

Maximising the continuity of synergistic cloud-aerosol-precipitation retrievals between the A-Train and EarthCare

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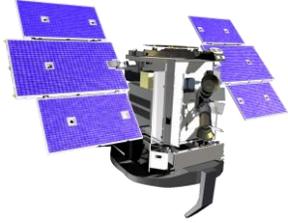


**National Centre for
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NATURAL ENVIRONMENT RESEARCH COUNCIL

CloudSat

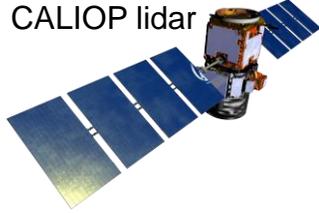
Cloud Profiling Radar (CPR)



A-Train

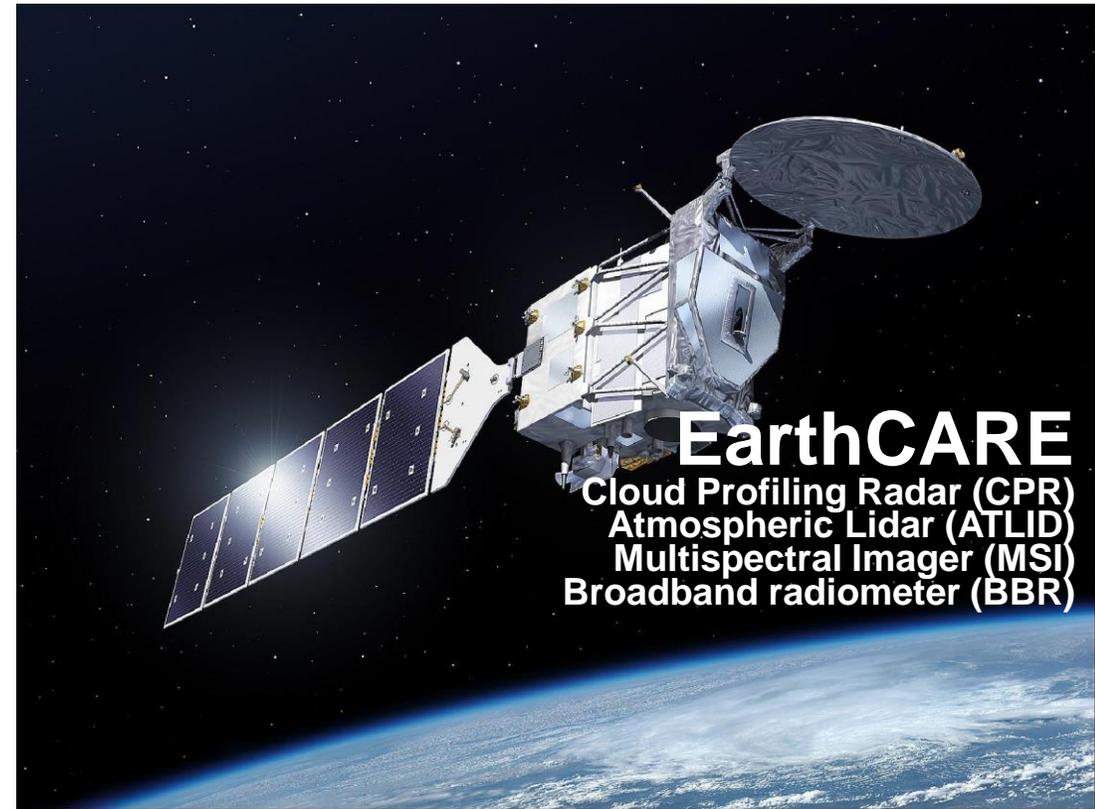
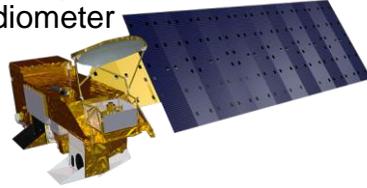
CALIPSO

CALIOP lidar



Aqua

MODIS imaging radiometer
CERES broadband radiometer



EarthCARE

Cloud Profiling Radar (CPR)
Atmospheric Lidar (ATLID)
Multispectral Imager (MSI)
Broadband radiometer (BBR)

ICE
PRECIP-COLUMN
RAIN-PROFILE
SNOW-PROFILE
CloudSat-only retrievals

CALIPSO
Ice cloud and aerosol
profile retrievals

MODIS
Passive cloud and aerosol
retrievals

DARDAR
CloudSat-CALIPSO-
MODIS retrieval
of ice cloud & snow

CCCM
CERES-CloudSat-CALIPSO-
MODIS composite retrievals
on CERES footprint

CWC-RVOD
CloudSat-MODIS retrieval
of ice & liquid clouds

CCM-CAP
Unified & synergistic
retrieval of cloud-
aerosol-precipitation

CAPTIVATE
← Clouds, Aerosol and Precipitation
from multiple Instruments using a
Variational Technique →

