

Effects of cloud variability on TROPOMI molecular and cloud property products

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- 1 Motivation
- 2 Inside-pixel variability effects
- 3 Impact on cloud property retrieval
- 4 Impact on ozone retrieval
- 5 Conclusions

Motivation

- The design of spectrometers aboard spaceborne platforms:
 - ▶ trade-off between spectral and spatial resolutions
 - ▶ guarantee of high signal-to-noise ratios
- Land/ocean optical imagers favor spatial over spectral resolution
 - ▶ designed to highly resolve surface properties
- Atmospheric composition sensors favor spectral over spatial resolution
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 - ▶ aerosol plumes
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Karen L. Nyberg @AstroKarenN

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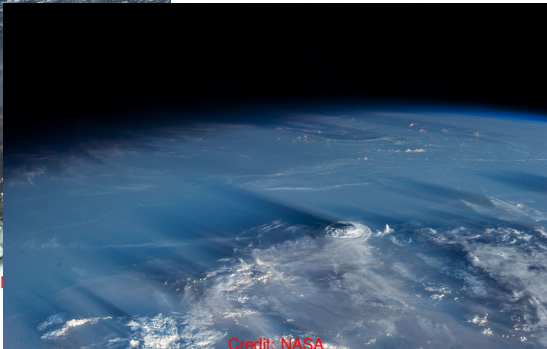
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Credit: NASA

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- Models do not account for photon horizontal transport
 - ▶ 1-dimensional (1D) horizontally homogeneous radiative transfer (RT) models
- variability and 3D RT effects:
 - ▶ bias modelled radiances
 - ▶ bias retrieved atmospheric products
 - ▶ ... but, by how much?

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Inside-pixel variability effects – Components

- heterogeneous cloud field
 - ▶ PaLM – A PARallelized Large-Eddy Simulation Model for Atmospheric and Oceanic Flows [Heinze et al. (2012)]
 - ▶ $6.4 \times 6.4 \text{ km}^2$ domain
 - ▶ $10 \times 10 \times 10 \text{ m}^3$ cells of liquid water content (LWC)

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⇒ radiance spectrum of a heterogeneous scene as seen by TROPOMI

Inside-pixel variability effects – Study

Goal

- Study the impact of non-resolved in-scene variability on:
 - ▶ simulated radiance spectra
 - ▶ molecular and cloud products

Design

- Select a high-resolved $10 \times 10 \times 10 \text{ m}^3$ PaLM cumulus field embedded in a $6.4 \times 6.4 \text{ km}^2$ domain
- Coarsen n -fold the original 3D LWC field at 9 different horizontal resolutions:
 - ▶ $10 \times 10 \text{ m}^2$, $20 \times 20 \text{ m}^2$, $40 \times 40 \text{ m}^2$, $80 \times 80 \text{ m}^2$, $160 \times 160 \text{ m}^2$, $320 \times 320 \text{ m}^2$, $640 \times 640 \text{ m}^2$, $1280 \times 1280 \text{ m}^2$, $6400 \times 6400 \text{ m}^2$
- Calculate spectra at each spatial resolution by
 - ▶ averaging over the whole spatial pixel domain
 - ▶ convolving the LBL RT calculations with the TROPOMI ILS

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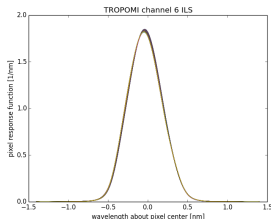
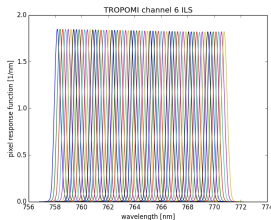
cloud property retrieval – Setup

- Observer at TOA = 60 km
- Viewing zenith angle: 15°
- Solar zenith angle: 0° , 30° , 60°
- Relative azimuth angle: 45°
- Surface height: 0 km
- Surface albedo = 0.1
- US standard atmosphere
- O₂ LBL absorption coefficients
- Rayleigh scattering
- Cloud PSD:
 - ▶ Gamma: $N_c=100$, $\alpha=7$
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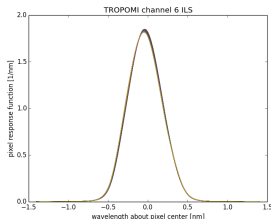
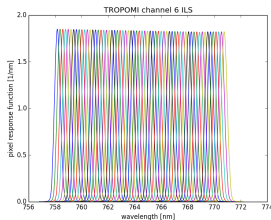
- TROPOMI channel 6 spectral sampling and resolution



