

# Cloud Retrieval for Passive Sensors in the UV-VIS-NIR for the Polar Sentinel-5P, Geostationary Sentinel-4 and Deep Space DSCOVR Missions

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Wissen für Morgen



## Motivation

-

**trace gas retrievals need cloud information**



# Missions



# Sentinel-5 Precursor and Sentinel-4

Sentinel-5 Precursor



## Orbit

sun-synchronous polar / geostationary

## Temporal resolution and coverage

daily global / hourly Europe

## Instrument name

TROPOMI / UVN

## Spatial resolution

$3.5 \times 5.5 \text{ km}^2$  /  $8 \times 8 \text{ km}^2$

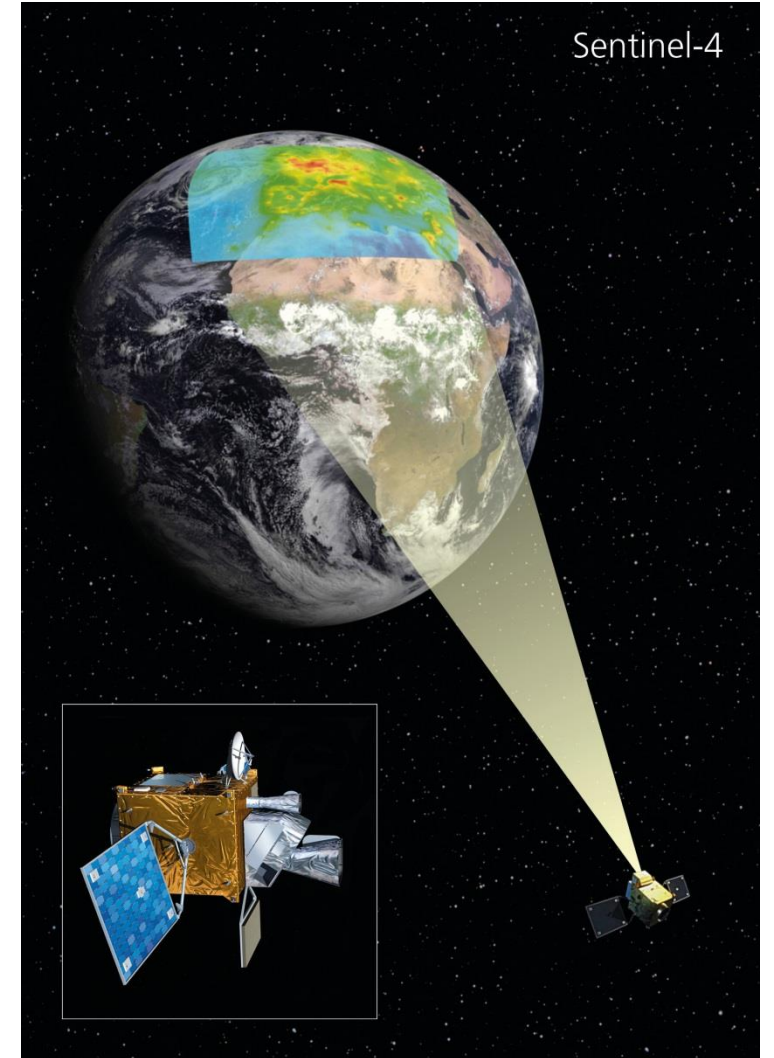
## Spectral coverage

UV-VIS-NIR-SWIR / UV-VIS-NIR

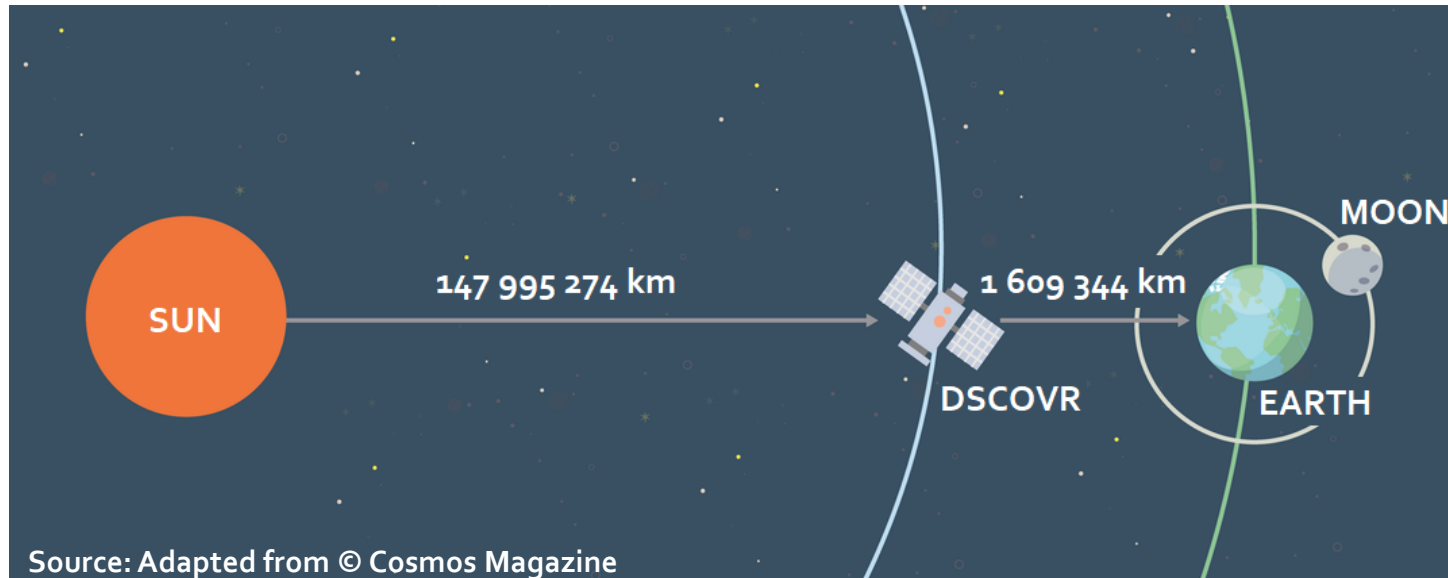
## Spectral resolution in the UVN

0.25-0.5 nm / 0.12-0.5 nm

Sentinel-4



# DSCOVOR – Deep Space Climate Observatory



Source: Adapted from © Cosmos Magazine



EPIC/DSCOVOR RGB  
images on 2015-07-16  
Source: NASA

## Orbit

Lagrange Point L1

## Temporal resolution and coverage

10-22 full disk images per day

## Instrument name

EPIC (Earth Polychromatic Imaging Camera)

## Spatial resolution

12 km at nadir

## Spectral coverage

10 channels across UV-VIS-NIR

## Spectral resolution in the UVN

bandwidth between 1-3 nm

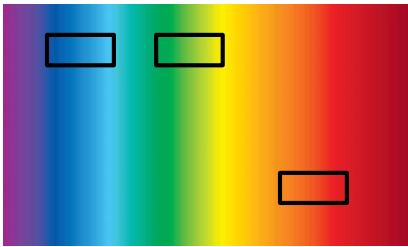


# The DLR Cloud Algorithms



# OCRA & ROCINN – Algorithm Overview

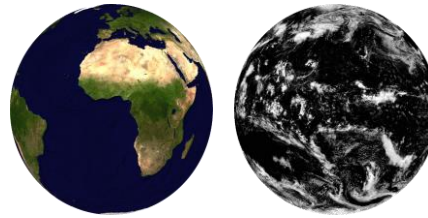
**OCRA**  
Optical Cloud  
Recognition Algorithm



