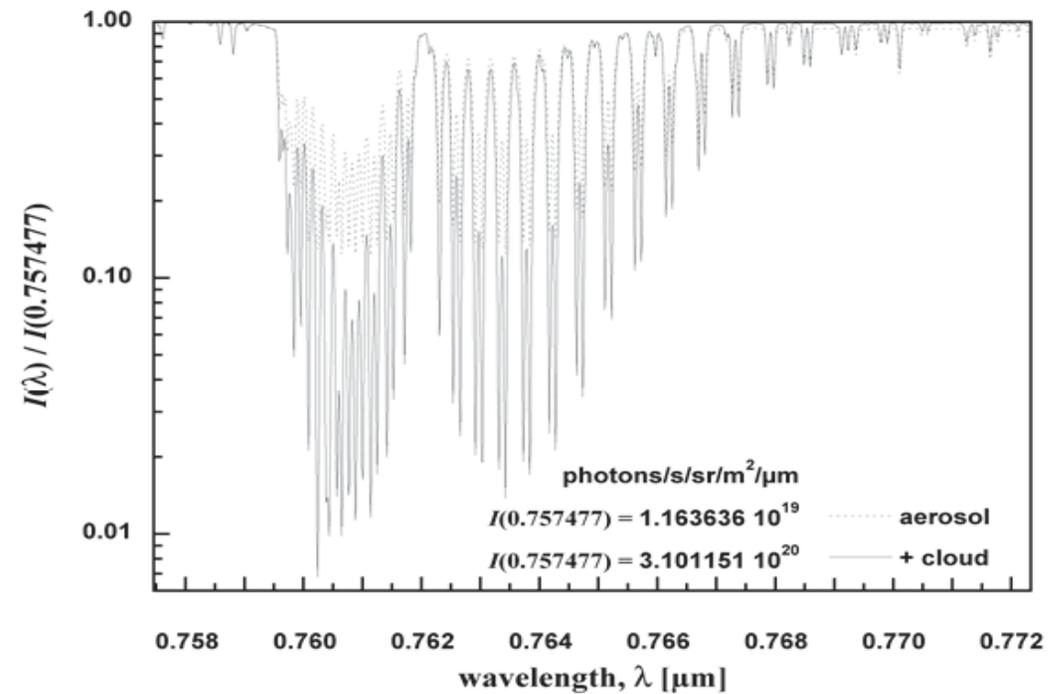
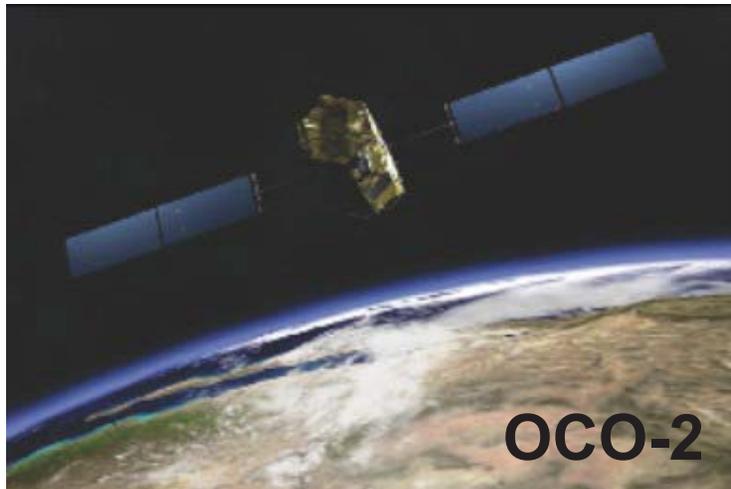
Quantum EM theory		↔	Classical EM	Remote sensing
Energy	$E = \hbar\omega = h\nu$	↔	wavelength $\lambda = c/\nu$	position along <i>spectral axis</i>
Momentum (collectively, pressure)	$\mathbf{p} = \hbar\mathbf{k} = \Omega E/c$	↔	direction $\Omega = k/k$	<i>escape direction at scene (or pixel position at detector)</i>

Photon Attributes / State Space

Quantum EM theory		↔	Classical EM	Remote sensing
Energy	$E = \hbar\omega = h\nu$	↔	wavelength $\lambda = c/\nu$	position along <i>spectral axis</i>
Momentum (collectively, pressure)	$\mathbf{p} = \hbar\mathbf{k} = \mathbf{\Omega} E/c$	↔	direction $\mathbf{\Omega} = \mathbf{k}/k$	<i>escape direction at scene (or pixel position at detector)</i>
Spin (angular momentum)	$S = \pm\hbar$	↔	polarization	calls for further filtering

That's it!

Hyper-spectral: OCO, AIRS, TES, etc.

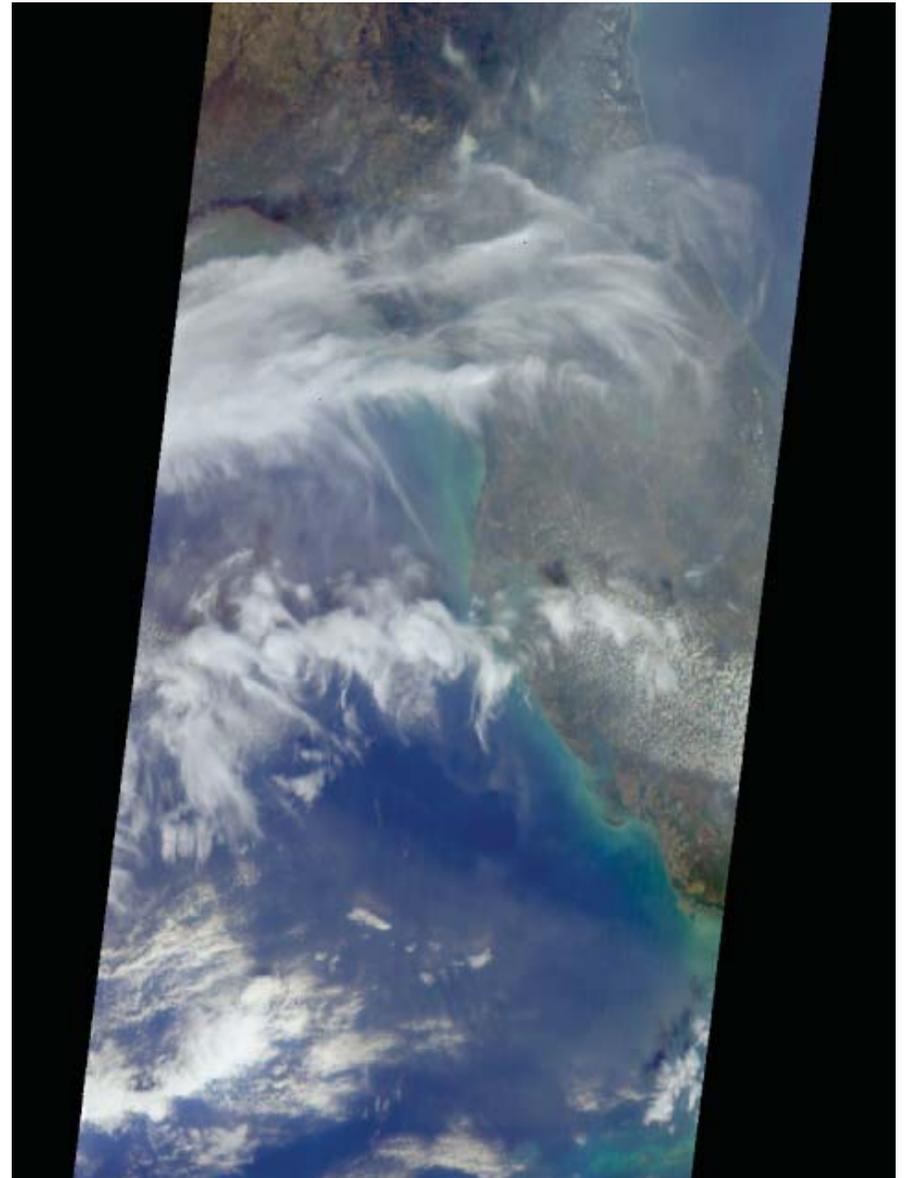
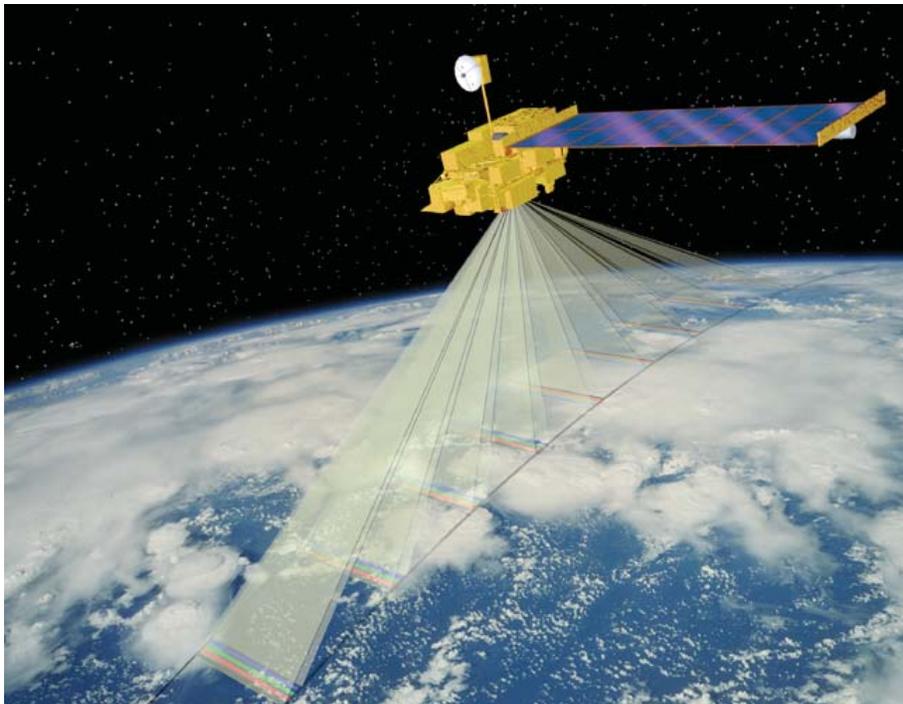


