

ORAC

Bringing three threads together

Gareth Thomas

Richard Siddans, Caroline Poulsen, Caroline Cox (RAL Space)

Adam Povey, Greg McGarragh, Haiyan Huang, Don Grainger (University of Oxford)

Matthias Jerg, Stefan Stapelberg (DWD)



**NERC National Centre
for Earth Observation**

NATURAL ENVIRONMENT RESEARCH COUNCIL



Deutscher Wetterdienst
Wetter und Klima aus einer Hand

Overview

- Introduction/reminder of what ORAC is
- Current status of ORAC
 - Aerosol cci
 - Cloud cci
 - Volcanic ash cloud retrieval
- *A posteriori* scene identification in ORAC
 - What is it?
 - Does it work?

What is ORAC

- “Optimal Retrieval of Aerosol & Cloud”, or “Oxford-RAL Aerosol & Cloud”.
 - Some in Cloud cci have also started calling it “CC4CL”
- Optimal estimation retrieval scheme for visible/IR imaging instruments
- There are aerosol, cloud and aerosol-SST variants in existence
- It has been applied to the (A)ATSR, SEVIRI, MODIS & AVHRR sensor families
- Is one of the four (A)ATSR algorithms in aerosol_cci and the “heritage channel” algorithm for cloud_cci

Aerosol cci

Baseline datasets (start of 2010): AOD September 2008

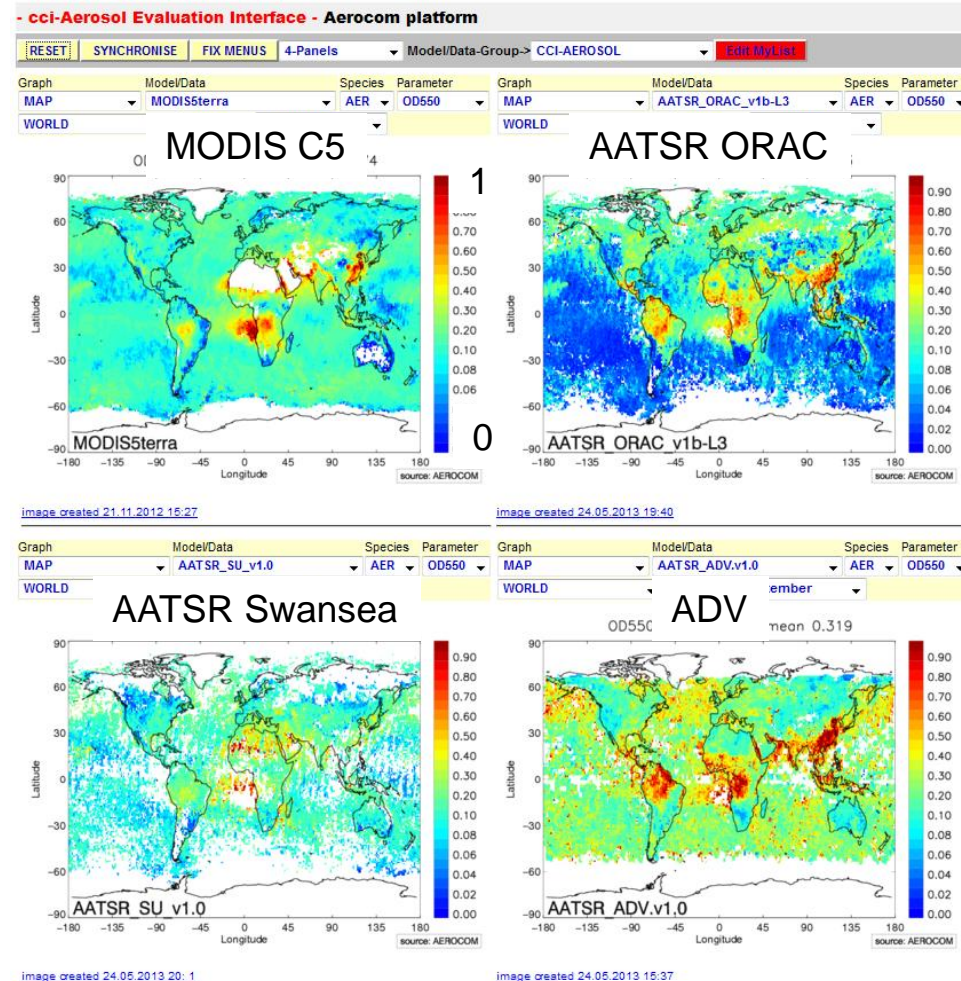
Unusually, all the algorithms that entered Phase I of the project continuing into Phase II.

- Main ECV products will be from (A)ATSR and POLDER
- Three “competing” (A)ATSR algorithms
 - ORAC
 - Swansea dual-view
 - Advanced Dual View (ADV) from FMI
- Swansea algorithm has been selected to provide the prototype ECV for the full ATSR-2/AATSR time-series

- Phase II will concentrate on
 - Improving uncertainty propagation
 - Improving validation and extending beyond AOD
 - Producing multiple long-term data sets

Public data sets *should be* here:

<http://www.icare.univ-lille1.fr/archive/index.php?dir=CCI-Aerosols/>



Aerosol cci

Final round-robin (end 2012): AOD September 2008

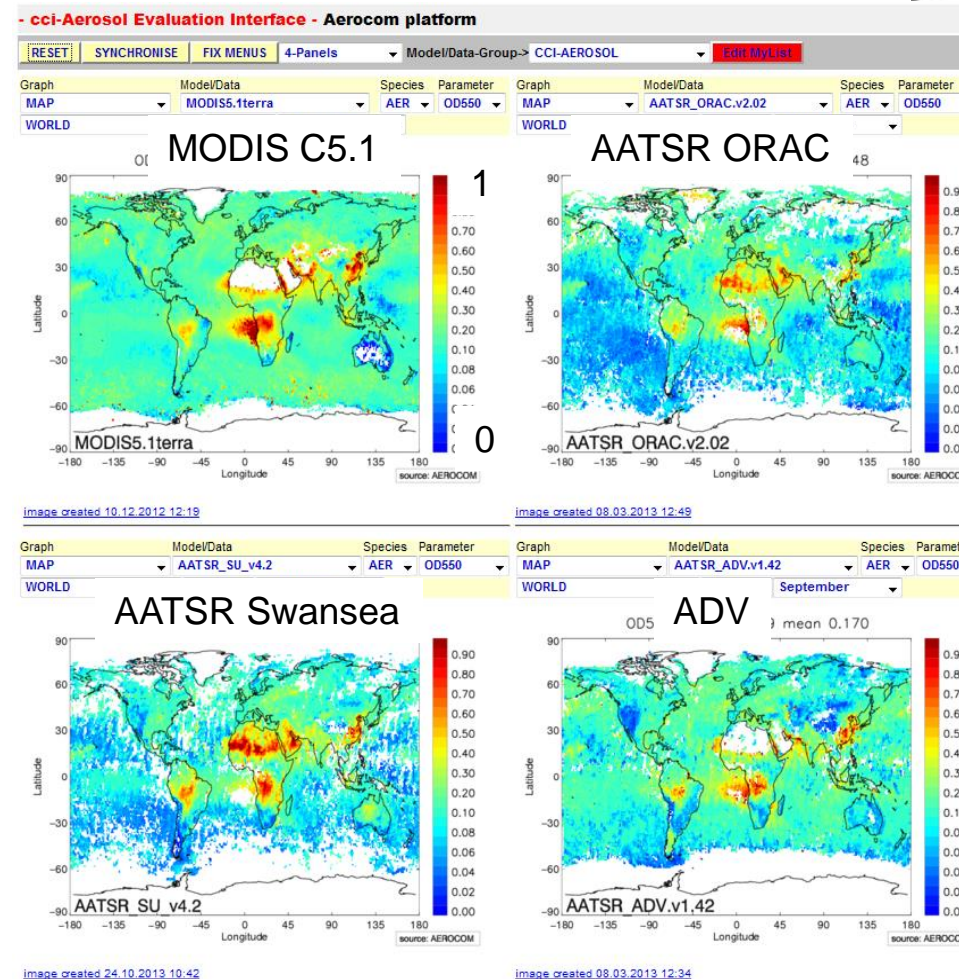
Unusually, all the algorithms that entered Phase I of the project continuing into Phase II.

- Main ECV products will be from (A)ATSR and POLDER
- Three “competing” (A)ATSR algorithms
 - ORAC
 - Swansea dual-view
 - Advanced Dual View (ADV) from FMI
- Swansea algorithm has been selected to provide the prototype ECV for the full ATSR-2/AATSR time-series

- Phase II will concentrate on
- Improving uncertainty propagation
 - Improving validation and extending beyond AOD
 - Producing multiple long-term data sets

Public data sets *should be* here:

<http://www.icare.univ-lille1.fr/archive/index.php?dir=CCI-Aerosols/>



Aerosol cci

Final round-robin (end 2012): AOD September 2008

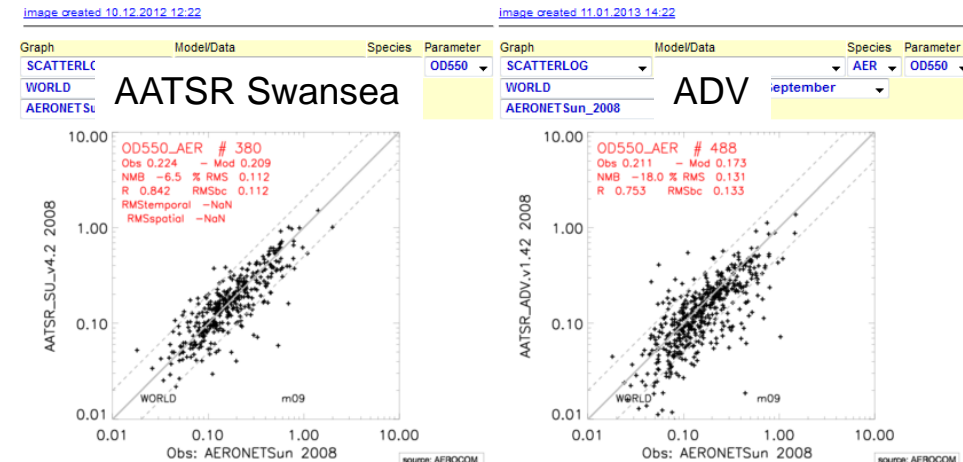
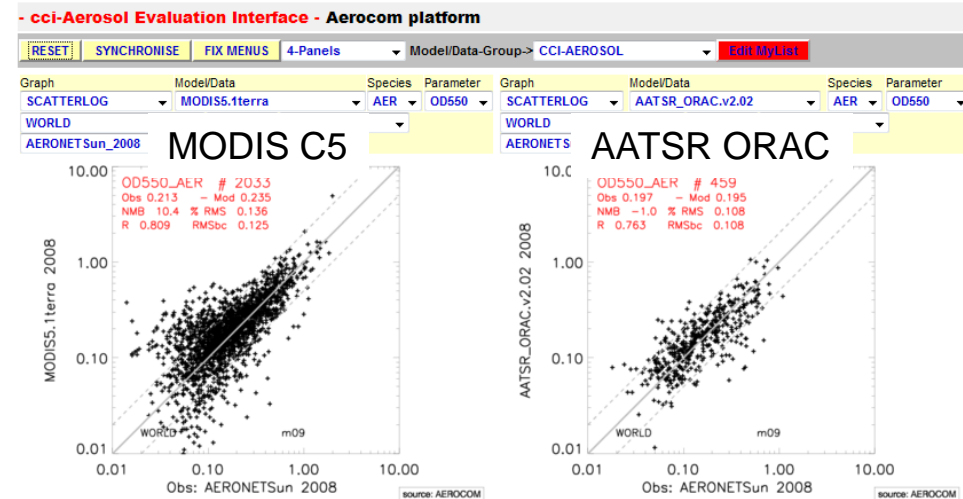
Unusually, all the algorithms that entered Phase I of the project continuing into Phase II.