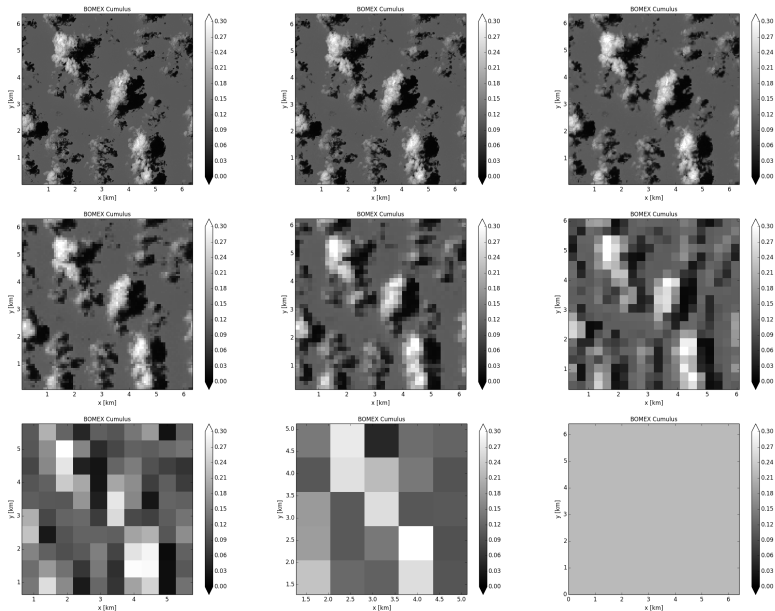
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- Actually, GOME2 ch4 spectral sampling and resolution

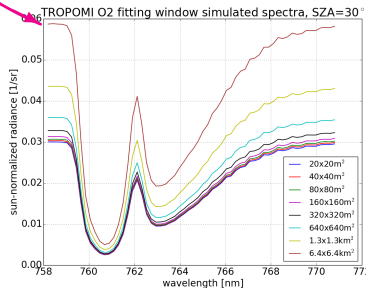
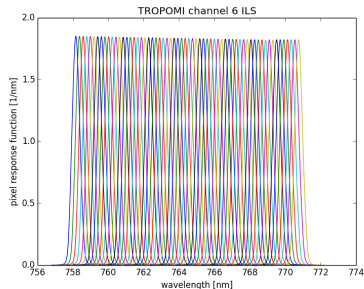
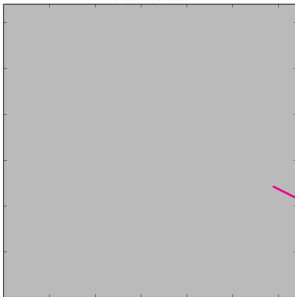


Spatially resolved radiances @ 758 nm

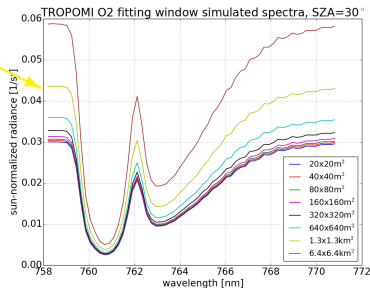
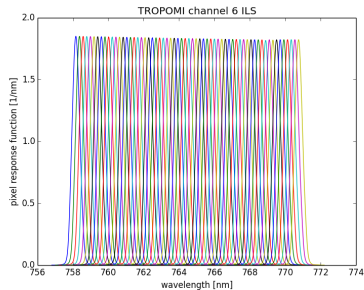
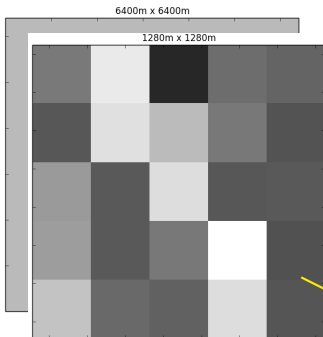


