



**Jet Propulsion Laboratory**  
California Institute of Technology

# Physics-Based Analysis of Cloud Information Content in EPIC/DSCOVR's O<sub>2</sub> A- and B-Band Channels

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# Outline

- **Physics-based approach**

- A.B. Davis, N. Ferlay, Q. Libois, A. Marshak, Y. Yang, and Q. Min. Cloud information content in EPIC/DSCOVR's oxygen A- and B-band channels: A physics-based approach. *J. Quant. Spectrosc. Rad. Transf.* **220**, 84-96 (2018).

<https://doi.org/10.1016/j.jqsrt.2018.09.006>

- **Optimal estimation approach**

- A.B. Davis, G. Merlin, C. Cornet, L. C.-Labonnote, J. Riédi, N. Ferlay, P. Dubuisson, Q. Min, Y. Yang, and A. Marshak. Cloud information content in EPIC/DSCOVR's oxygen A- and B-band channels: An optimal estimation approach. *J. Quant. Spectrosc. Rad. Transf.* **216**, 6-16 (2018).

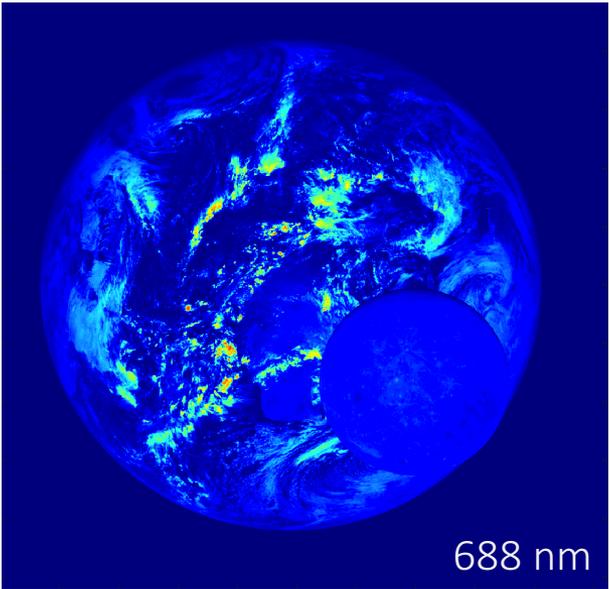
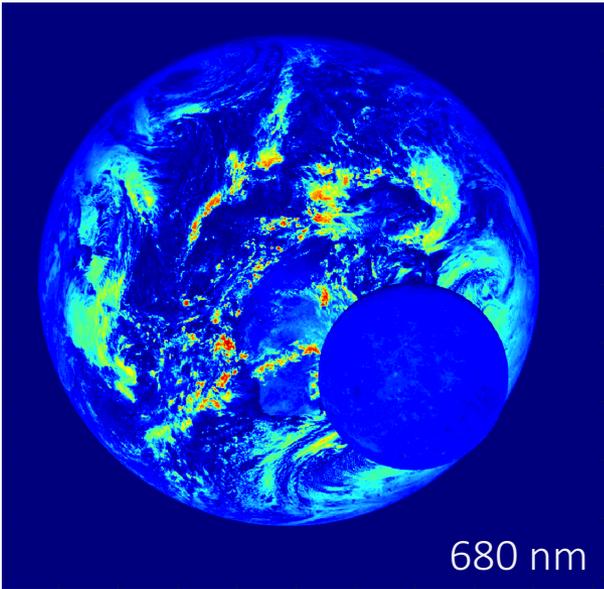
<https://doi.org/10.1016/j.jqsrt.2018.05.007>

- **Conclusions/outlooks**

# EPIC's A- and B-band channels, at Nov'17 new Moon

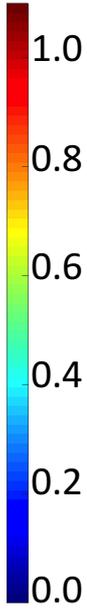
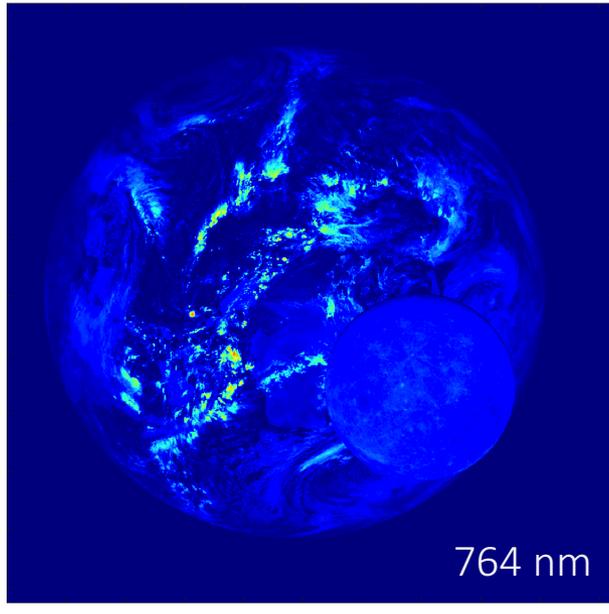
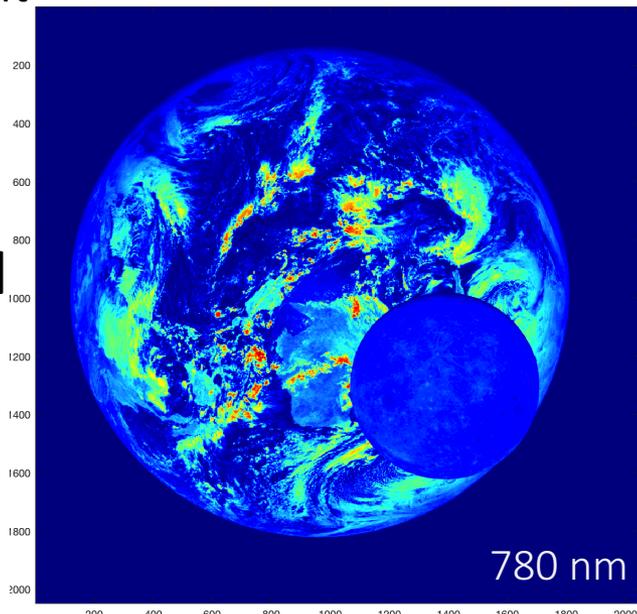
EPIC/DSCOVER  
O<sub>2</sub> channels:

**B-band**  
 $\tau_{O_2} \approx 0.3$



Reference on left  
( $\tau_{O_2} = 0$ )

**A-band**  
 $\tau_{O_2} \approx 0.6$



2017-11-19

# EPIC/DSCOVR O<sub>2</sub> channels, at Nov'17 new Moon

EPIC/DSCOVR  
O<sub>2</sub> channels:

**B-band**

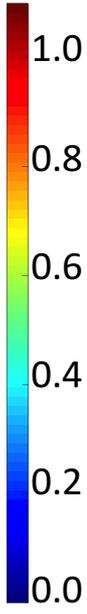
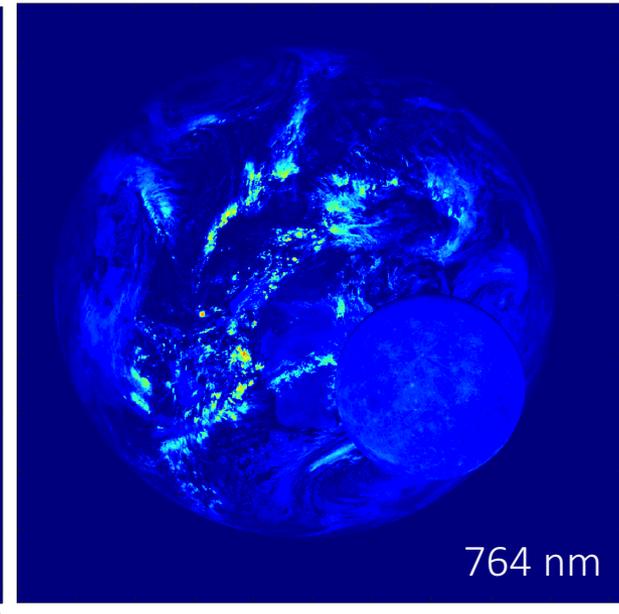
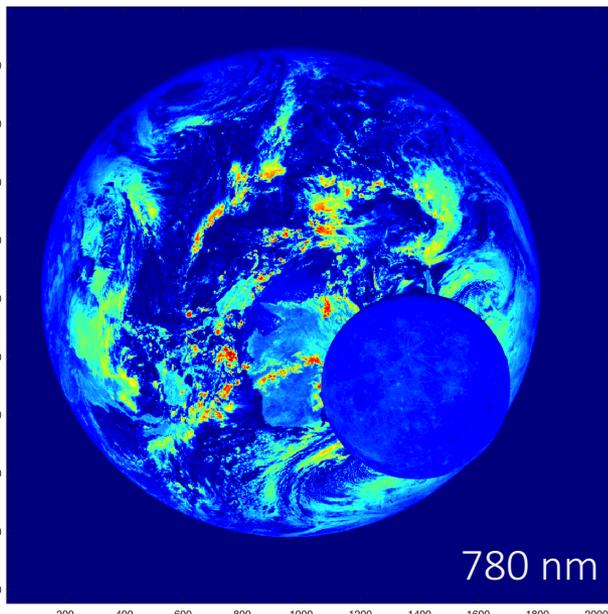
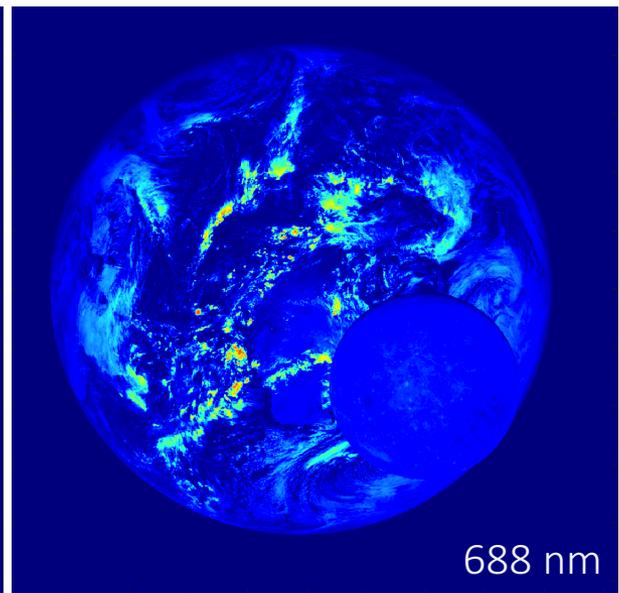
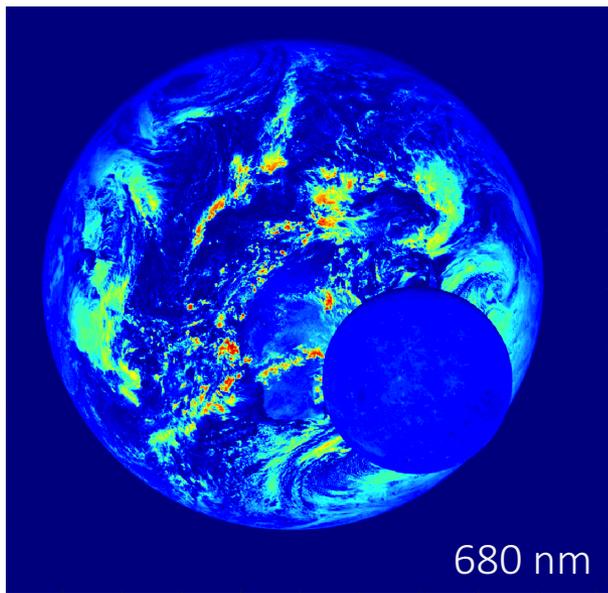
$\tau_{O_2} \approx 0.3$