ACM-CAP
Unified & synergistic
retrieval of cloud-
aerosol-precipitation

CCM-RAD
Radiative fluxes,
heating rates and
radiative closure
at nadir

CCM-3D
3D scene reconstruction
CCM-RT
Radiative transfer

ACM-3D
3D scene reconstruction
ACM-RT
Radiative transfer

ACMB-DF
Top-of-atmosphere
radiative closure
assessment

ATLID+CPR+MSI ↓

BBR ↓



input products

CALIPSO

CALIOP
attenuated
backscatter

β

CloudSat

Radar reflectivity
& path-integrated
attenuation (PIA)

Z, PIA

MODIS

Solar and thermal
infrared radiances

I, T_B

ECMWF

Meteorological
parameters &
surface properties
on A-Train track

T, q, O_3
 T_s, ε, A

DARDA

R
Synergistic
target
Classification

$x_j = f(T)$

surface properties

+ error estimates
and detection status
for each measurement

observation vector y

CCM-CAP

Cloud-Aerosol-
Precipitation from
CloudSat-CALIPSO-MODIS

output
product

CAPTIVATE

state vector x

First iteration: a-priori state vector x_a

ice and snow α, N_0, s	liquid cloud L, N	rain R, N_s	melting layer X_m	aerosol N
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forward model $H(x)$

Inc. multiple scattering & hydrometeor fallspeed

radar $Z, V_D,$ PIA	lidar β	radiometer I, T_B
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cost function J