- Main ECV products will be from (A)ATSR and POLDER
- Three “competing” (A)ATSR algorithms
 - ORAC
 - Swansea dual-view
 - Advanced Dual View (ADV) from FMI
- Swansea algorithm has been selected to provide the prototype ECV for the full ATSR-2/AATSR time-series

- Phase II will concentrate on
 - Improving uncertainty propagation
 - Improving validation and extending beyond AOD
 - Producing multiple long-term data sets

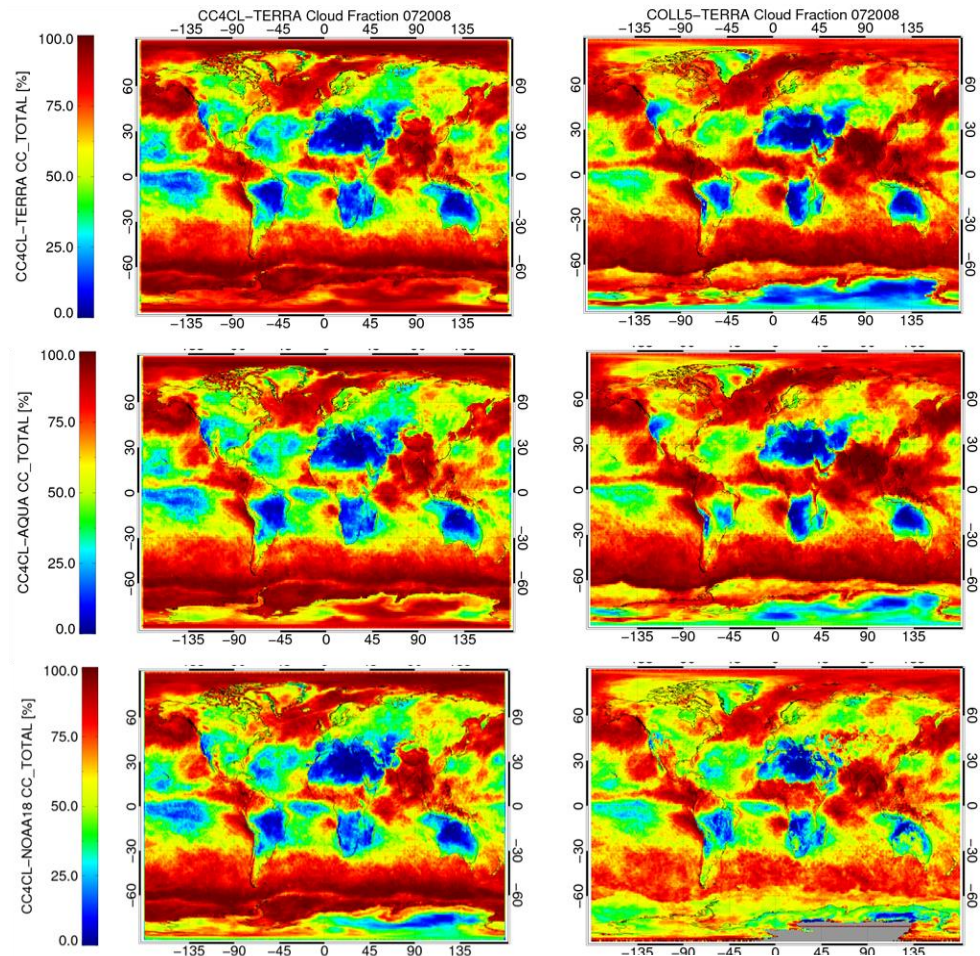
Public data sets *should be* here:

<http://www.icare.univ-lille1.fr/archive/index.php?dir=CCI-Aerosols/>



Cloud cci

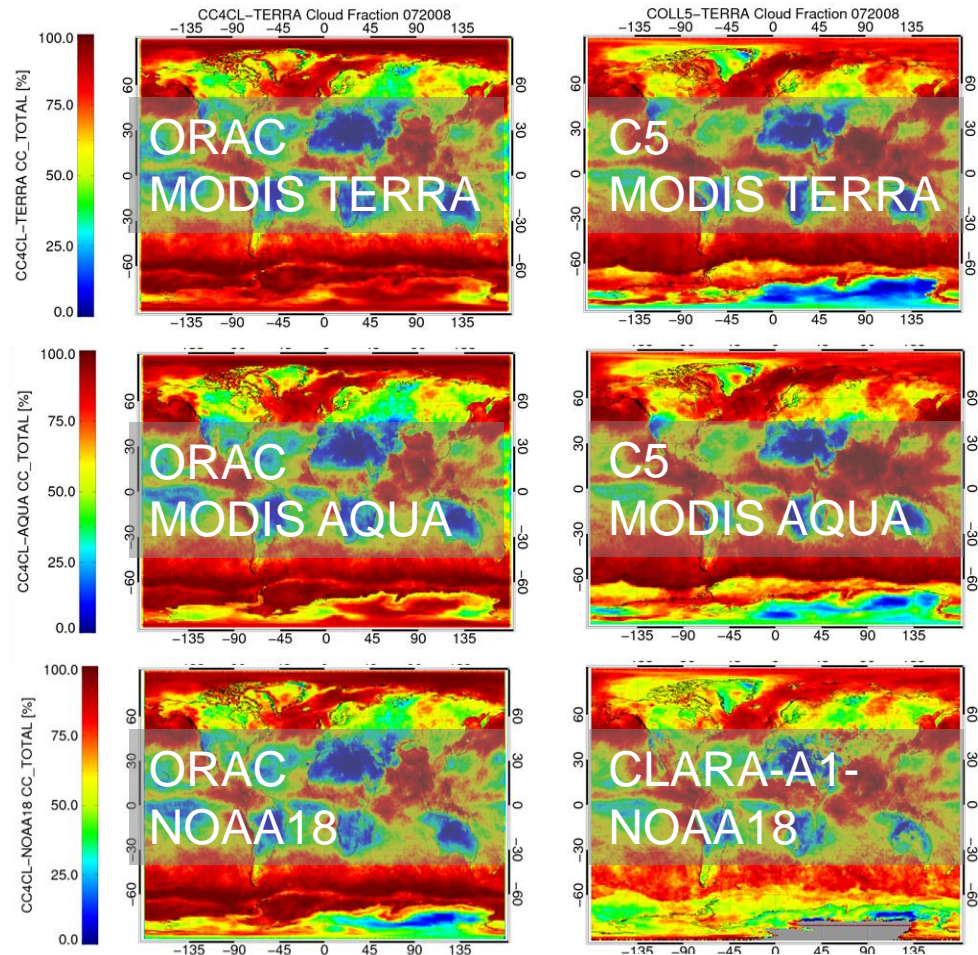
- ORAC has been used to produce global cloud products for 2007-2009 from
 - MODIS (Aqua+Terra) [DWD]
 - AVHRR (NOAA-15, 16, 17) [DWD]
 - AATSR [RAL]
- Development of the retrieval system is still ongoing
 - Addressing bugs/problems revealed by first processing
 - Optimising the retrieval for speed



Plots: Stefan Stapelberg (DWD)

Cloud cci

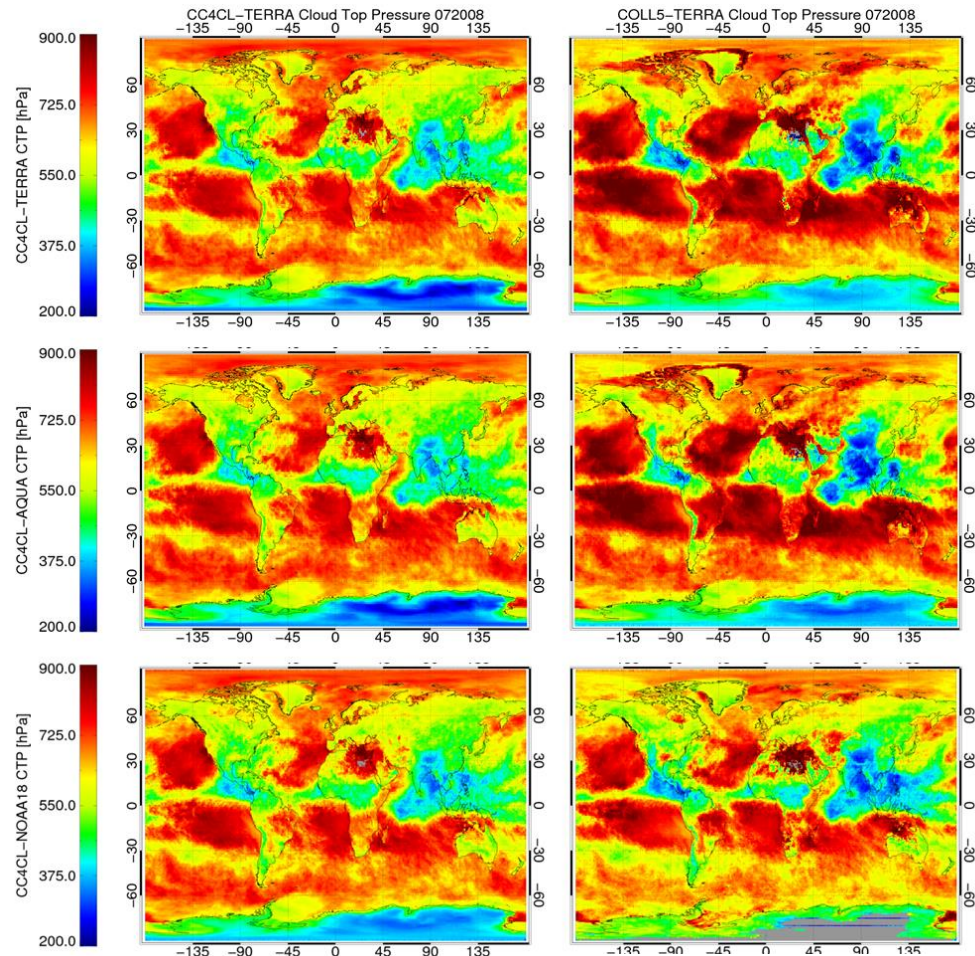
- ORAC has been used to produce global cloud products for 2007-2009 from
 - MODIS (Aqua+Terra) [DWD]
 - AVHRR (NOAA-15, 16, 17) [DWD]
 - AATSR [RAL]
- Development of the retrieval system is still ongoing
 - Addressing bugs/problems revealed by first processing
 - Optimising the retrieval for speed



Plots: Stefan Stapelberg (DWD)

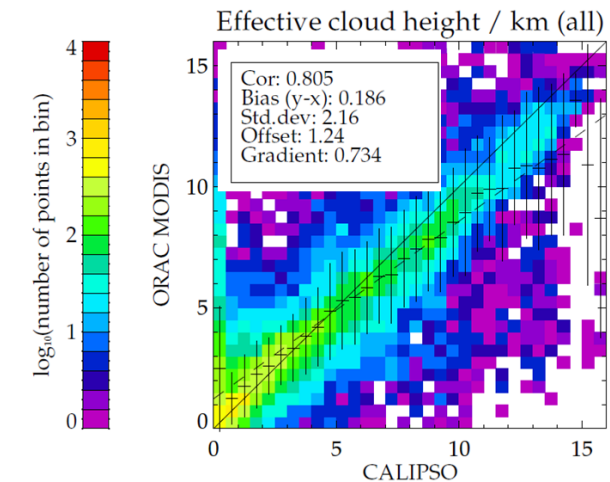
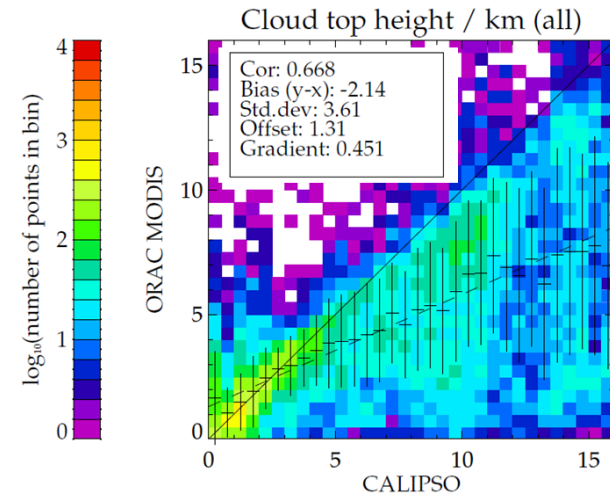
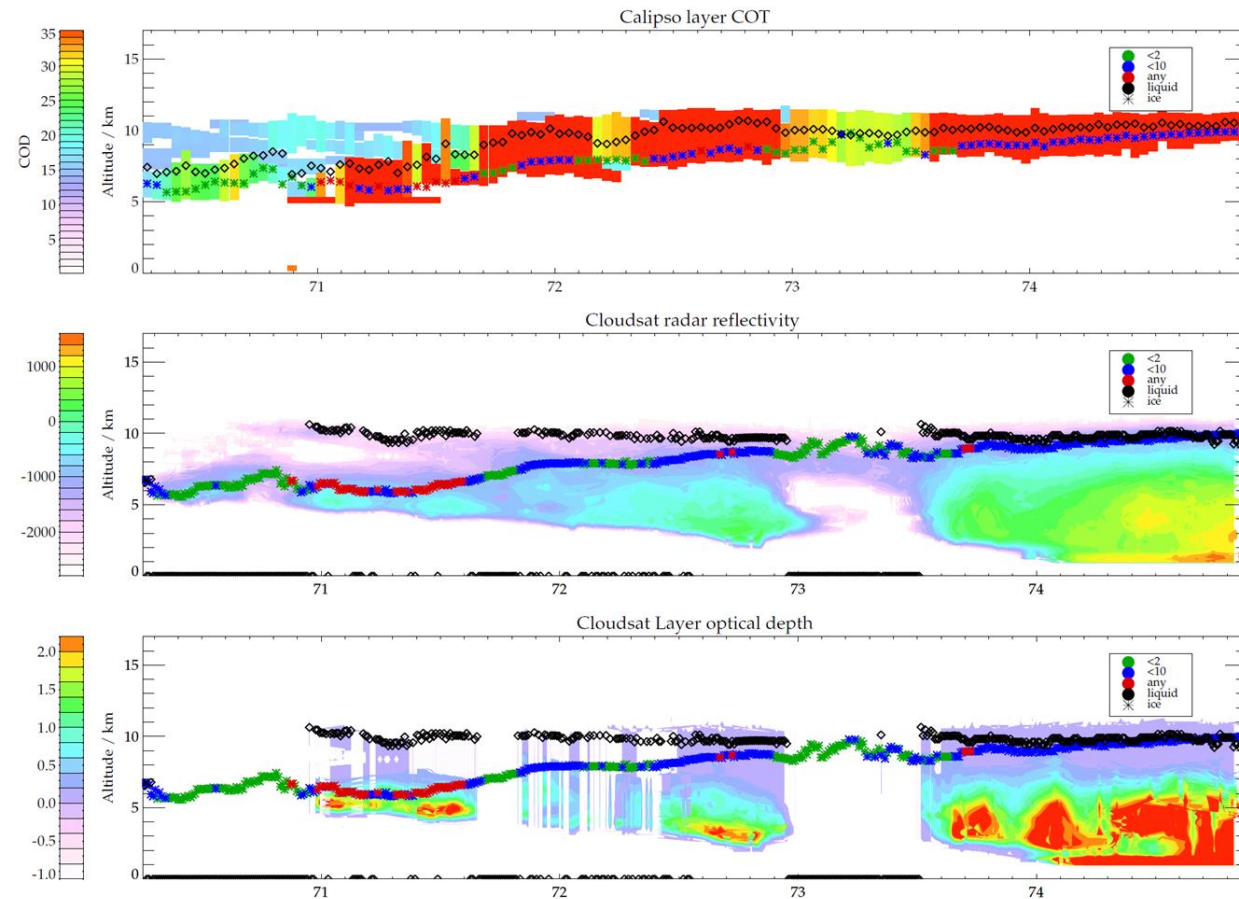
Cloud cci

