



# Liquid cloud properties from ground-based remote sensing: Still a need for research?

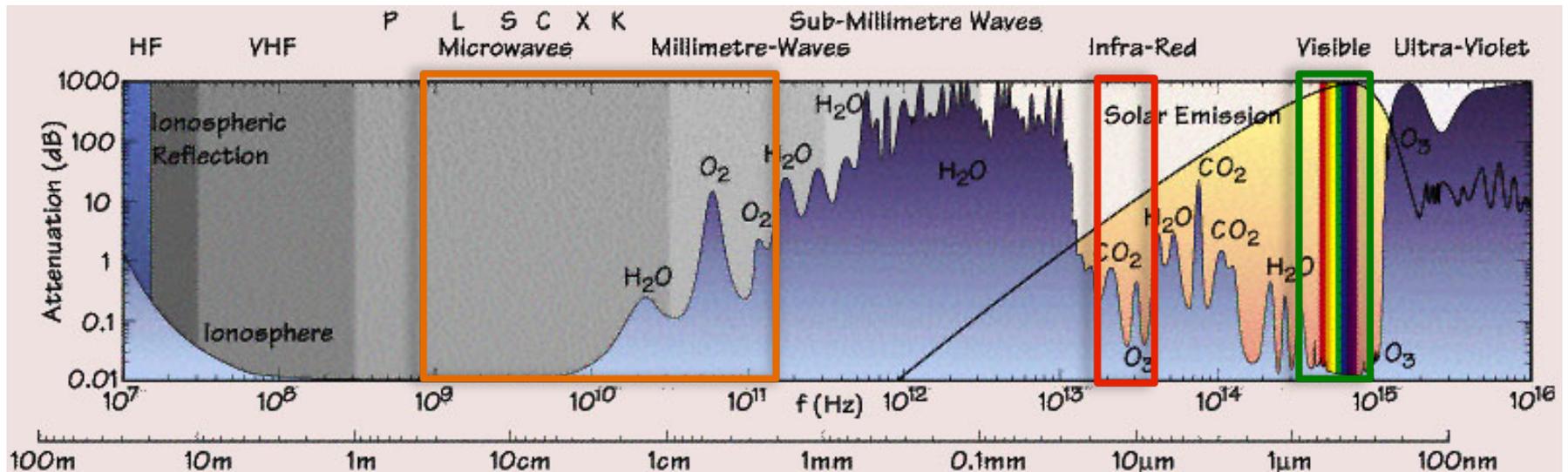
Kerstin Ebell  
(showing the results of many other people)

University of Cologne, Institute of Geophysics and Meteorology



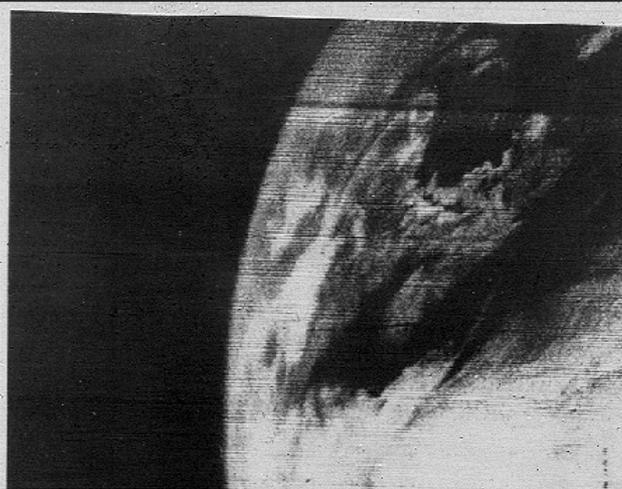
# Remote-sensing

- most remote sensing methods use electromagnetic waves; important wavelengths in so-called window regions: atmosphere is especially transparent
- windows in the region of visible radiation (**optical window**), the maximum of terrestrial radiation (**atmospheric window**) and in the **radio window**
- Remote sensing instruments deployed on various platforms, e.g. satellites, aircraft, balloons, surface

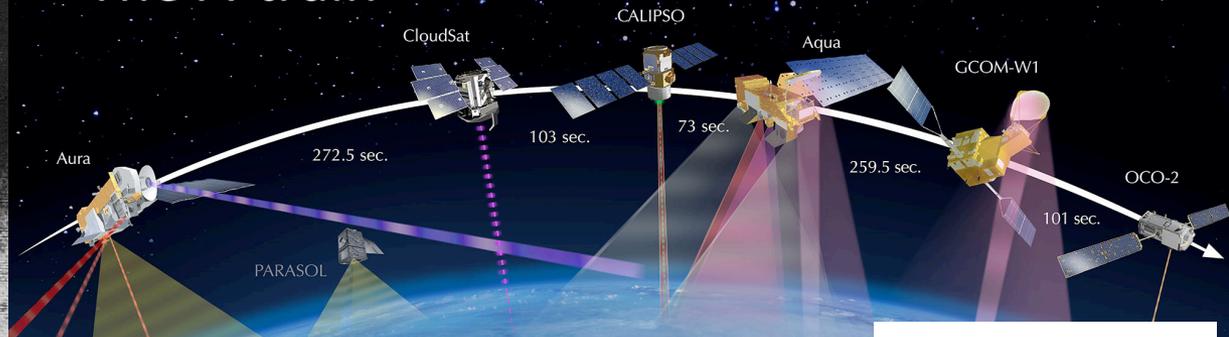


# Cloud observations from satellites...

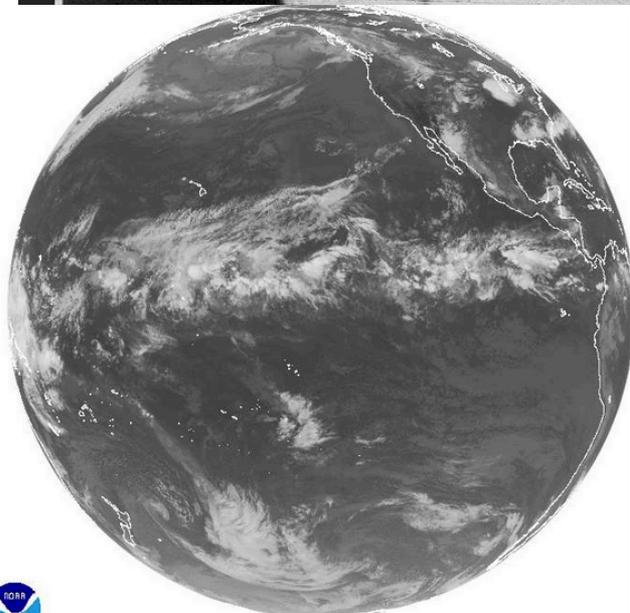
FIRST TELEVISION PICTURE FROM SPACE  
TIROS I SATELLITE  
APRIL 1, 1960



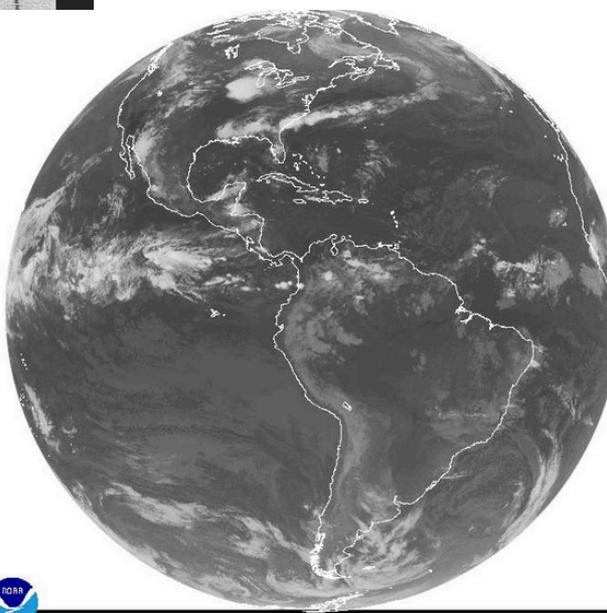
## The A-train



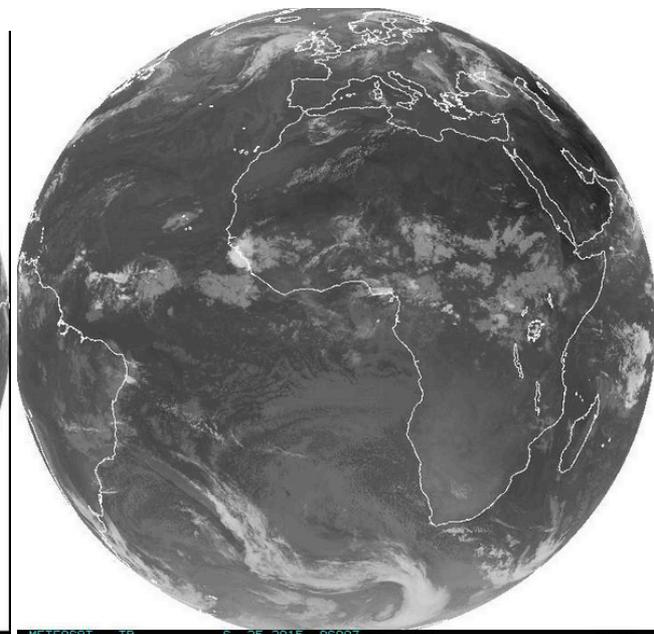
<http://atrain.nasa.gov/>



ARM Summer Training 201! GOES-West, NOAA

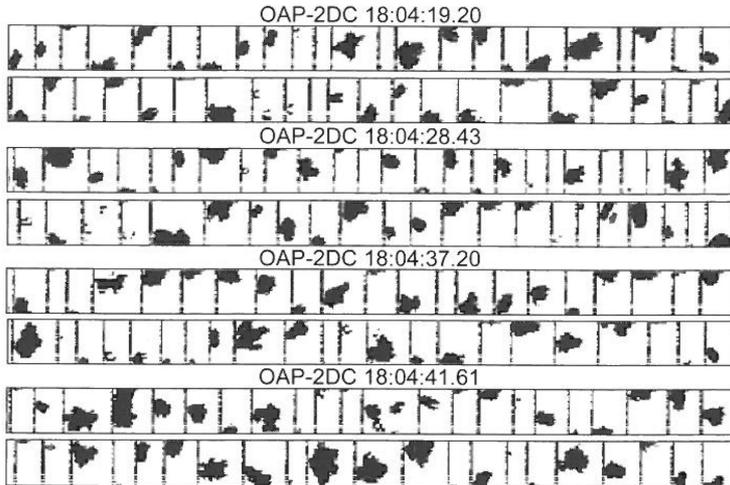


GOES-East, NOAA



Meteosat, Eumetsat

# ... and in-situ



## Optical Array Probes



**Nevezorov Probe**



Heintzenberg and Charlson (2009)