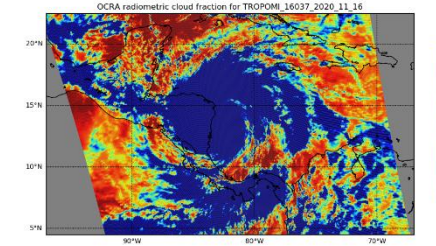
color space approach

**ROCINN**  
Retrieval of Cloud Information  
using Neural Networks

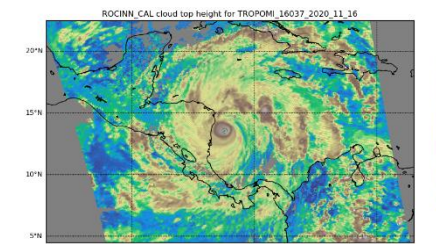
neural network approach



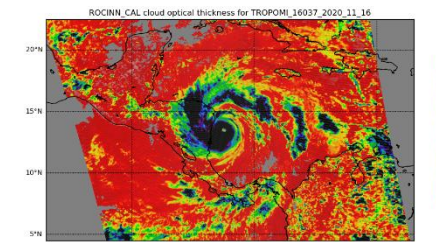
clear-sky composite



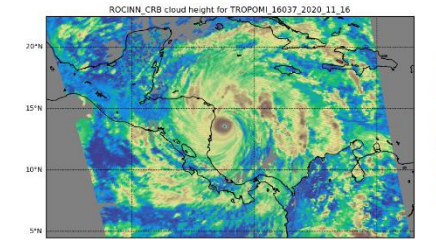
Radiometric  
cloud fraction



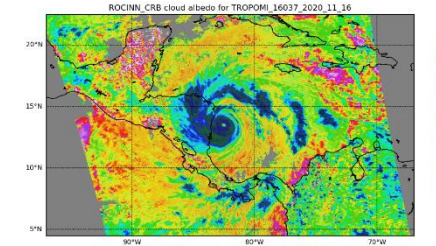
cloud top  
height



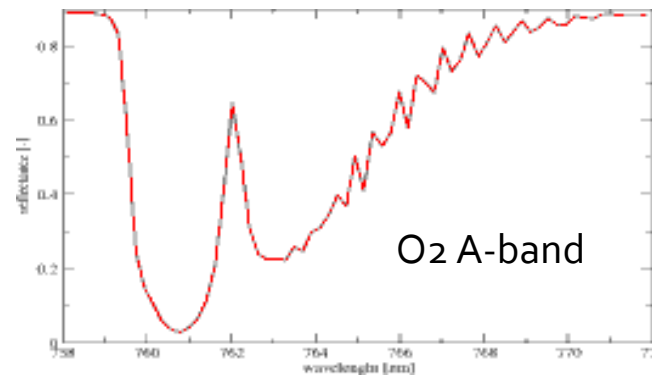
cloud opt.  
thickness



eff. cloud  
height



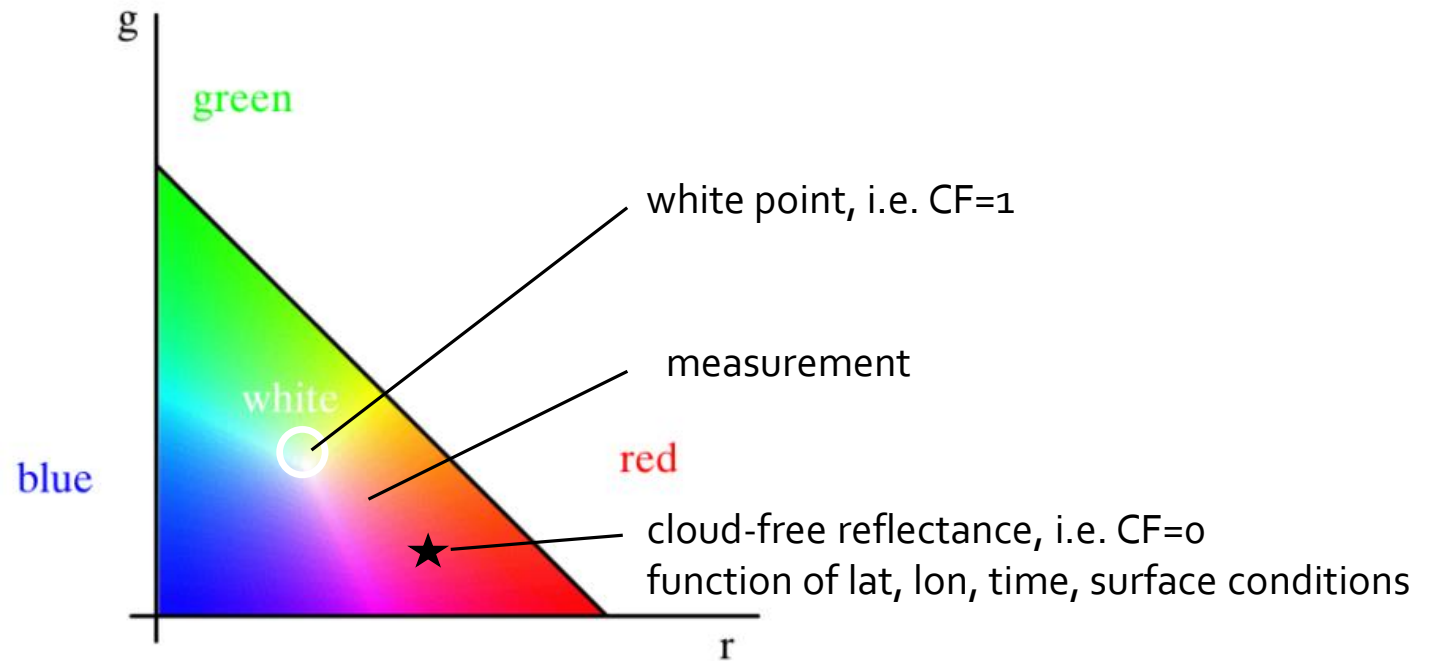
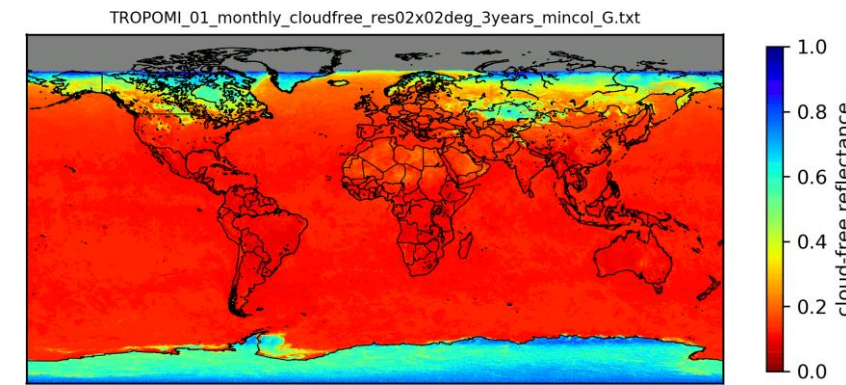
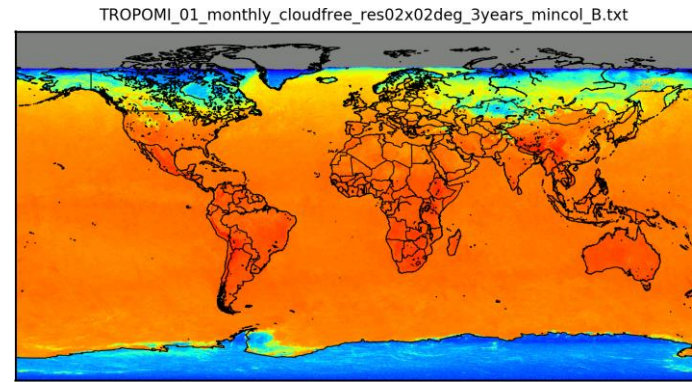
cloud albedo