- ORAC has been used to produce global cloud products for 2007-2009 from
 - MODIS (Aqua+Terra) [DWD]
 - AVHRR (NOAA-15, 16, 17) [DWD]
 - AATSR [RAL]
- Development of the retrieval system is still ongoing
 - Addressing bugs/problems revealed by first processing
 - Optimising the retrieval for speed



Plots: Stefan Stapelberg (DWD)

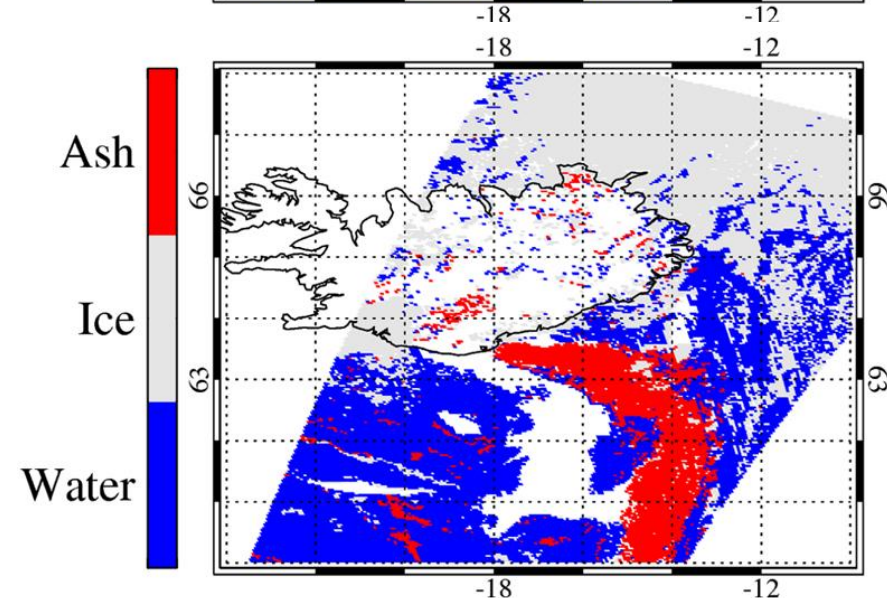
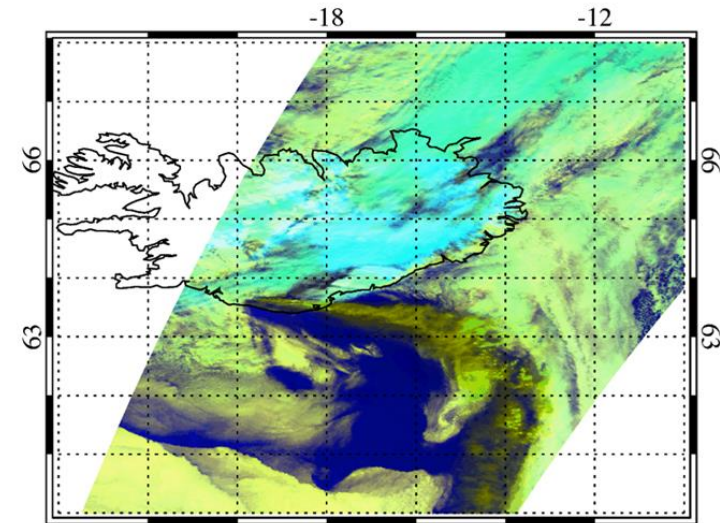
The cloud-top-height problem



coloc.ATS.TOA.1PRUPA20080921.082333.000065272072.00178.34300.2963.1-512.66-71.gv3.spi.11110001.v3p18

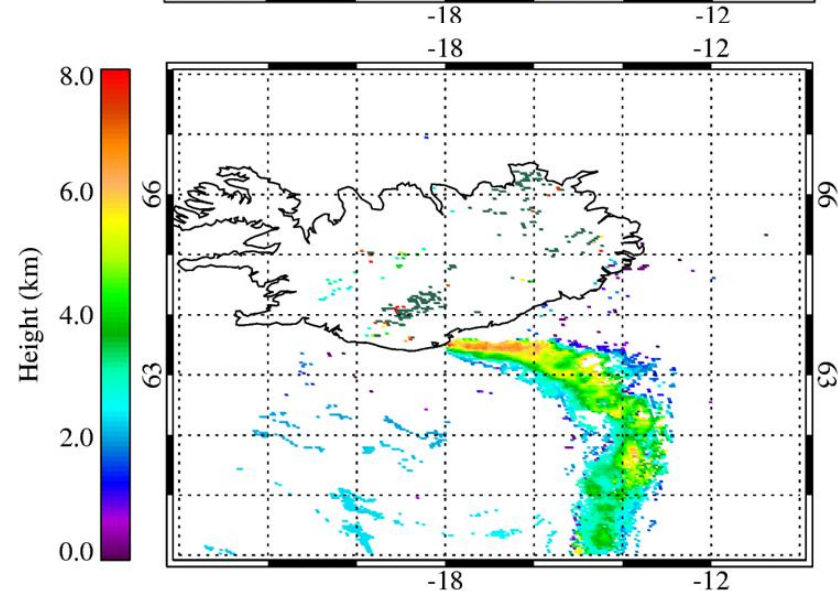
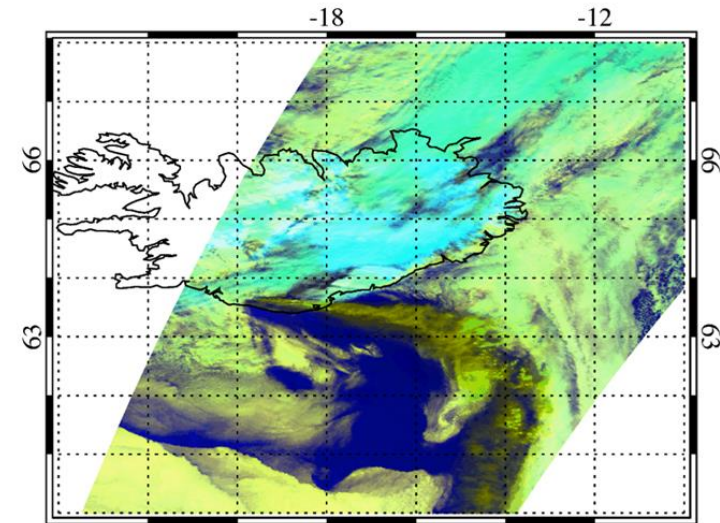
Volcanic ash

- ORAC has now been extensively applied to the retrieval of volcanic ash properties
- The OE method is well suited to this problem, as ash is generally easily distinguished from cloud
- Two main problems remain:
 - Ash optical properties
 - The multi-layer issue



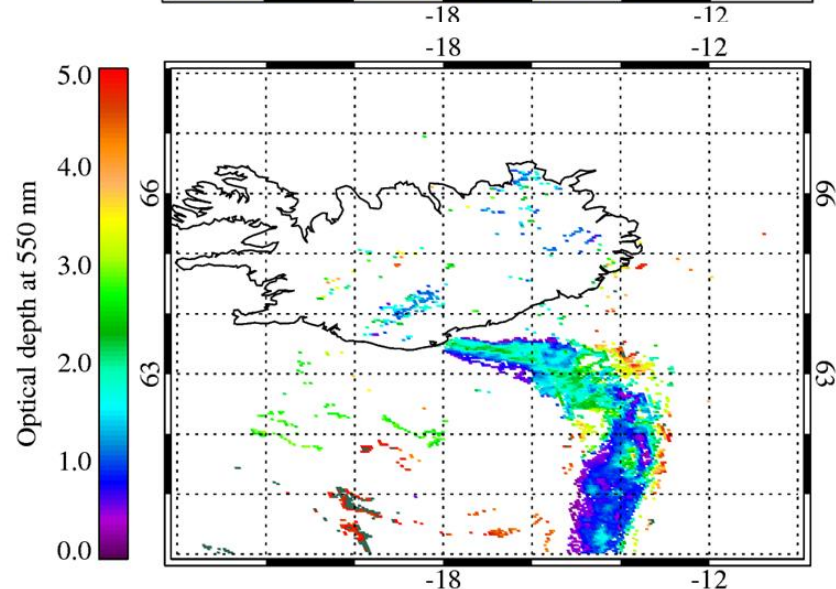
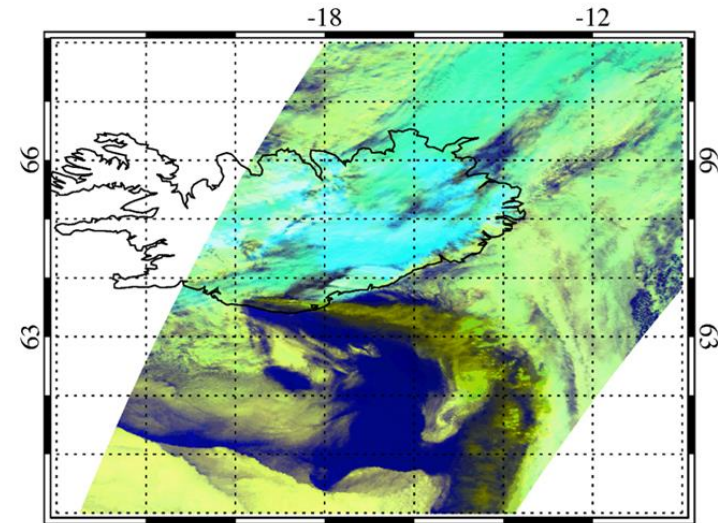
Volcanic ash

- ORAC has now been extensively applied to the retrieval of volcanic ash properties
- The OE method is well suited to this problem, as ash is generally easily distinguished from cloud
- Two main problems remain:
 - Ash optical properties
 - The multi-layer issue



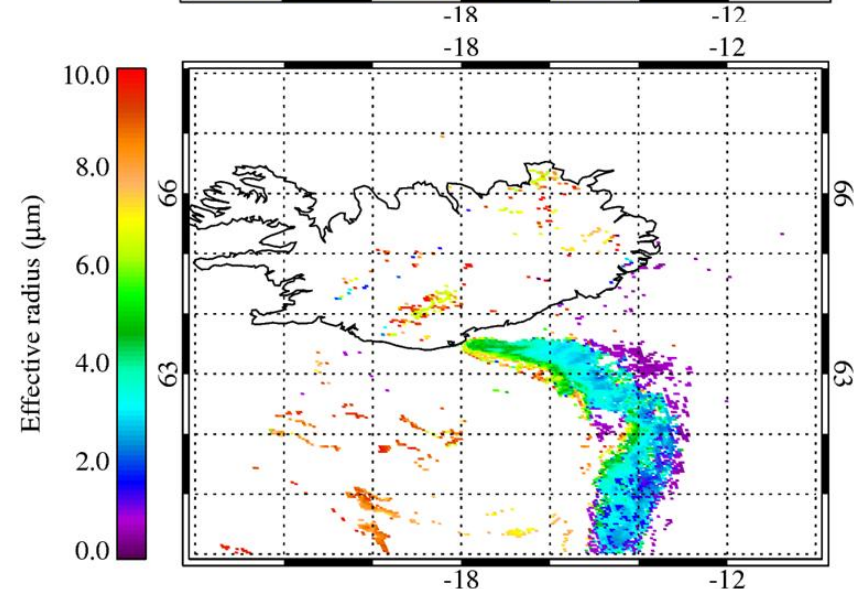
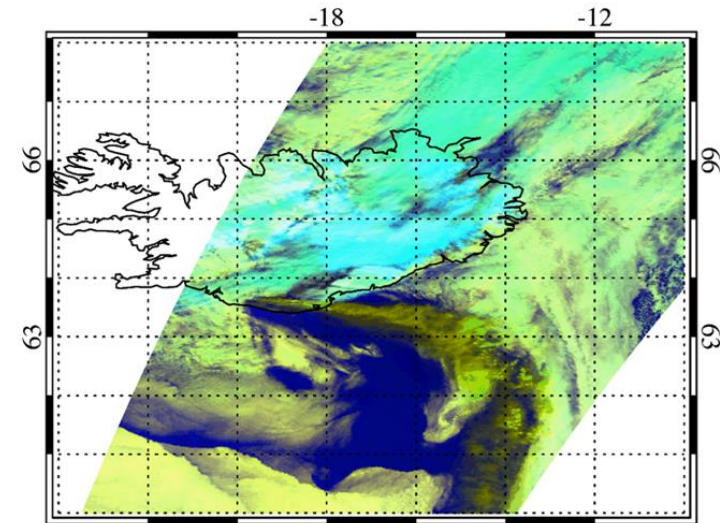
Volcanic ash