## Forward Scattering Spectrometer Probe (FSSP)



Laser beam registers every single particle. Forward scattering as measure of size ( $0.5 - 50 \mu\text{m } \varnothing$ )

Total and liquid water content  
→ ice water content

# Why don't satellites and in-situ measurements give us everything we need?

## Why ground-based?

detailed **physical process studies** through

- more direct signals
- less surface disturbances
- high temporal resolution
- interchangeable configurations

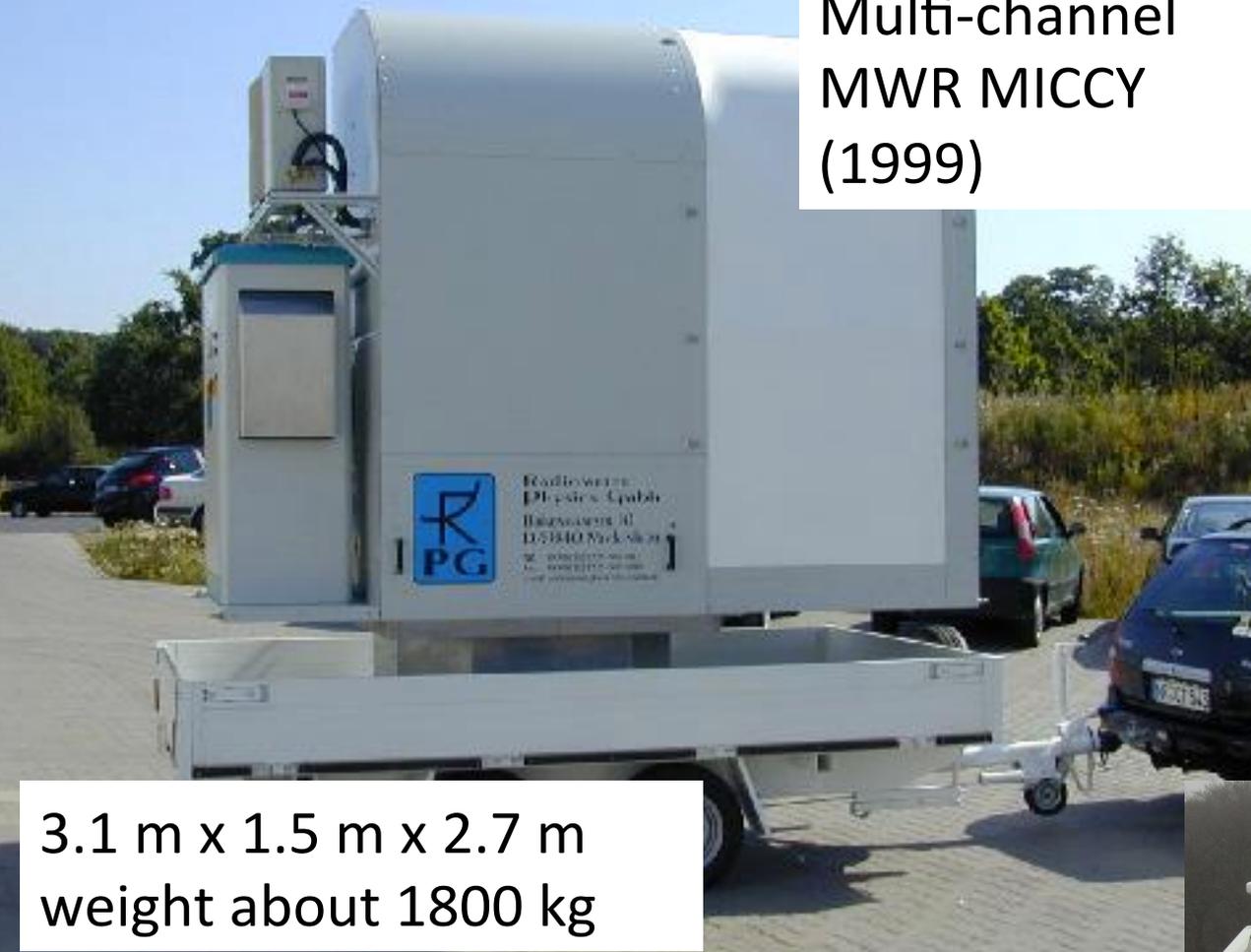
→ develop **prototypes** for future satellite missions

- **boundary layer** observations → complementary to satellites
- **continuous observations**: diurnal cycle, mesoscale meteorology

- atmospheric ground-based remote sensing **technology** developing fast (24/7, unmanned, affordable ...)

→ last decade: leaps in technological development

# Multi-channel MWR MICCY (1999)



3.1 m x 1.5 m x 2.7 m  
weight about 1800 kg

0.63 × 0.36 × 0.9 m  
weight about 60 kg

# Multi-channel MWR HATPRO (~2007)



# Why do I care about water clouds?

cirrus and cirrostratus cloud amount: 19.6 %

deep convective cloud amount: 2.6 %

International Satellite Cloud  
Climatology Project (ISCCP)  
(Rossow and Schiffer, 1999)  
→ <http://isccp.giss.nasa.gov/>

middle-level cloud amount: 19.0 %

low-level cloud amount: 27.5 %

# Radiative impact of low-level clouds

Global annual mean full sky cloud-induced radiative flux changes ( $W m^2$ ) at TOA

Cloud type	TL = TOTAL
<b>Cirrus</b>	1.3
Cirrostratus	-2.4
Deep convective	-3.3
Altostratus	-1.7
Altostratus	-6.3
Nimbostratus	-2.7
<b>Cumulus</b>	-4.6
<b>Stratocumulus</b>	-11.5
Stratus	-2.2
Sum (true)	-33.4

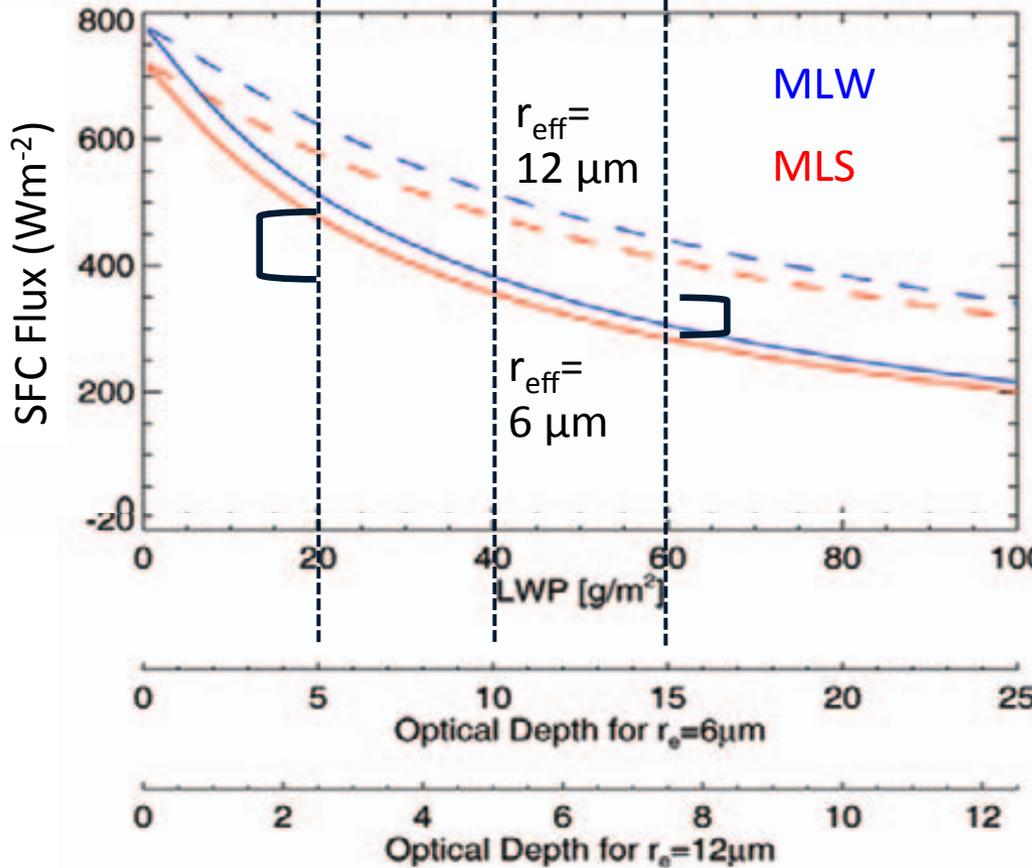
from Chen et al. (2000)

← based on ISCPP data and radiative transfer model calculations

💧 low clouds provide 55 % to 60 % to the annually averaged net cloud effect at the TOA