O2 A-band spectra

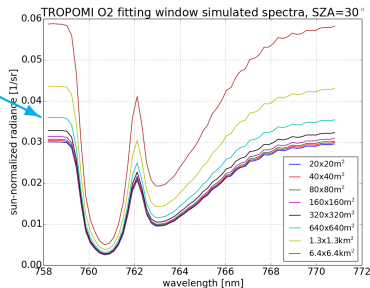
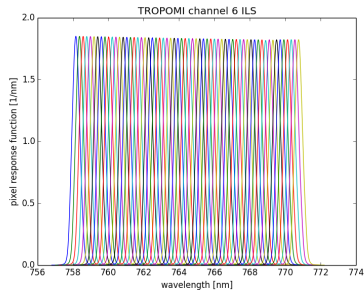
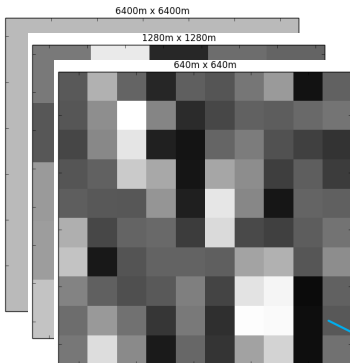
6400m x 6400m



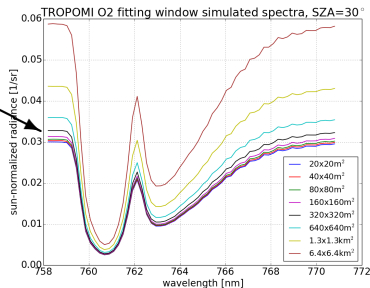
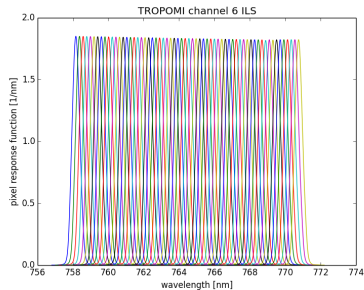
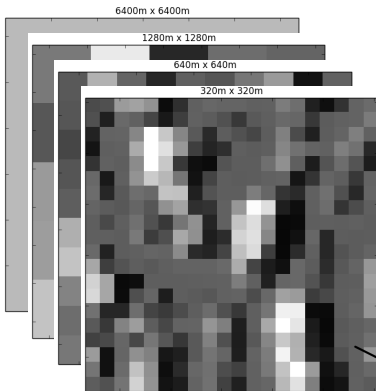
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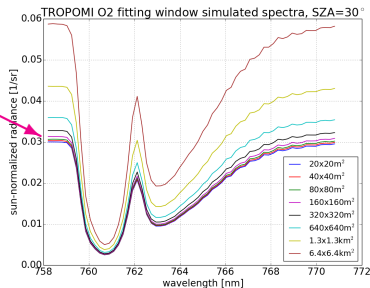
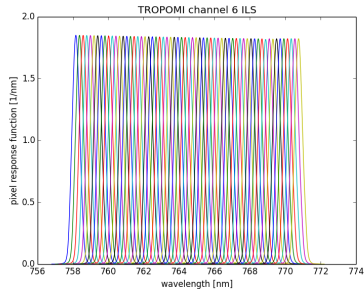
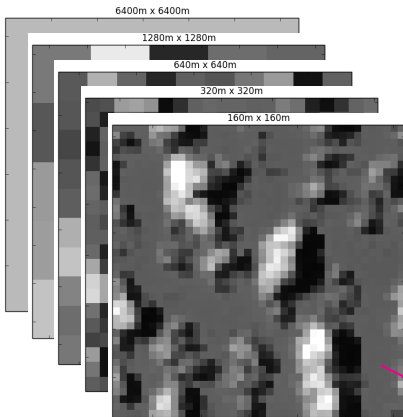
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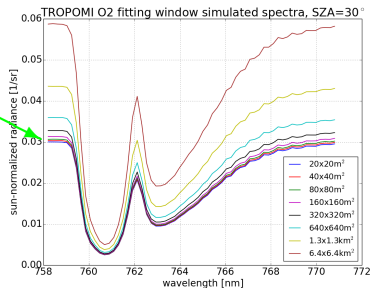