- ORAC has now been extensively applied to the retrieval of volcanic ash properties
- The OE method is well suited to this problem, as ash is generally easily distinguished from cloud
- Two main problems remain:
 - Ash optical properties
 - The multi-layer issue



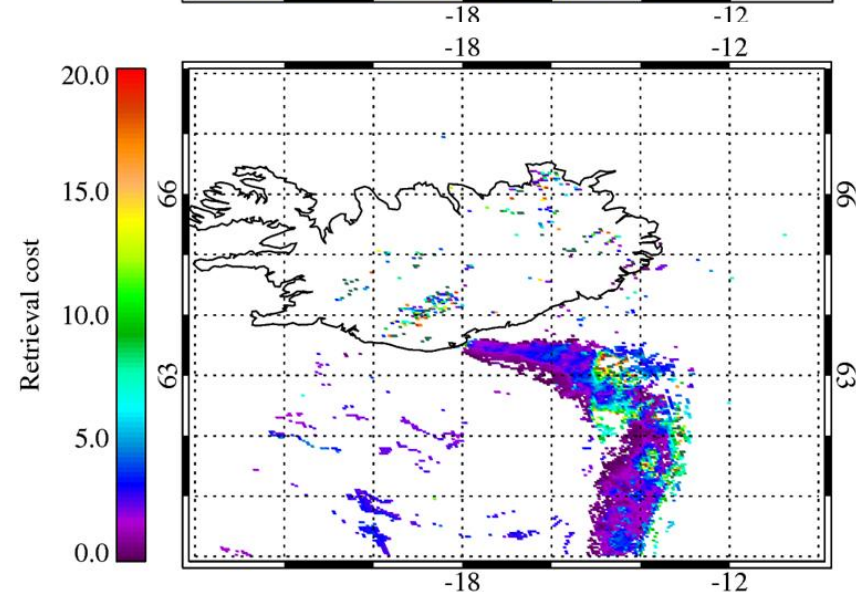
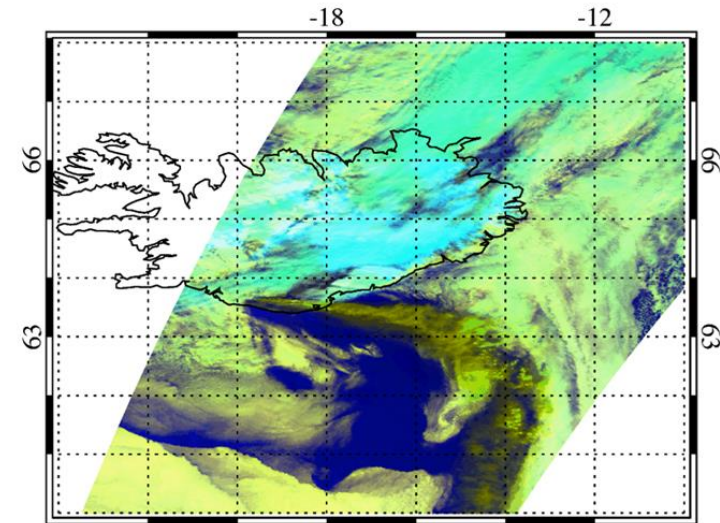
Volcanic ash

- ORAC has now been extensively applied to the retrieval of volcanic ash properties
- The OE method is well suited to this problem, as ash is generally easily distinguished from cloud
- Two main problems remain:
 - Ash optical properties
 - The multi-layer issue

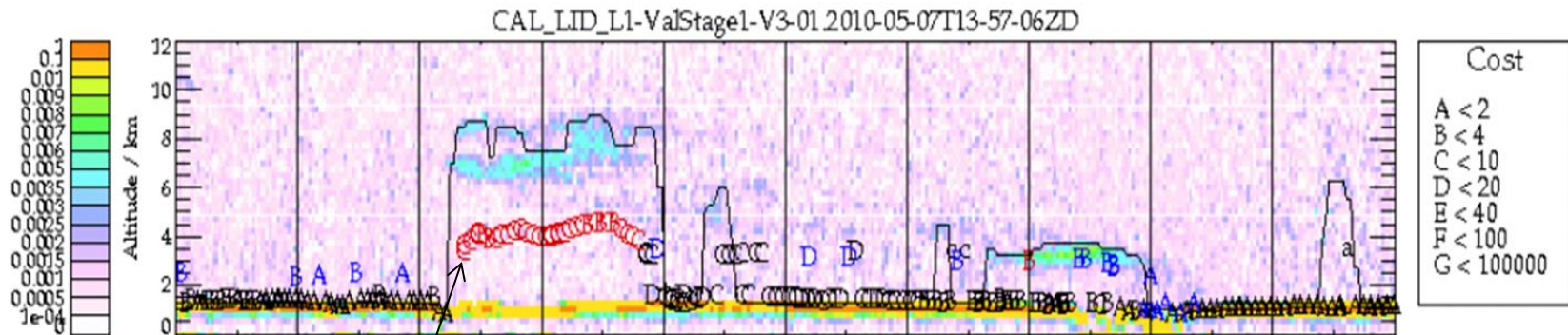


Volcanic ash

- ORAC has now been extensively applied to the retrieval of volcanic ash properties
- The OE method is well suited to this problem, as ash is generally easily distinguished from cloud
- Two main problems remain:
 - Ash optical properties
 - The multi-layer issue

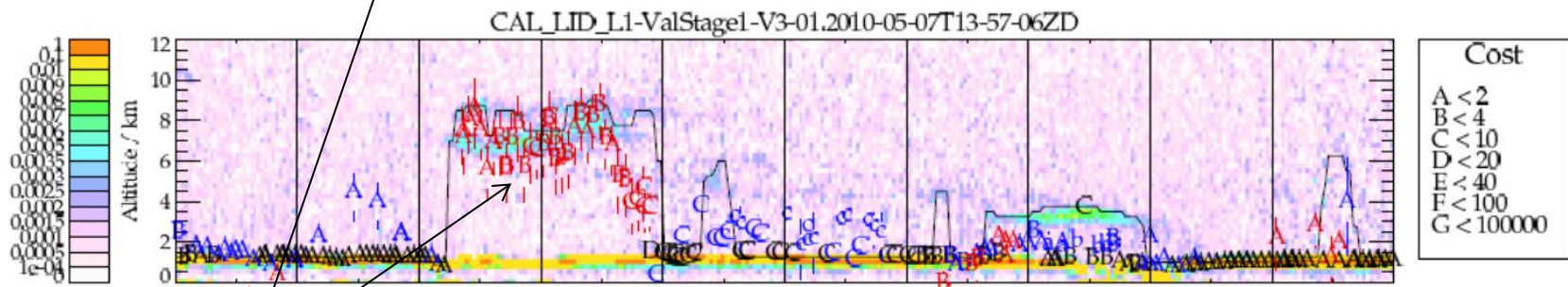


Multi-layer retrieval



Single-layer forward model

Cloud Layer Below

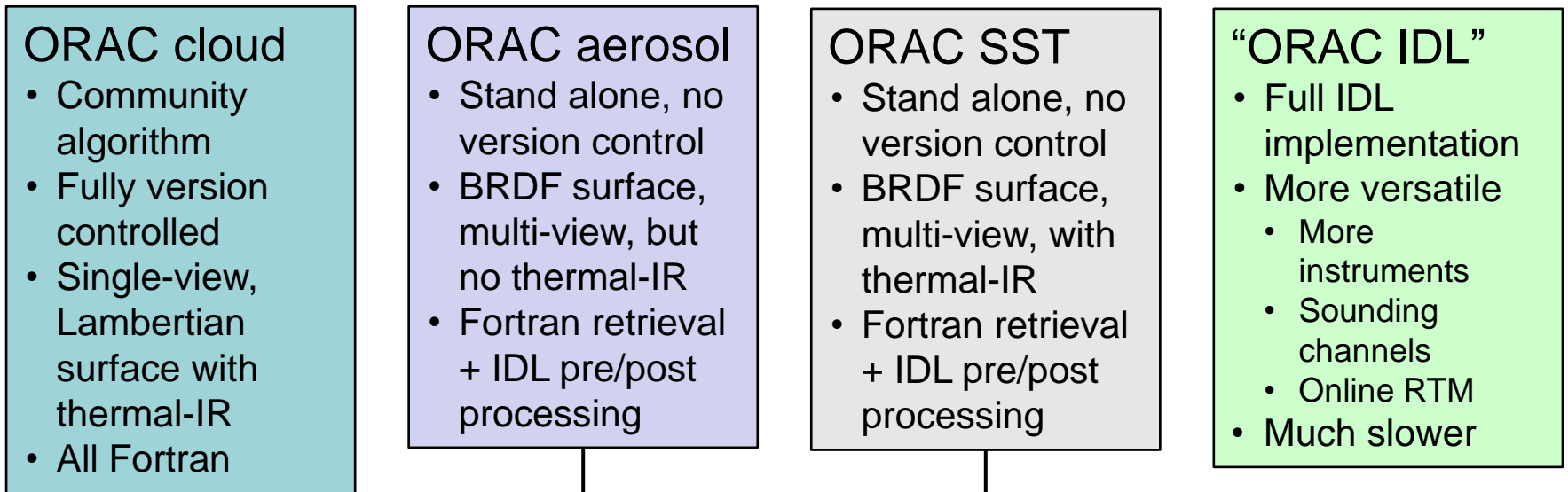


Dual-layer forward model

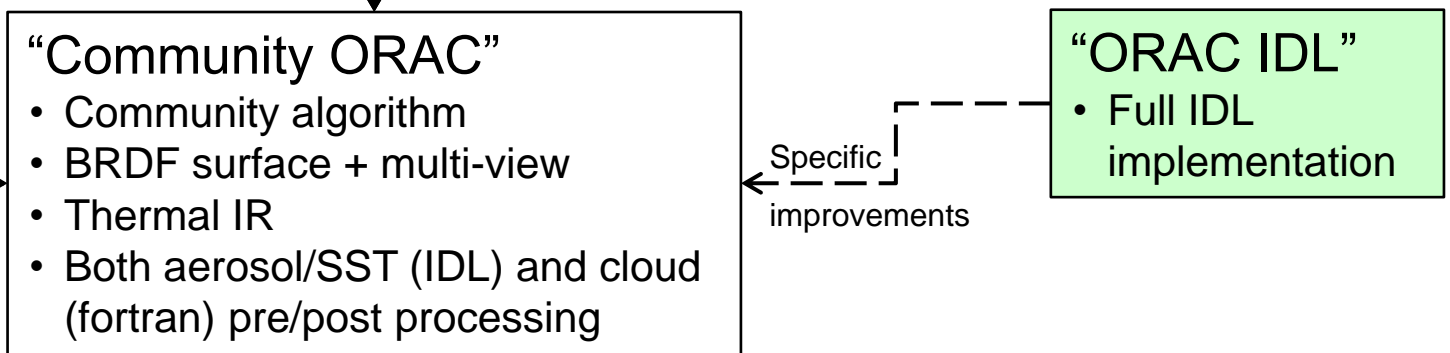
ORAC-SEVIRI retrievals

What's next for ORAC

NOW



12 MONTHS



A posteriori scene identification

- OE retrieval provides statistics on the quality of the fit
 - In particular the retrieval cost is directly related to the conditional probability of the retrieved state given the measurement (for a particular set of assumptions):

$$J = -2 \ln P(\mathbf{x}|\mathbf{y})$$

- Can we use this information to distinguish between cloud and aerosol (and different cloud/aerosol types)?