# Radiative flux sensitivity to LWP

Shortwave SFC flux



LWP:  $40 \text{ g/m}^2$

$-20 \text{ g/m}^2$

$\rightarrow \Delta \text{SW}_{\text{SFC}} +150 \text{ Wm}^{-2}$

$+20 \text{ g/m}^2$

$\rightarrow \Delta \text{SW}_{\text{SFC}} -70 \text{ Wm}^{-2}$

- small errors in LWP  $\rightarrow$  large errors in radiative impact of these clouds
- radiative fluxes are very sensitive to changes in LWP, if LWP is low

# What do we want to know?

- ◆ **macrophysics:** cloud yes/no, location in atmospheric column, vertical and horizontal extension, cloud fraction, overlap
- ◆ **microphysics:**
  - ◆ phase: liquid, ice, liquid+ice, melting ice,..
  - ◆ column integrated water amount (liquid water path LWP), vertical distribution of water, size of droplets → information on drop size distribution  $N(D)$

$$\text{kth moment of DSD: } M_k = \int_0^{\infty} D^k N(D) dD$$

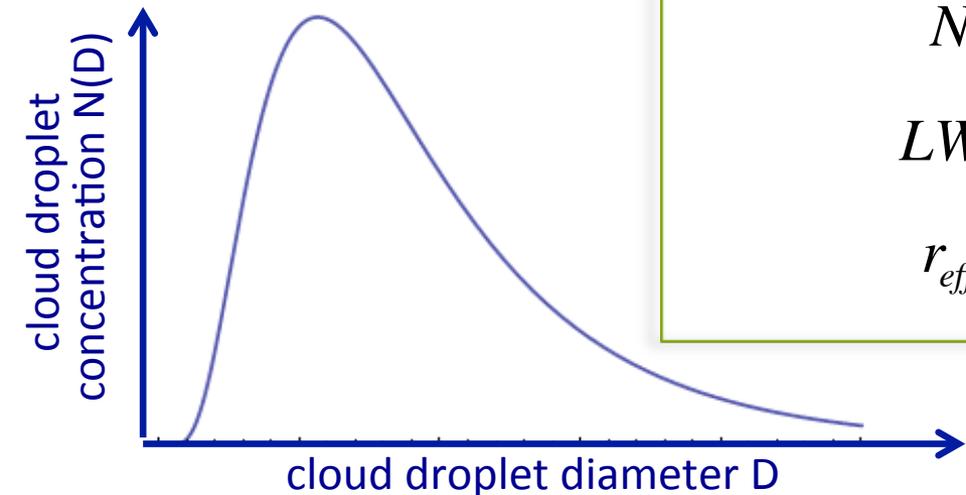
$$N_t = M_0$$

$$LWC = \frac{\pi\rho_l}{6} \int_0^{\infty} D^3 N(D) dD = \frac{\pi\rho_l}{6} M_3$$

$$r_{eff} = \frac{M_3}{M_2}$$

→ radiative properties can be parameterized by moments, e.g.:

$$\text{if } \lambda \ll r$$
$$\tau_c = \frac{3LWP}{2\rho_l r_{eff}}$$

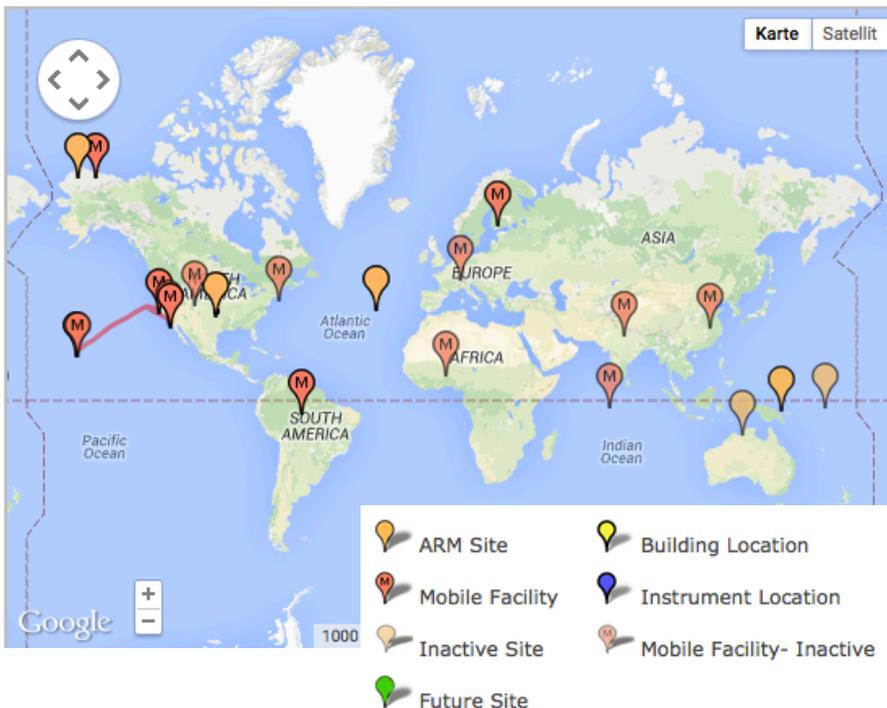


# The rise of the „supersites“

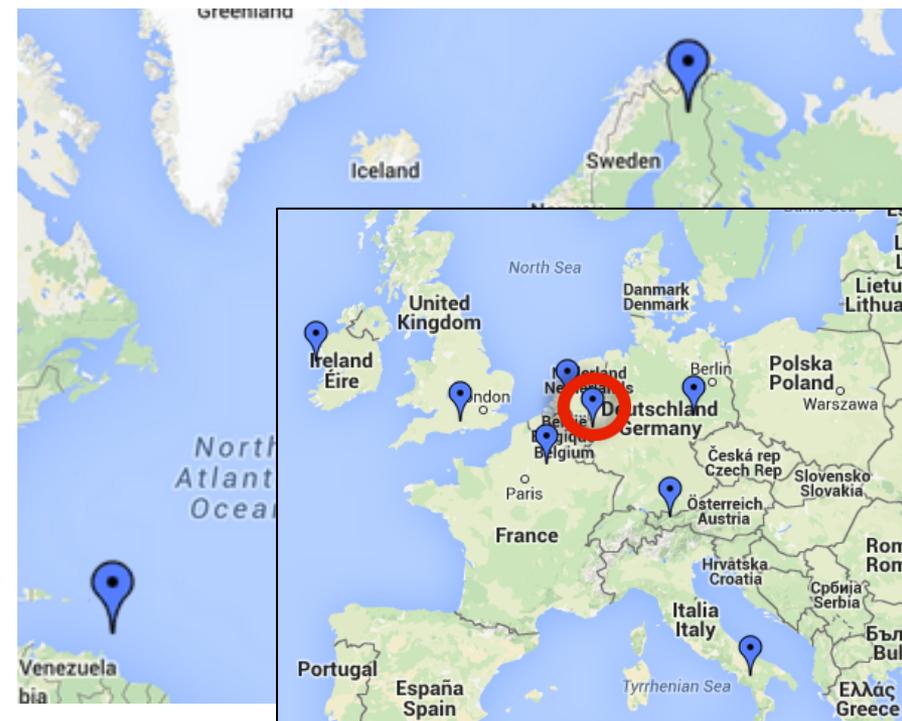
supersite = atmospheric profiling observatory that has instruments to derive vertical (line of sight) profiles of temperature, wind, humidity, aerosol, clouds and precipitation