Multi-angle/multi-spectral: MISR

Aerosols: use radiometry

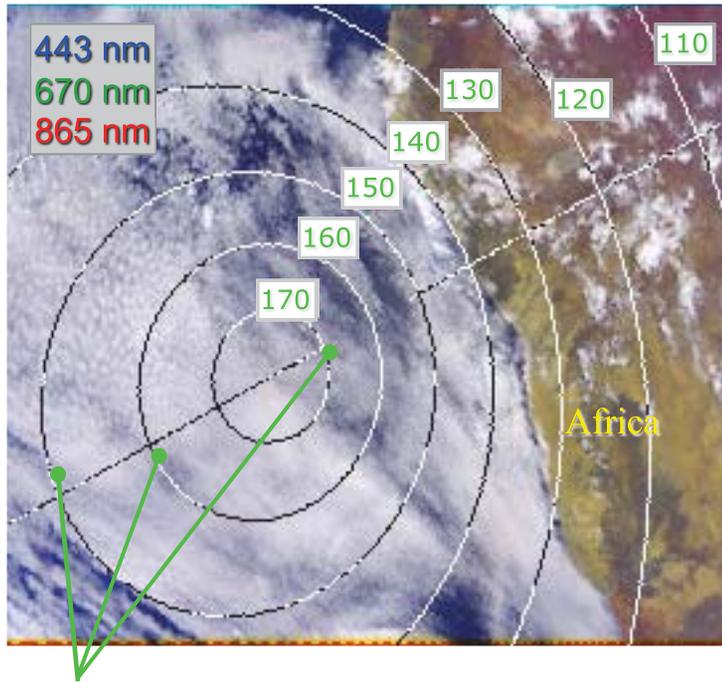
Clouds: use geometry

Ä in operational pipeline



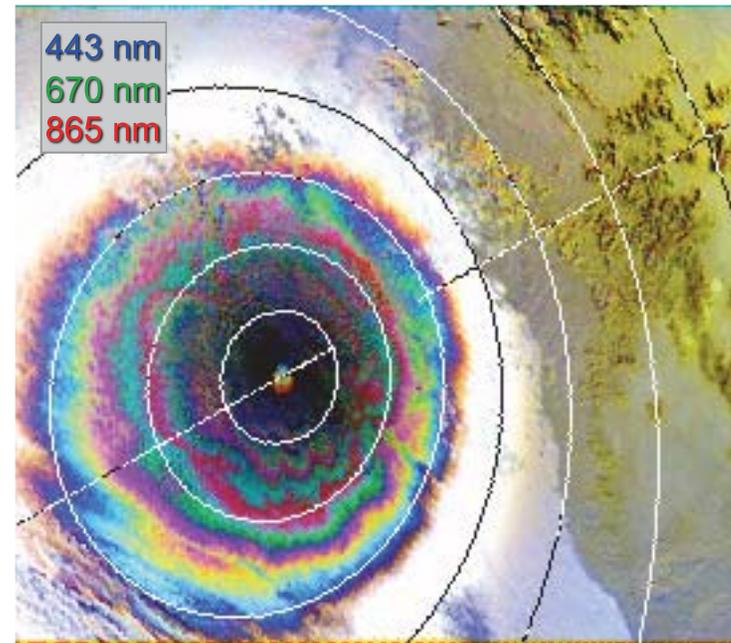
Multi-angle/multi-spectral with polarization diversity: POLDER

Stratocumulus over the ocean



Scattering angles

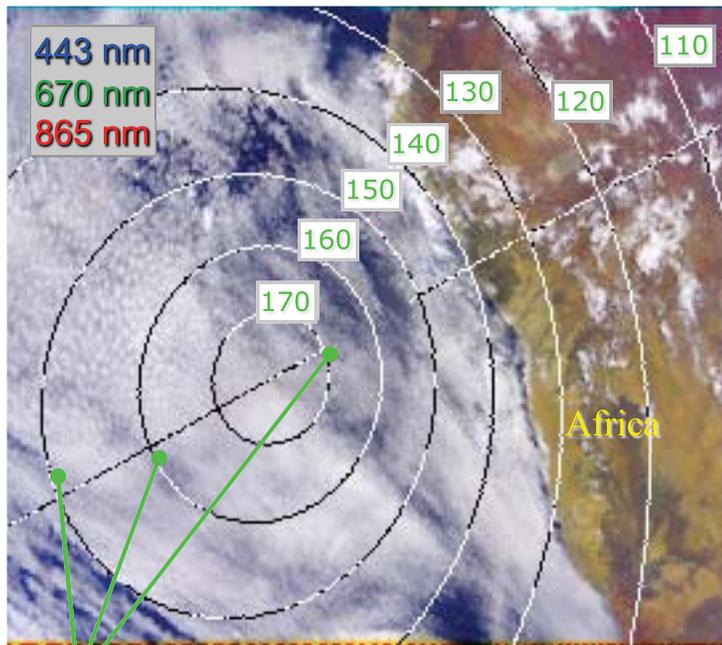
Same scene in polarized light



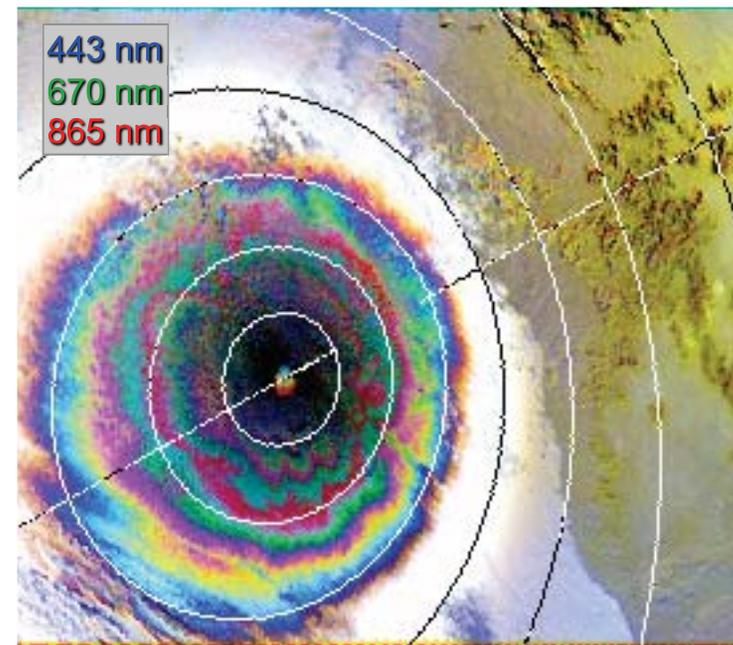
Source: François-Marie Bréon, LSCE, France

Multi-angle/multi-spectral with polarization diversity: POLDER

Stratocumulus over the ocean



Same scene in polarized light



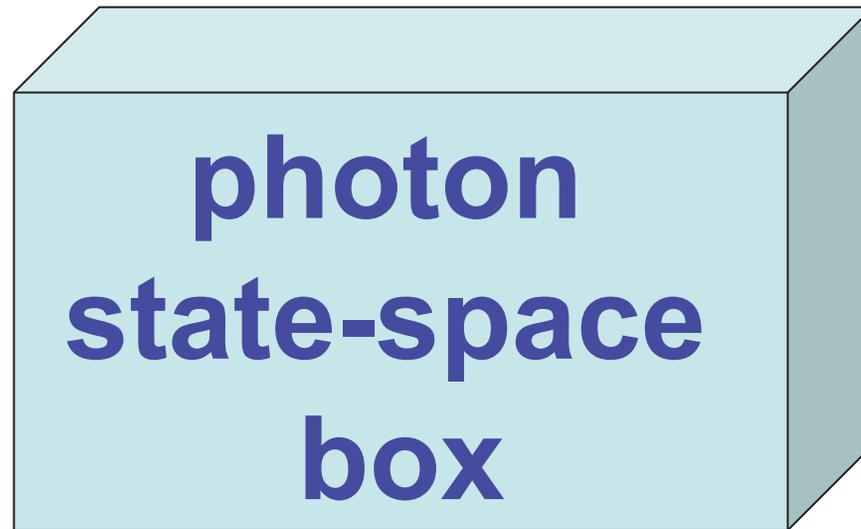
Source: François-Marie Bréon, LSCE, France



(hyper-angular, mono-pixel)

$\bar{\text{A}}$ and beyond!

Emerging paradigms from outside of the \bar{A}



Two examples of 3D particulate atmosphere tomography
using MISR data in
multi-angle/multi-pixel (and multi-spectral) retrievals \bar{A}