χ^2 test

- Measurement cost function:

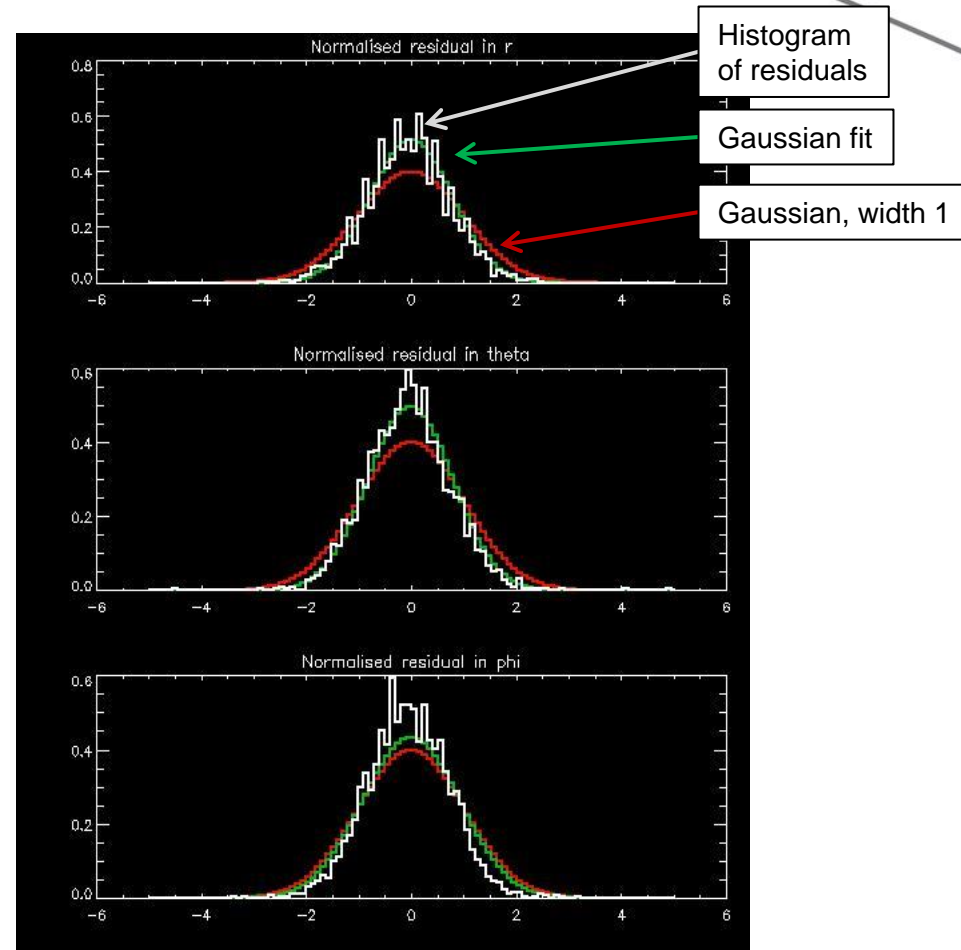
$$J_m = [\mathbf{y} - \mathbf{f}(\mathbf{x})] \mathbf{S}_y^{-1} [\mathbf{y} - \mathbf{f}(\mathbf{x})]$$

will be a random sample from a normal distribution with a standard deviation of 1, with degrees of freedom equal to the number of measurements, m .

- Thus, it should follow a χ^2 distribution with m degrees of freedom and each J_m value can thus provide a probability that the retrieval is consistent with the measurement
- *Assumes that the covariance matrix, \mathbf{S}_y , is an accurate representation of the uncertainty in the system and that the forward model, $\mathbf{f}(\mathbf{x})$, is a good representation of the physics of the measurement.*
- Similar argument can be applied to the a priori cost.

A test case

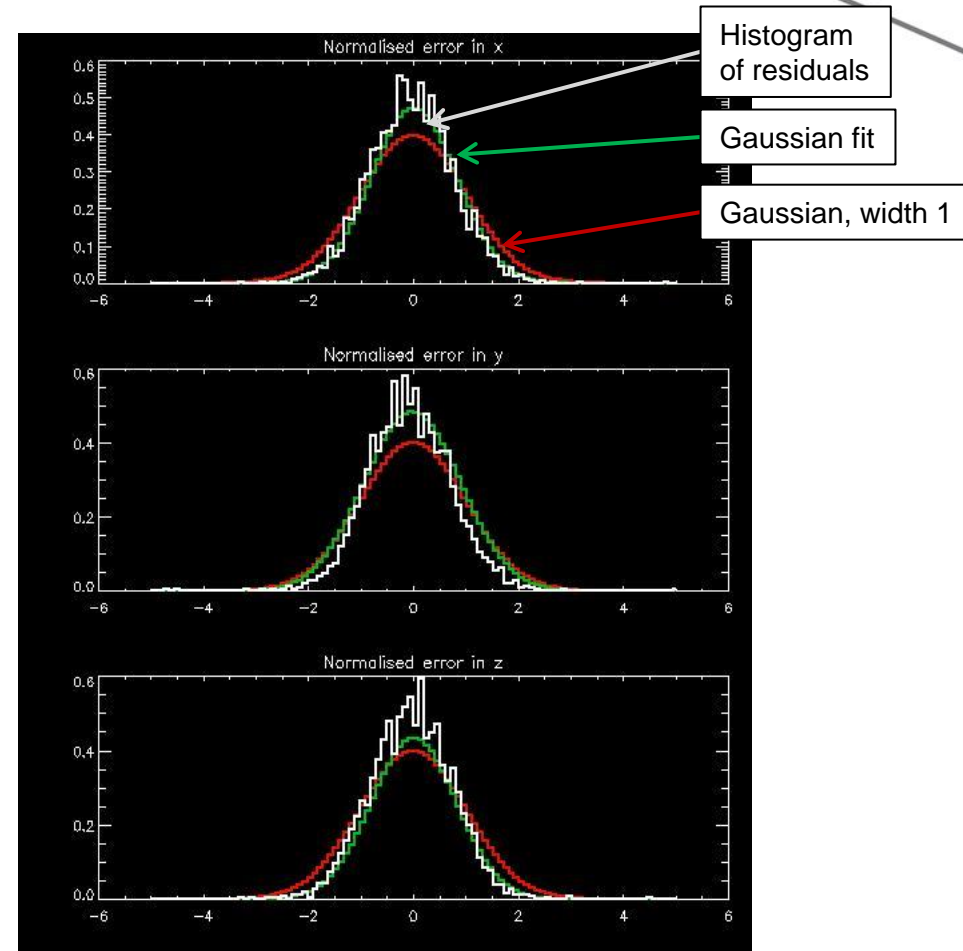
- Simple numerical retrieval of three parameters from 3 measurements
 - Forward model is transform from Cartesian to polar coordinates
- Gaussian noise with standard deviation of 0.01 added to forward model



Distribution of measurement residuals, normalised by uncertainty

A test case

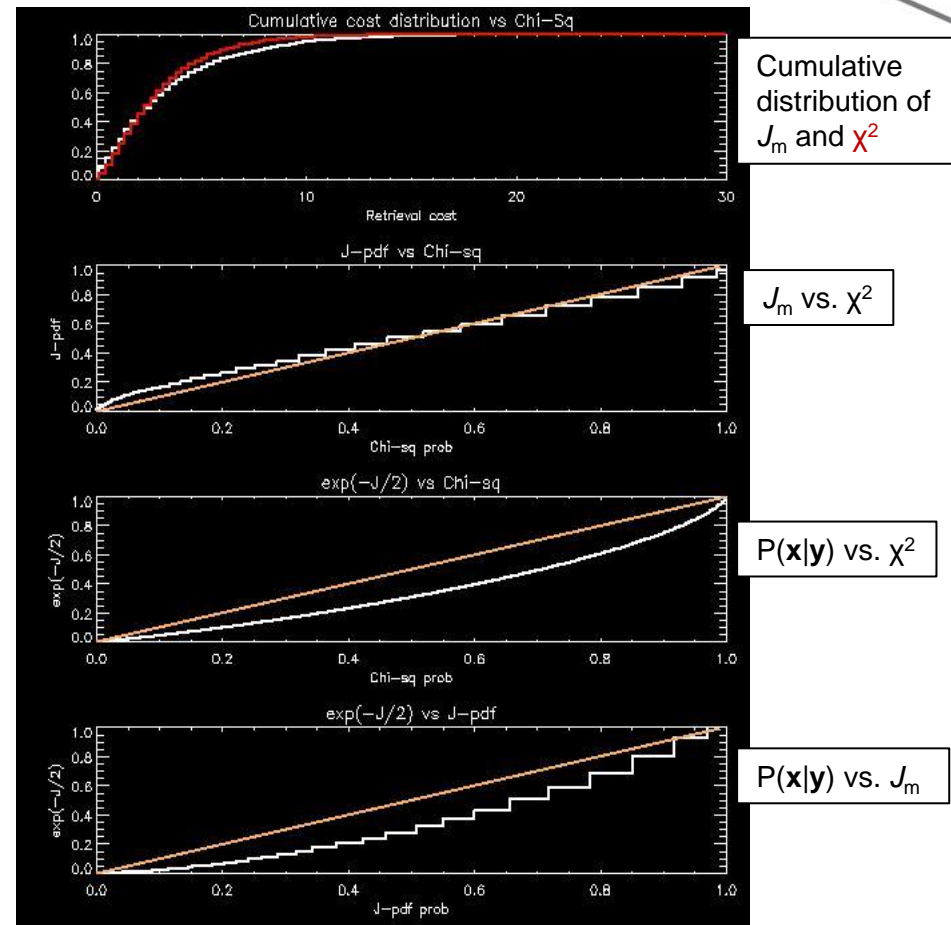
- Retrieved state and measurement both agree very well with theoretical distribution
- Retrieval works!



Distribution of state vector residuals, normalised by retrieved uncertainty

A test case

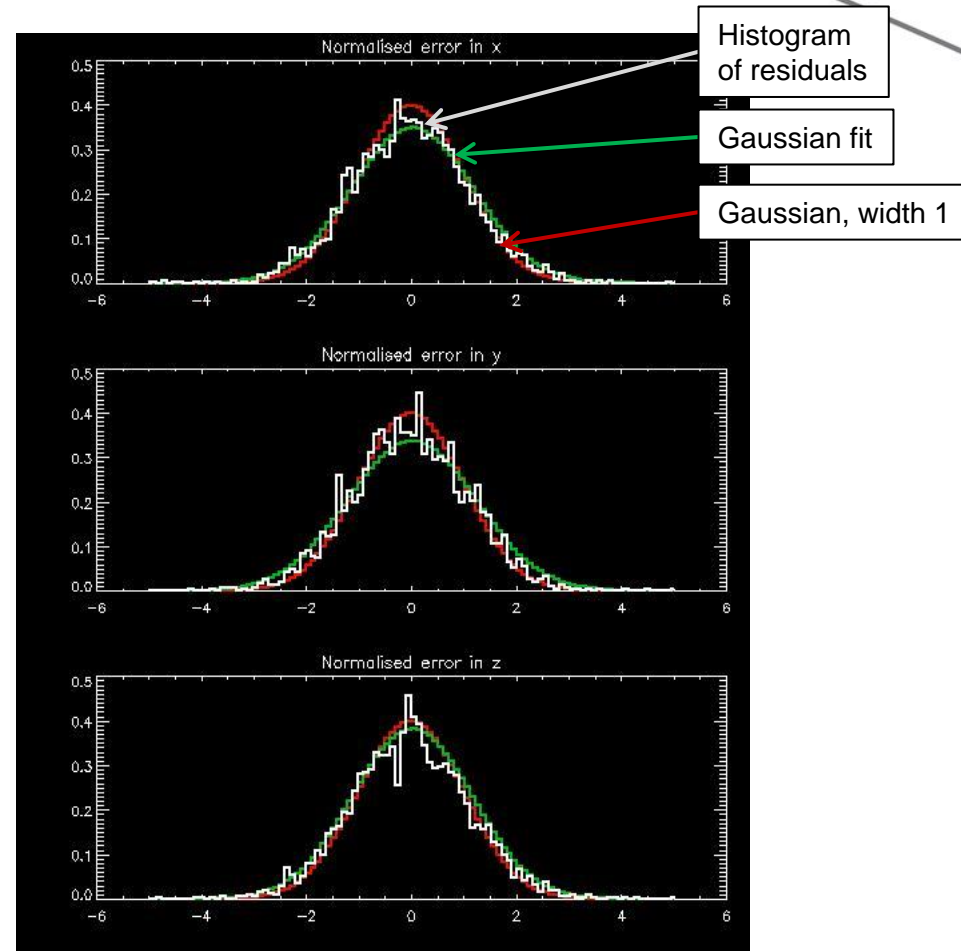
- Cumulative distribution of cost is very close to expected χ^2 distribution
- Note that the conditional probability $P(\mathbf{x}|\mathbf{y})$ is pretty close to the χ^2 probability, but they are not the same



A test case

Assumed uncertainty
50% too small....

- Retrieval still works
- Even retrieved uncertainty is acceptable...