→ long-term, continuous cloud observations

## ARM Climate research facilities



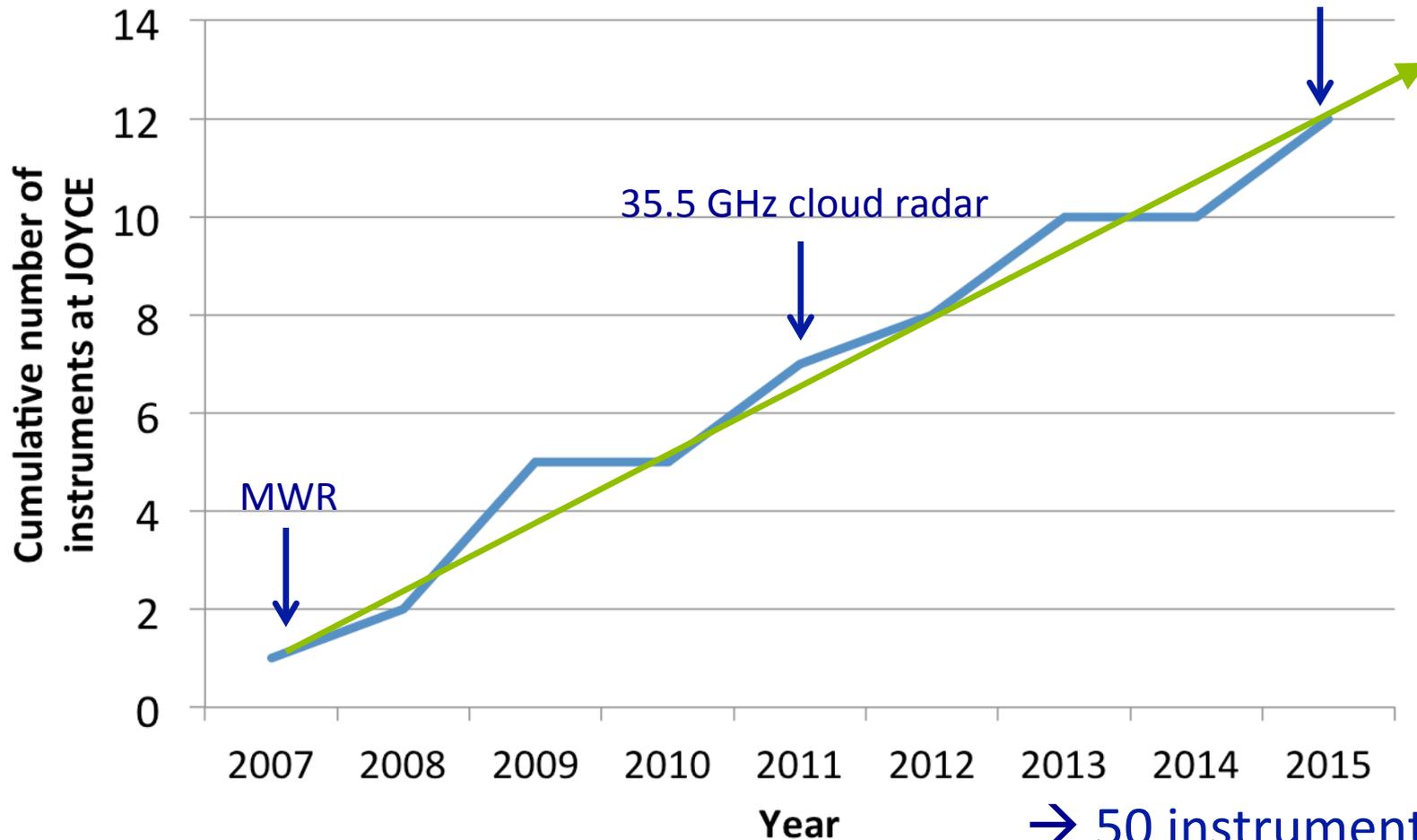
## European sites/initiatives



# Jülich Observatory for Cloud Evolution



94 GHz cloud radar  
& MFRSR



→ 50 instruments in 2043!

# JOYCE instruments

- Scanning 35 GHz cloud radar MIRA<sup>1</sup>
- Scanning 14 channel microwave radiometer<sup>2</sup> with IR pyrometer<sup>3</sup>
- Scanning Doppler wind lidar<sup>4</sup>
- Atm. emitted radiance interferometer<sup>5</sup>
- Total Sky Imager TSI<sup>6</sup>
- Laser ceilometer CT25K and CHM15k<sup>7</sup>
- Micro Rain Radar<sup>8</sup>, sodar<sup>9</sup>
- Cimel sun photometer
- Radiation sensors<sup>10</sup>
- 120 m meteorological mast<sup>11</sup> including eddy covariance station  
+ multifilter rotating shadowband radiometer and FM-CW 94 GHz cloud radar in 2015!



# Liquid cloud property check list

- vertically resolved cloud mask?
- phase identification?
- liquid water path?
- vertical distribution of liquid water?
- effective radius of cloud droplets?

# Ceilometer/Micropulse lidar

- active instrument: sends laser pulses (optical spectrum, e.g. 905 nm or 532 nm) and measures backscattered light → profiles of backscatter coefficient
- backscatter proportional to  $D^2$
- very sensitive to small particles → cloud base height = ceiling (aviat.)

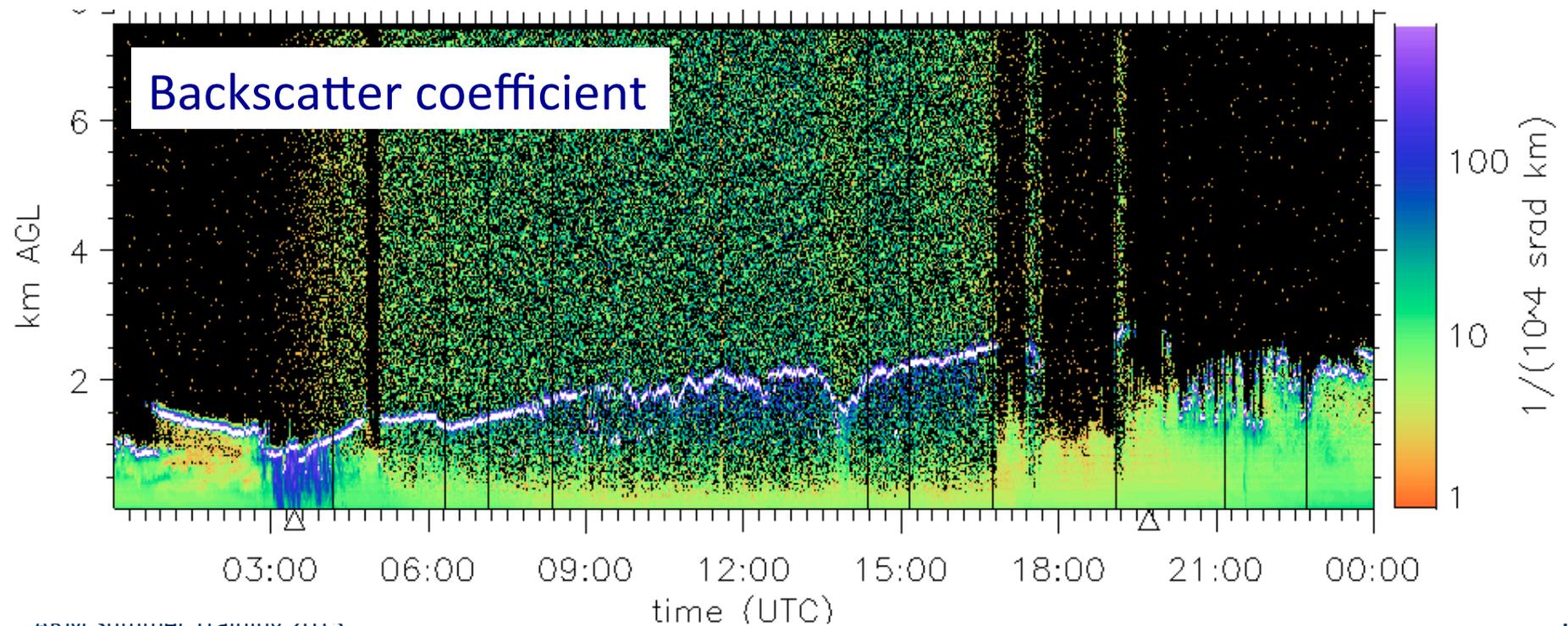


[www.arm.gov](http://www.arm.gov)