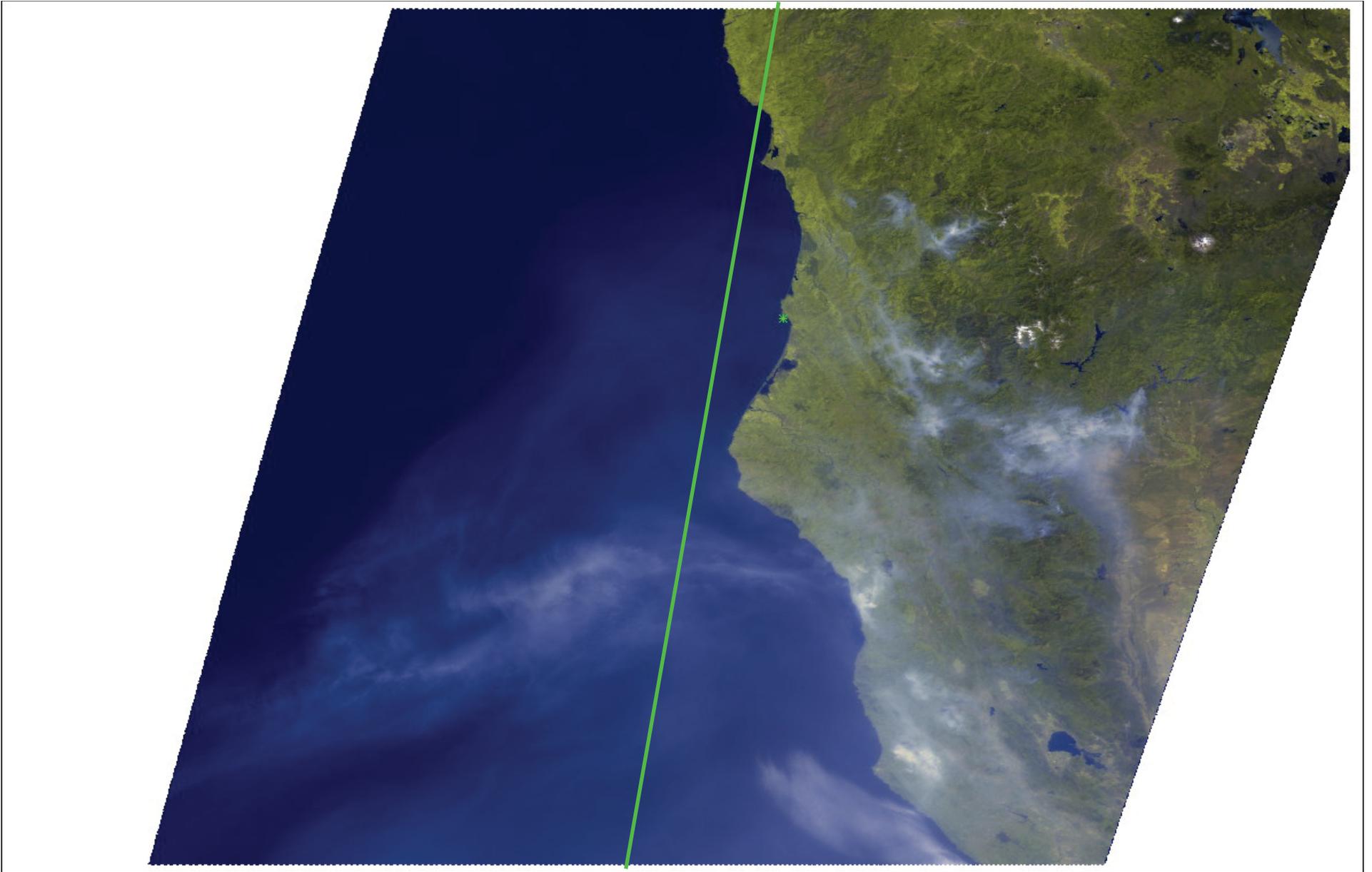
Aerosol Extinction Field Reconstruction

**multi-pixel/multi-angle and multi-spectral
algorithm**

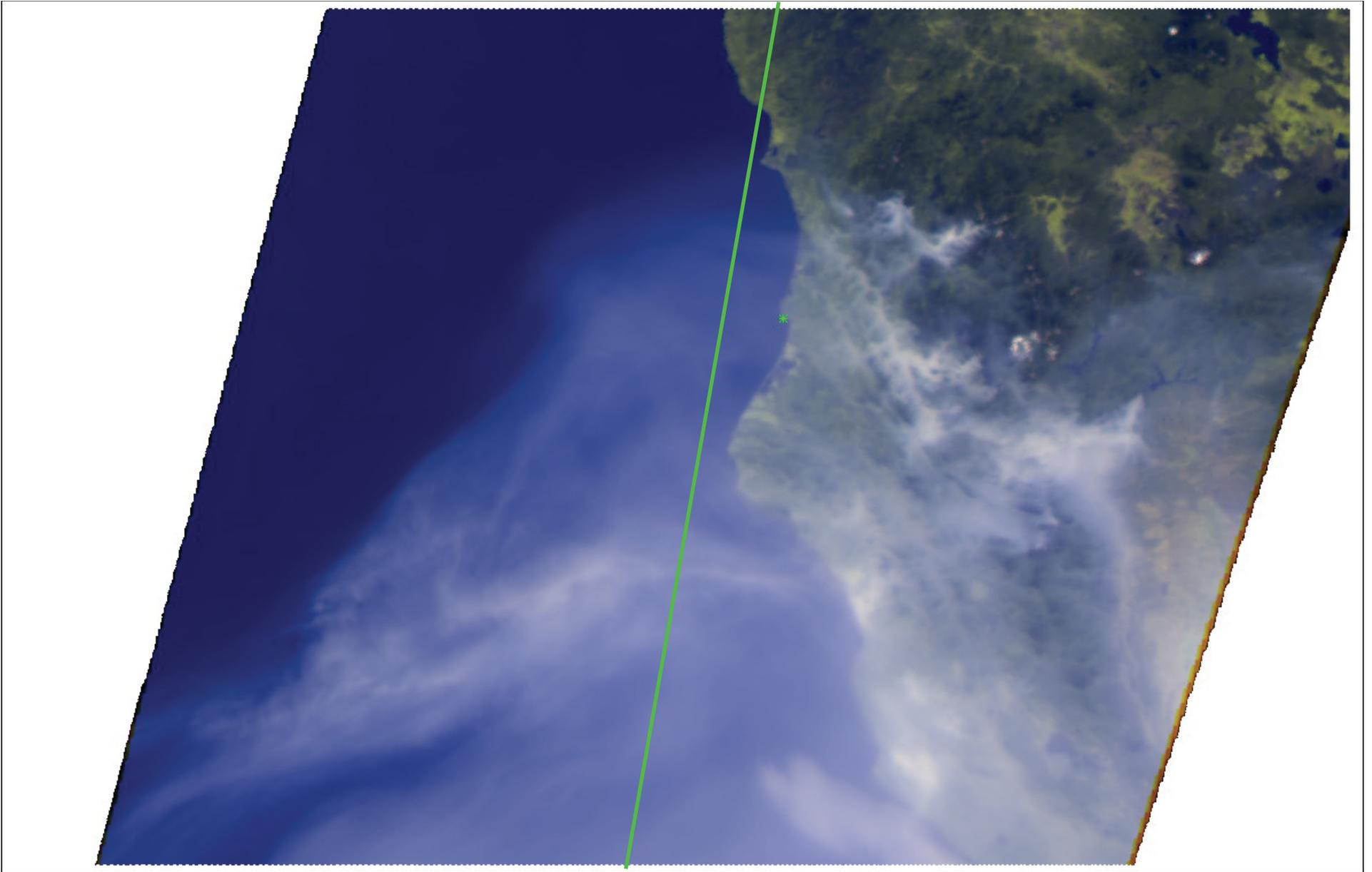
**Medical Imaging Analog:
Computed (X-ray) Tomography**



**Dry Lightning Caused Fires in Northern California
June 27, 2008**

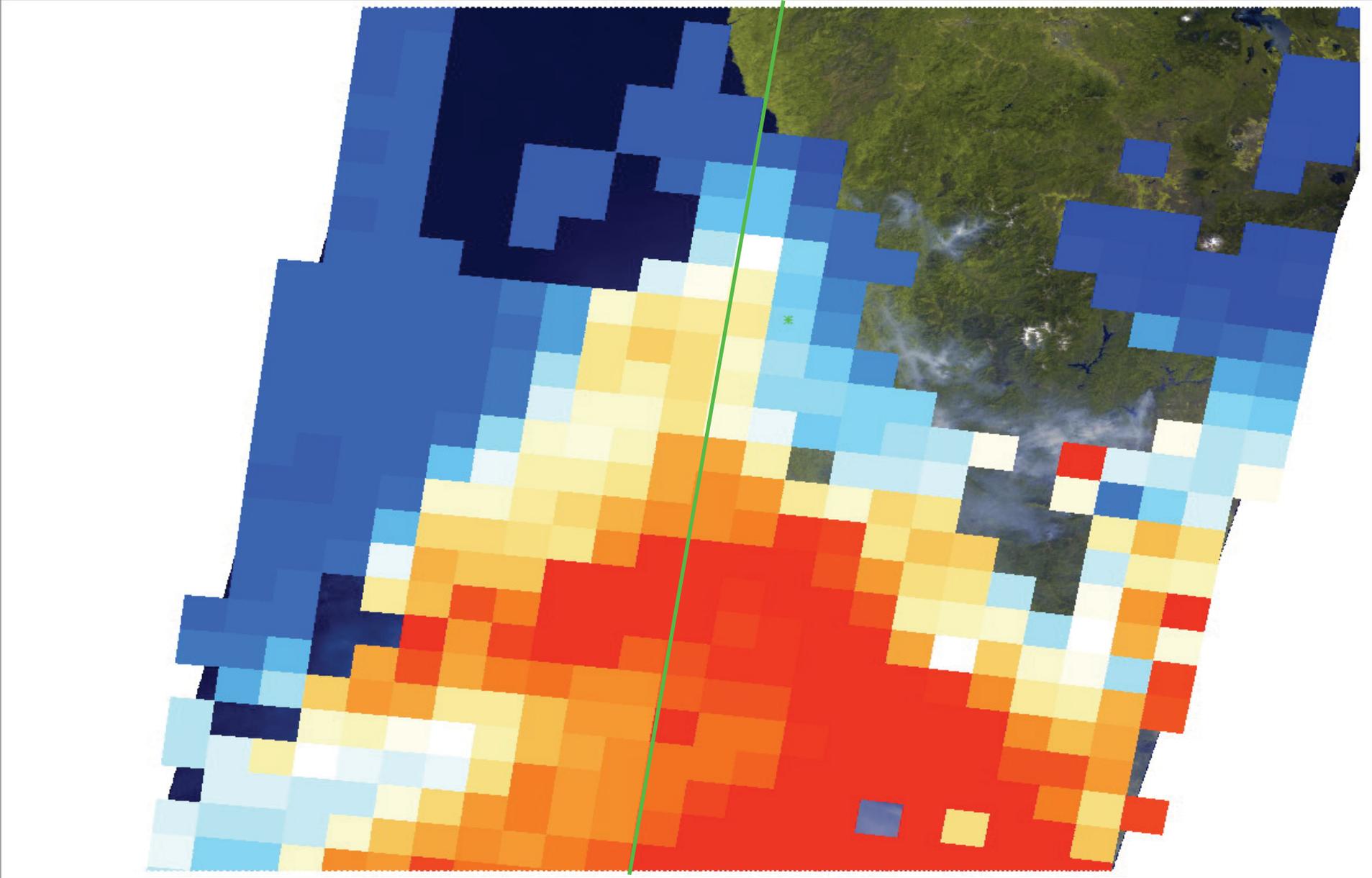


An



Df

Green Band Aerosol Optical Depth



“ART” reconstruction

- Convert model aerosol optical depths (AODs) to “tilted” AODs to account for atmospheric path by dividing by cosine of viewing angle.
- Construct vector, \mathbf{y} , of MISR tilted AODs (335 elements)
- Define another [unknown] vector, \mathbf{x} , for a regular grid of extinction values (9 rows x 49 columns = 441 elements) at $\bar{A}1$ km resolution.
- Construct a matrix \mathbf{S}_{ij} (335 x 441 elements S_{ij}) that maps \mathbf{x} to \mathbf{y} : $\mathbf{y} = \mathbf{S} \mathbf{x}$.
 - This matrix is based on the calculated geometric path of each camera ray through the regular grid
- Solve [ill-posed problem] using a bound-constrained linear least-squares minimization algorithm.