Hurricane Iota  
©NASA worldview

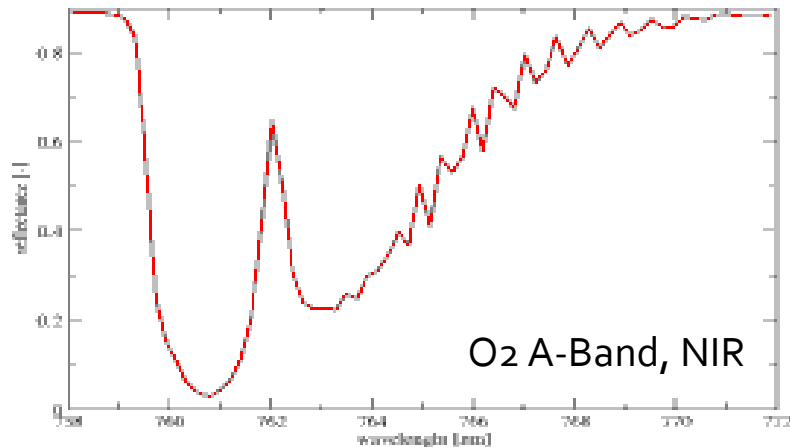
# OCRA Overview

- generate cloud-free reflectance composite maps
- map measured reflectances to RGB color space
- assume cloud to be „white“ in RGB space
- measured reflectance will be between white point and cloud-free point
- Radiometric cloud fraction scaled between cloud-free and white point



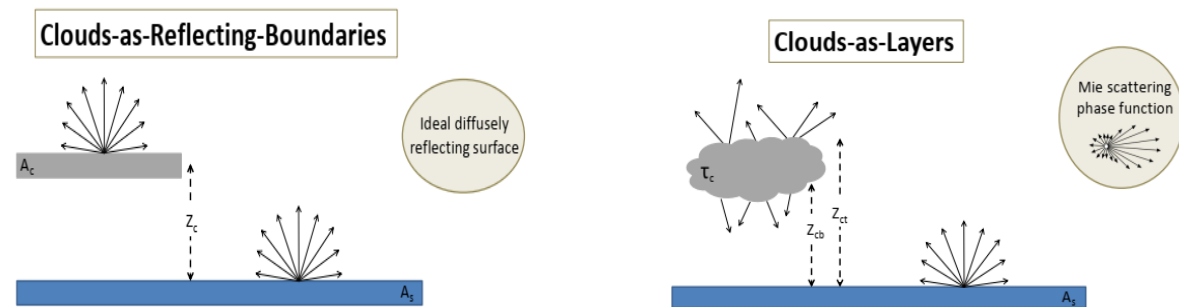


# ROCINN Overview



Fitting window: [758-771] nm

- two cloud models:
  - CRB**: clouds as reflecting boundaries (Lambertian reflector)
  - CAL**: clouds as layers (Mie-scattering liquid water droplets)



- forward model uses VLIDORT RT code
- neural network parametrized using smart sampling
- inversion using Tikhonov regularization

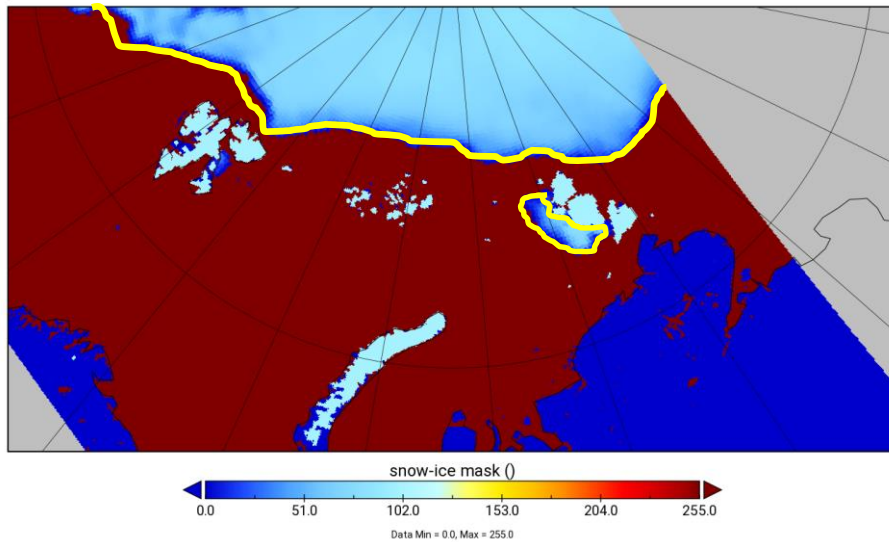
Loyola et al. (2018): The operational cloud retrieval algorithms from TROPOMI on board Sentinel-5 Precursor, *Atmos. Meas. Tech.*, 11, 409-427. <https://doi.org/10.5194/amt-11-409-2018>  
 Loyola et al. (2016): Smart sampling and incremental function learning for very large high dimensional data, *Neural Networks*, Vol. 78, 75-87. <https://doi.org/10.1016/j.neunet.2015.09.001>



# recent improvements (I)

- surface albedo climatology is replaced by daily surface albedo retrieval (GE\_LER) using TROPOMI measurements
- map is updated daily (G<sub>3</sub>\_LER), but only for those grid cells which have a cloud fraction < 0.05

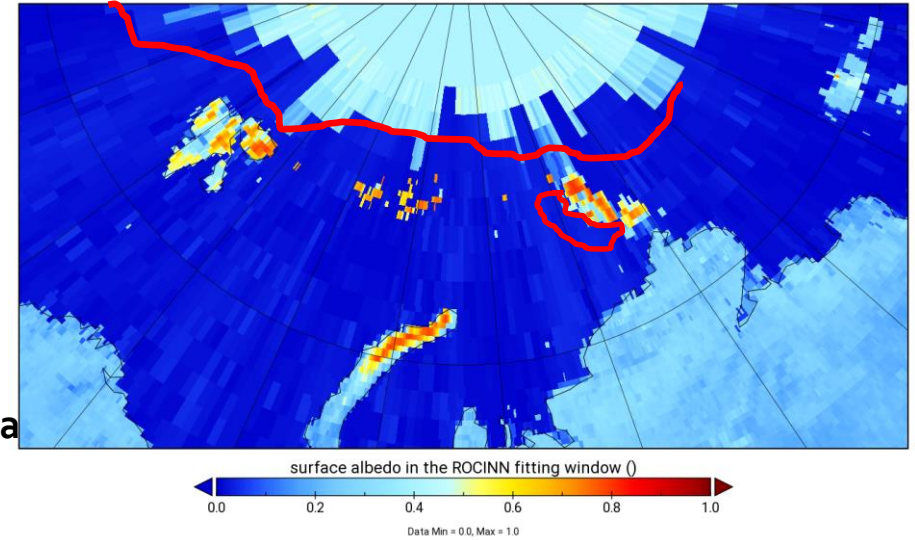
ECMWF snow-ice mask with NISE definitions, 2020-08-09, orbit 14625