Compare $y - H(x)$ & $x - x_a$
Check for convergence

minimizer

Update the
state vector

if converged

if not
converged

calculate retrieval error

Error covariances & averaging
kernels for retrieved and derived
variables

Kalman smoother

for selected
state variables

next
profile

1-way on first (backward) pass
and 2-way on second (forward) pass

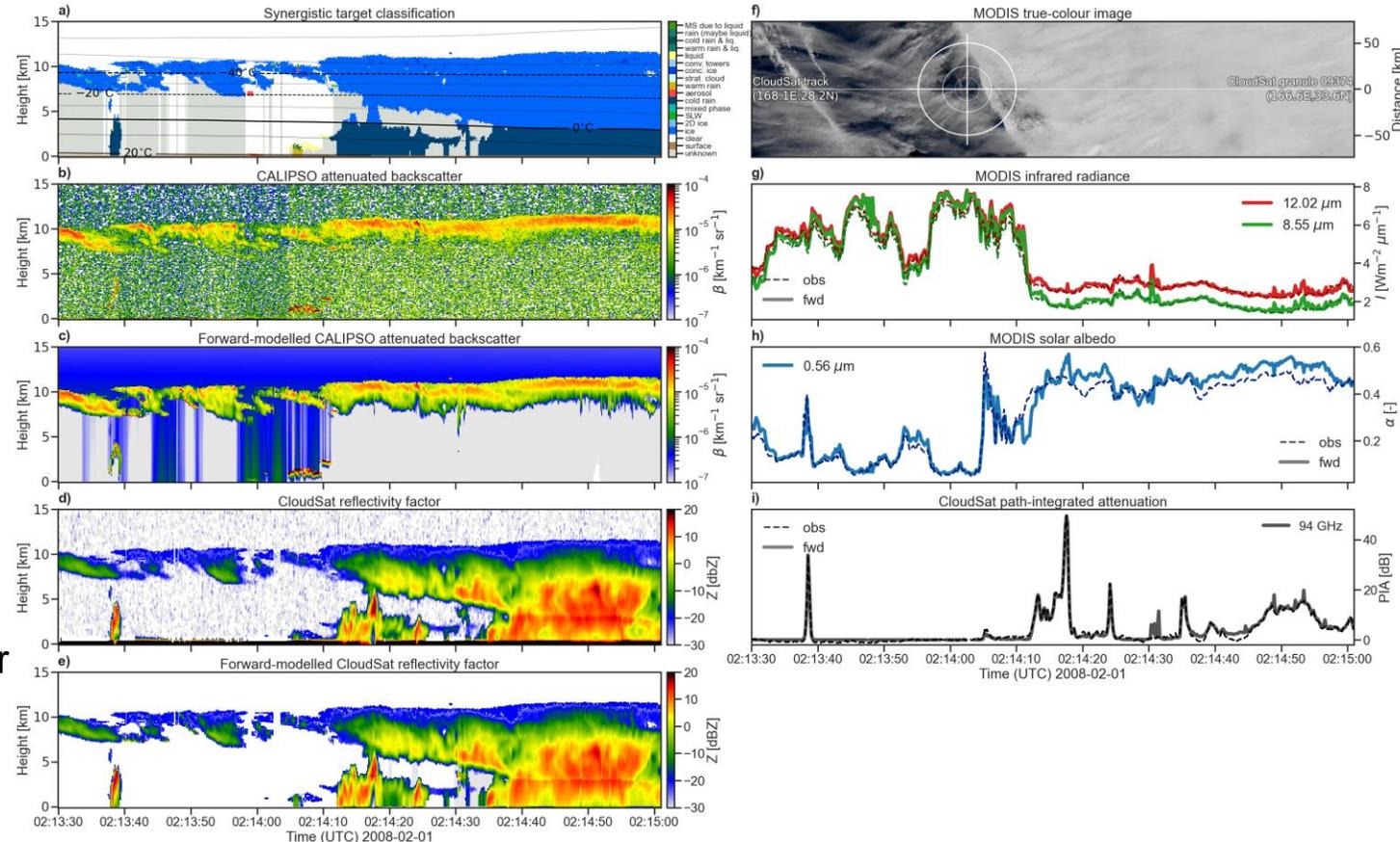
CAPTIVATE: key features

- Advantages of synergistic retrievals are well-established:

- radar-lidar synergy is greatest for ice cloud; most other components rely on one or the other active sensor
- But some measurements act as integrated constraints with contributions from different components of the atmosphere (radar PIA from liquid cloud and rain; solar radiances from ice and liquid cloud)

- Advantages of a unified retrieval in complex and layered cloud scenes

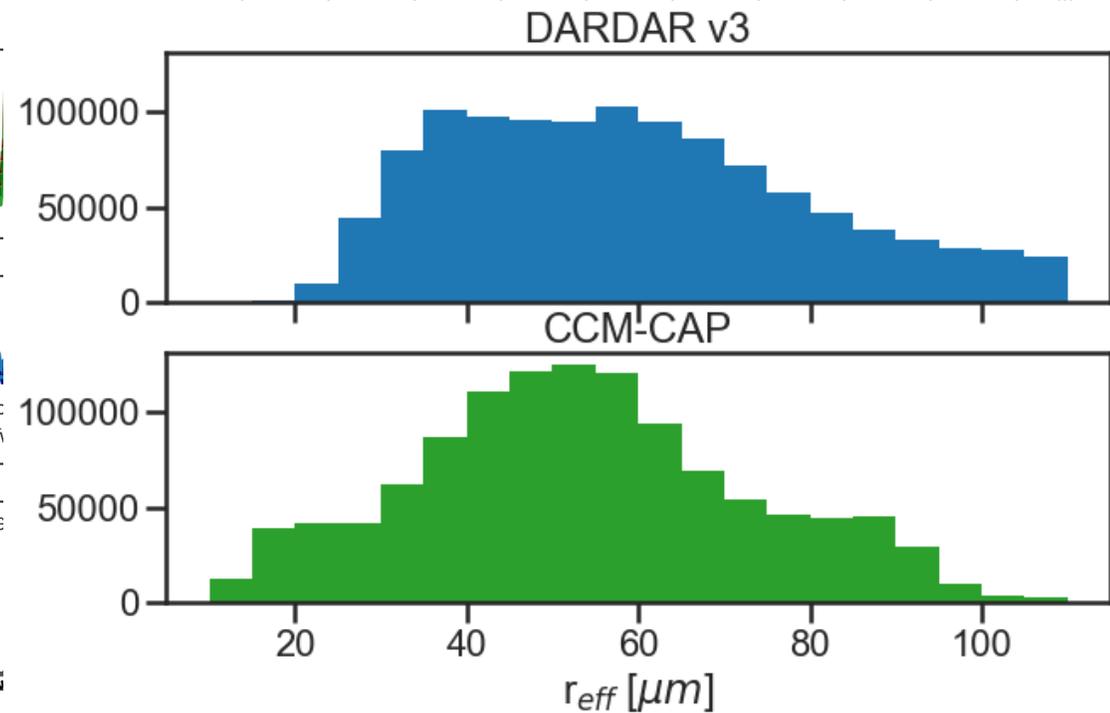
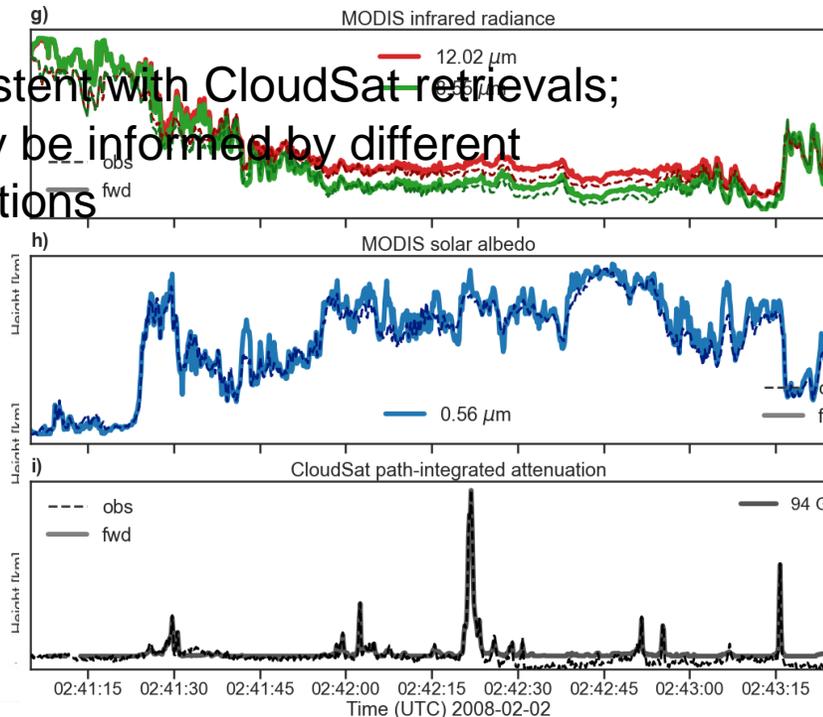
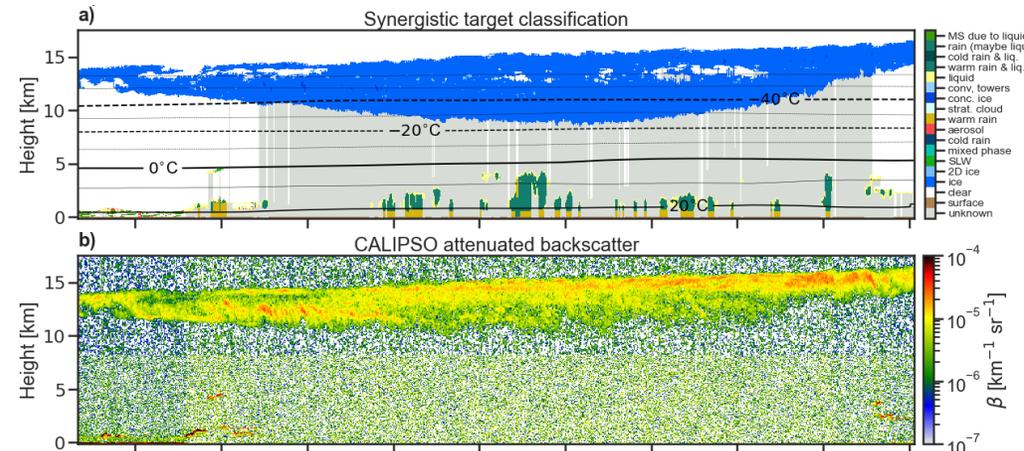
- Mass flux conserved across the melting layer
- Mixed-phase/supercooled liquid layers
- Ice clouds over warm liquid cloud
- Drizzle/rain colocated with liquid cloud