Magic Square

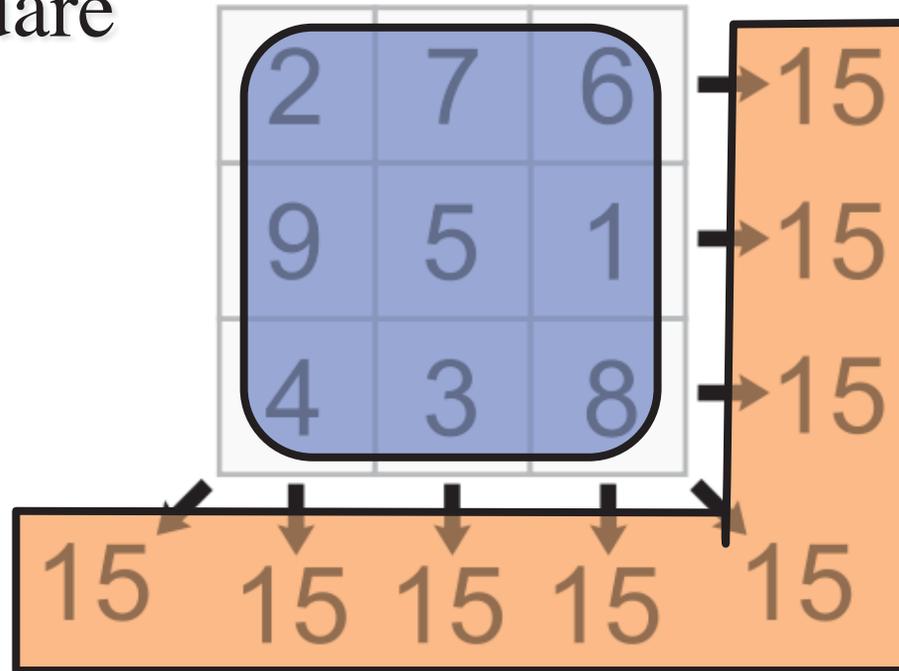
2	7	6	→ 15
9	5	1	→ 15
4	3	8	→ 15

15 ↙ ↓ ↓ ↓ ↘ 15

15 15 15 15 15

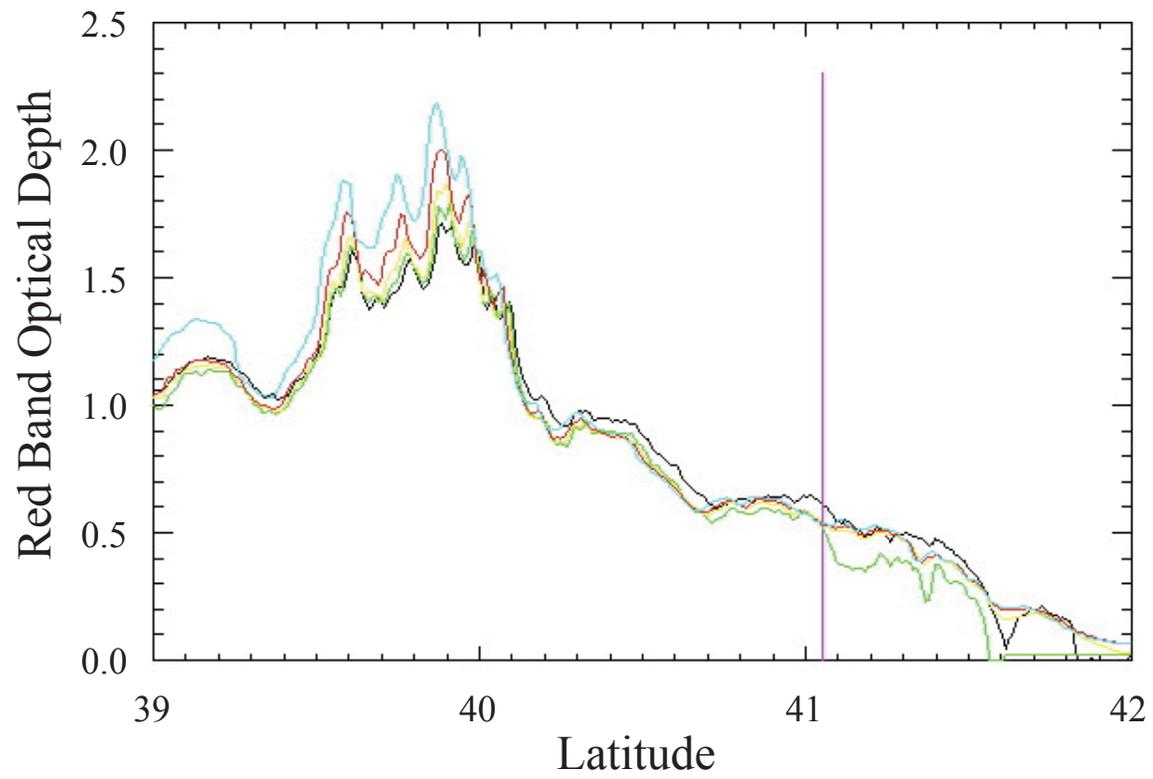
Magic Square

Vector x



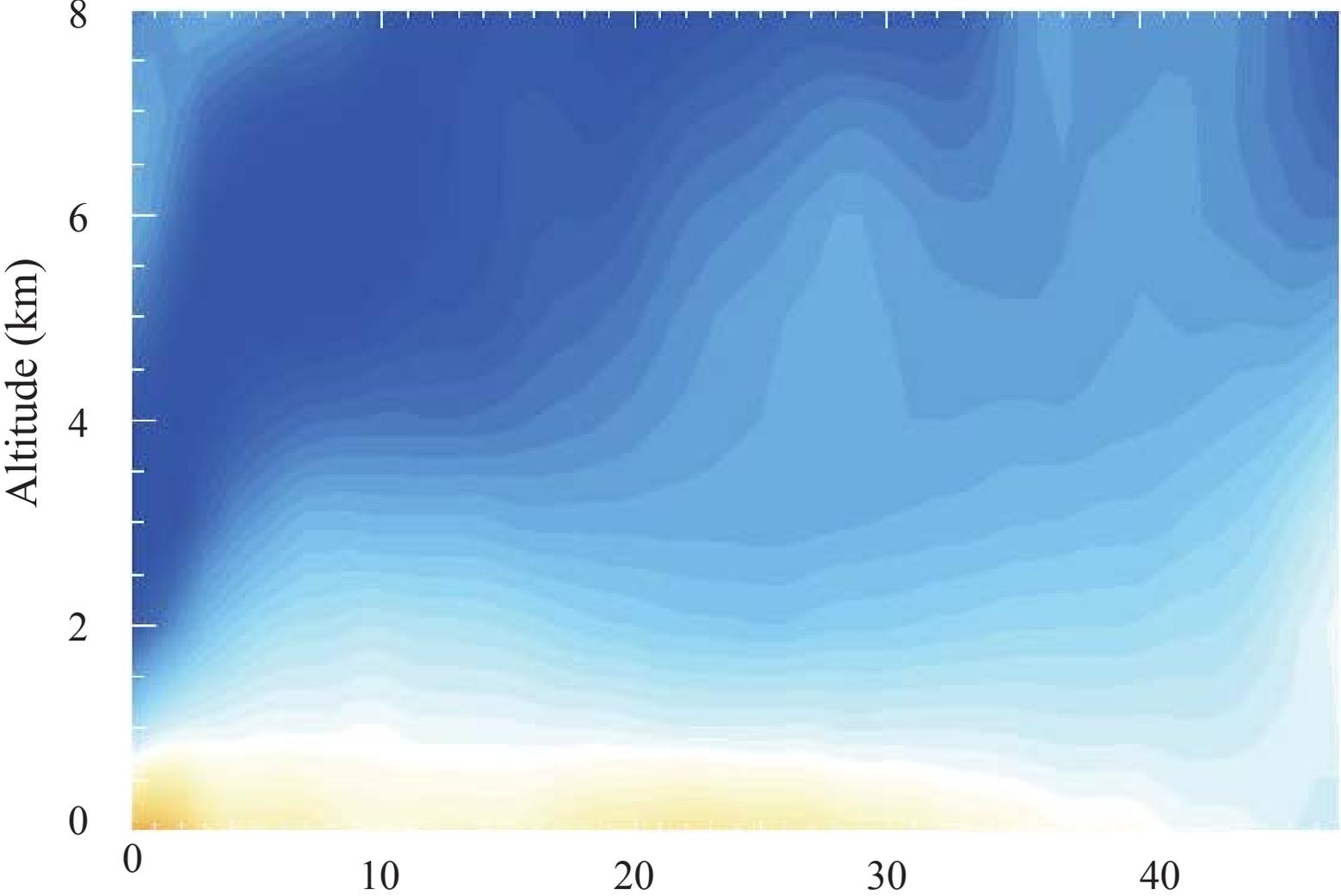
Vector y

Medical Imaging Analog:
Computed Tomography (CT)