[www.arm.gov](http://www.arm.gov) 15

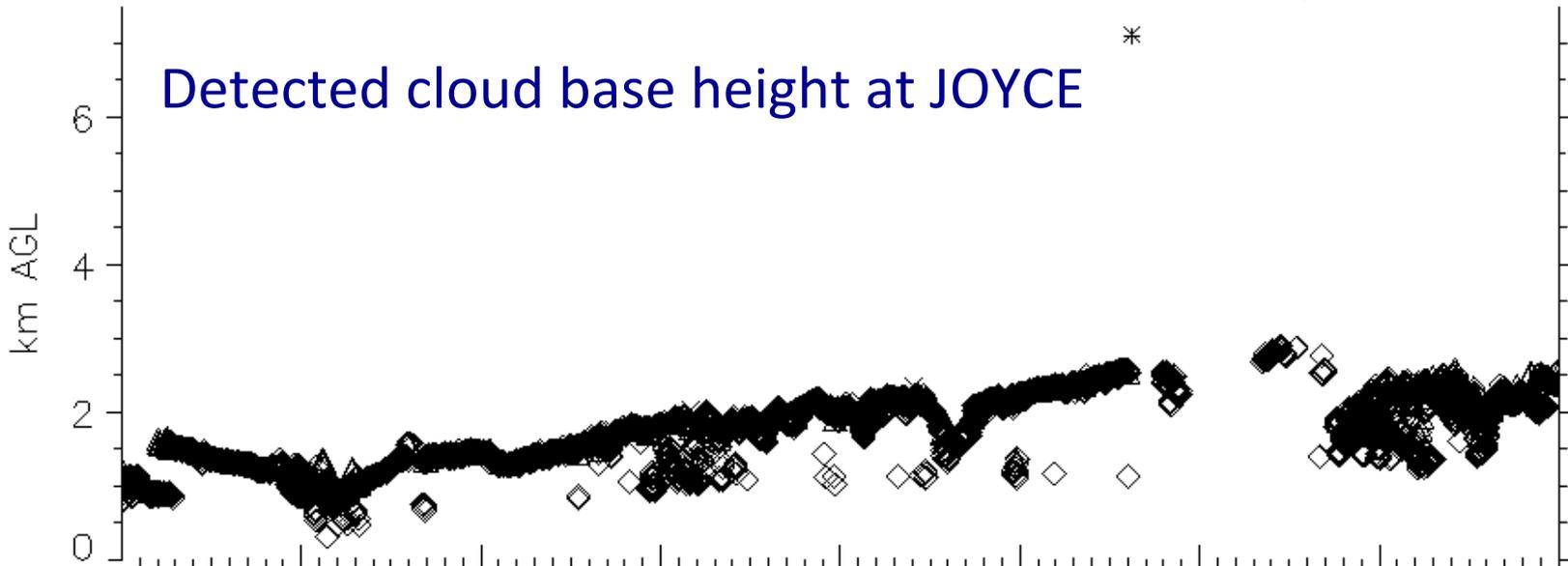
# Cloud detection/cloud base height



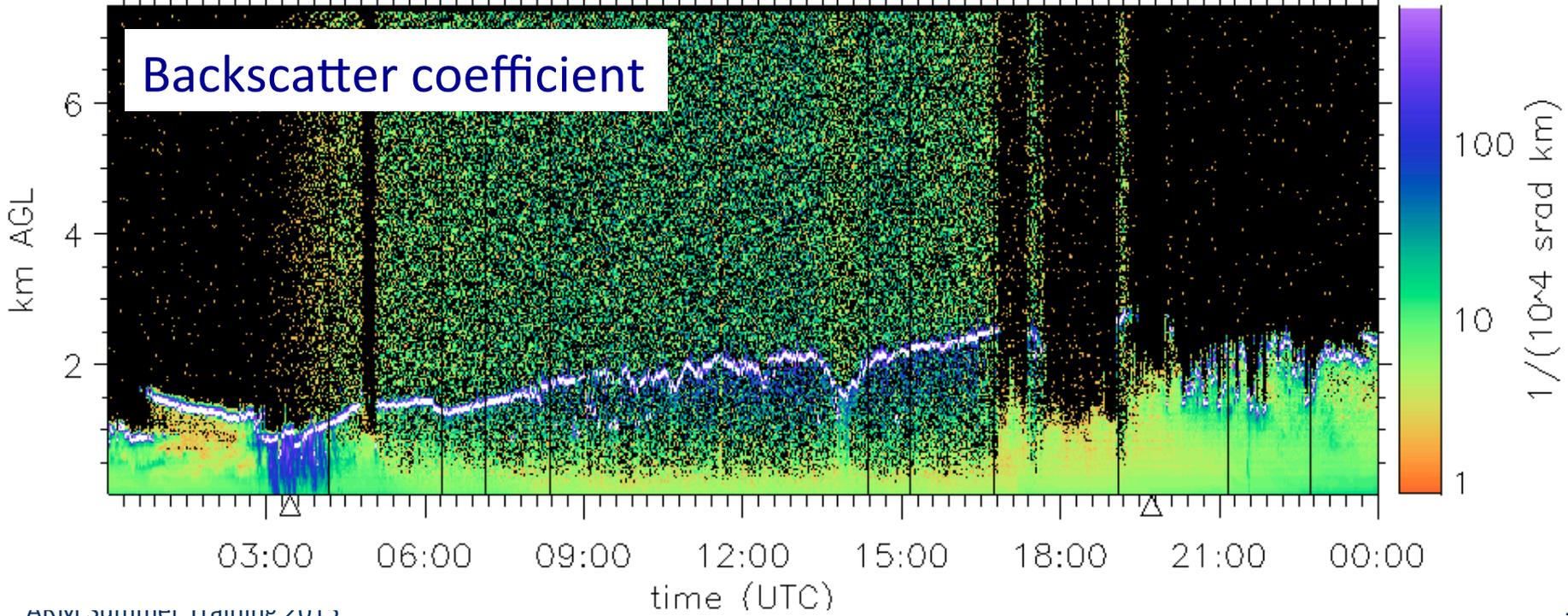
### Detected cloud base height at JOYCE

Cloud stats.:

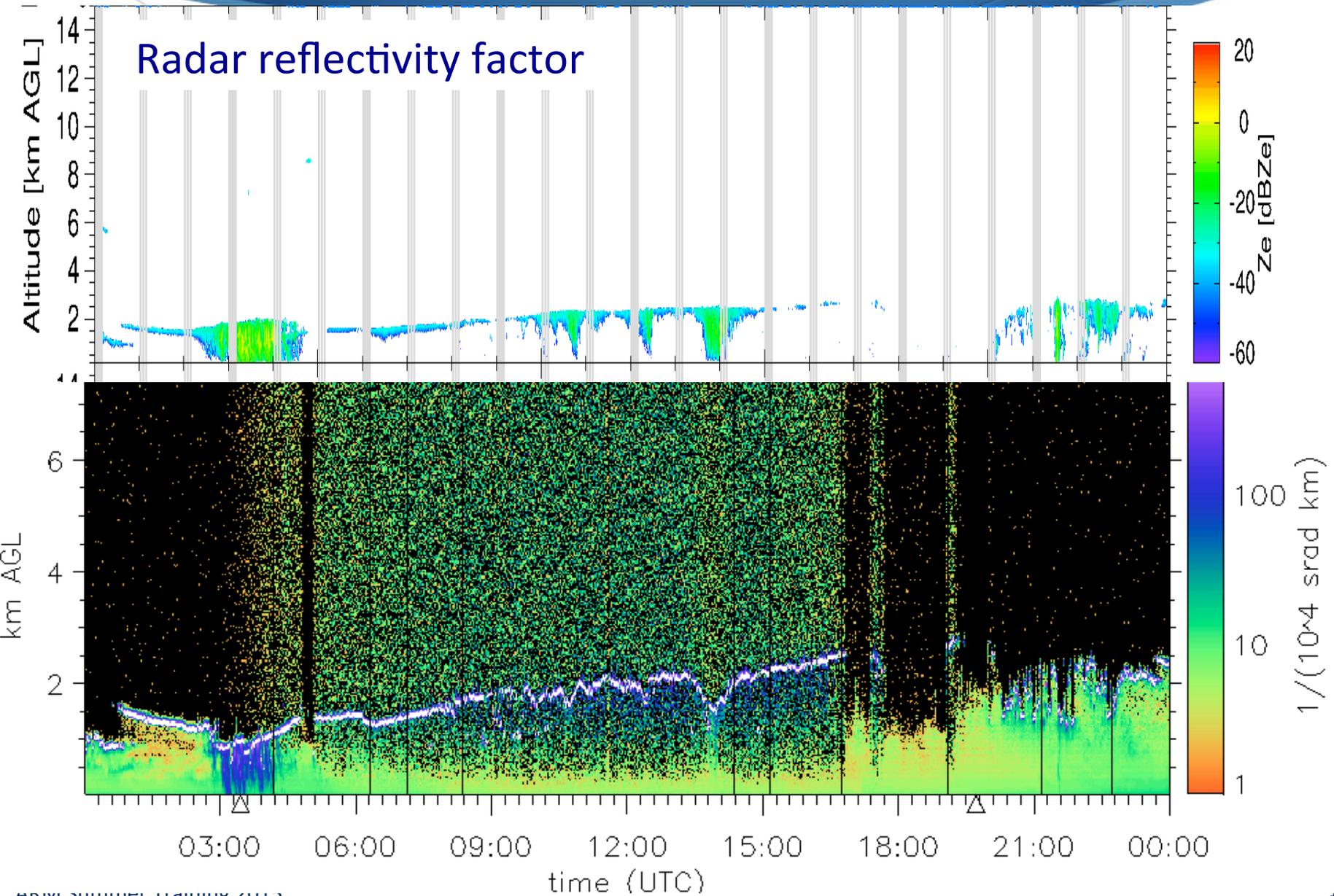
◇	1st	84.8%
△	2nd	5.7%
□	3rd	0.2%
+	vvis	0.1%
×	max	0.1%



### Backscatter coefficient



# Cloud radar vs. ceilometer



# Cloud radar

- ◆ active instrument: sends laser pulses (microwave spectrum, e.g. ~35 GHz, ~94 GHz)  
→ sensitive towards cloud droplets
- ◆ backscatter proportional to  $D^6$   
→ few drizzle/rain drops dominate radar signal
- ◆ measures backscattered signal  
→ Doppler spectrum and moments: profiles of radar reflectivity, mean Doppler velocity, Doppler spectral width, linear depolarization ratio

KAZR



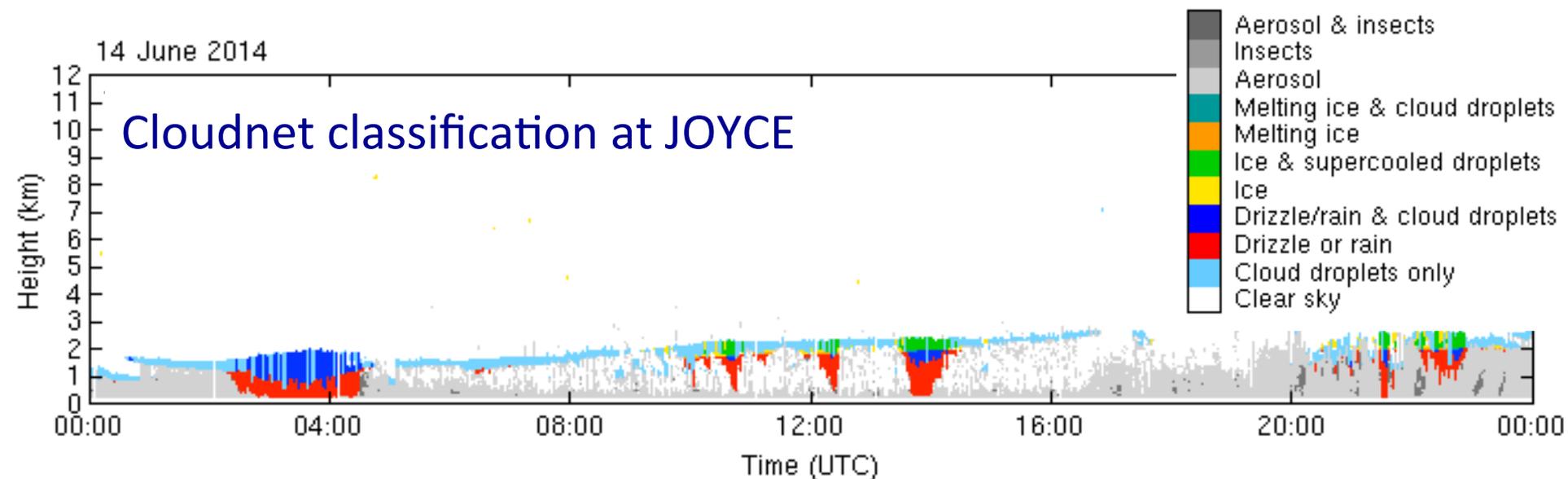
[www.arm.gov](http://www.arm.gov)

KA/W-SACR



# Cloudnet cloud classification

- ◆ synergistic product based on measurements from **radar and ceilometer/lidar**
  - + **additional information: MWR, radiosondes/NWP model data**
- lidar backscatter, radar reflectivity, change of radar reflectivity with height, LDR, (change in) Doppler velocity, temperature profile



→ more on Cloudnet and products [www.cloud-net.org](http://www.cloud-net.org) or ask Ewan!

also other, e.g. Shupe (2007), Active Remote Sensing of Clouds (ARSCL) VAP →  $Z_{\text{hydrom.}}$

# Liquid cloud property check list

- ✔ vertically resolved cloud mask ✔
- ✔ phase identification ✔
- liquid water path?
- vertical distribution of liquid water?
- effective radius of cloud droplets?