Reference on left  
( $\tau_{O_2} = 0$ )

**A-band**

$\tau_{O_2} \approx 0.6$

effective optical thicknesses



2017-11-19

# EPIC's A- and B-band channels, in full spectral detail

EPIC/DSCOVER

O<sub>2</sub> channels:

**B-band**

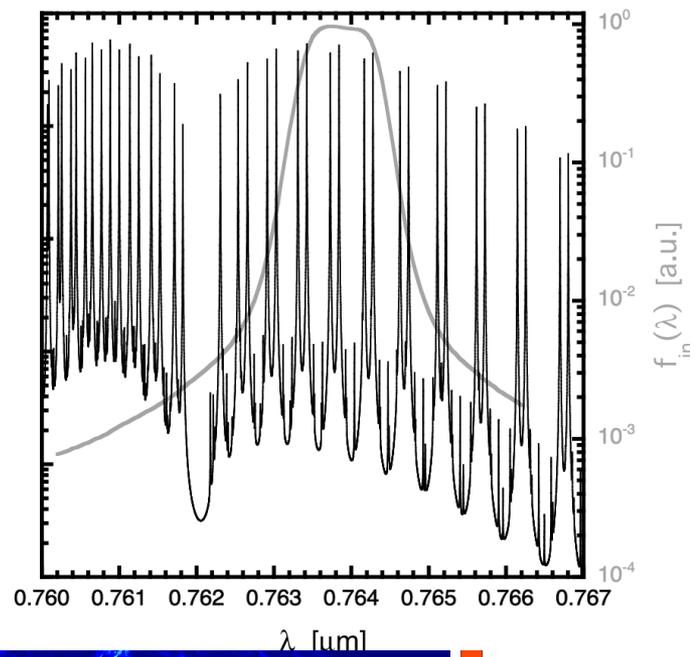
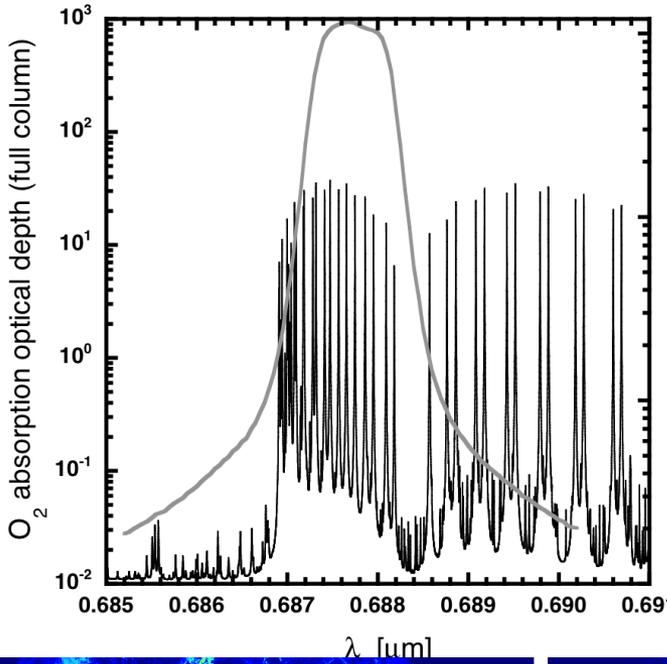
$\tau_{O_2} \approx 0.3$

Reference on left  
( $\tau_{O_2} = 0$ )