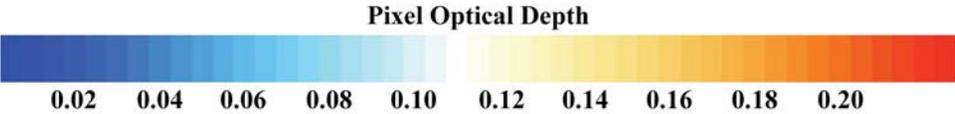


Vector y

Coarse/smooth 3D spatial distribution of aerosol extinction



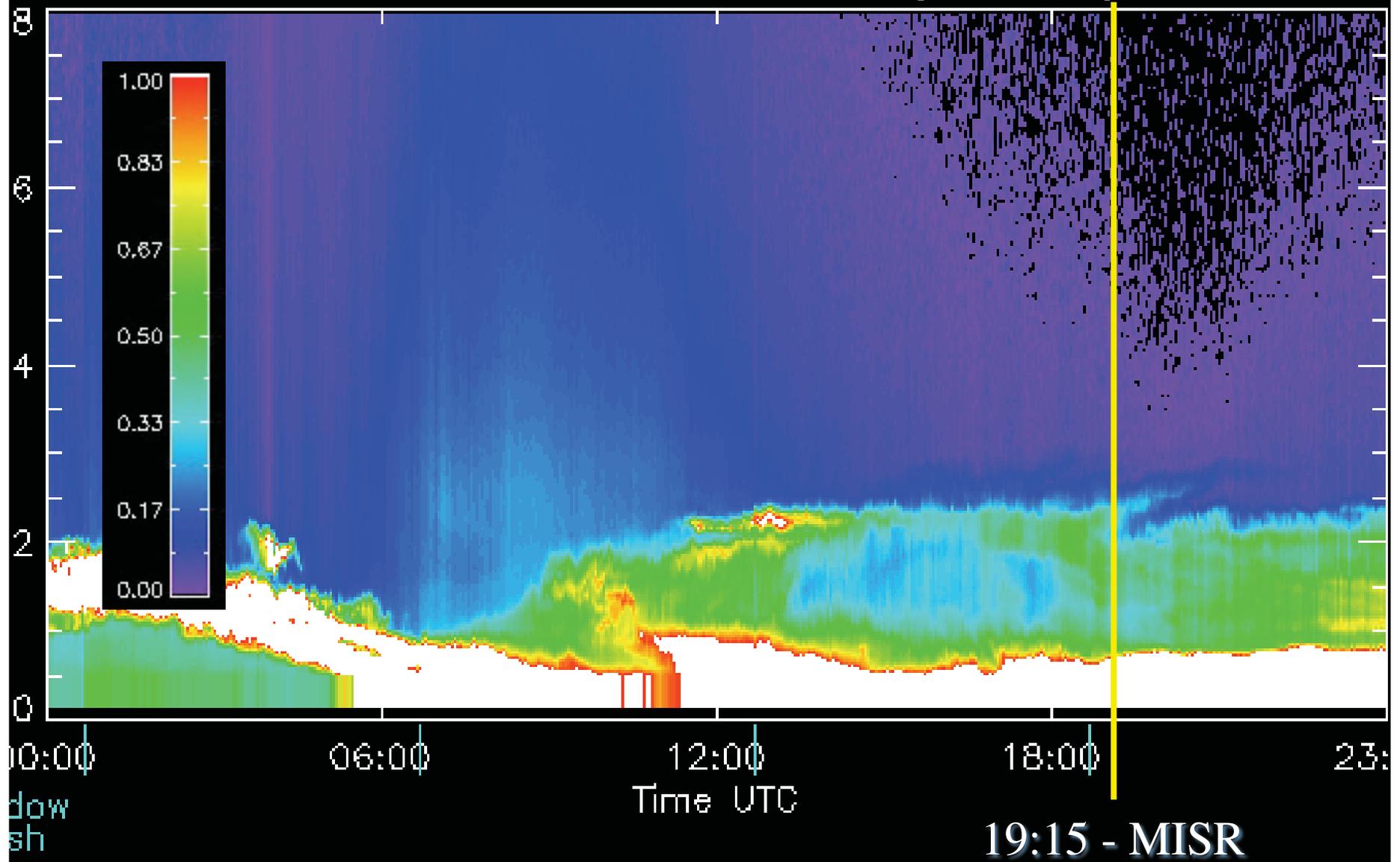
Vector x



PLNET Level 1.0 Data: Trinidad_Head 20080627 (v0, MPL40405)

Co
Prelim

Normalized Relative Backscatter (527.0 nm)

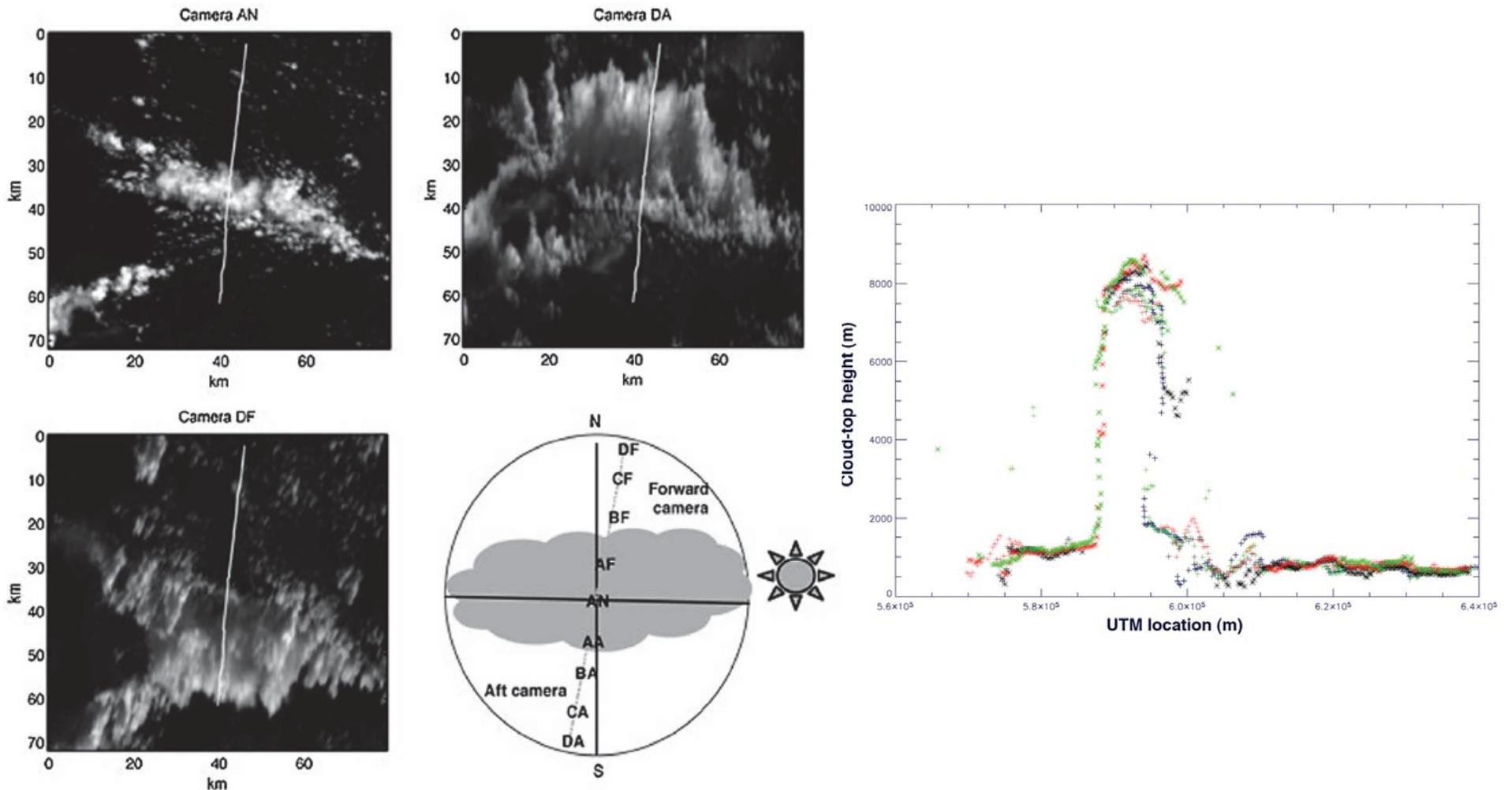


3D (Cu-type) Clouds

**multi-pixel/multi-angle [and multi-spectral]
algorithm**

**Medical Imaging Analog:
Diffuse Optical Tomography**

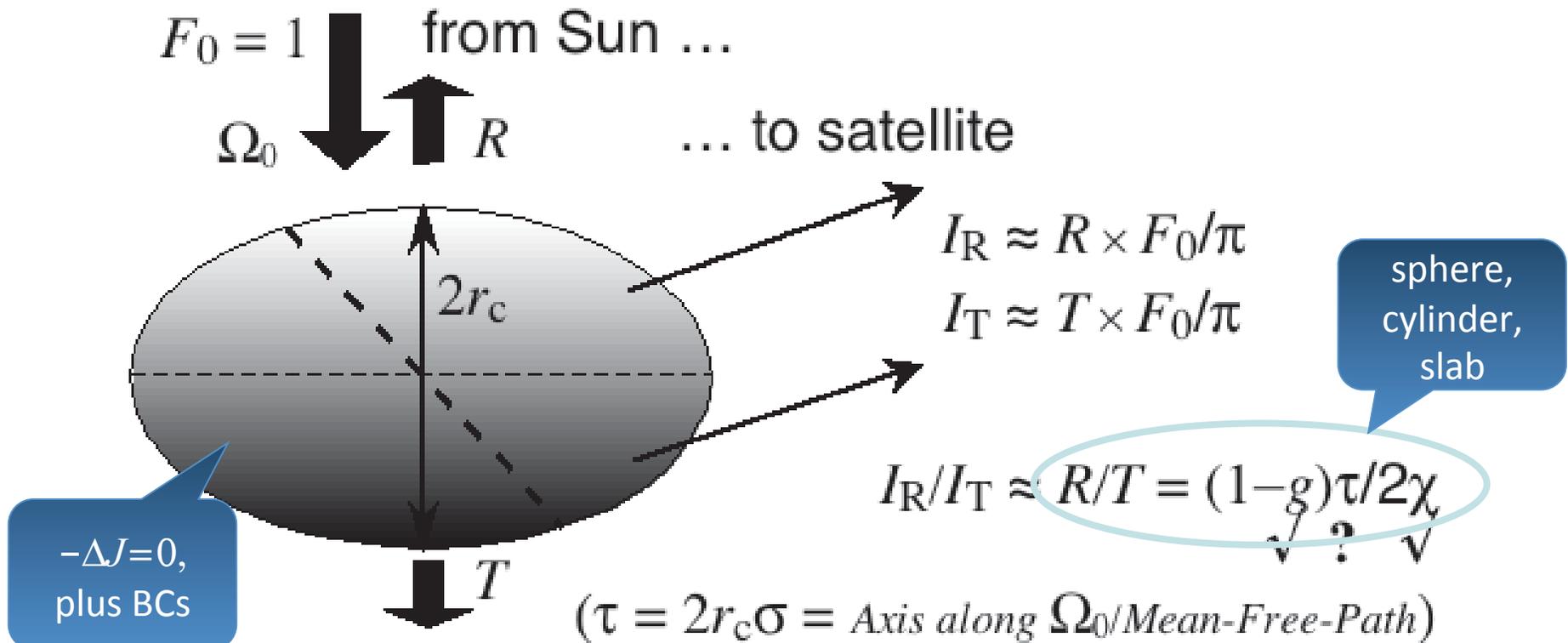
A previously-studied cloud mass \bar{A}



Cornet, C., and R. Davies, Use of MISR measurements to study the radiative transfer of an isolated convective cloud: Implications for cloud optical thickness retrieval. *J. Geophys. Res.*, **113** (D4): D04202 (2008).