CCM-CAP: Ice and snow

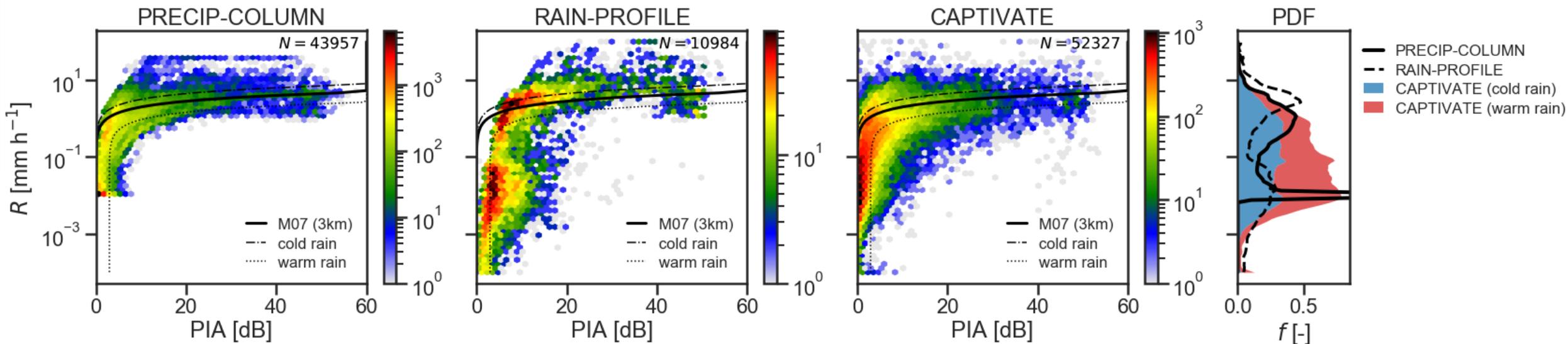
- Retrievals of optically thin ice at cloud-top are under-constrained (lidar-only)
 - Assimilating the radar noise floor provides an upper limit on the forward-modelled radar reflectivity of ice clouds retrieved from lidar-only measurements
 - Results in an overall reduction in effective radius compared to previous retrievals (e.g. DARDAR), consistent with very high-sensitivity airborne radar-lidar retrievals, but not previously