**A-band**

$\tau_{O_2} \approx 0.6$

effective optical thicknesses



200  
400  
600  
800  
1000  
1200  
1400  
1600  
1800  
2000



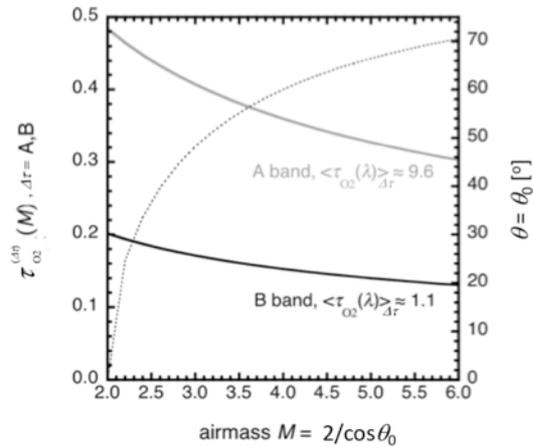
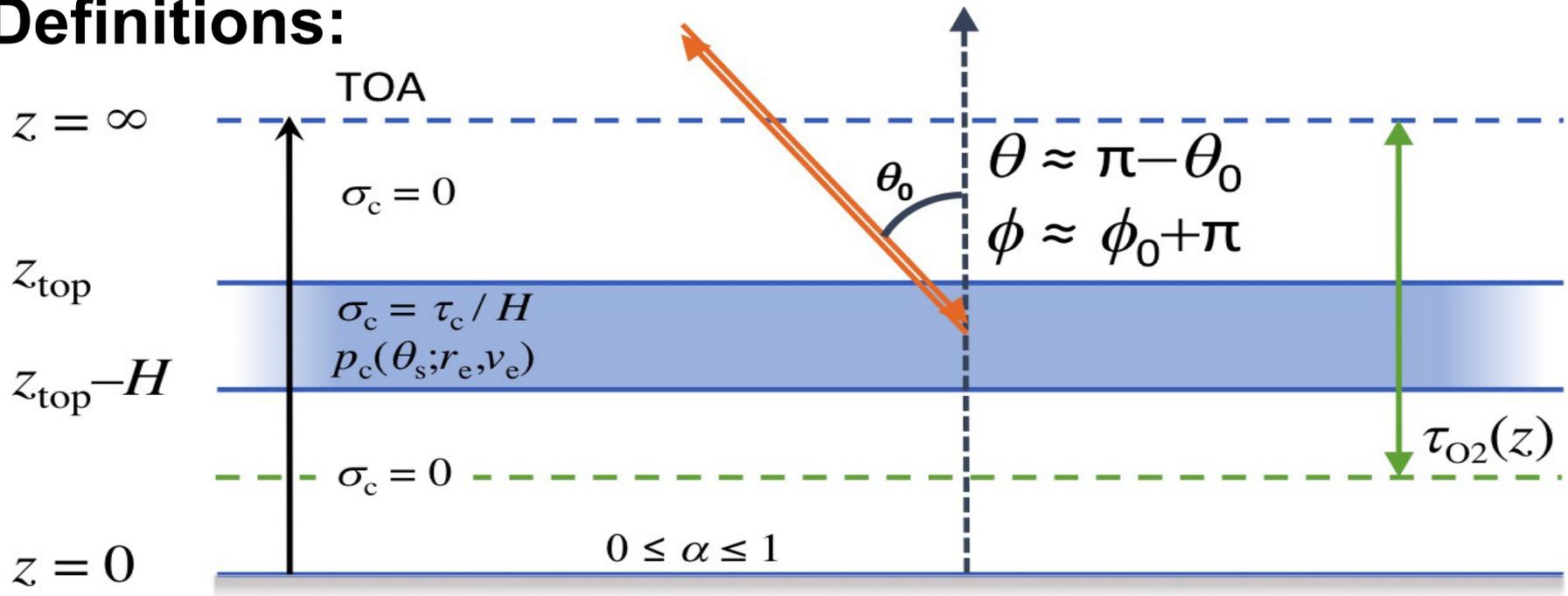
$$\tau_{O_2}^{(\Delta\lambda)}(M) = -\frac{1}{M} \log \left( \int_{\Delta\lambda} \exp[-M\tau_{O_2}(\lambda; 0)] f_{in}^*(\lambda) d\lambda \right)$$

so-called "airmass" factor

$$M = \frac{1}{\cos \theta_0} + \frac{1}{\cos \theta} = \frac{1}{\mu_0} + \frac{1}{\mu}$$

# Physics-based modeling approach

## Definitions:



$$\tau_{\text{O}_2}^{(\Delta\lambda)}(M) = -\frac{1}{M} \log \left( \int_{\Delta\lambda} \exp[-M\tau_{\text{O}_2}(\lambda; 0)] f_{\text{in}}^*(\lambda) d\lambda \right)$$

so-called “airmass” factor

$$M = \frac{1}{\cos\theta_0} + \frac{1}{\cos\theta} = \frac{1}{\mu_0} + \frac{1}{\mu},$$

# Physics-based modeling approach

## Assumptions:

- plane-parallel geometry (1D RT is OK)
- optical thick clouds (asymptotic/diffusion theory)
- dark surface (water)
- use exponential pressure profile, as needed:

$$p(z) = p(0)\exp[-z/H_{\text{mol}}] \rightarrow \tau_{\text{O}_2}(\lambda; z) = \tau_{\text{O}_2}(\lambda; 0)\exp[-z/H_{\text{mol}}]$$

- use “effective” O<sub>2</sub> optical depth for each band ...  
 ... i.e., it's OK to use  $\tau_{\text{O}_2}^{(\Delta\lambda)}(M)$  here

$$-\log r_\lambda(\Omega; \Omega_0, \tau_c) \approx (1/\mu_0 + 1/\mu) \overbrace{\tau_{\text{O}_2}(\lambda; z_{\text{top}})} + (\mu + \mu_0) \underbrace{[\tau_{\text{O}_2}(\lambda; z)]_{z_{\text{top}}-H}^{z_{\text{top}}}} \times (1 + \underbrace{C(\tau_c, g, \mu_0)})$$