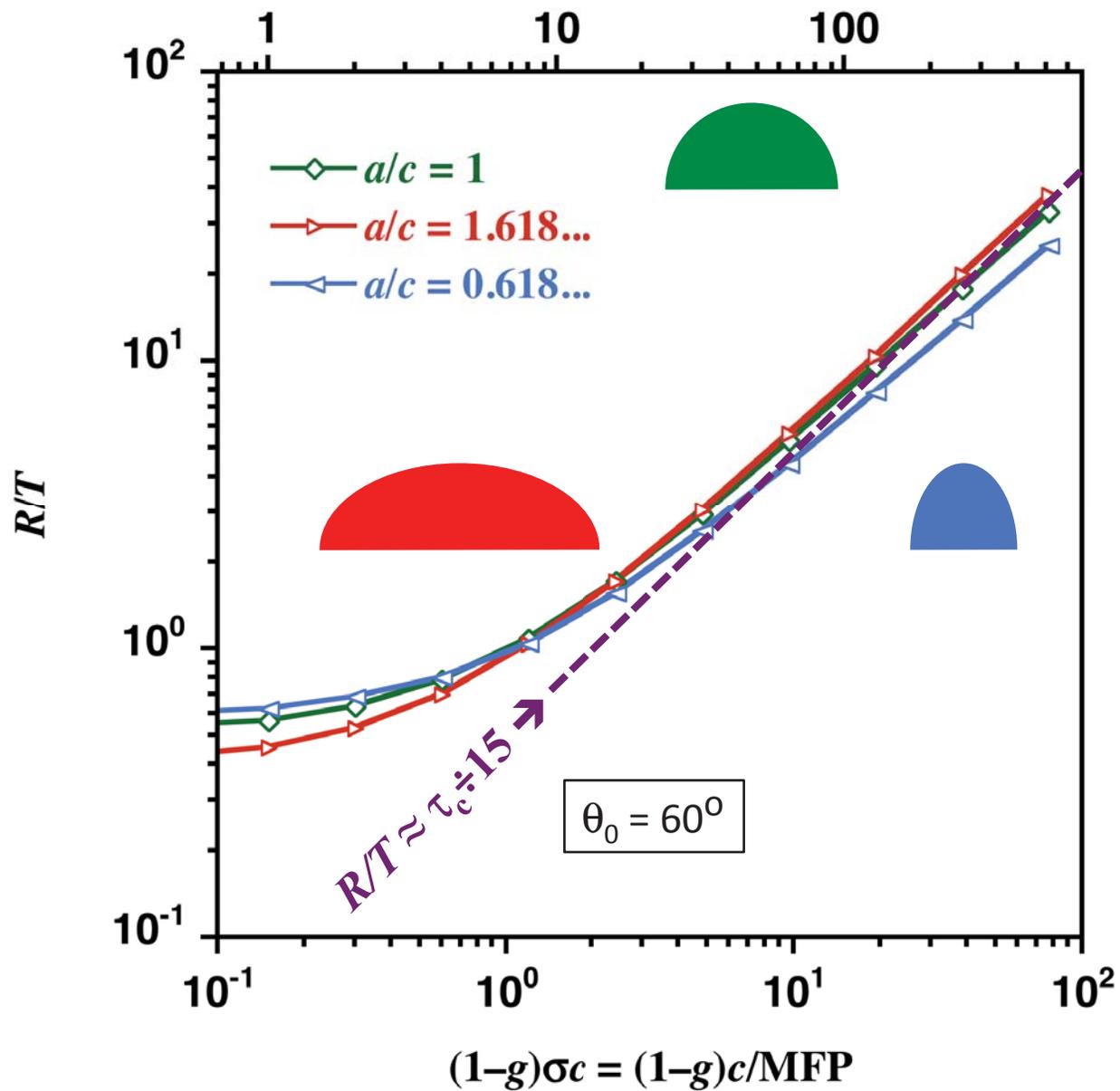
Let there be a spheriodal cloud!

\bar{A}

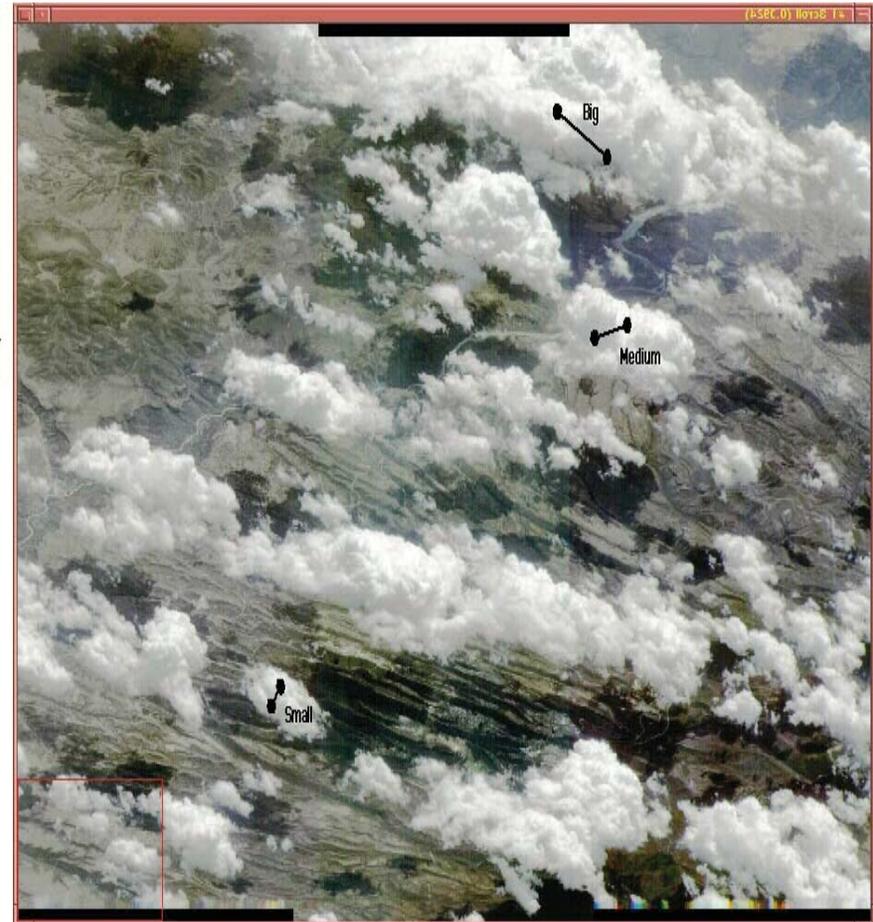
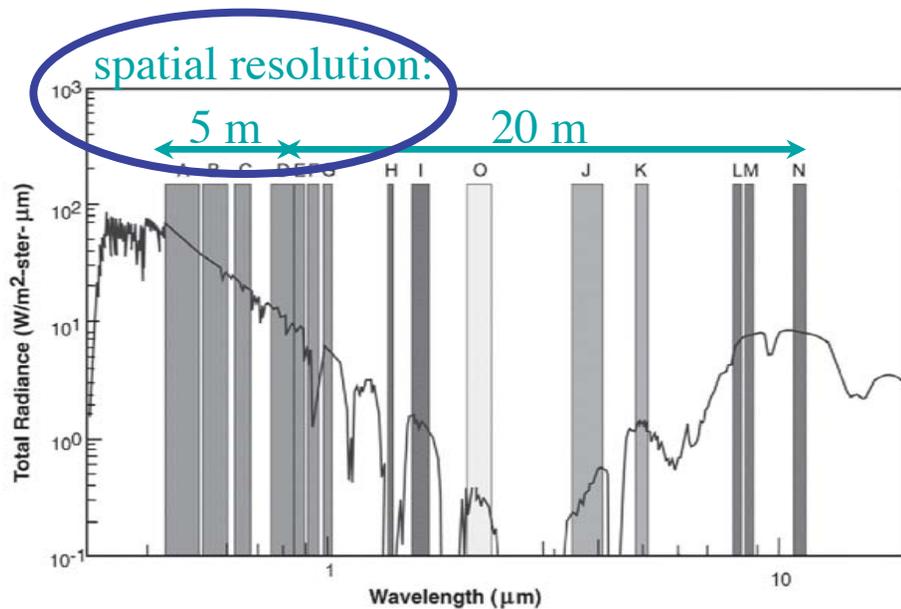


Davis, A. B., 2002: Cloud remote sensing with sideways-looks: Theory and first results using Multispectral Thermal Imager (MTI) data, in *S.P.I.E. Proceedings, vol. 4725: "Algorithms and Technologies for Multispectral, Hyperspectral, and Ultraspectral Imagery VIII,"* Eds. S. S. Shen and P. E. Lewis, S.P.I.E. Publications, Bellingham (Wa), pp. 397-405.

$$\tau_c = \sigma c = c/\text{MFP} \quad (\text{when } g = 0.85)$$



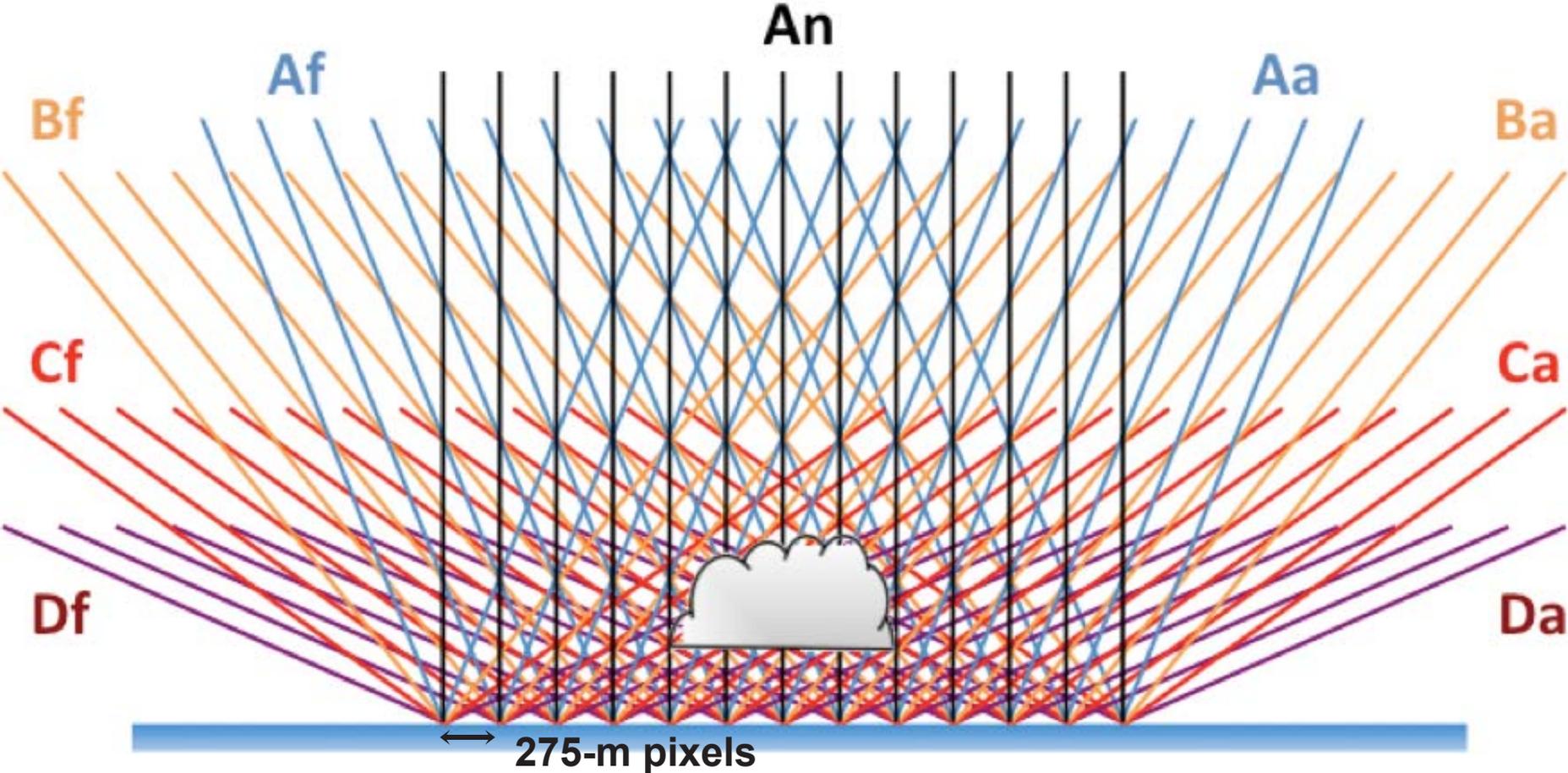
Preliminary results for an MTI scene (Los Alamos, NM)