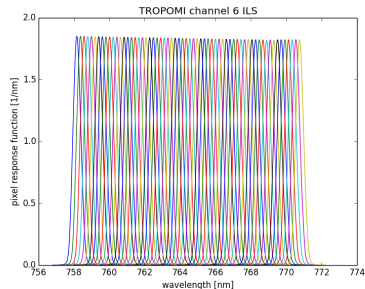
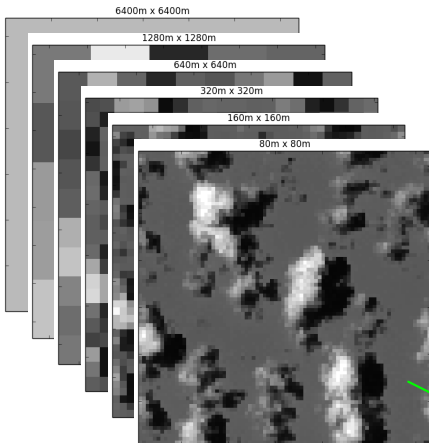
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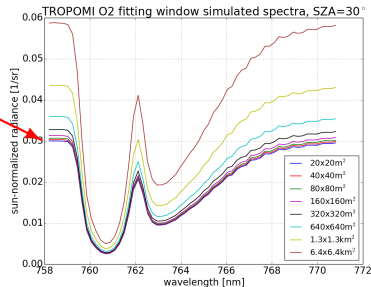
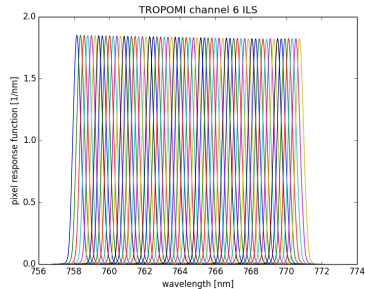
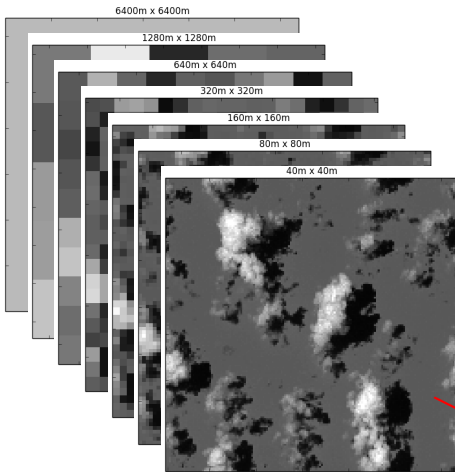
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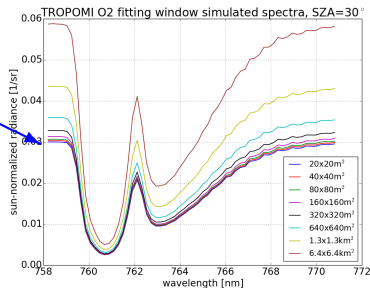
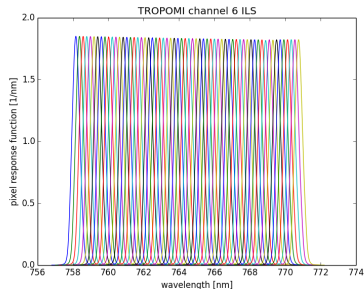
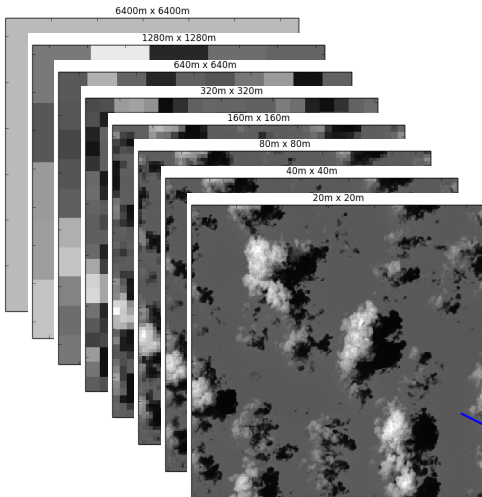
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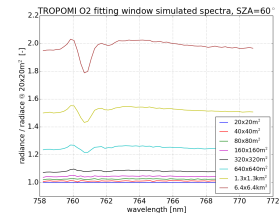
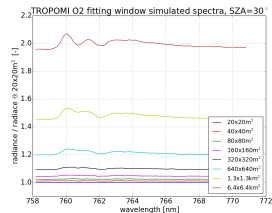