DOAS ratios:

$I_{\text{abs},\lambda}/I_{\text{ref},\lambda}$ ,  $\lambda = \text{A-,B-bands}$

... and here

pre-asymptotic  
correction term

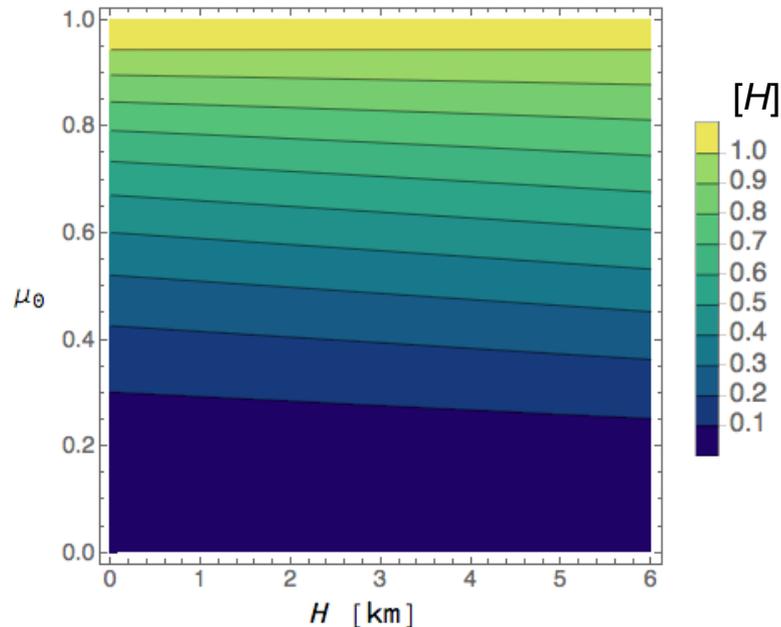
(Differential Optical Absorption Spectroscopy)

# Physical insights ...

retrieval bias from ignoring in-cloud path length

$$z_{\text{top}}^{(\text{app})} \approx z_{\text{top}} - \overbrace{H_{\text{mol}} \log \left[ 1 + \mu_0 \mu \left( e^{H/H_{\text{mol}}} - 1 \right) \times \left( 1 + C(\tau_c, g, \mu_0) \right) \right]}^{\text{retrieval bias from ignoring in-cloud path length}}$$
$$\approx -\mu_0 \mu H \times \left( 1 + C(\tau_c, g, \mu_0) \right)$$

Normalized CTH bias  $\left( z_{\text{top}} - z_{\text{top}}^{(\text{app})} \right) / H$  from (19) for EPIC  
( $\mu_0 = \mu$ ), and  $H_{\text{mol}} = 8$  km.

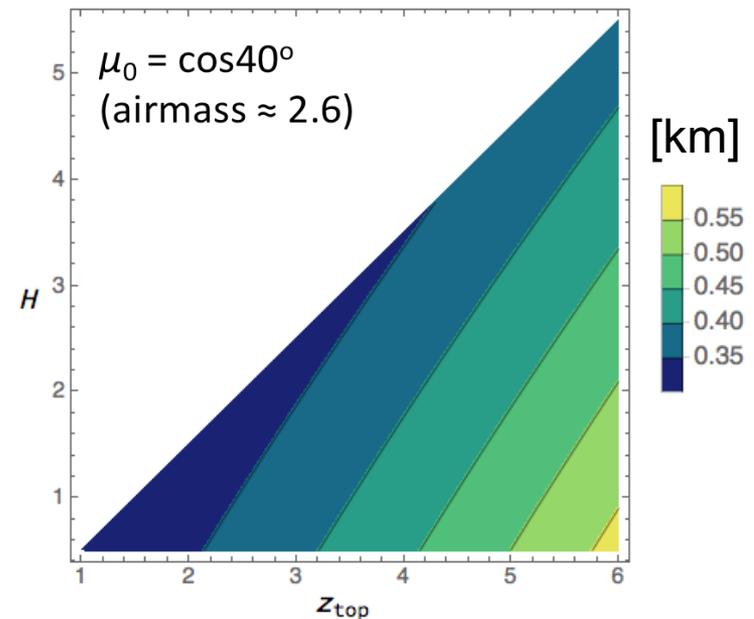
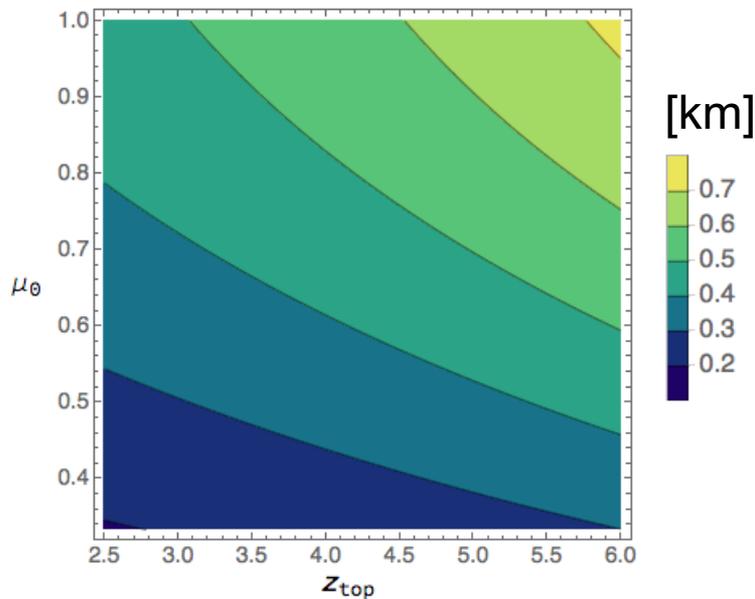