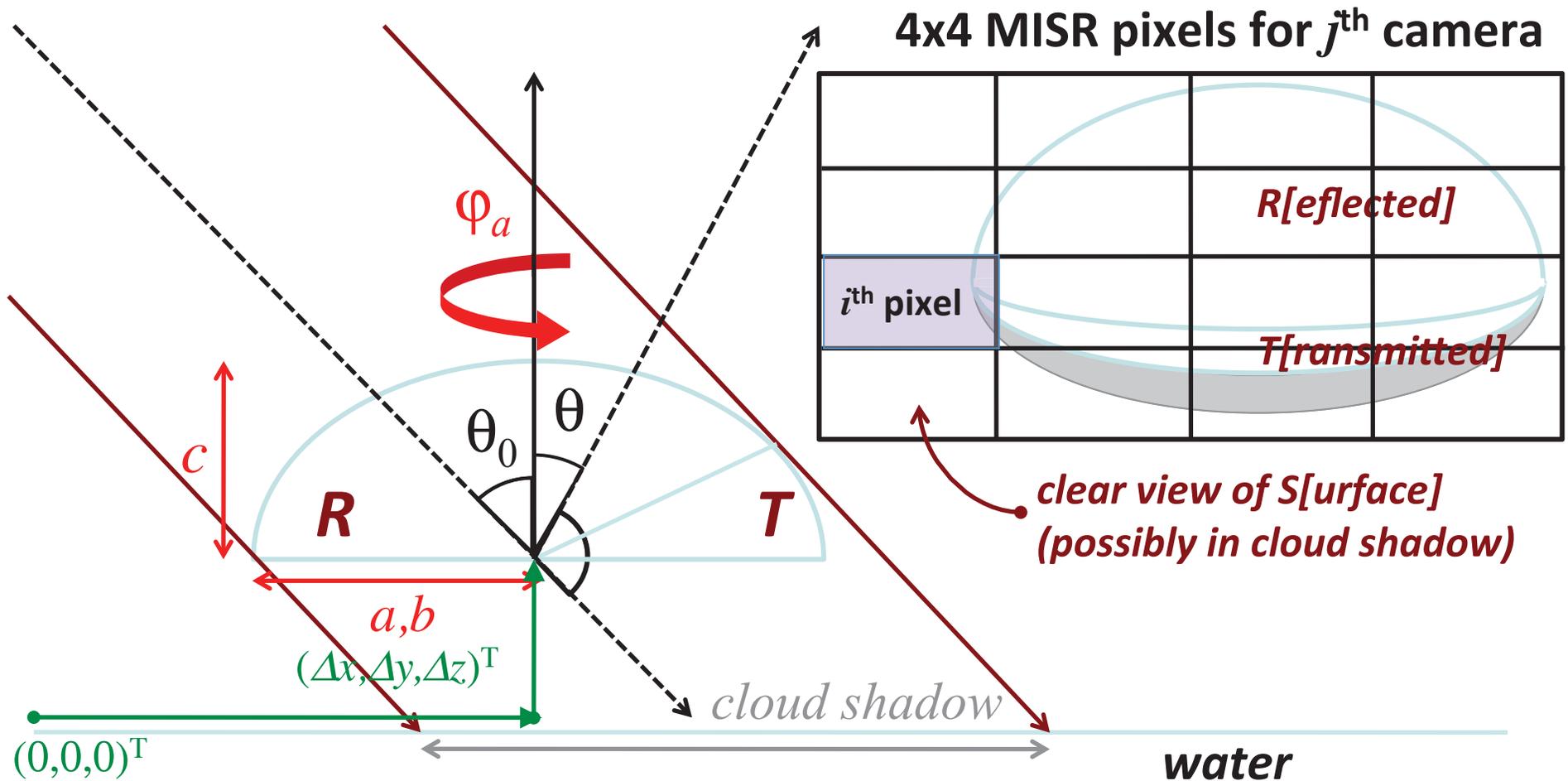
Cloud → “D” data ↓	“Big”		“Medium”		“Small”	
	R-region	T-region	R-region	T-region	R-region	T-region
mean:	255.89	58.091	257.37	56.816	195.79	70.770
st-dev/mean (%):	2.1	11	6.0	5.5	7.6	5.4
minimum:	244.40	45.861	228.34	51.128	167.00	65.461
maximum:	272.33	70.634	282.50	66.459	227.21	82.429
# of pixels (-):	272	624	132	340	81	120
I_R/I_T range (-):	3.5–5.9		3.4–5.5		2.0–3.5	
τ_{eff} range (-):	33–56		33–52		19–33	



Definition of cloud-mask volume using MISR's nine push-broom cameras



Very fast 3D cloud RT using radiosity & (in this case) a geometric primitive

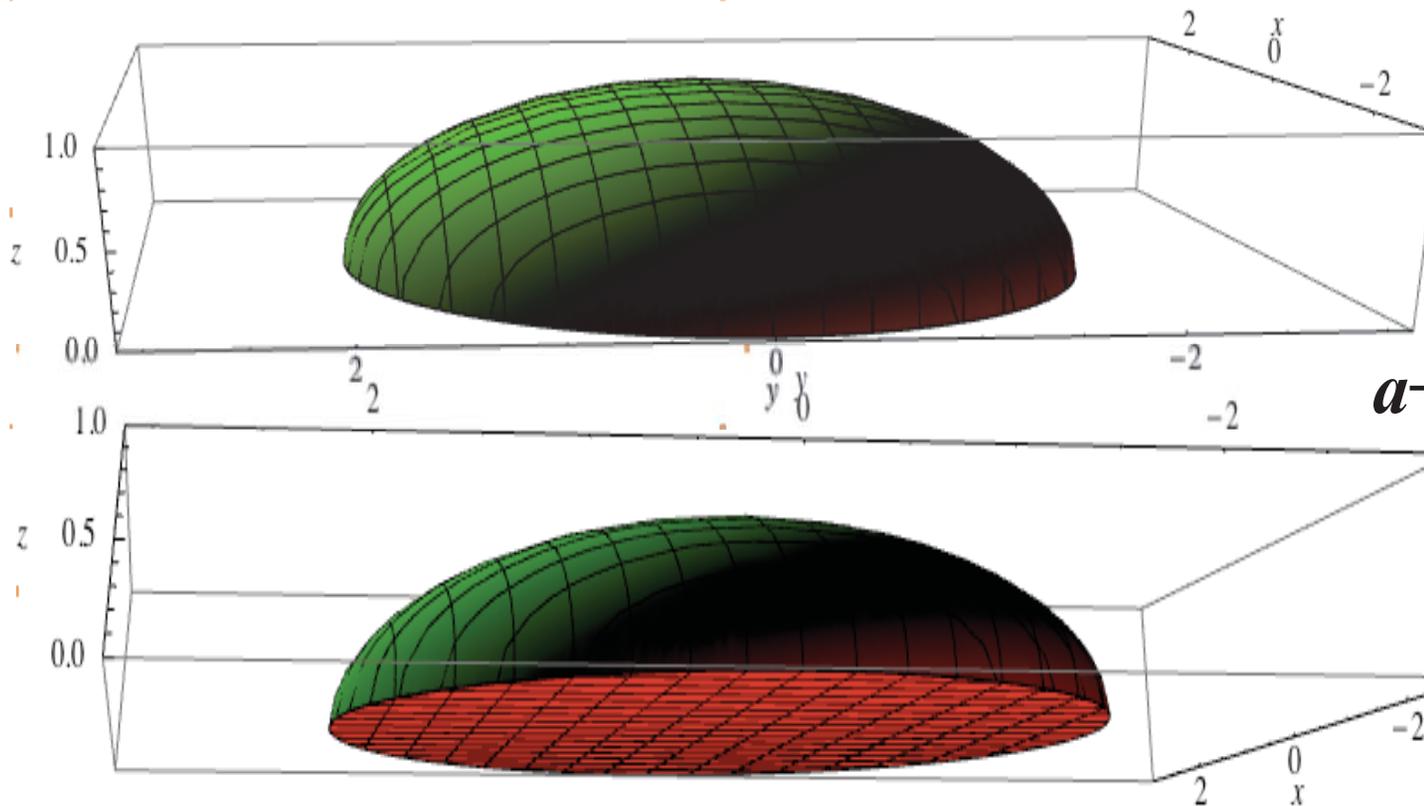


Multi-pixel/multi-angle radiosity-based radiance model:

$$I_j(\mathbf{i}) = \text{gain} \times [f_{R_j}(\mathbf{i}) R + f_{T_j}(\mathbf{i}) T + f_{S_j}(\mathbf{i}) S_j]$$

where $f_{F_j}(\mathbf{i})$ ($F = R, T, S$) depend on geometry, i.e., $\{ a, b, c, \varphi_a; \Delta x, \Delta y, \Delta z \}$.