- Ice water path consistent with CloudSat retrievals; higher snow rates may be informed by different microphysical assumptions



CCM-CAP: Drizzle & rain

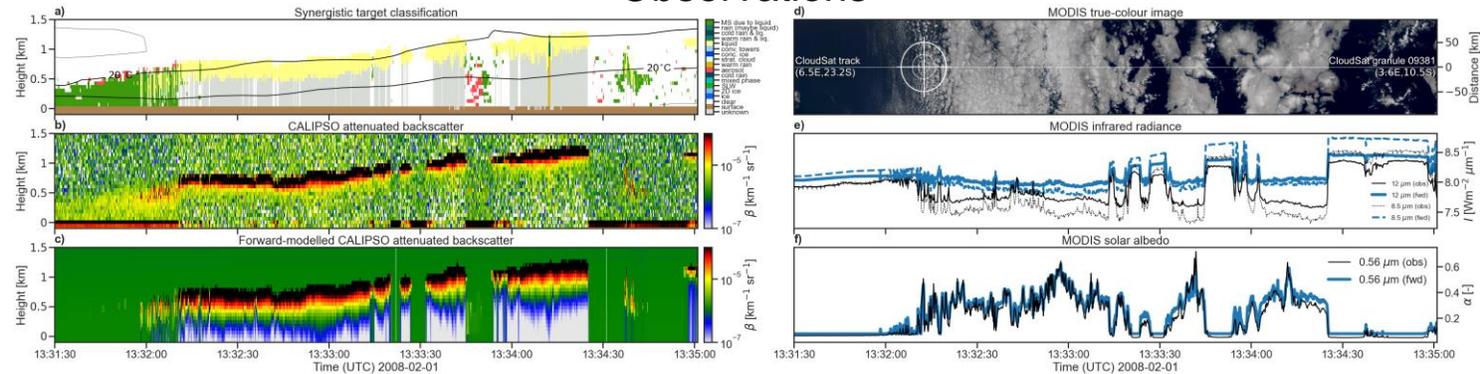
- Rain drop size distribution (DSD) scales as a function of rain rate (Abel & Boutle, QJRMS, 2012), but can also retrieve deviations from this relation when constrained by observations
 - PIA can provide a strong constraint on retrieving DSDs with high concentrations of small drops in warm rain (Mason et al., *ACP*, 2017)
- Rain rate and liquid cloud both contribute to PIA
- Compared with CloudSat rain products
 - PRECIP-COLUMN doesn't resolve features within the profile (e.g. evaporation or collision-coalescence), but is well-constrained by PIA
 - RAIN-PROFILE has distinct attenuating and non-attenuating regimes