## Climatology:

- does not capture the actual ice border and sea ice

MERIS climatology, August

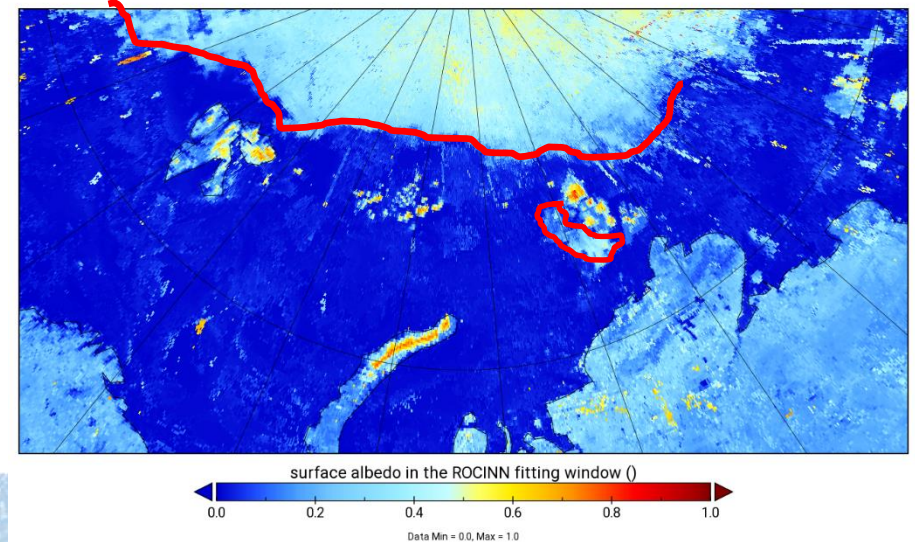


## Daily G<sub>3</sub>\_LER maps:

- better spatial resolution

- better representation of actual surface conditions

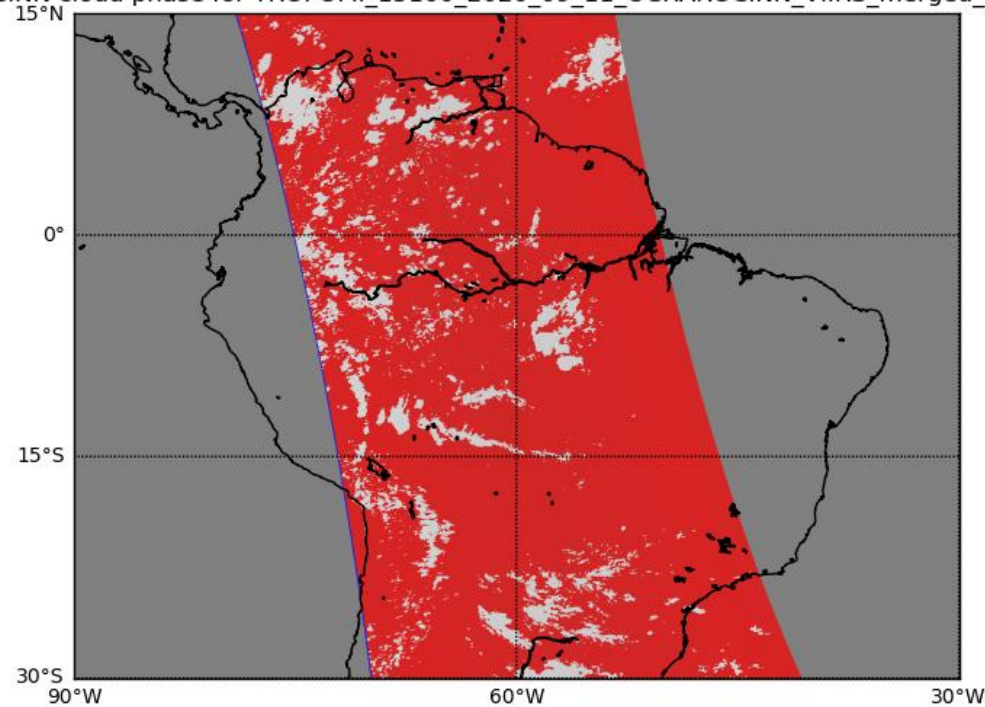
2020-08-09, G<sub>3</sub>\_LER, CLOUD\_OFFL



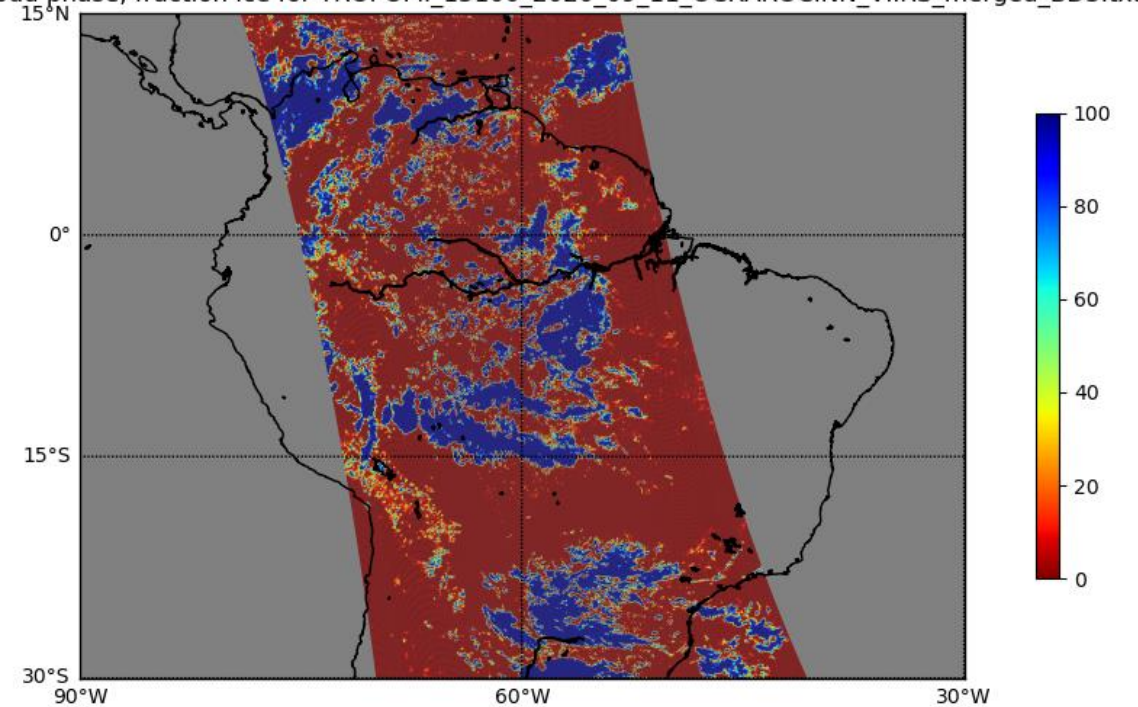
## recent improvements (II)

- addition of an ice cloud detection flag to identify scenes not parameterized yet by ROCINN (only liquid clouds)

ROCINN cloud phase for TROPOMI\_15100\_2020\_09\_11\_OCRAROCINN\_VIIRS\_merged\_VIIRS cloud phase, fraction ice for TROPOMI\_15100\_2020\_09\_11\_OCRAROCINN\_VIIRS\_merged\_BD3.txt



operational ROCINN ice cloud detection



VIIRS *cloud\_phase\_ice* fraction from the CloudPhase EDR



## recent improvements (III)

- ROCINN\_CAL: ice cloud parameterization (VLIDORT 2.8.3)

Spurr, R. J. D. VLIDORT: A linearized pseudo-spherical vector discrete ordinate radiative transfer code for forward model and retrieval studies in multilayer multiple scattering media, *JQSRT*, 102, 316-42, 10.1016/j.jqsrt.2006.05.005, 2006

Bryan Baum et al.: Ice cloud single-scattering property models with the full phase matrix at wavelengths from 0.2 to 100  $\mu\text{m}$ ., *JQSRT*, 10.1016/j.jqsrt.2014.02.029, 2014

- Bulk properties and phase function of ice crystals of an effective diameter:  $D_{\text{eff}} = 25 \mu\text{m}$  at 760 nm

- Simulated ice clouds have been retrieved with:

- ROCINN\_CAL for water clouds
  - CTH errors between 0.3 and 5 km
  - COT errors up to 90%

- ROCINN\_CAL for ice clouds
  - CTH errors up to 0.1 km
  - COT errors up to 10%

- When ice clouds are present, the **ROCINN\_CAL\_ice** implementation will considerably reduce the uncertainties