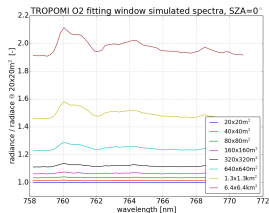
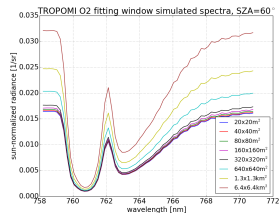
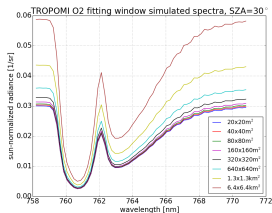
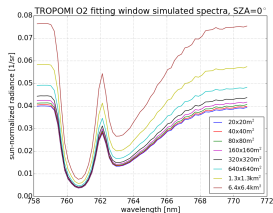
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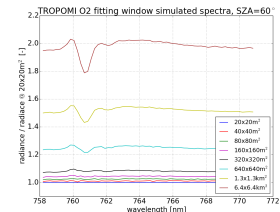
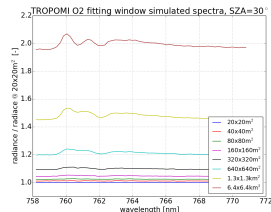
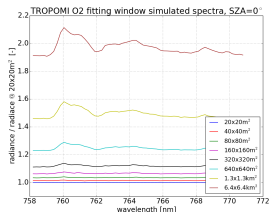
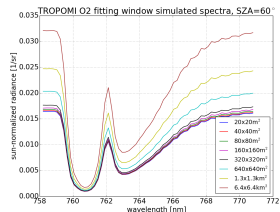
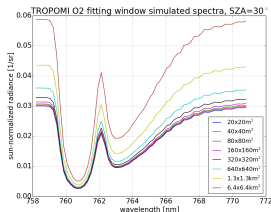
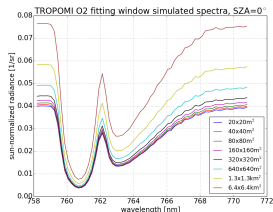
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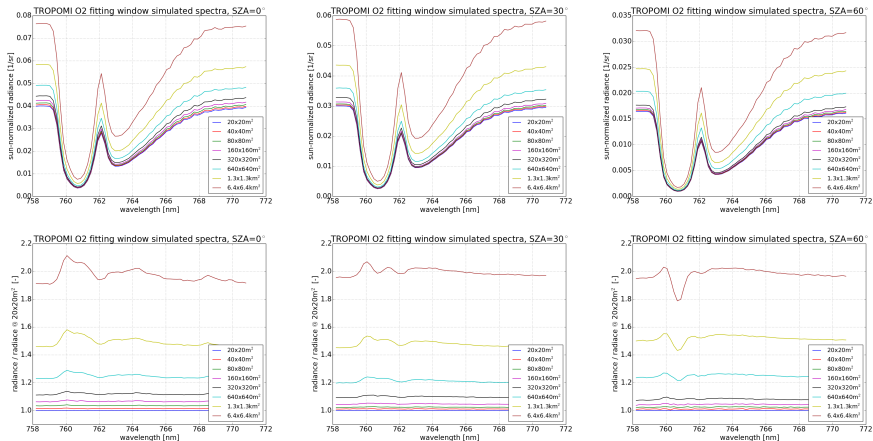