# Microwave radiometer

- ▶ passive instrument measuring at different frequencies:
  - ▶ channels along the water vapor abs line (22 GHz) + 1 window channel at 31.4 GHz for cloud information
  - ▶ channels along oxygen abs. complex at 60 GHz provide vertical temperature information

## Products & accuracies

- Integrated Water Vapor (IWV):  $0.6 \text{ kg m}^{-2}$
- Liquid Water Path (LWP):  $20 \text{ g m}^{-2}$
- Hum. Profiles:  $0.4\text{-}0.8 \text{ g m}^{-3}$   
(2 degrees of freedom for signal)
- Temp. profiles  $0.5\text{-}1.0 \text{ K}$   
(4 degrees of freedom for signal)



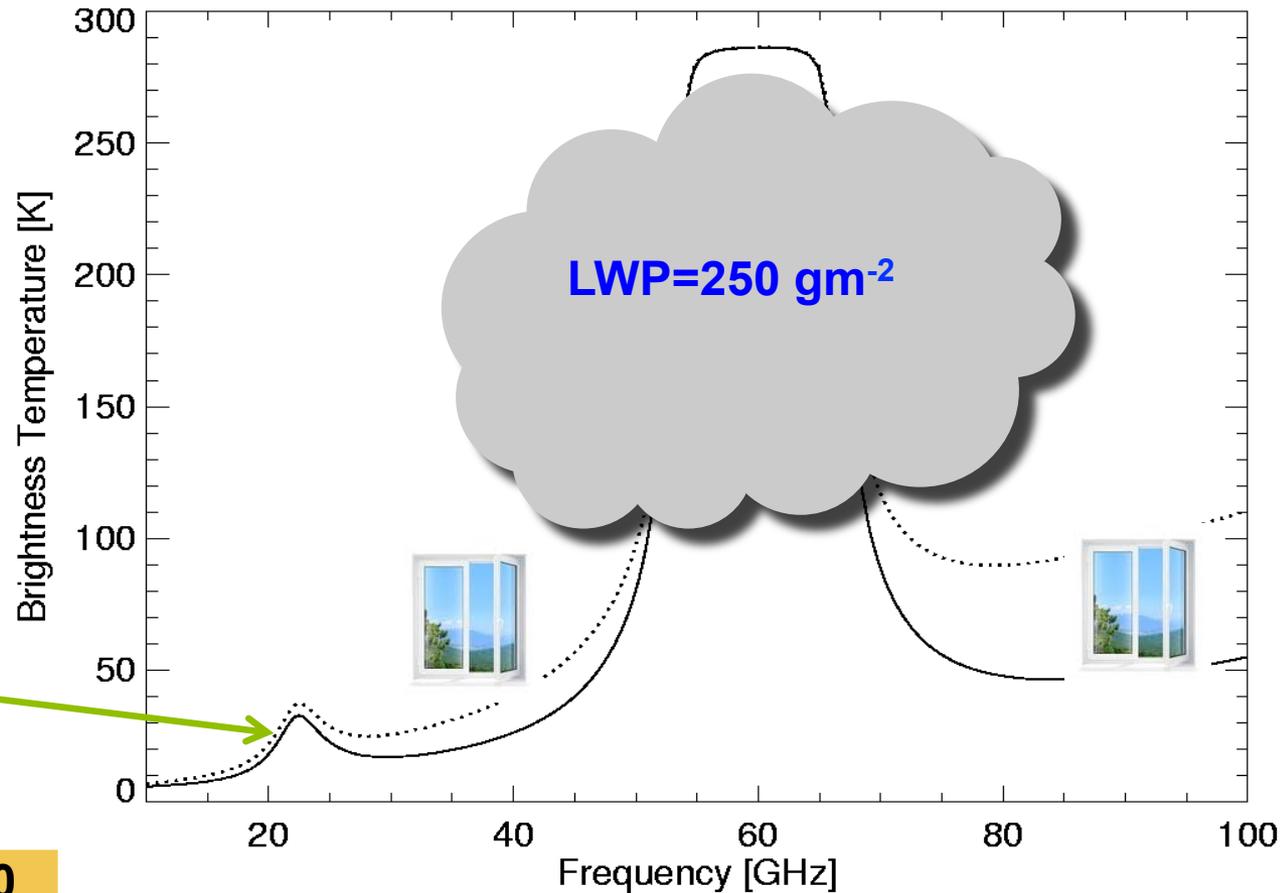
# Retrieval of LWP and IWV from MWR

- Integrated Water Vapour (IWV)
- Liquid Water Path (LWP)

→ both integrated column amounts

10 kg m<sup>-2</sup> correspond to a liquid column of 1cm

Amplitude of H<sub>2</sub>O line is proportional to IWV



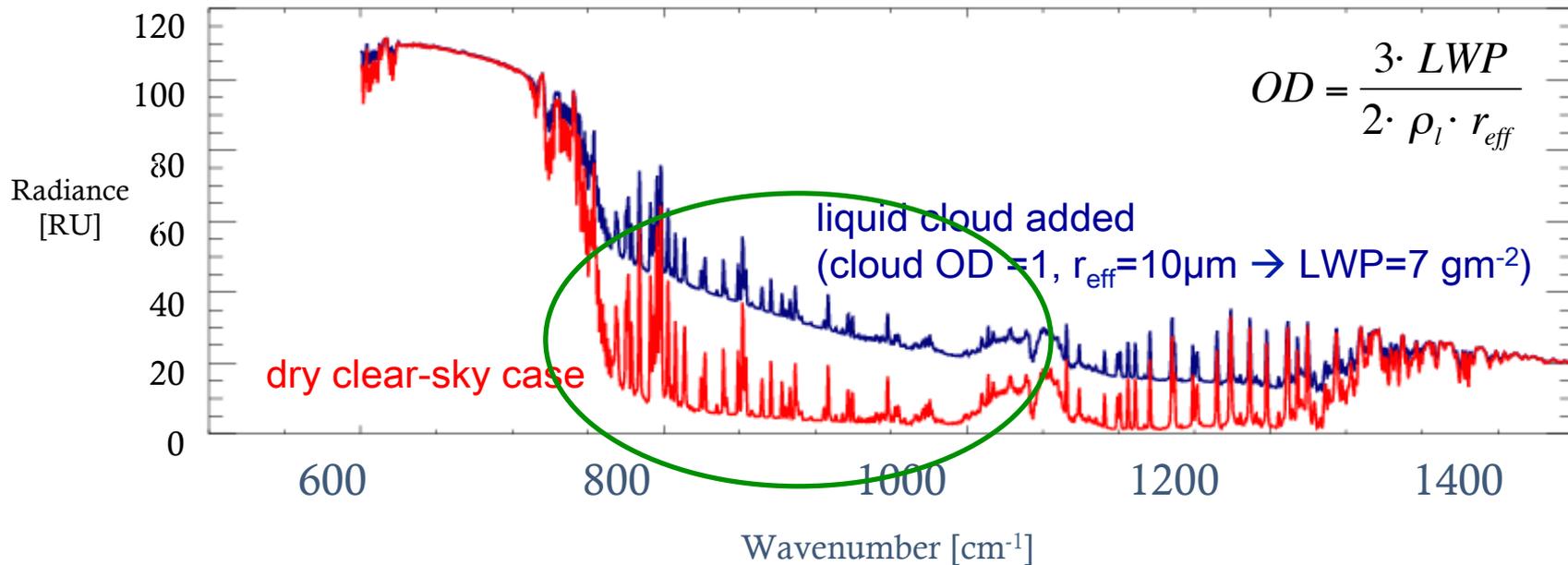
$\nu$ / GHz	31	36	90
$\Delta TB$ / K	9	12	50

$$LWP = a_1 + b_1 * TB_{31} - c_1 * TB_{24}$$

$$IWV = a_2 - b_2 * TB_{31} + c_2 * TB_{24}$$

# IR spectrometer

## Liquid clouds in the IR



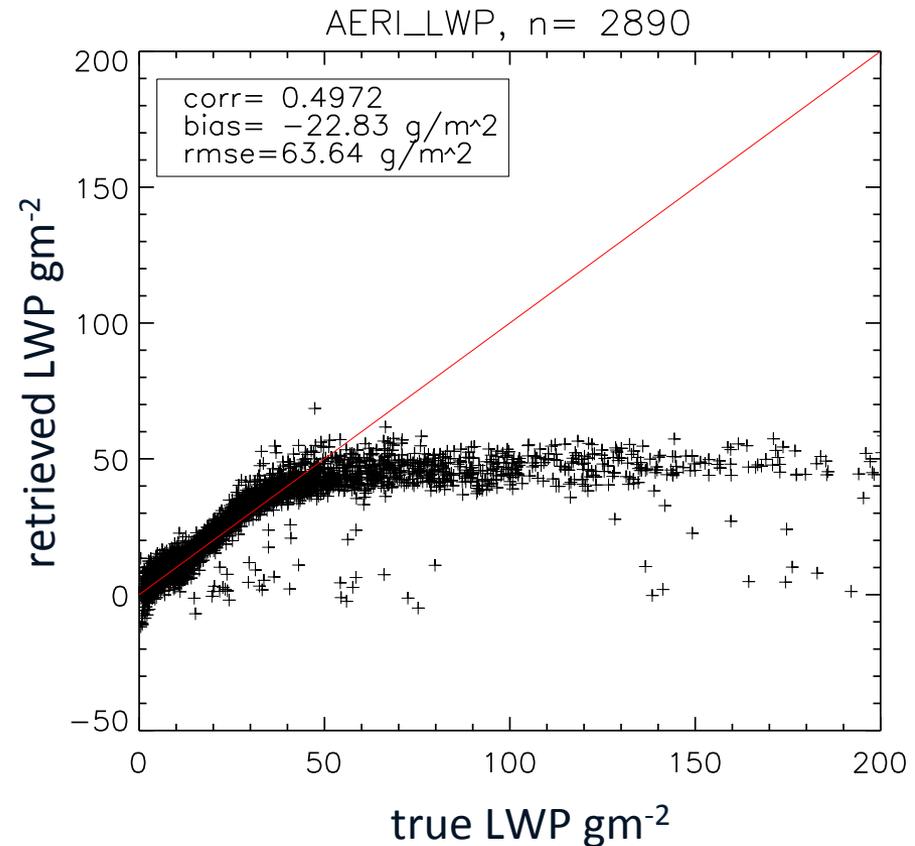
- **Offset** is sensitive to cloud optical depth (OD), **slope** to effective radius ( $r_{eff}$ )