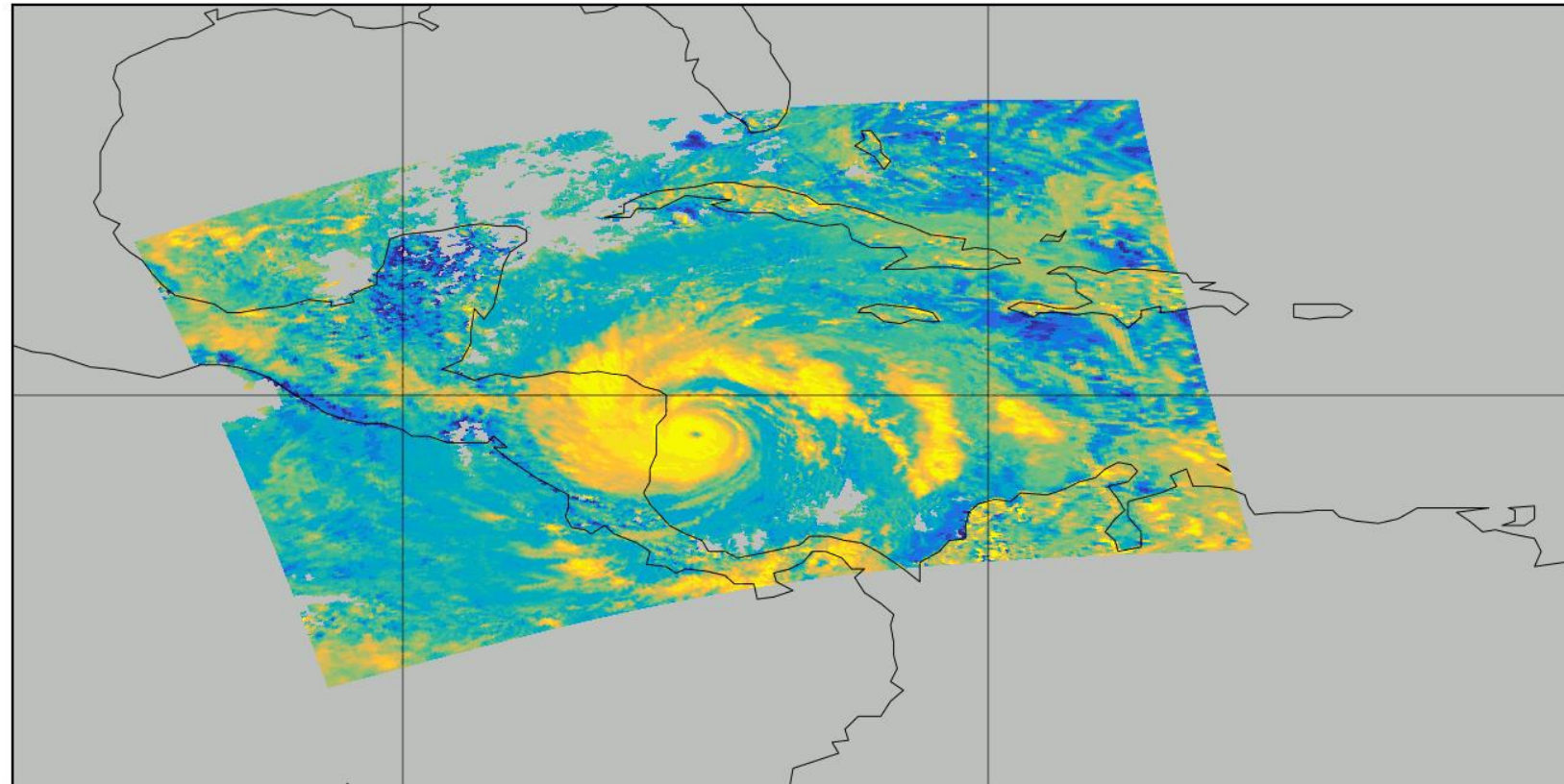


## Application Examples: S5P

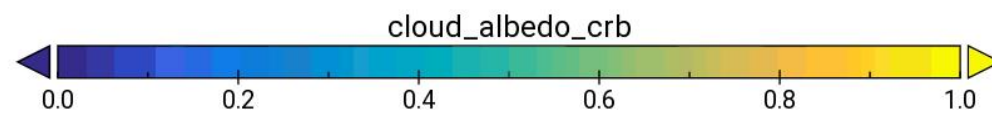


# S5P – operational cloud products

Hurricane Iota, 2020-11-16, orbit 16037

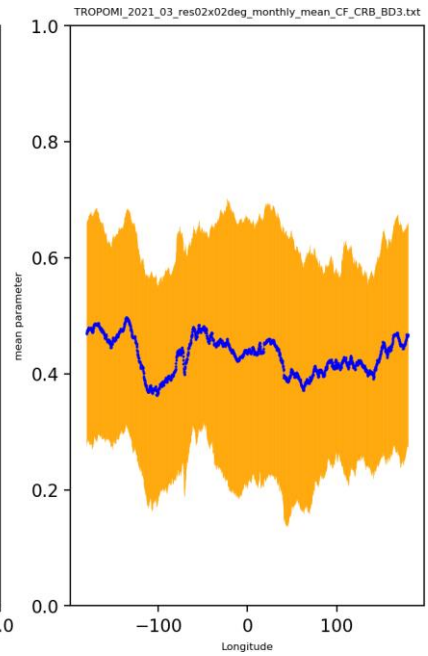
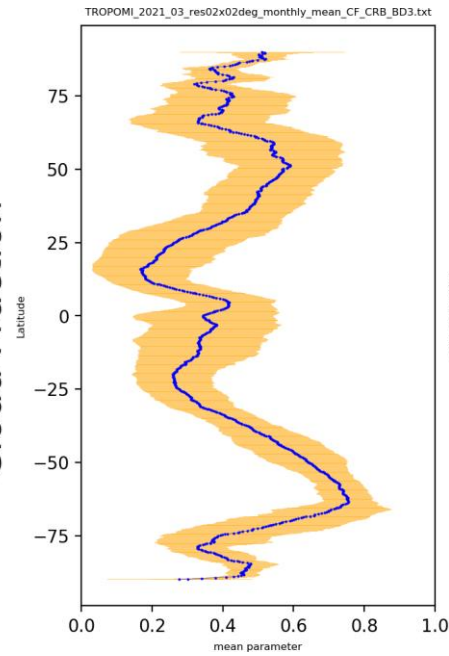
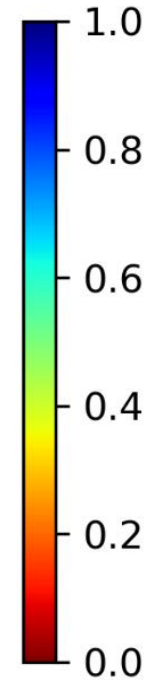
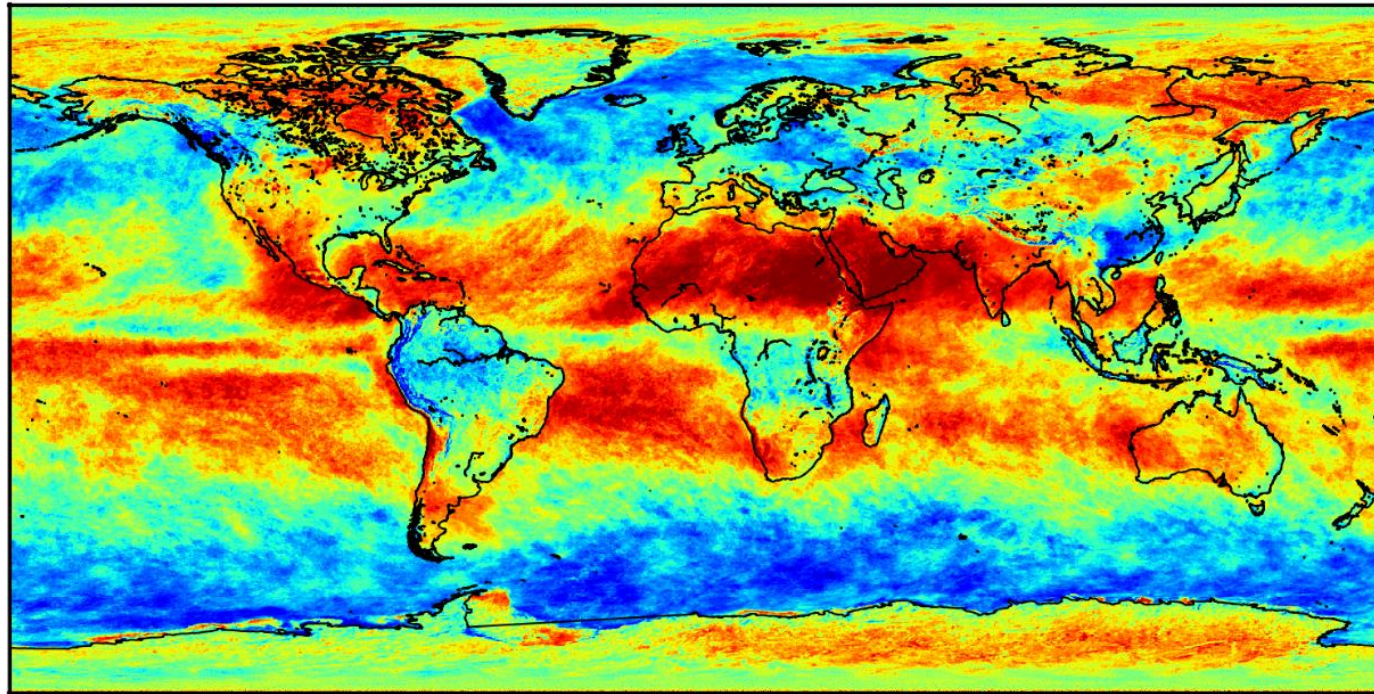