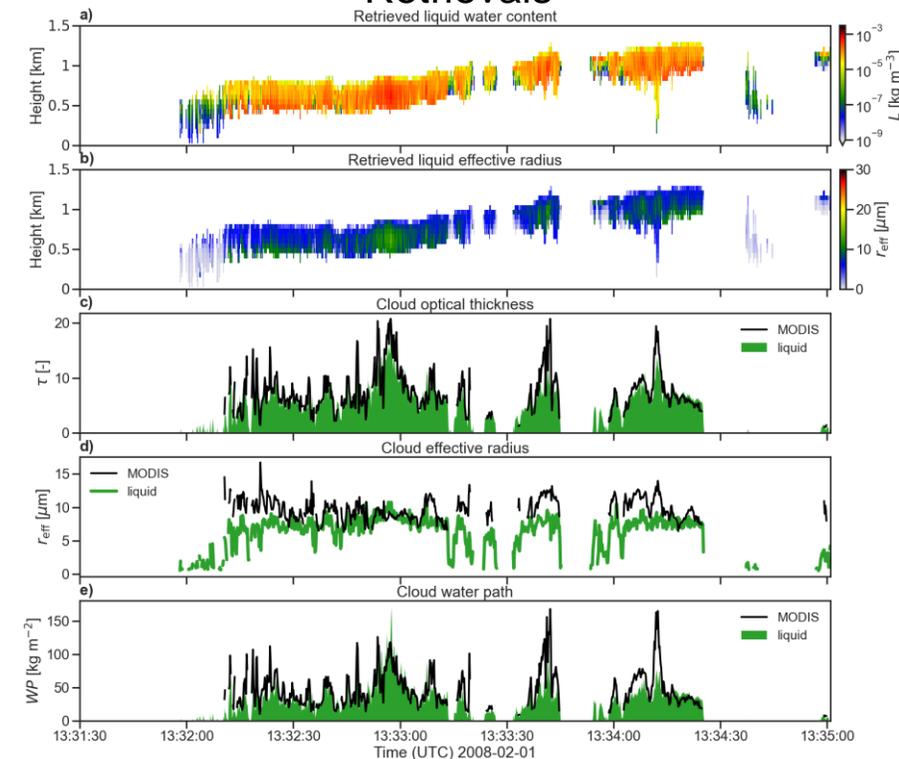
CCM-CAP: liquid cloud

- Retrievals of non-precipitating liquid cloud closely resemble MODIS cloud retrievals
- But it's likely that a majority of liquid cloud goes undetected by spaceborne radar-lidar
 - radar is sensitive to precipitating hydrometeors, while lidar is quickly extinguished near cloud-top
 - Missing liquid clouds have a strong shortwave/solar radiance signal, so can't be ignored from a cloud-radiation perspective
 - Have demonstrated some success by coarsely represent missing liquid cloud with the assumption that liquid cloud is colocated with rain.