# IR spectrometer

- multivariate regression based retrieval using AERI radiances

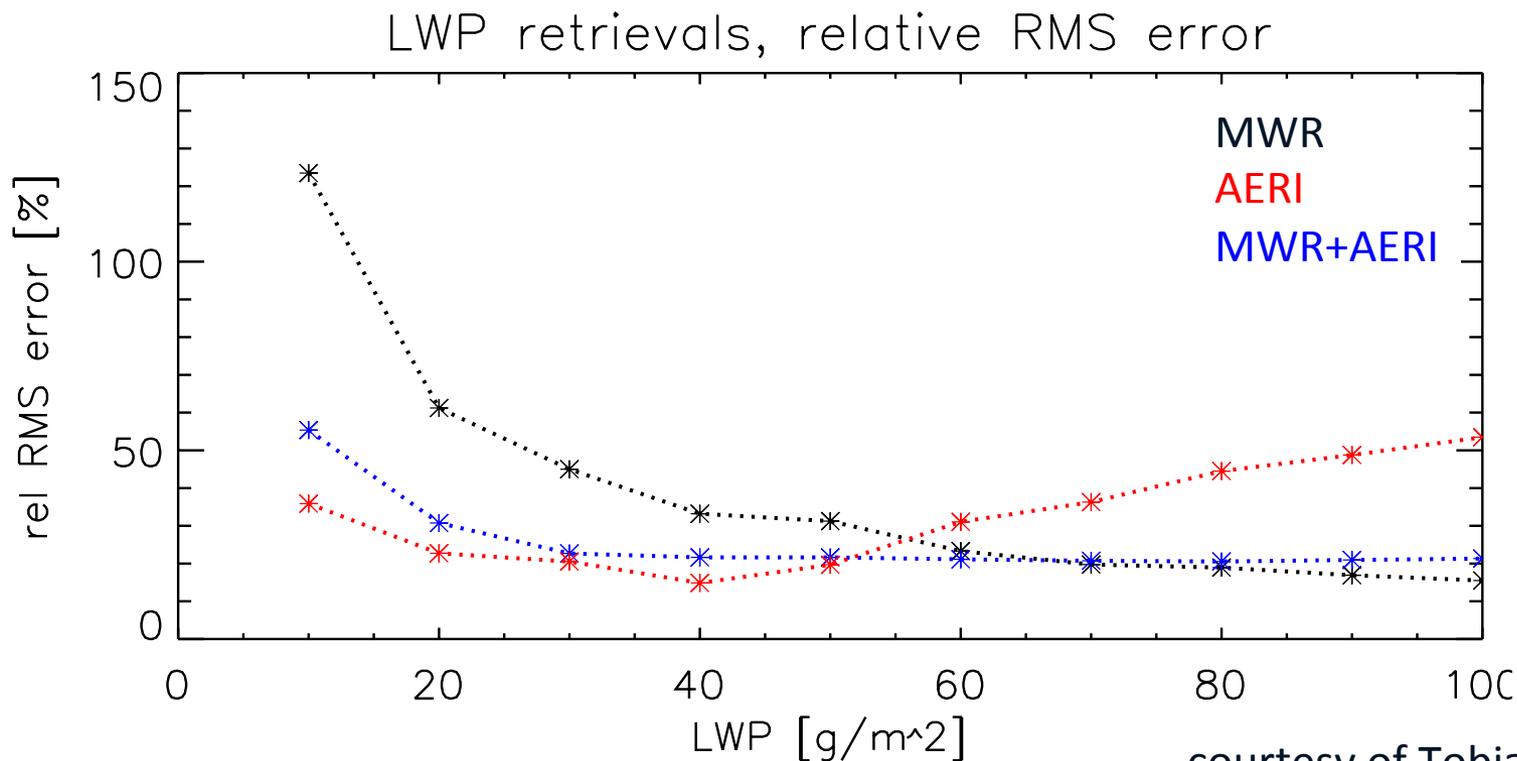
$$LWP = c_0 + \mathbf{c}_1 \mathbf{I}_{AERI} + \mathbf{c}_2 \mathbf{I}_{AERI}^2$$

- Liquid clouds saturate the signal at  $LWP \approx 60 \text{ gm}^{-2}$
- MWR needed for thicker clouds



# Combination of MWR + AERI

- results of a regression-based approach by T. Marke



courtesy of Tobias Marke

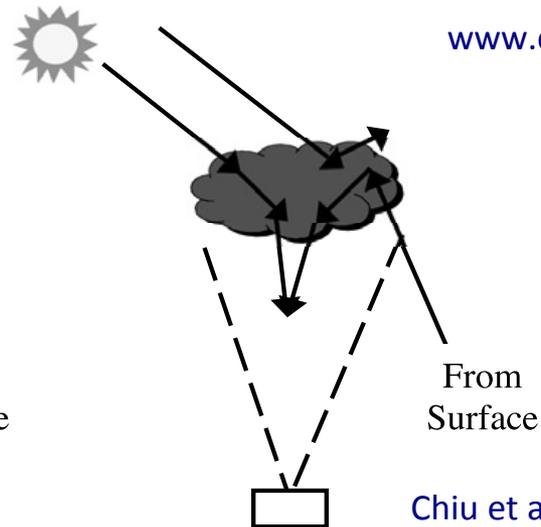
for variational-based approach see Turner (2007)

# Sunphotometer

- ☘ sun/sky radiometers with a  $1.2^\circ$  field-of-view (FOV) that measure radiance at wavelengths of 440, (500), 675, 870, 1020, (1640) nm  
→ clear-sky: microphysical and optical properties of aerosol  
*Aerosol Robotic Network (AERONET)*  
<http://aeronet.gsfc.nasa.gov/>



- ☘ cloudy sky: „cloud mode“: zenith radiance measurements 440, 870, 1640 →  $COT$ ,  $r_{eff}$

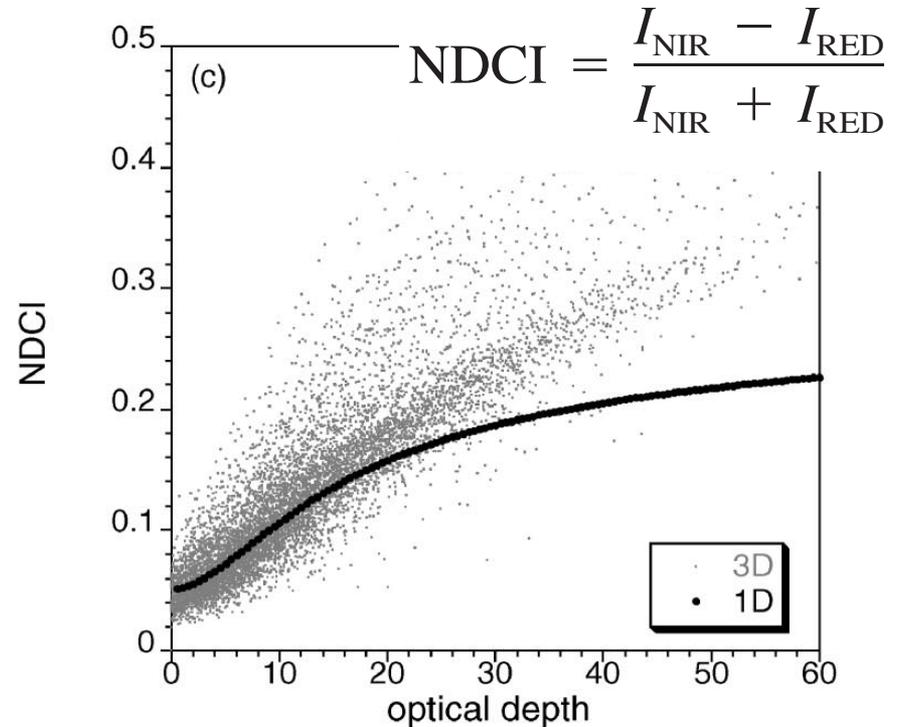
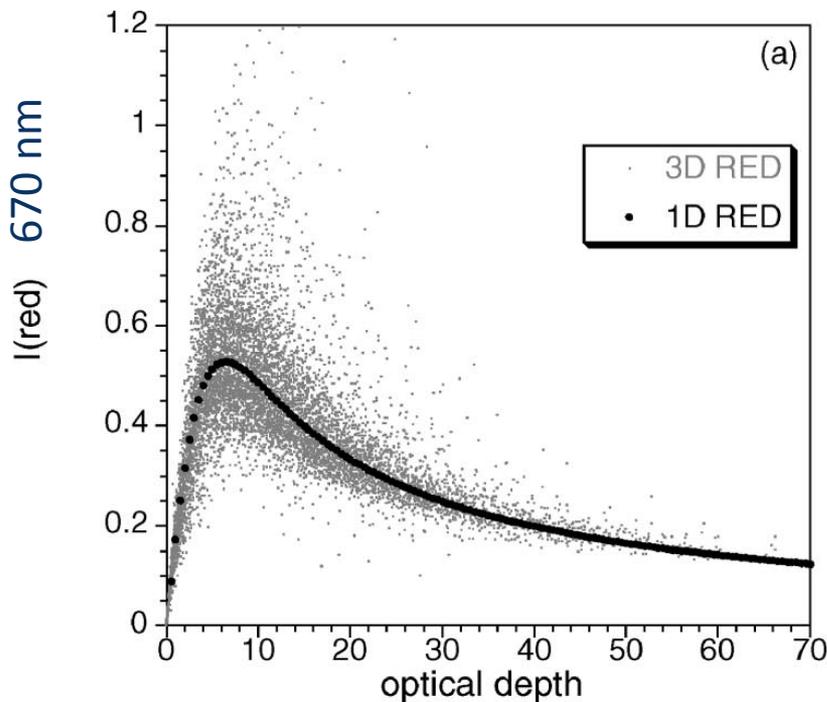