Hurricane Iota  
©NASA worldview



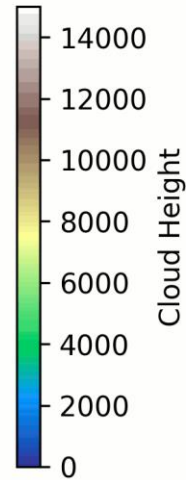
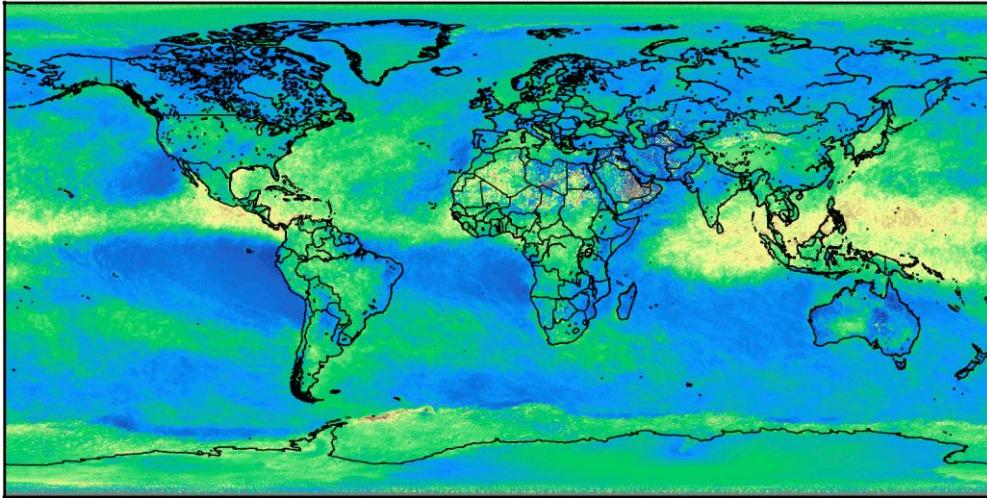
# S5P – monthly mean maps and zonal/meridional means

TROPOMI\_2021\_03\_res02x02deg\_monthly\_mean\_CF\_CRB\_BD3.txt

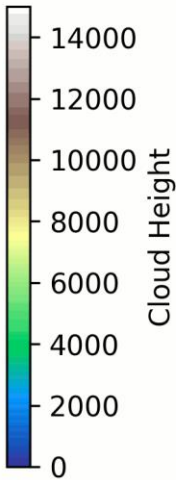
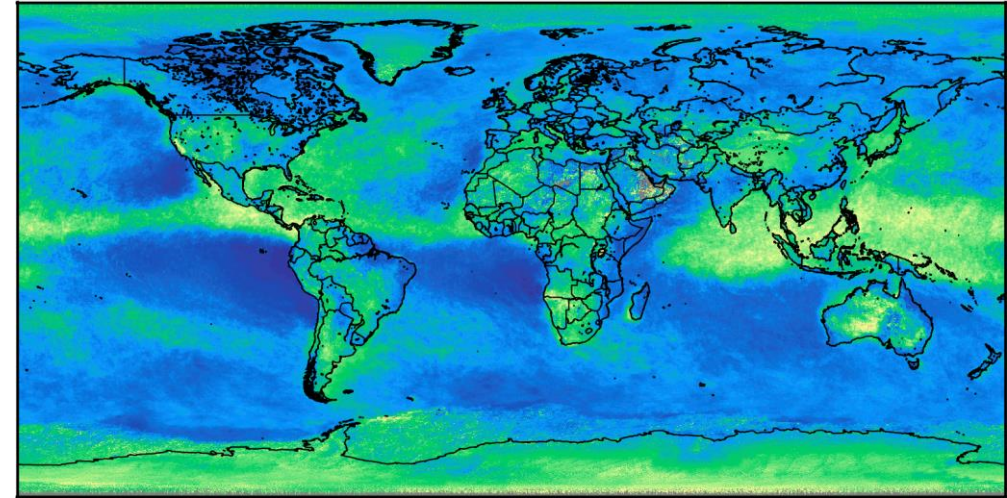


# S5P – yearly temporal evolution

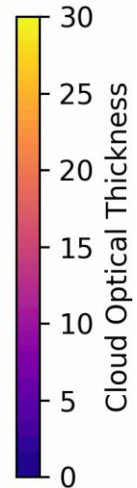
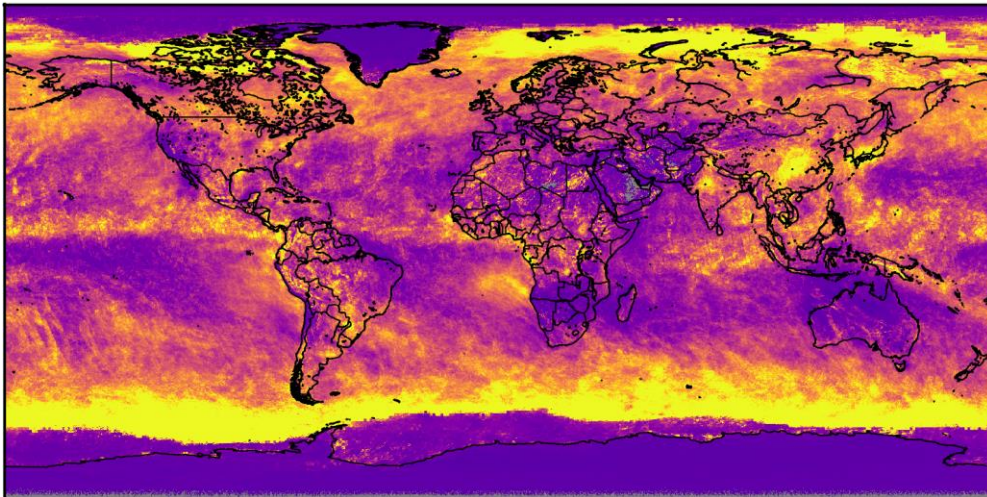
TROPOMI\_2018\_09\_res02x02deg\_monthly\_mean\_CTH\_CAL.txt