Observations

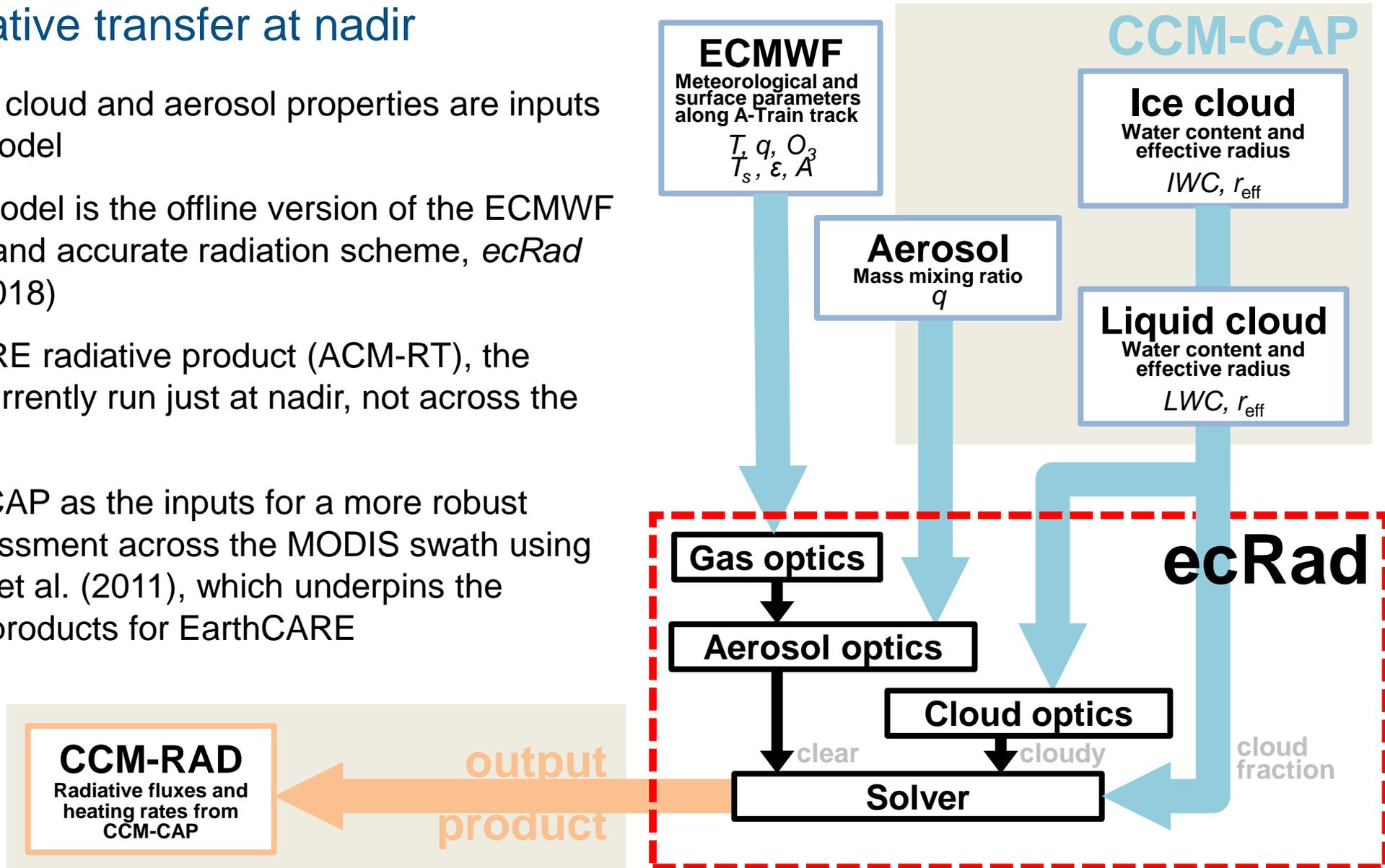


Retrievals



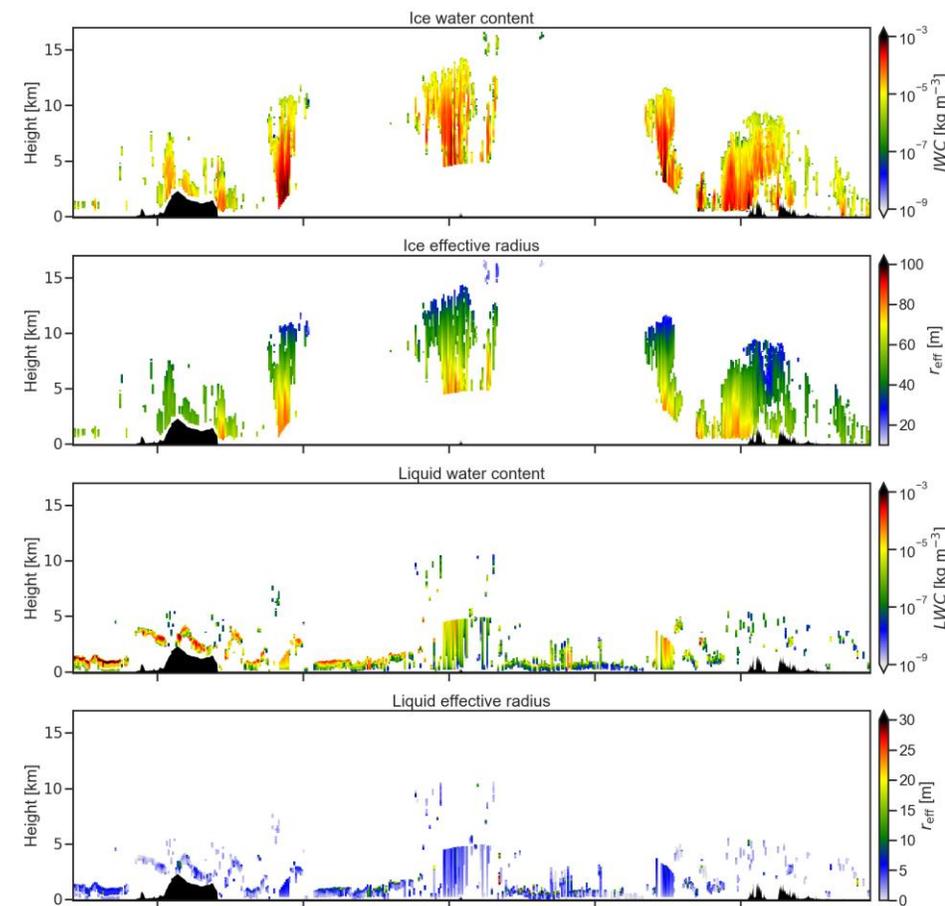
CCM-RAD: radiative transfer at nadir

- CCM-CAP retrieved cloud and aerosol properties are inputs to radiative transfer model
- Radiative transfer model is the offline version of the ECMWF forecast model's fast and accurate radiation scheme, *ecRad* (Hogan and Bozzo, 2018)
- Unlike the EarthCARE radiative product (ACM-RT), the radiative transfer is currently run just at nadir, not across the MODIS swath
- We also use CCM-CAP as the inputs for a more robust radiative closure assessment across the MODIS swath using the method of Barker et al. (2011), which underpins the ACM-3D & ACM-RT products for EarthCARE