O2 A-band spectra



- Coarser cloud fields bias the radiances to higher values

O2 A-band spectra



- Coarser cloud fields bias the radiances to higher values
- The spectral shape also depends on spatial resolution

Cloud property retrieval – Forward model

- Independent pixel approximation:
 - ▶ cloudy and clear sky contributions
- Clouds and surface treated as Lambertian reflectors
- Rayleigh scattering
- O₂ absorption
- geometrical cloud fraction
 - ▶ portion of vertical columns where cloud optical thickness is greater than 0

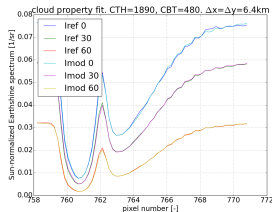
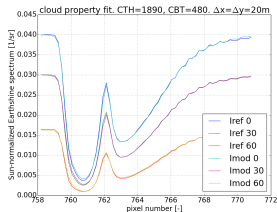
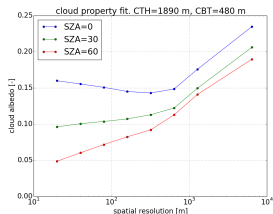
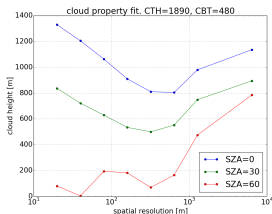
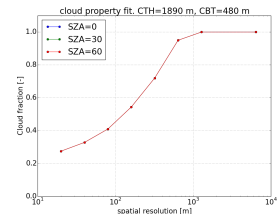
$$I(\textcolor{red}{CH}, \textcolor{red}{CA}, CF) = CF I_{cld}(CH, CA) + (1 - CF) I_{clr}(SH, SA)$$

CF: Cloud fraction

CH: Cloud height, CA: Cloud albedo

SH: Surface height, SA: Surface albedo

Cloud property retrieval – Results



- Overall good convergence, independently of spatial resolution
- CA increases for decreasing spatial resolution
- CH well below geometrical cloud top height
 - ▶ for high solar inclination ($SZA=60^\circ$), CH below cloud bottom height

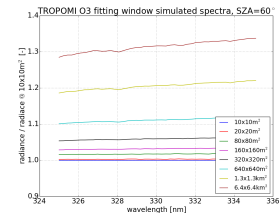
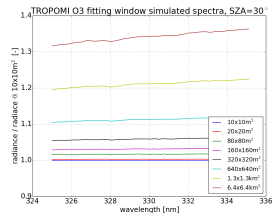
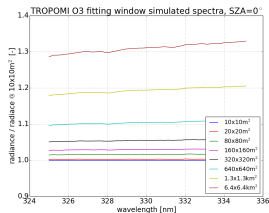
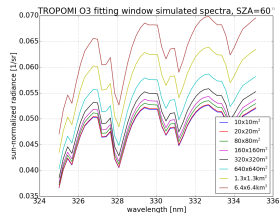
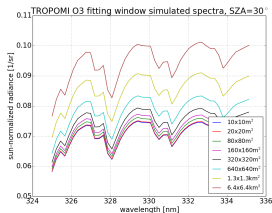
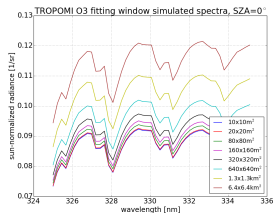
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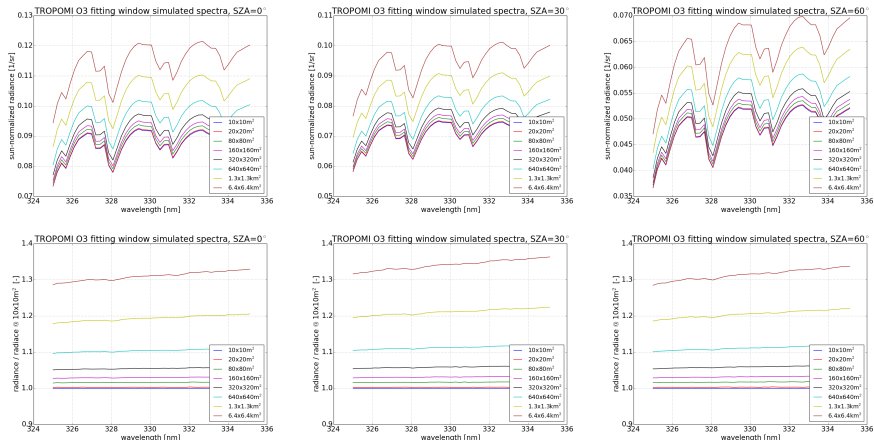
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- Relative azimuth angle: 45°
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- Surface albedo = 0.1
- US standard atmosphere
- O₃ cross sections [Brion]
- NO₂ cross sections [Vandaele]
- CSs interpolated at all atm. levels
- Rayleigh scattering
- No Raman scattering
- Cloud PSD:
 - ▶ Gamma: $N_c=100$, $\alpha=7$
- Mie scattering
 - ▶ full cloud phase function
- TROPOMI channel 3:
 - ▶ Gaussian slit function
 - ▶ Spectral sampling: 0.22 nm
 - ▶ Spectral resolution (FWHM): 0.54 nm