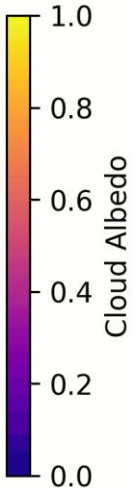
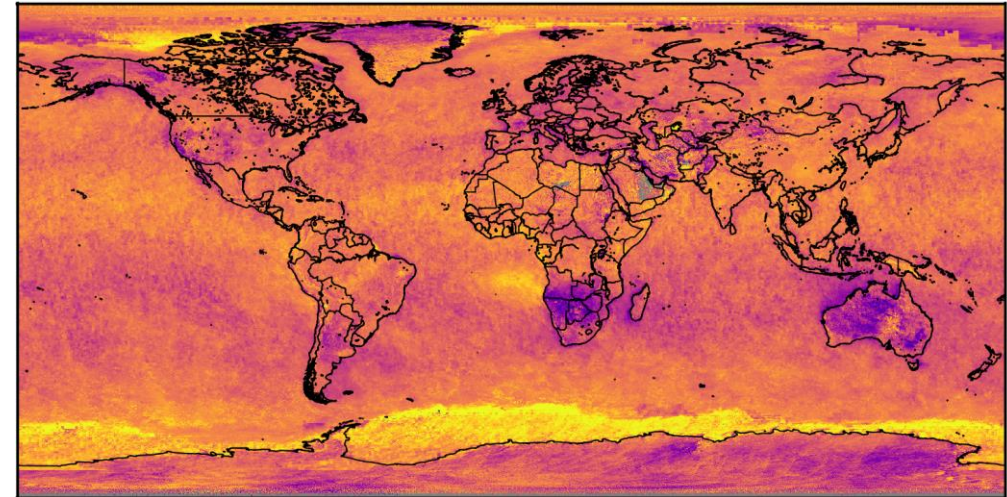
TROPOMI\_2018\_09\_res02x02deg\_monthly\_mean\_CH\_CRB.txt



TROPOMI\_2018\_09\_res02x02deg\_monthly\_mean\_COT\_CAL.txt

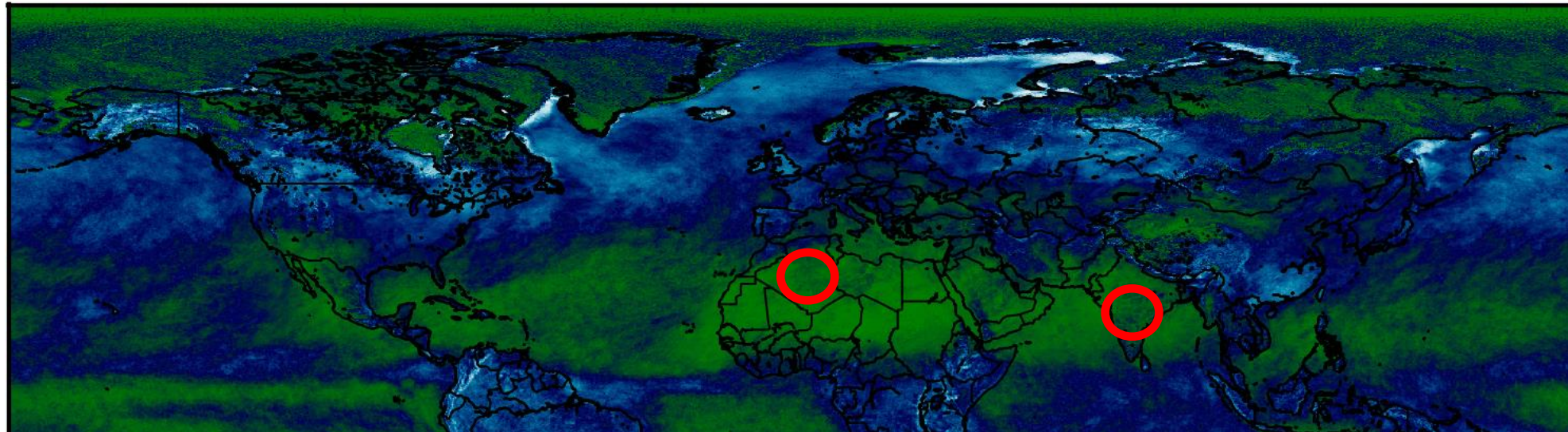


TROPOMI\_2018\_09\_res02x02deg\_monthly\_mean\_CA\_CRB.txt

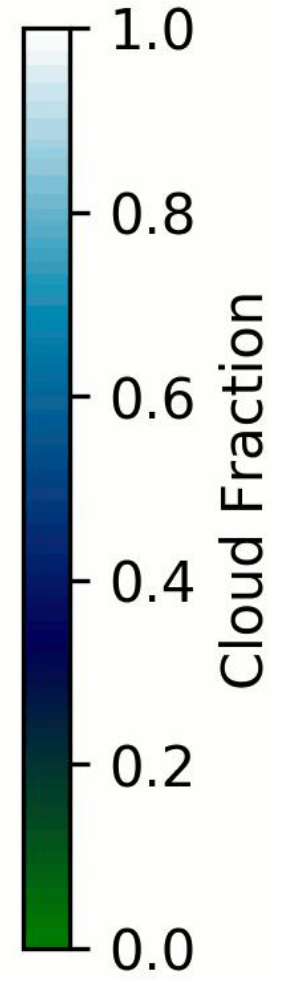
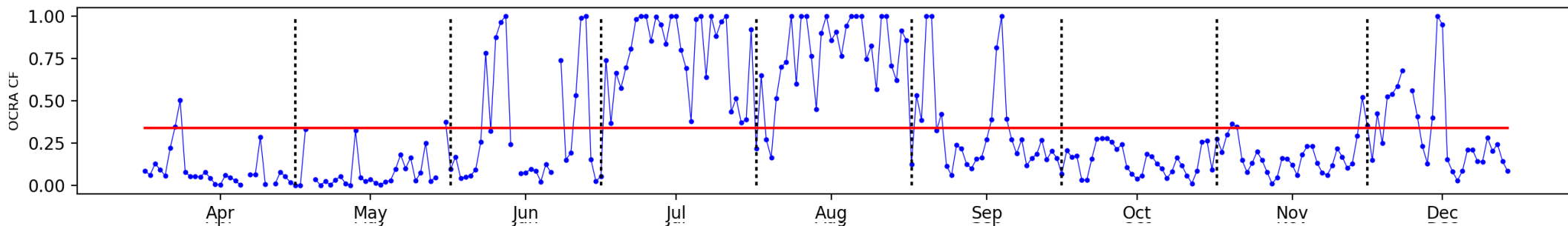