# Physical insights ...

random retrieval uncertainty resulting from sensor error, assuming 1.5% on DOAS ratio

$$\Delta z_{\text{top}}^{(\text{app})} \approx \frac{0.015 H_{\text{mol}}}{\tau_{\text{O}_2}^{(\Delta\lambda)}} \times \frac{e^{z_{\text{top}}/H_{\text{mol}}}}{2\mu_0 (\mu_0^{-2} + e^{H/H_{\text{mol}}} - 1)}$$

$$H = 2 \text{ km}$$



# Statistical (optimal estimation) approach

## Bayes' theorem:

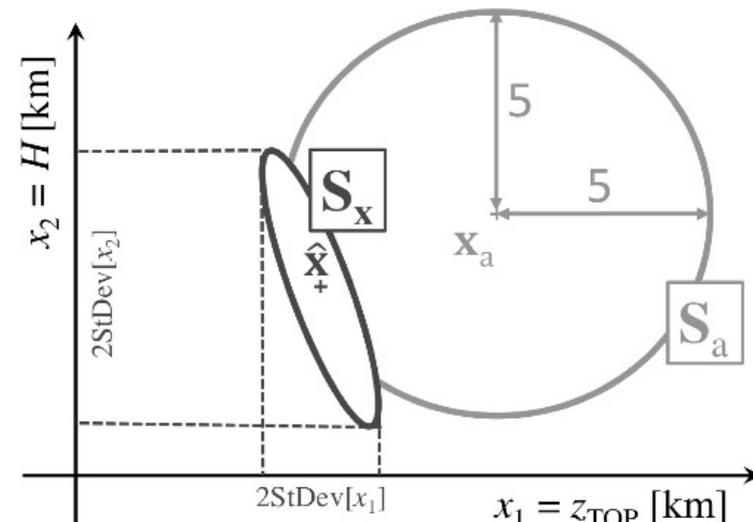
PDF of total cost function =

PDF of forward model prediction error on observations  $\mathbf{y}$   
 $\times$  PDF of prior uncertainty on state vector  $\mathbf{x}$

likelihood of  $\mathbf{y}$ , given  $\mathbf{x}$     prior uncertainty on  $\mathbf{x}$

$$\underbrace{p(\mathbf{x}|\mathbf{y})}_{\text{posterior uncertainty on } \mathbf{x}, \text{ given } \mathbf{y}} = \underbrace{p(\mathbf{y}|\mathbf{x})}_{\text{likelihood of } \mathbf{y}, \text{ given } \mathbf{x}} \underbrace{p(\mathbf{x})}_{\text{prior uncertainty on } \mathbf{x}} / \underbrace{p(\mathbf{y})}_{\text{unimportant}}$$

sensor  $\downarrow$     model  $\downarrow$   
errors in  $\mathbf{y}$  space:  $\mathbf{S}_y = \mathbf{S}_\varepsilon + \mathbf{S}_b$





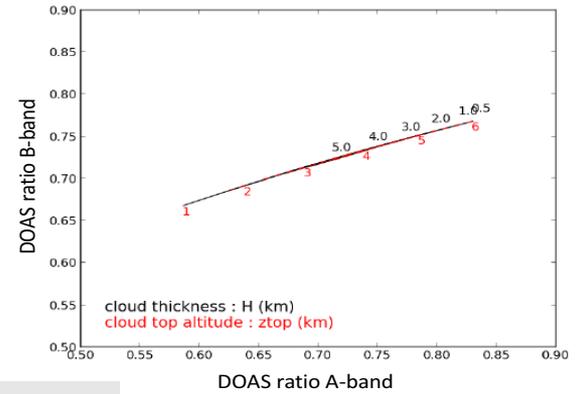
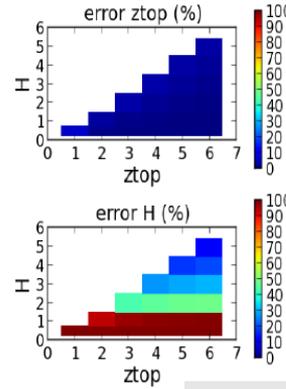
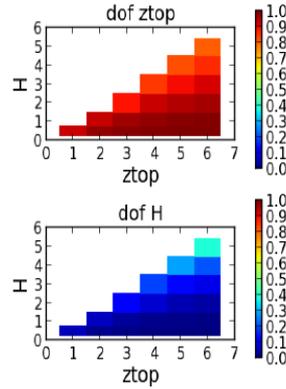
# Posterior uncertainty of the retrieved cloud properties

$0 \leq \text{pDOF}_i \leq 1$ , for  $i = 1, 2$  ( $z_{\text{top}}, H$ ) and  $\sigma_{pi} = \sigma_{ai}(1 - \text{DOF}_i)^{1/2}$  for the standard deviations.