Ozone fitting window spectra

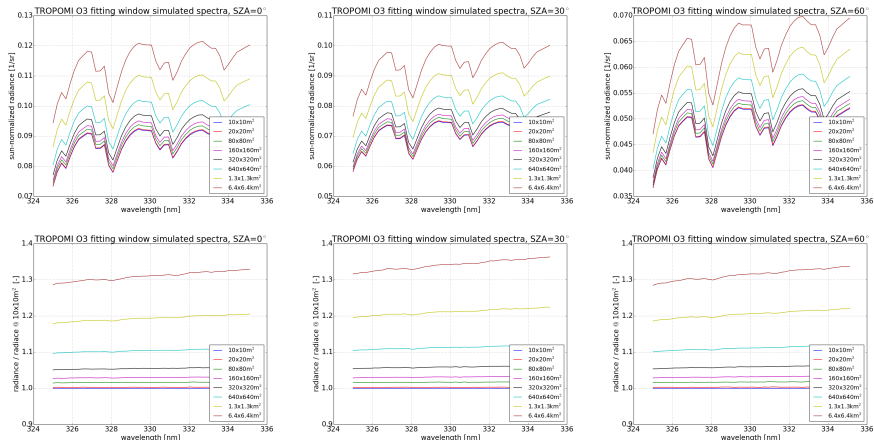


Ozone fitting window spectra



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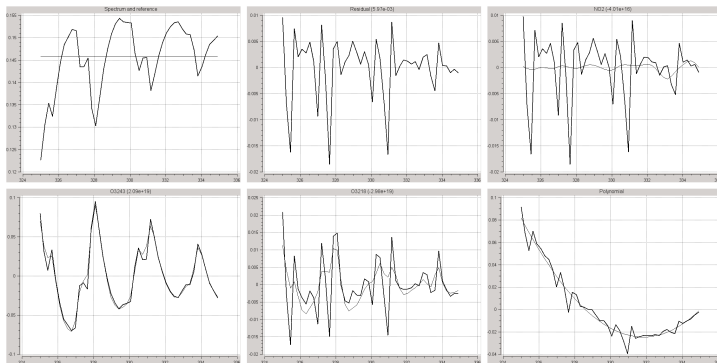
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- The spectral shape **slightly** depends on spatial resolution

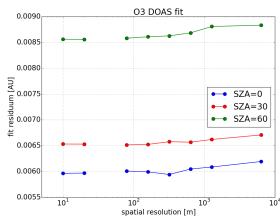
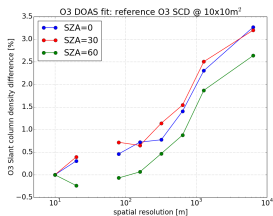
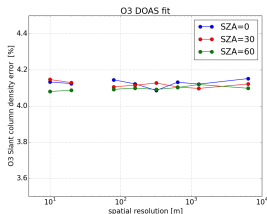
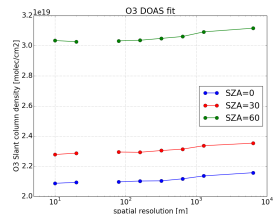
Ozone retrieval

- Differential optical absorption spectroscopy (DOAS)
- QDOAS software package developed @ BIRA
- Results presented as O3 slant column densities (SCD)



- Overall good convergence, independently of spatial resolution and illumination
- low residua

Ozone retrieval – Results



- Ozone SCD increases for decreasing spatial resolution
- Impact of scene heterogeneity on ozone SCDs is relatively small, below the fit error (4%)
- However, the effect can be remarkable for tropospheric ozone

Outline

- 1 Motivation
- 2 Inside-pixel variability effects
- 3 Impact on cloud property retrieval
- 4 Impact on ozone retrieval
- 5 Conclusions

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 - ★ minimum at resolution of some hundreds of meters, but difficult to generalize the results

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- ▶ It can be remarkable for tropospheric ozone
- ▶ We expect higher inhomogeneous clouds to have a larger effect on ozone
- ▶ We also expect the heterogeneity effects to be larger for tropospheric gases: H₂O, CO, CH₄, . . .

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