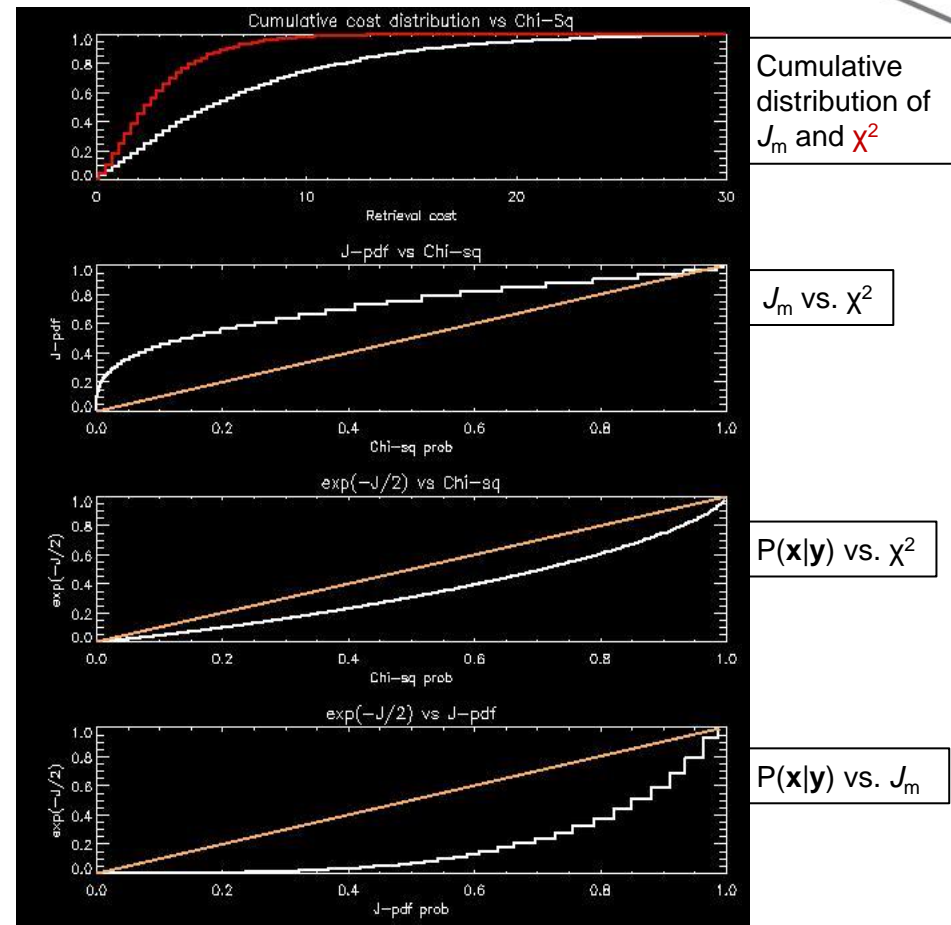


Distribution of state vector residuals, normalised by retrieved uncertainty

A test case

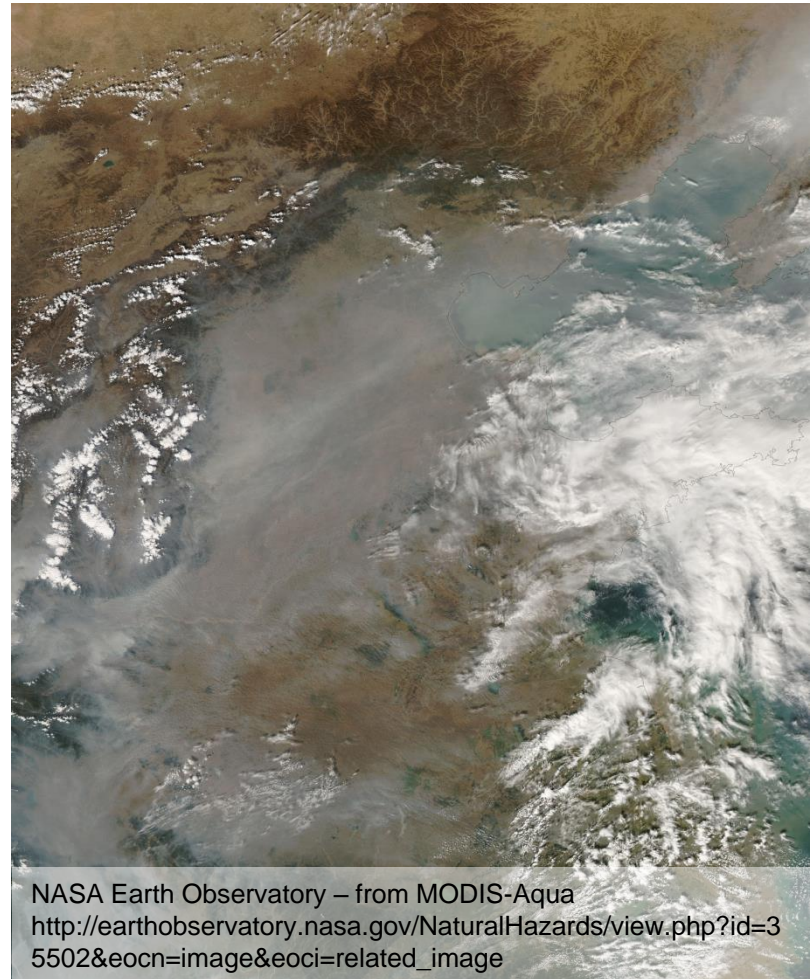
Assumed uncertainty
50% too small....

- χ^2 comparison breaks down
 - Too many high-cost retrievals
- However, the results are still qualitatively useful
 - Better states still provide higher χ^2 probabilities

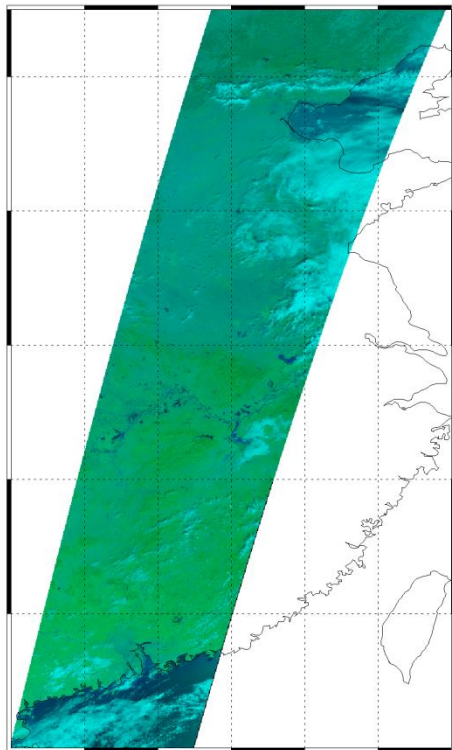


Application to ORAC processor

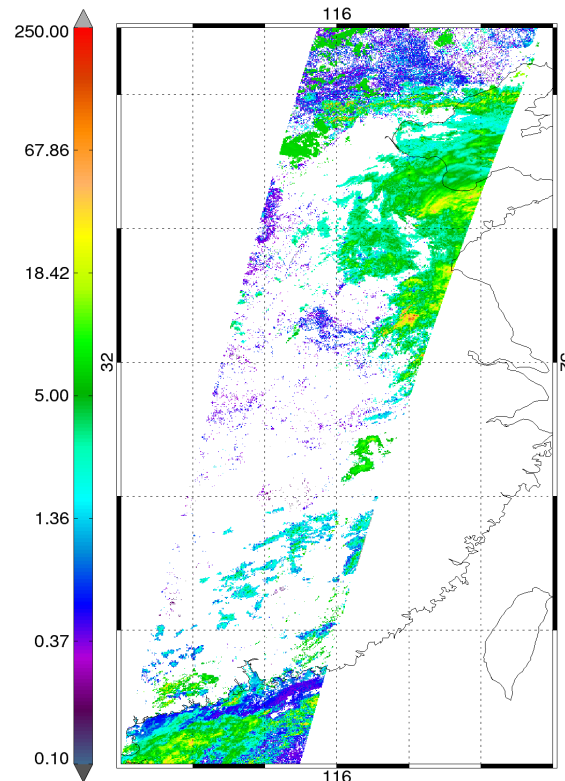
- Chinese haze event on 16-Oct-2008
- A good example of where traditional cloud flagging might struggle!
- AATSR processed:
 - Cloud_cci product
 - Aerosol_cci
 - “Bayesian” retrieval using cloud_cci processor



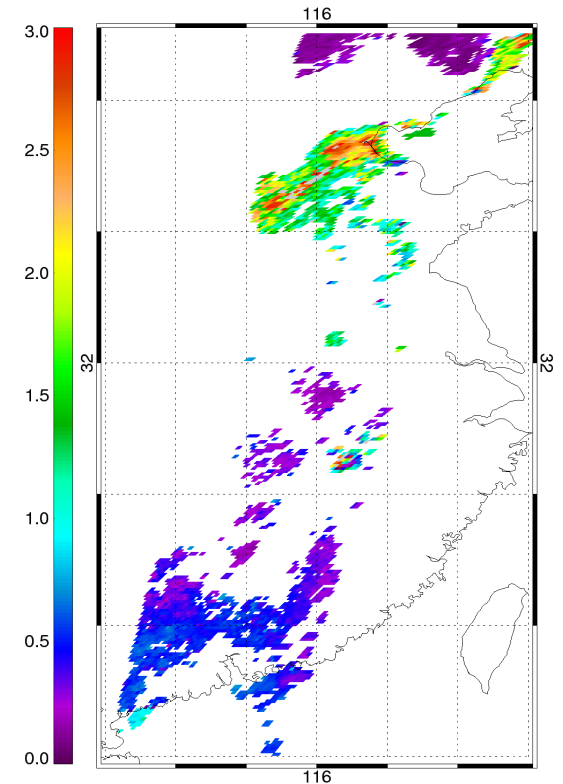
Cloud and Aerosol cci consistency



AATSR false colour

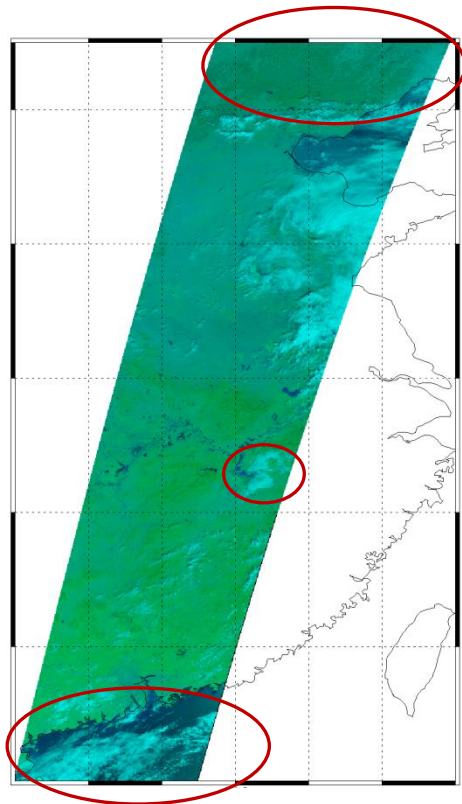


Cloud_cci L2 (ORAC)
cloud optical depth

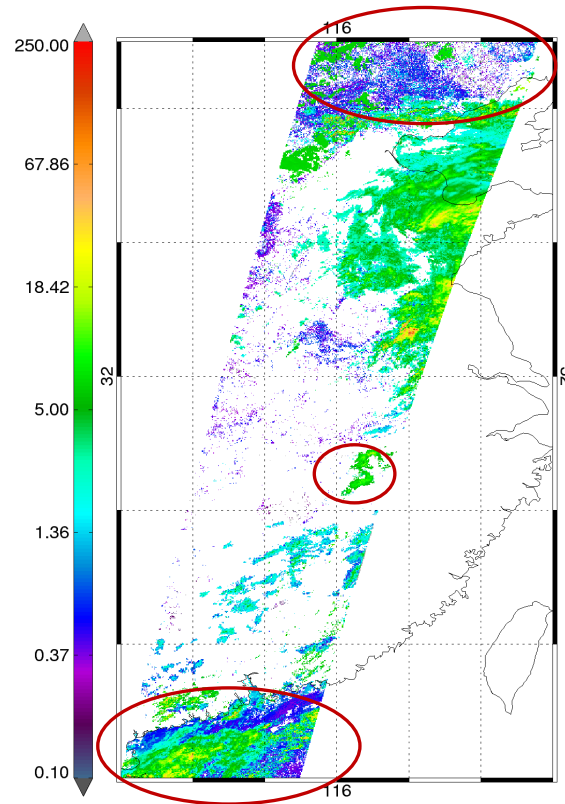


Aerosol_cci L2 (ORAC)
aerosol optical depth

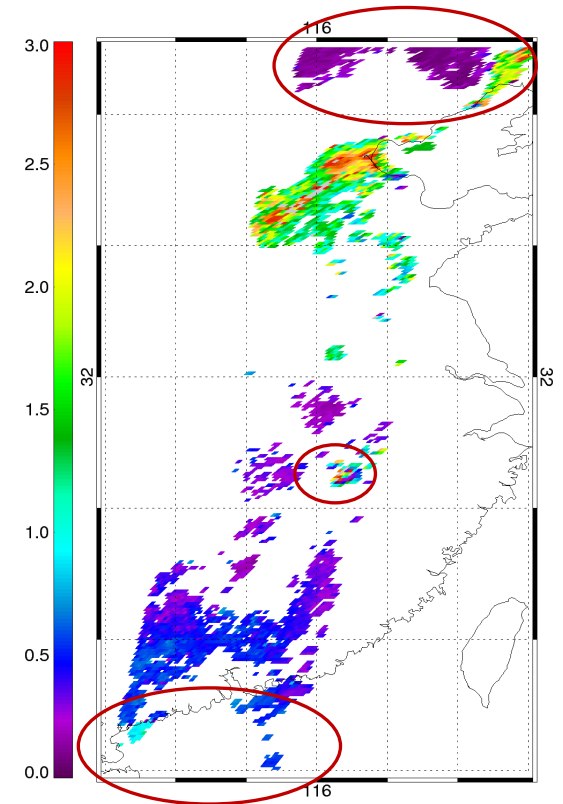
Cloud and Aerosol cci consistency



AATSR false colour

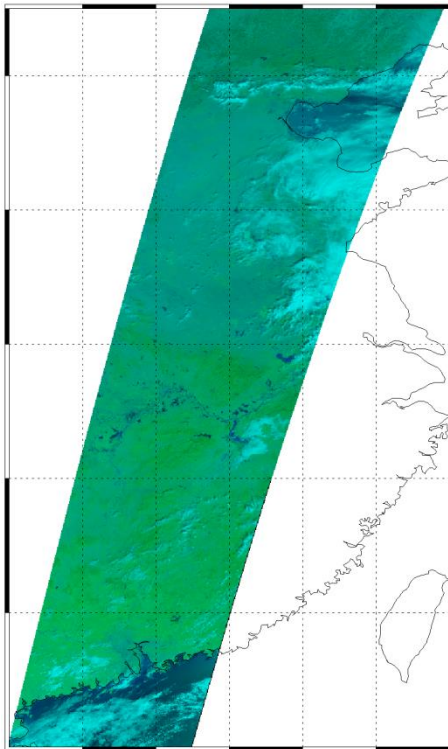


Cloud_cci L2 (ORAC)
cloud optical depth



Aerosol_cci L2 (ORAC)
aerosol optical depth

Bayesian approach...