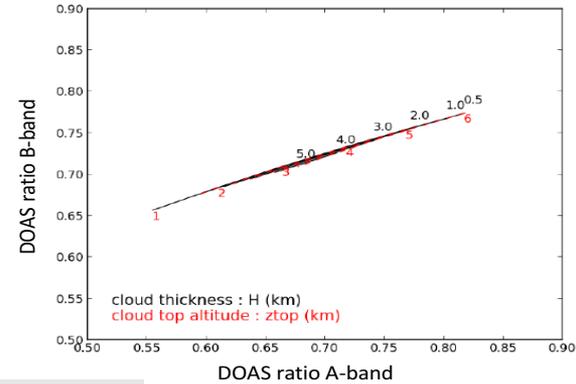
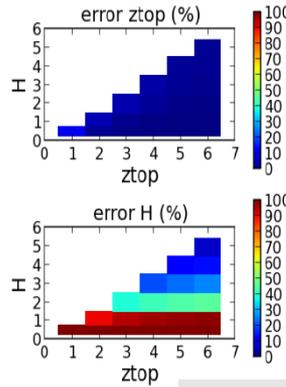
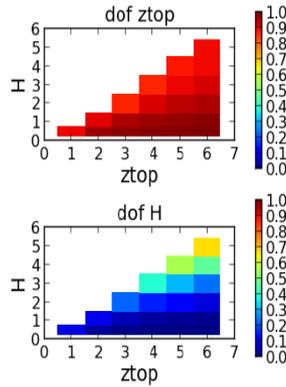
Errors expressed here in %:  $100 \sigma_i/x_i$

Nakajima/King-type plots of DOAS ratios

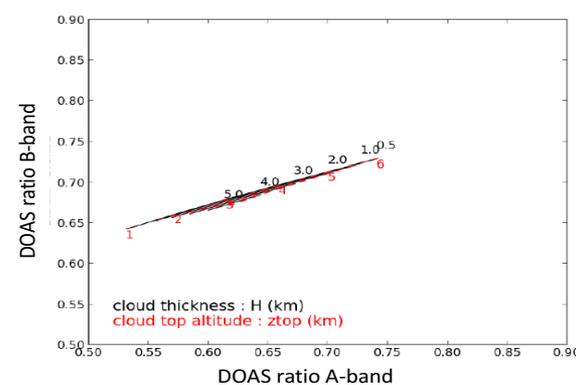
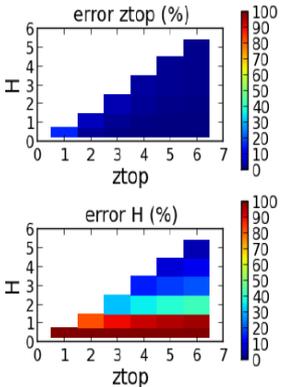
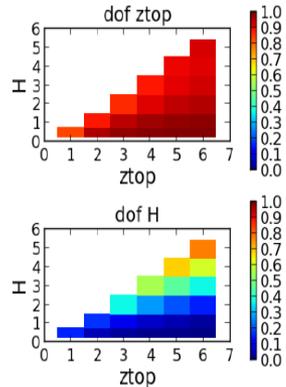
$\tau_c = 1, \alpha = 0$



$\tau_c = 32, \alpha = 0$



$\tau_c = 32, \alpha = 0.8$



# Conflict with Yang et al. (2013)?

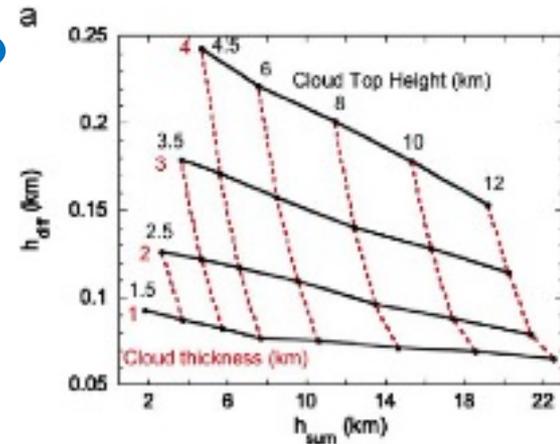
cloud optical thickness (COT)  $\tau_c = 30$

→ not really ...

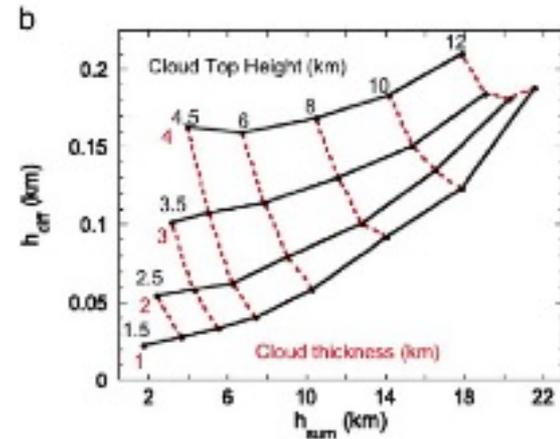
ECH: “effective” cloud height

$$h_{\text{diff, sum}} = \text{CEH}_A \pm \text{CEH}_B$$

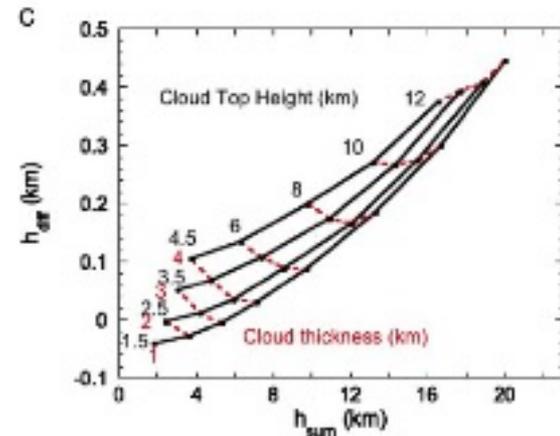
Factoring in sensor error,  
uncertainty on  $h_{\text{diff}}$  is  $\sim$  vertical axes  
uncertainty on  $h_{\text{sum}}$  is  $\ll$  horizontal axes