Test Case: *Scalene Hemi-Ellipsoids*

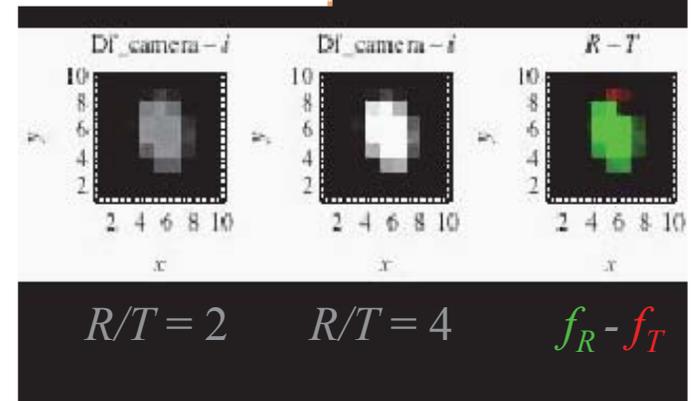


$$a \div b \div c = 3 \div 2 \div 1$$

$$a \div b \div c = [(1 + \sqrt{5})/2]^{2 \div 1 \div 0} \approx 2.618 \div 1.618 \div 1$$

on a 10 x 10 grid

NB. "Naïve" camera model used here →

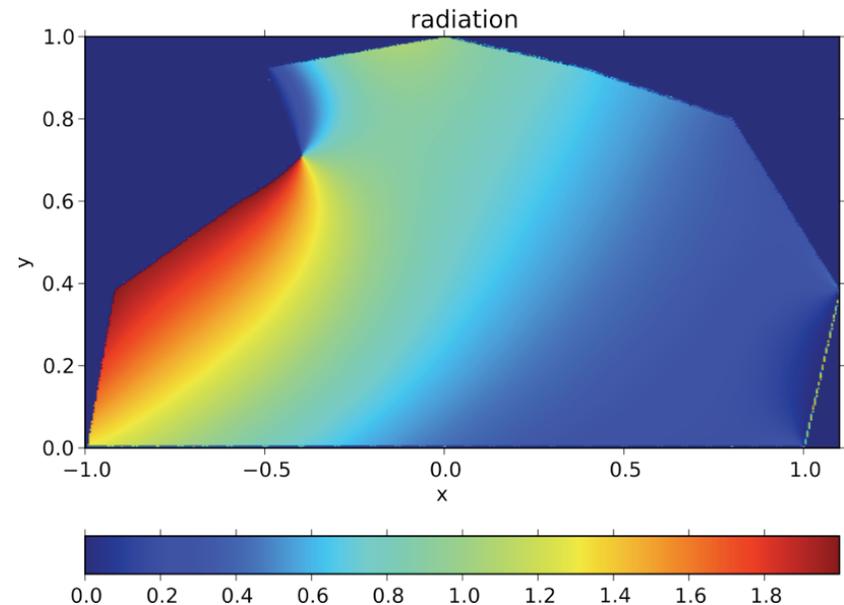
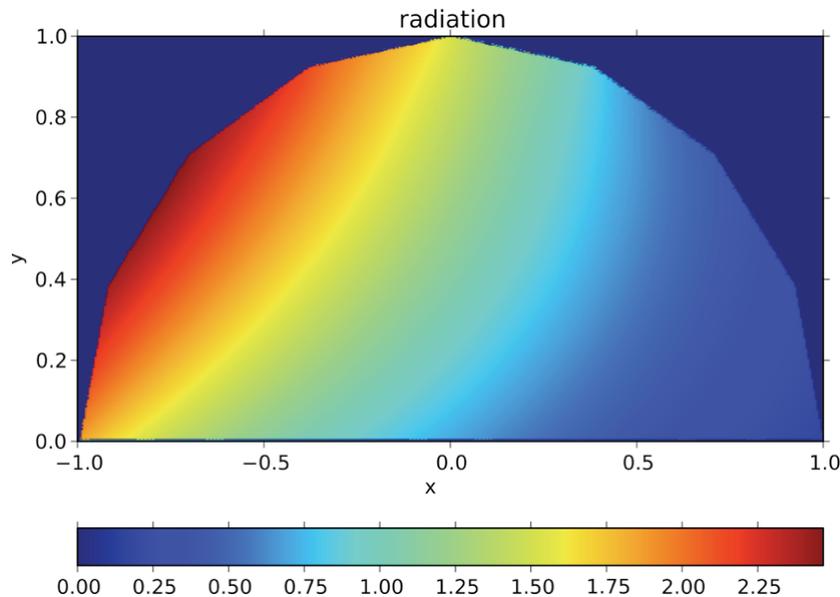


Convergence \bar{A} to self?

- **OK** *Forward model:* Random quadrature implementation, hence numerical noise (3% in this case)
- **Caution?** *Inverse model:* Powell minimization (no need for derivatives), from *Numerical Recipes* [Press et al., 1993]
- **Problem!**

I_R/I_T ratio + 7 more parameters:			"gain"	$2 \times \max\{a, b\}$	$2 \times \min\{a, b\}$	$ c $ [km]	φ_s [°]	δx [km]	δy [km]	N_{iter}	converged?
truth \rightarrow [1 st guess] \searrow			1	$2.618\dots = GR^2$	$1.618\dots = GR$	0.5	0	1.5	1.5	∞	
$\tau_c \approx 15 \times$	R/T (truth)	R/T [3 (ret'd)]	1.2	2	2	1	45	1.25	1.25]	0	
30	2	1.949	1.026	2.618	1.616	0.510	0.027	1.501	1.499	9	✓
75	5	5.303	0.943	2.663	1.586	0.444	0.723	1.491	1.500	7	✓
150	10	10.01	1.003	2.616	1.616	0.497	-0.08	1.499	1.500	13	✓
225	15	17.67	0.849	2.623	1.618	0.499	89.77	1.499	1.500	12	✓
300	20	27.80	0.719	2.621	1.616	0.492	90.26	1.499	1.501	9	✓
375	25	-0.517	-47.5	2.661	1.611	0.224	91.447	1.446	1.505	6	✗

Need a *fast* 3D RT for cloud masses!



$$\nabla^2 J = 3\sigma_t \sigma_a J \quad (\text{Helmoltz's elliptical PDE})$$

Boundary Condition : $\frac{1}{4}[J + 3\underbrace{\chi F \cdot n}_{\text{outward normal}}](\mathbf{x}) \begin{cases} -F_0 \Omega_0 \cdot \mathbf{n}(\mathbf{x}), & \text{if } > 0 \text{ (an illuminated facet);} \\ 0, & \text{otherwise.} \end{cases}$

where (Fick's law) $F(\mathbf{x}) = \frac{-1}{3(\sigma_t + \sigma_a)} \nabla J$

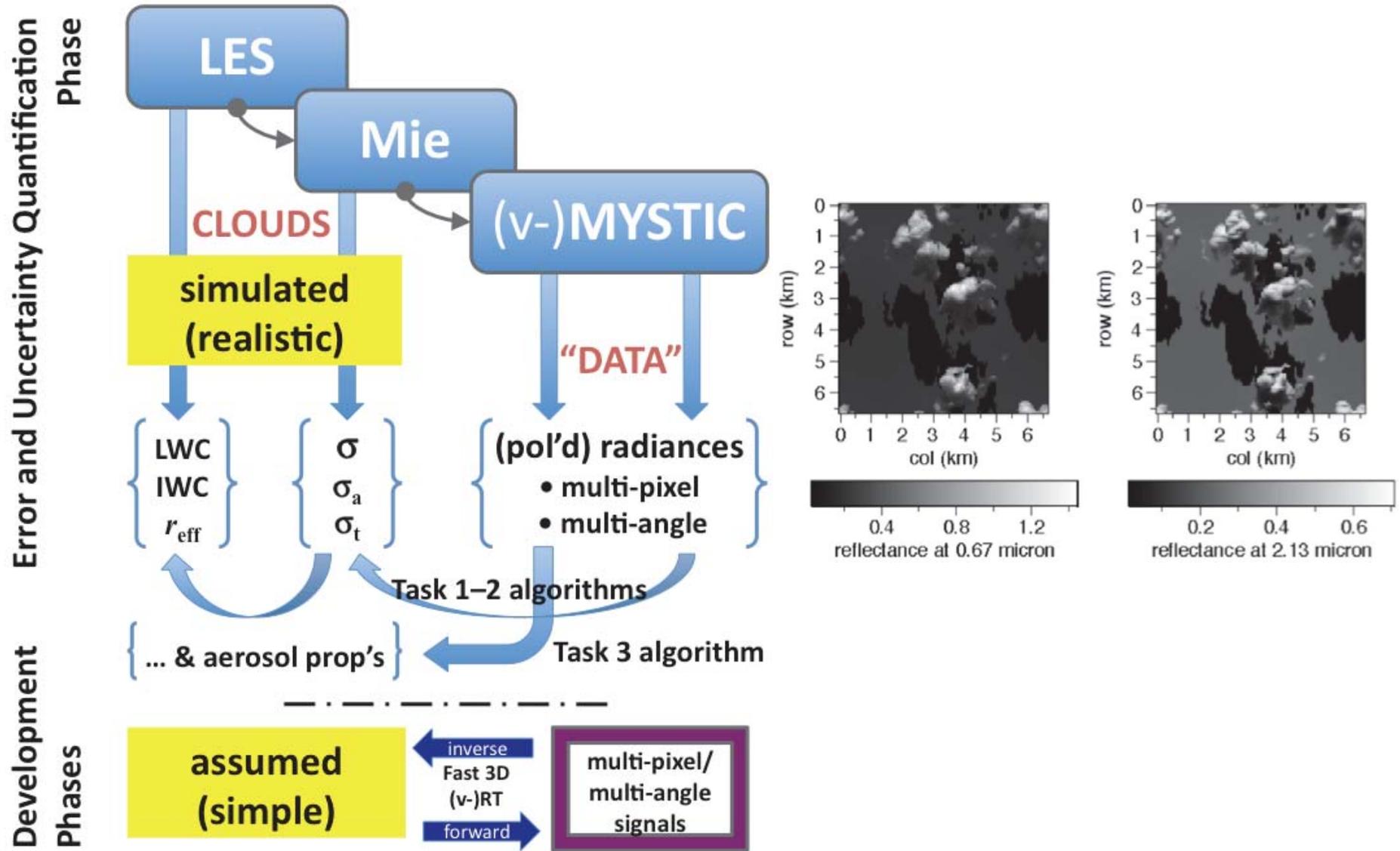
Outgoing Fluxes? $\frac{1}{4F_0}[J - 3\underbrace{\chi F \cdot n}_{\text{outward normal}}](\mathbf{x}) \begin{cases} R \\ T \end{cases}$

$$J(\mathbf{x}) = \oint_{\partial C} [J(\mathbf{x}') \partial_n G(\mathbf{x}', \mathbf{x}) - G(\mathbf{x}', \mathbf{x}) \partial_n J(\mathbf{x}')] d\mathbf{x}',$$

$$\nabla^2 G = \delta(\mathbf{x}' - \mathbf{x}), \text{ hence } G(\mathbf{x}', \mathbf{x}) = -1/4\pi \|\mathbf{x}' - \mathbf{x}\|.$$

Medical Imaging Analog:
Diffuse Optical Tomography

Error & Uncertainty Quantification



Summary

- **“Mono-pixel” retrieval methodology is reaching its fundamental limit with multi-angle/multi-spectral photo-polarimetry**
- **Two emerging classes of retrieval algorithm worth growing:**
 - **Time-domain exploitation “beyond ranging”**
 - **Multi-pixel techniques, including tomography**
- **Cross-fertilization with bio-medical imaging**
- **MISR is blazing the multi-pixel path forward**

Thank you!

Questions?

Acknowledgments:

Support from NASA/SMD/ESD/Radiation Sciences