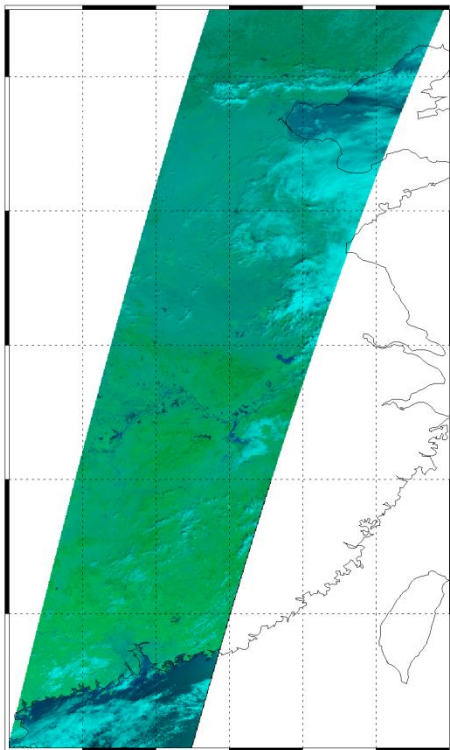


AATSR false colour

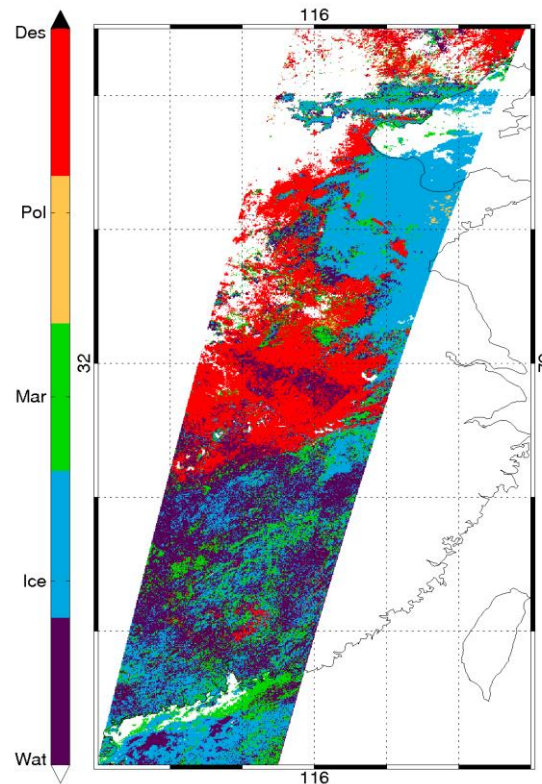
ORAC cloud_cci processor re-run on the scene shown:

- Run with:
 - Water & ice cloud
 - Desert dust (OPAC with a non-spherical coarse mode)
 - Maritime class (OPAC at 80%RH)
 - Pollution (OPAC polluted continental)
- Used OPAC rather than aerosol_cci classes because the thermal IR properties needed

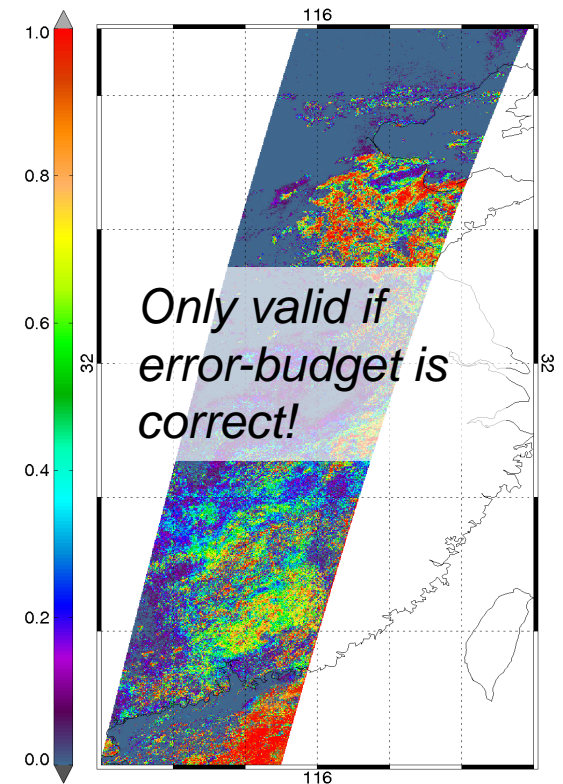
χ^2 results



AATSR false colour

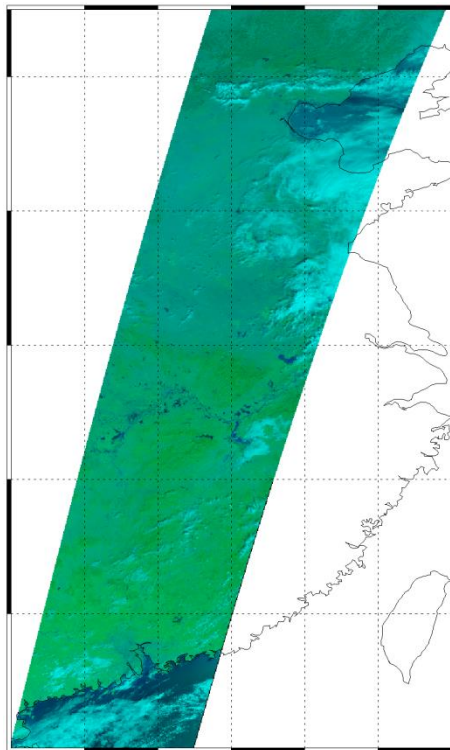


Best-type according to χ^2 test.



χ^2 probability of best-type.

Interpreting the results

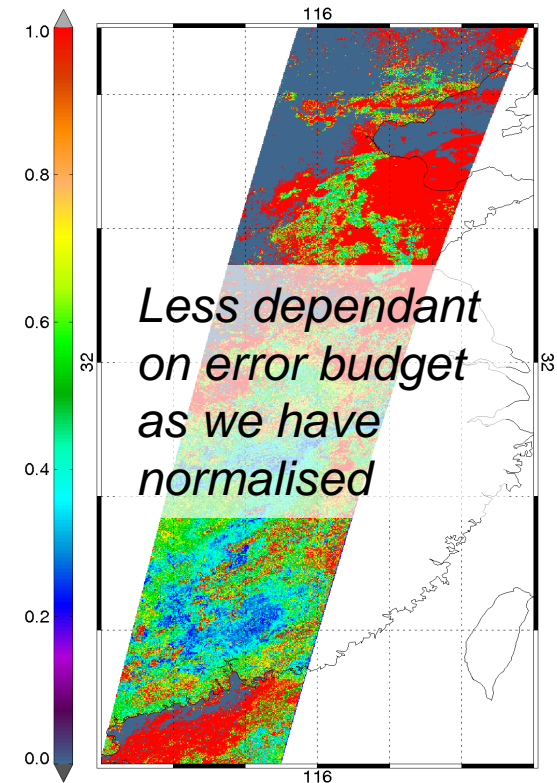


AATSR false colour

Of the available types, how certain are we of the best fitting?

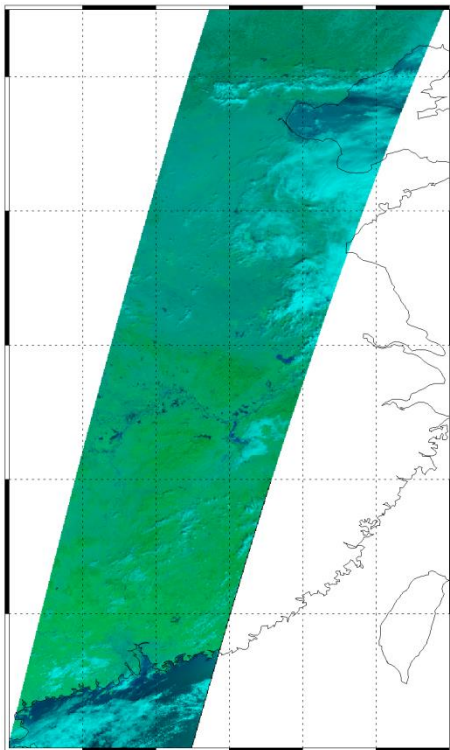
- Normalise the probability:

$$P_n = P_b / [\sum_i P_i]$$
- This can be used as a “cloud mask”

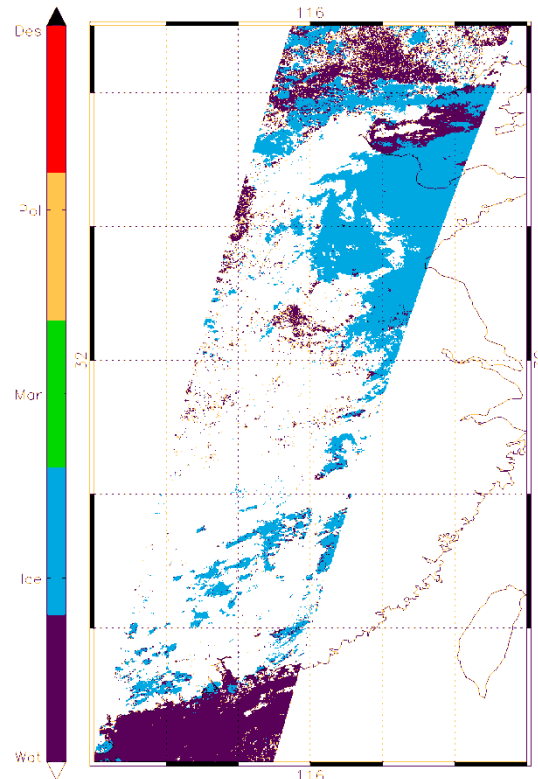


Normalised χ^2 probability of best-type.

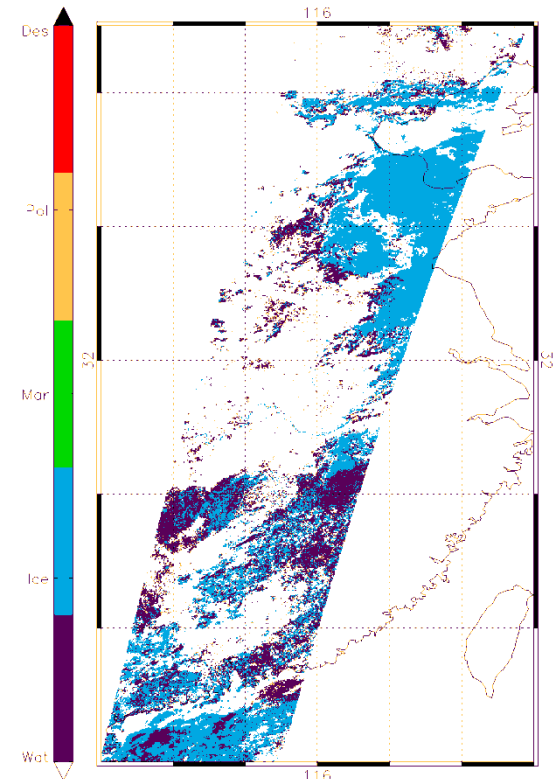
Does it work?



AATSR false colour



Cloud_cci cloud phase



χ^2 cloud phase. Either:

- $P_n > 0.75$ of water or ice
- Sum of $P_n > 0.85$ for water and ice

Concluding remarks

- A lot more work is to be done with the Bayesian scene identification
- The method shows both promise and potential problems. In this case:
 - Is not “tricked” by Chinese haze or sediment laden coastal waters.
 - The latter in particular seems to be a problem with the neural net mask.
 - Can fit very thin water cloud to (what appear to be) some clear sky pixels.
 - We don’t really get a cloud mask.
 - The question we are asking is “is our forward model consistent with observations”?