$\tau_c = 10$

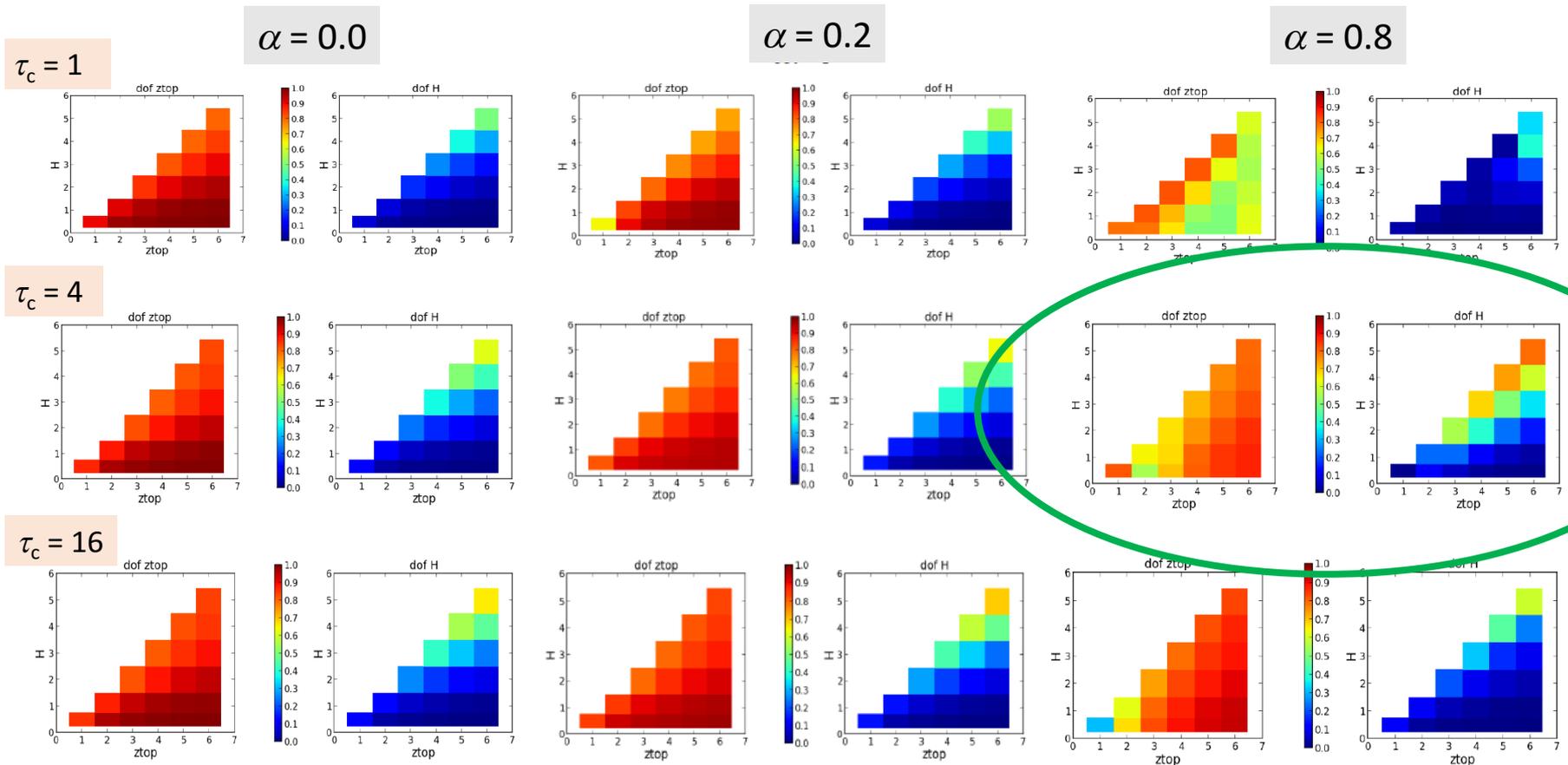


$\tau_c = 5$



# A sweet spot for $z_{\text{top}}$ and $H$ retrieval?

→ moderately opaque clouds over bright surfaces ...



... a possible application to arctic clouds?

# Summary/outlook

- **Optimal estimation approach**
  - computational (exact) 1D RT model
  - Rodgers' [2000] statistical formalism
- **Physics-based approach**
  - analytical (but approximate) 1D RT model
  - physical insights about biases and sensitivities
- **Both approaches ...**
  - use derivatives of signals w.r.t. cloud properties (a.k.a. Jacobians)
  - account for sensor noise
  - lead to the conclusion that, under most circumstances, only the cloud top height can be reliably retrieved