# S5P – time series for any given geolocation



India



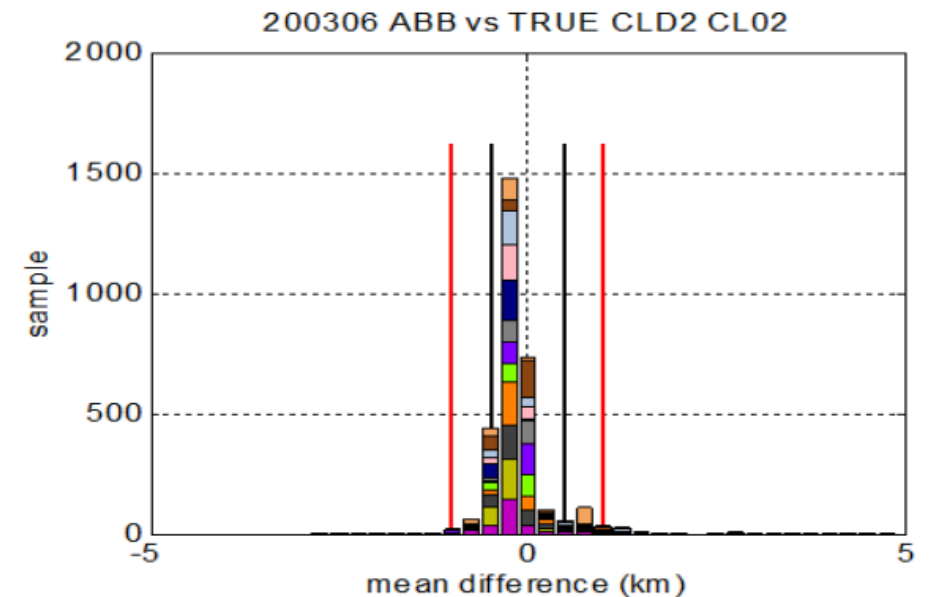
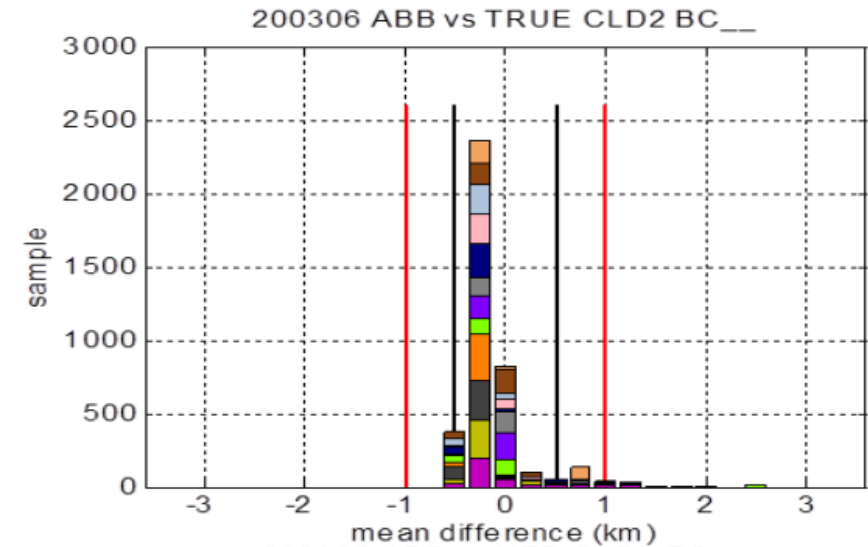
## Application Examples: S4

- no real data yet, launch early 2024
- synthetic data



## S<sub>4</sub> – ROCINN applied to synthetic test data

- Test data for 17 locations across the S<sub>4</sub> FOV
- Examples are shown for cloud top height (plots taken from S<sub>4</sub>L2 Verification report)
- Retrievals are within mission requirements for both
  - fully cloudy scenes (top panel)
  - partially cloudy scenes (bottom panel)



## Application Examples: DSCOVR



## DSCOVER – OCRA applied to EPIC

- Aggregation of daily maps in intervals of +/- 14 days with a small-percentile kernel and 0.2 deg resolution

Clear-sky maps for EPIC channels (780, 551, 388) nm

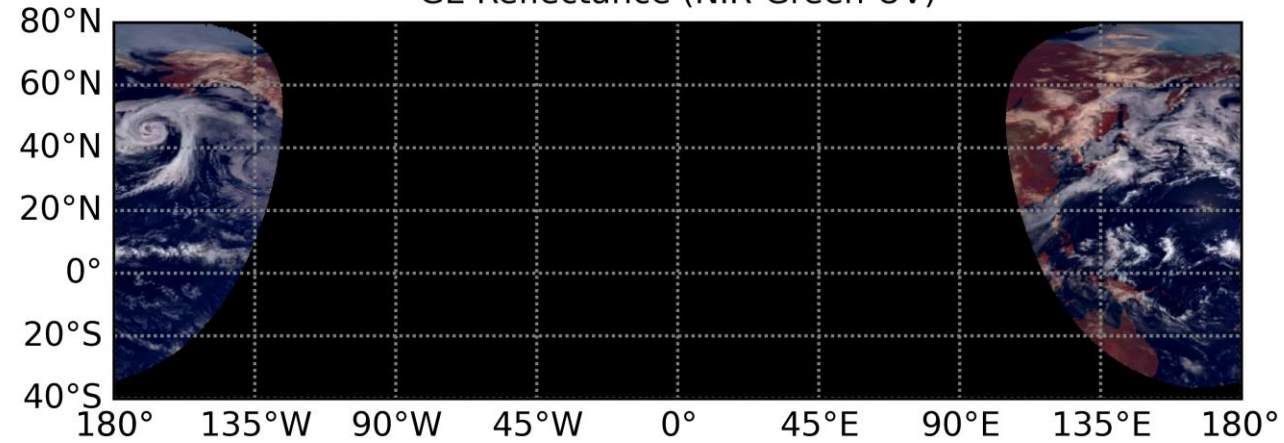


# DSCOVER – OCRA applied to EPIC

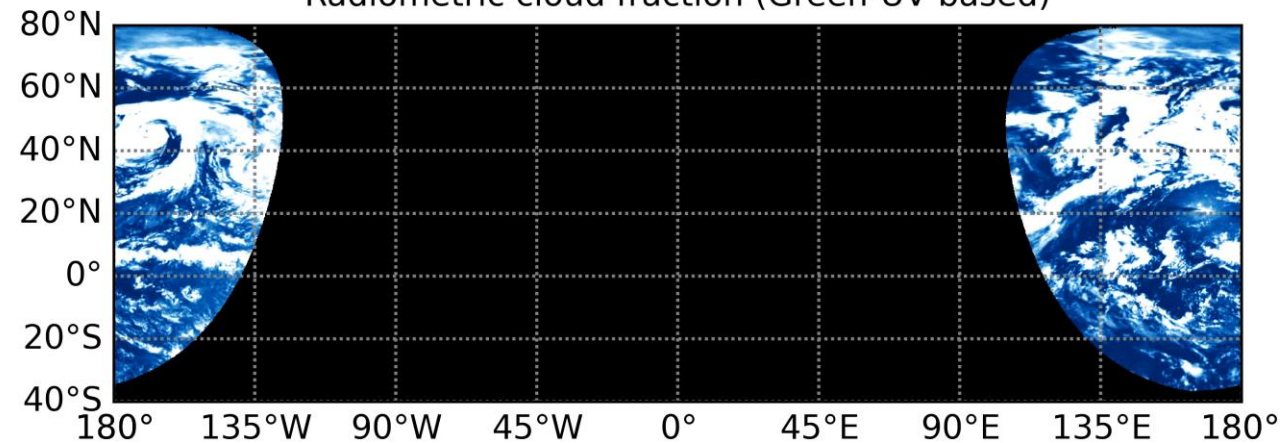
- True color (top) and OCRA radiometric cloud fraction (bottom) for 15 June 2018

2018-06-15 01:04:37 UTC

GE Reflectance (NIR-Green-UV)



Radiometric cloud fraction (Green-UV based)



# Conclusion and Outlook

## Conclusion

- OCRA/ROCINN has been **successfully implemented for Sentinel-5P** and is operational since more than three years and will also be used **operationally for the geostationary Sentinel-4**
- OCRA/ROCINN has been successfully applied to EPIC and OCRA to GEMS
- Validation against VIIRS, ground-based data (CLOUDNET) and synthetic data show good agreement

## Outlook

- OCRA/ROCINN cloud data records are already available for GOME, SCIAMACHY, GOME-2A/B/C
- Generate a **consistent, homogeneous multi-sensor cloud properties dataset** starting in 1995 and adding to the above also EPIC, S5P, GEMS and the upcoming S4, S5



***Thank you for your attention!***

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