Extra slides

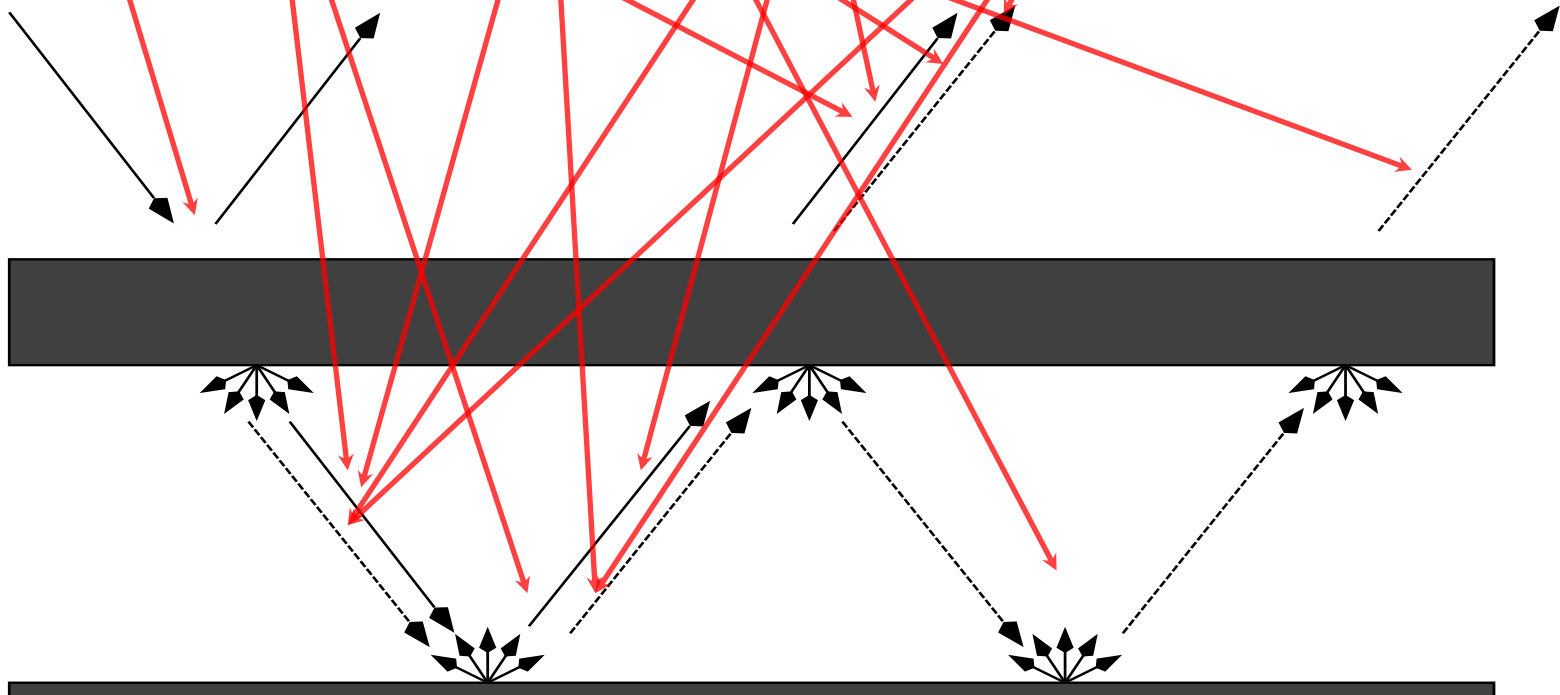
ORAC surface BRDF treatment

The approach taken is to treat direct and diffuse transmission separately:

$$R_{\text{TOA}} = R_{\text{bb}} + T_{\text{bb}\downarrow} \rho_{\text{bb}} T_{\text{bb}\uparrow} + T_{\text{bb}\downarrow} \rho_{\text{bd}} T_{\text{db}\uparrow} + T_{\text{bd}\downarrow} \rho_{\text{db}} T_{\text{bb}\uparrow} + T_{\text{bd}\downarrow} \rho_{\text{dd}} T_{\text{db}\uparrow} +$$

$$T_{\text{bb}\downarrow} \rho_{\text{bd}} R_{\text{dd}} \rho_{\text{db}} T_{\text{bb}\uparrow} + T_{\text{bb}\downarrow} \rho_{\text{bd}} R_{\text{dd}} \rho_{\text{dd}} T_{\text{db}\uparrow} + T_{\text{bd}\downarrow} \rho_{\text{dd}} R_{\text{dd}} \rho_{\text{db}} T_{\text{bb}\uparrow} + T_{\text{bd}\downarrow} \rho_{\text{dd}} R_{\text{dd}} \rho_{\text{dd}} T_{\text{db}\uparrow} +$$

...



ORAC surface BRDF treatment

- The forward model currently used in ORAC results from making the approximation:

$$T_{bd}^{\downarrow} \rho_{db} T_{bb}^{\uparrow} + T_{bd}^{\downarrow} \rho_{dd} T_{db}^{\uparrow} \approx T_{bd}^{\downarrow} \rho_{dd} T_{tb}^{\uparrow}$$

where $T_{tb}^{\uparrow} = T_{bb}^{\uparrow} + T_{db}^{\uparrow}$ is the total transmission of the atmosphere in the viewing direction.

- Applying this approximation, collecting terms and applying a series limit, results in:

$$R_{TOA} = R_{bb} + T_{bb}^{\downarrow} (\rho_{bb} - \rho_{dd}) T_{bb}^{\uparrow} + \frac{(T_{bb}^{\downarrow} \rho_{bd} + T_{bd}^{\downarrow} \rho_{dd}) T_{db}^{\uparrow}}{1 - \rho_{dd} R_{dd}}$$

ORAC surface BRDF treatment

- The constraints required to define the different surface reflectance terms (and the wavelength variation in surface reflectance, for single view retrievals) come from the a priori value
- Thus, we need a way of setting an a priori surface reflectance that includes an accurate representation of the BRDF
- LAND: MODIS BRDF (MCD43B)
- OCEAN: An sea surface reflectance model, incorporating:
 - reflectance off wave slopes from Cox and Munk statistics + ECMWF 10m wind components
 - whitecap reflectance (again as a function of 10m wind)
 - volumetric scattering, using chlorophyll-a and CDOM products from GlobCOLOUR
 - Sayer et al., Atmos. Meas. Tech., 3, 813-838, 2010.

The Swansea-style forward model

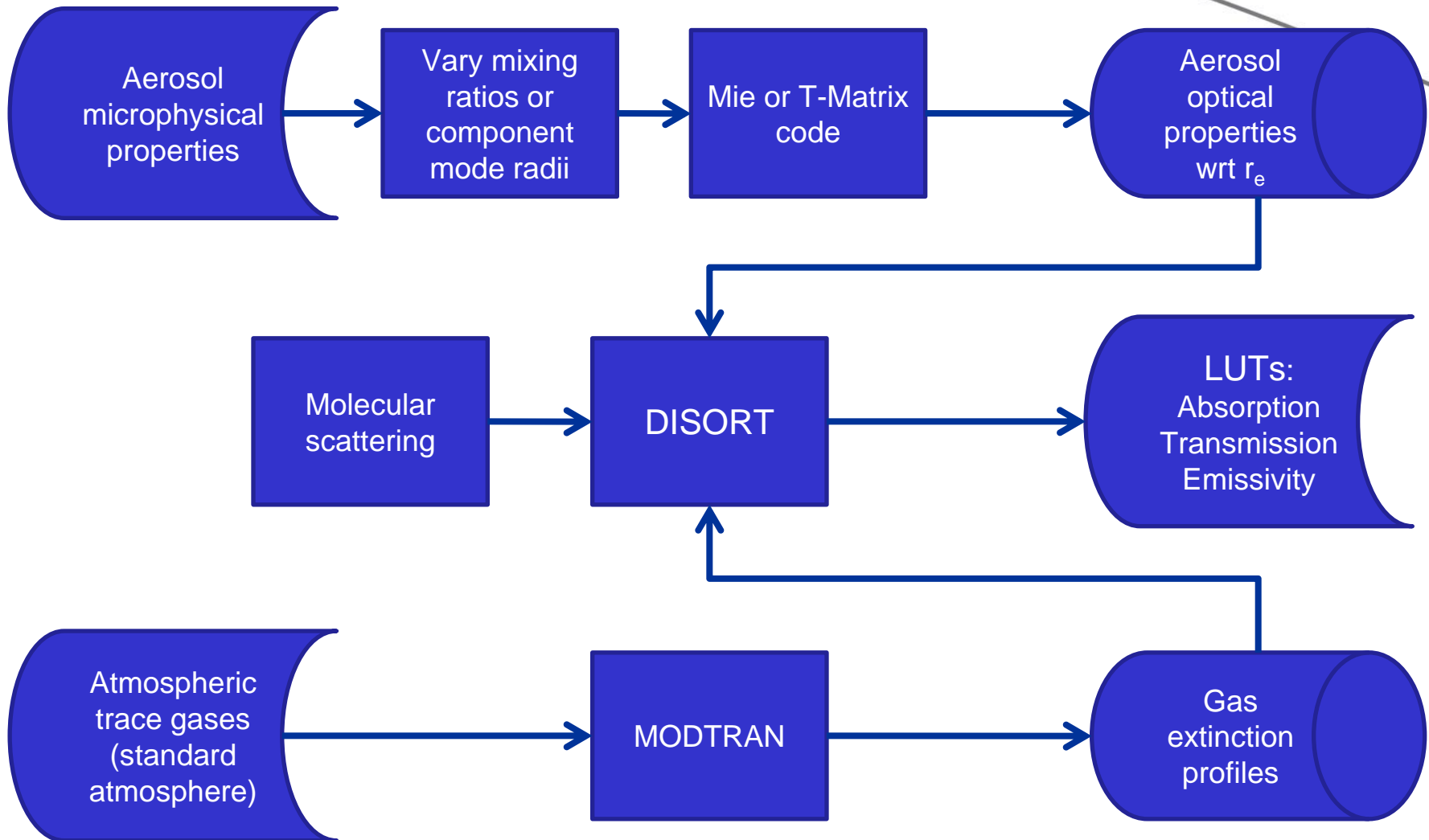
- The Swansea forward model consists of two parts:
 - An equation for the surface reflectance, as a function of 6 model parameters (for AATSR) and the diffuse fraction of downwelling radiance:

$$R_{\text{surf}}(\lambda, \omega) = (1 - D(\lambda)) \left[P(\omega)s(\lambda) + \frac{g\gamma s(\lambda)}{1 - g} \right] + D(\lambda) \frac{\gamma s(\lambda)}{1 - g}$$
$$g = (1 - \gamma)s(\lambda)$$

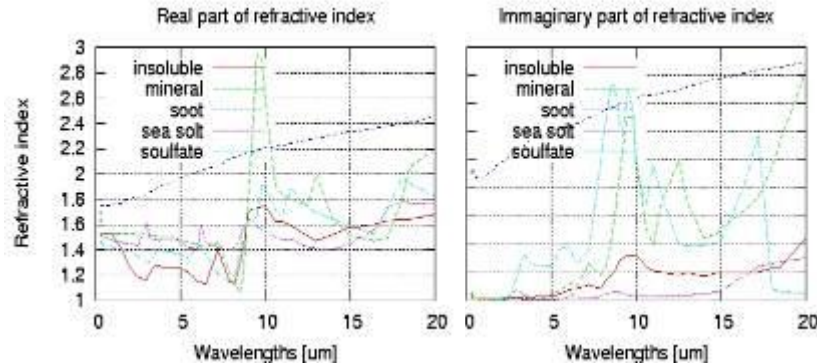
- The standard equation for TOA radiance widely used in aerosol and cloud retrieval algorithms:

$$R_{\text{toa}} = R_{\text{atm}}(\tau, r_e, \omega_s, \omega_v) + \tilde{T}(\tau, r_e, \omega_s, 2\pi) \frac{R_{\text{surf}}}{1 - R_{\text{surf}} R_{\text{atm}}(\tau, r_e, 2\pi, 2\pi)} \tilde{T}(\tau, r_e, 2\pi, \omega_v)$$

Aerosol models



Aerosol models



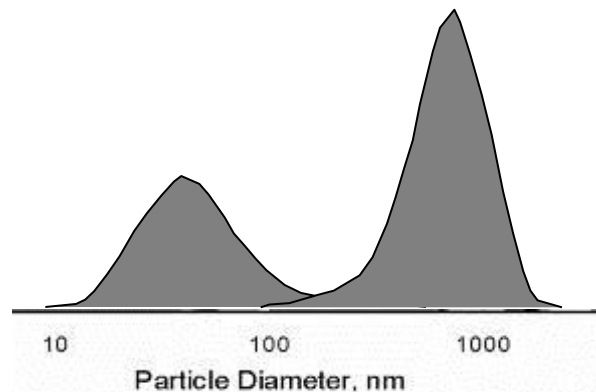
Every component is characterized by:

Spectral refractive index $n(\lambda) + i k(\lambda)$

Mode radius r_m and spread σ
log normal size distribution by number

Aspect ratio distribution

$$N(r) = \frac{C}{\ln(\sigma)r\sqrt{2\pi}} \exp\left\{-\frac{[\log(r/r_m)]^2}{2\ln^2(\sigma)}\right\}$$



Changing the mixing ratio between components we obtain the optical properties corresponding to different effective radii

$$r_e = \frac{\int_0^\infty r^3 n(r) dr}{\int_0^\infty r^2 n(r) dr}$$



$K_{\text{ext}}(\lambda)$

$0 < \omega(\lambda) < 1$

$P(\theta, \lambda)$