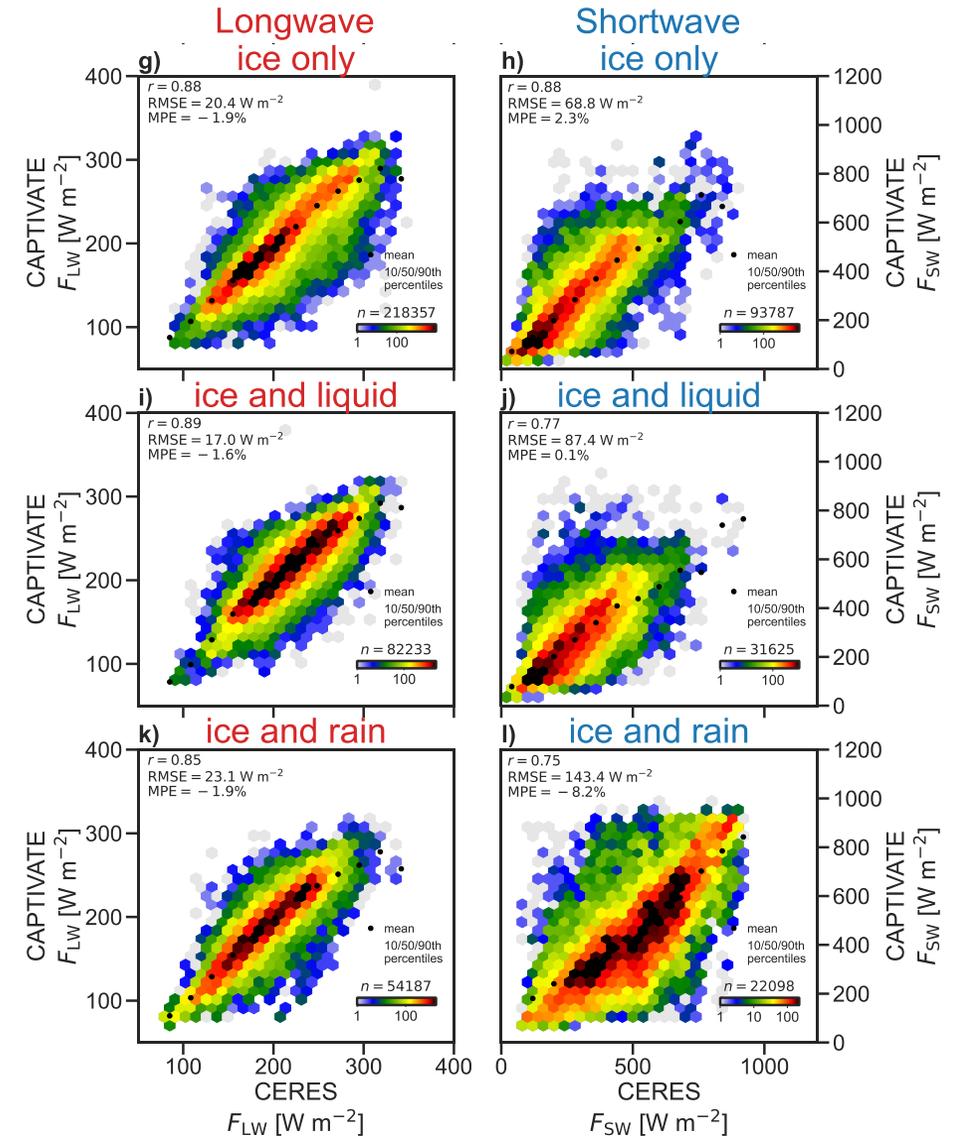
CCM-RAD: radiative fluxes & heating rates

- ecRad computes LW & SW radiative fluxes through the atmosphere
- Radiative heating rates calculated from net radiative flux over each layer
- Top-of-atmosphere (TOA) fluxes vs CERES measurements
- LW radiative closure well-constrained; determined by cloud-top height/temperature
- SW radiative closure more sensitive to cloud phase, retrieved properties and surface albedo
- Assumption of liquid cloud in rain improves shortwave fluxes at TOA, but drastically changes heating rates in the lower atmosphere
 - Cloud base height assumption from spaceborne radar-lidar could be evaluated using surface fluxes



CCM-RAD: radiative closure

- With coarse distinctions between cloud regimes; more careful selection could be made to target specific processes
- Radiative closure assessment provides a built-in evaluation of biases in the radiative properties of clouds retrieved by CCM-CAP. This can be used to quantify improvements, e.g. in microphysical properties of ice clouds
- Increased noise attributed to inherent mismatch between the ~1km scale of CloudSat-CALIPSO measurements and the ~20km CERES footprint.
 - Should be substantially reduced by carrying out across-track radiative closure assessment (Barker et al. 2011 & ACM-RT)
- SW closure more challenging than LW



Summary

- CCM-CAP: unified and synergistic retrieval of clouds, aerosols and precipitation from the A-Train
 - Advantages over a patchwork of single-instrument and synergistic retrievals, especially in complex and layered scenes
 - Continuity with EarthCARE synergistic retrievals
 - Current applications:
 - Testing and development of EarthCARE synergistic retrievals
 - Evaluation of ECMWF integrated forecast system
 - Forward-modelling of novel microwave/radar sensors
- Downstream products: radiative fluxes and heating rates
 - CCM-RAD at nadir using ECMWF radiation scheme; Radiative closure vs CERES provides a built-in evaluation of CCM-CAP
 - Also providing CCM-CAP as input to 3D radiative transfer methodology (Barker et al.
- Paper and initial dataset to follow
 - Starting with 1 year of A-Train data (2008)
 - To expand to first 5 years of A-Train (2006 to 2011); period of optimal CloudSat-CALIPSO performance
 - Looking for users!
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