[www.cimel.fr](http://www.cimel.fr)

Chiu et al. (2006)

# Sunphotometer

- principle: spectral contrast in surface reflectance  
→ most vegetated surfaces are dark at red wavelengths and bright at NIR wavelengths → different surface-cloud interactions



- need to take into account for 3D-effects via a „radiatively effective“ cloud fraction  $A_c$

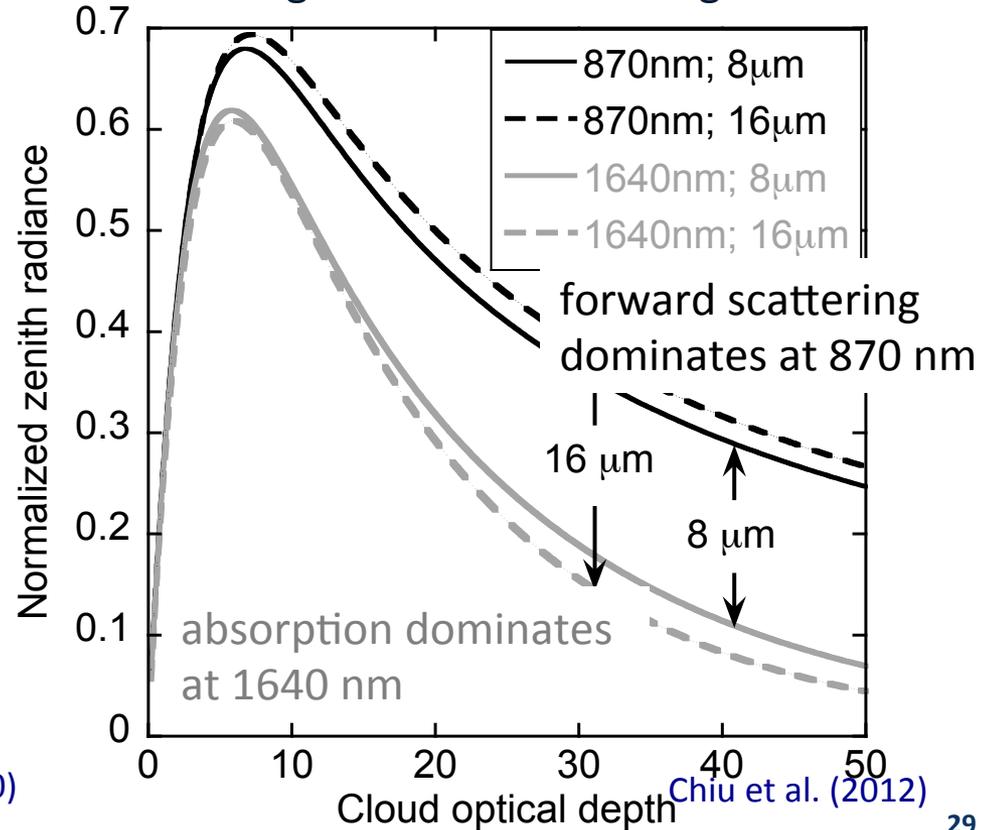
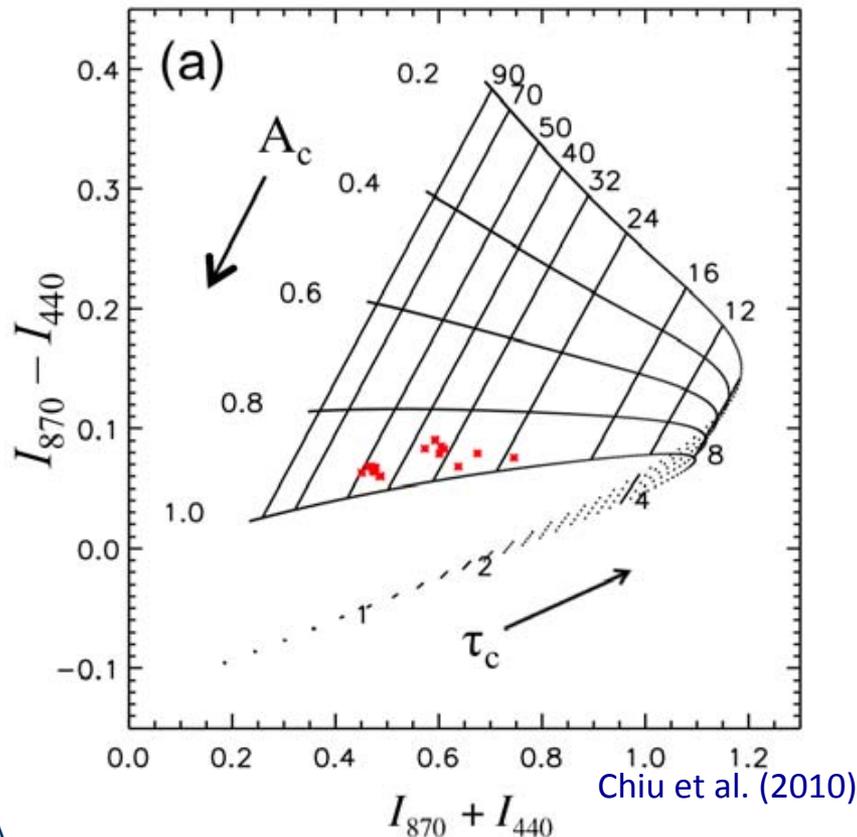
Marshak et al. (2004)

# Sunphotometer

- look-up table approach: for different COT and „effective“ cloud fraction  $A_c$ , precalculate zenith radiance measurements at 440 and 870
- inclusion of 1640 nm  $\rightarrow r_{\text{eff}}$

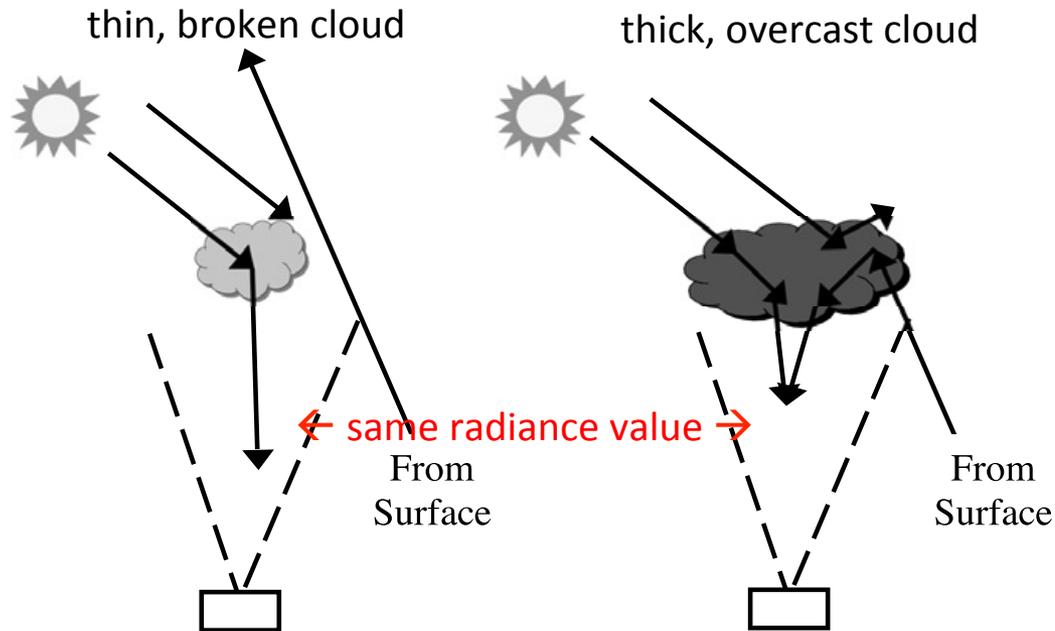
the larger  $r_{\text{eff}}$

- the stronger absorption  $\rightarrow I$  reduced
- the stronger forward scattering  $\rightarrow I$  increased



# Sunphotometer

## ☘ clear-sky contamination



Chiu et al. (2006)

- method works best for overcast cloud scenes
- COT error 15-25 %,  $r_{\text{eff}}$  error 11-22 %
- ice clouds above ok, as long as liquid fraction  $> 0.2$  (25% error in COT)

# Liquid cloud property check list

- ✔ vertically resolved cloud mask ✔
- ✔ phase identification ✔
- ✔ liquid water path ✔ cloud optical thickness, layer-averaged  $r_{\text{eff}}$  ✔
- ✔ vertical distribution of liquid water?
- ✔ effective radius of cloud droplets?

→ active remote sensing methods

# Cloud radar: Z-LWC relations

kth moment of DSD:

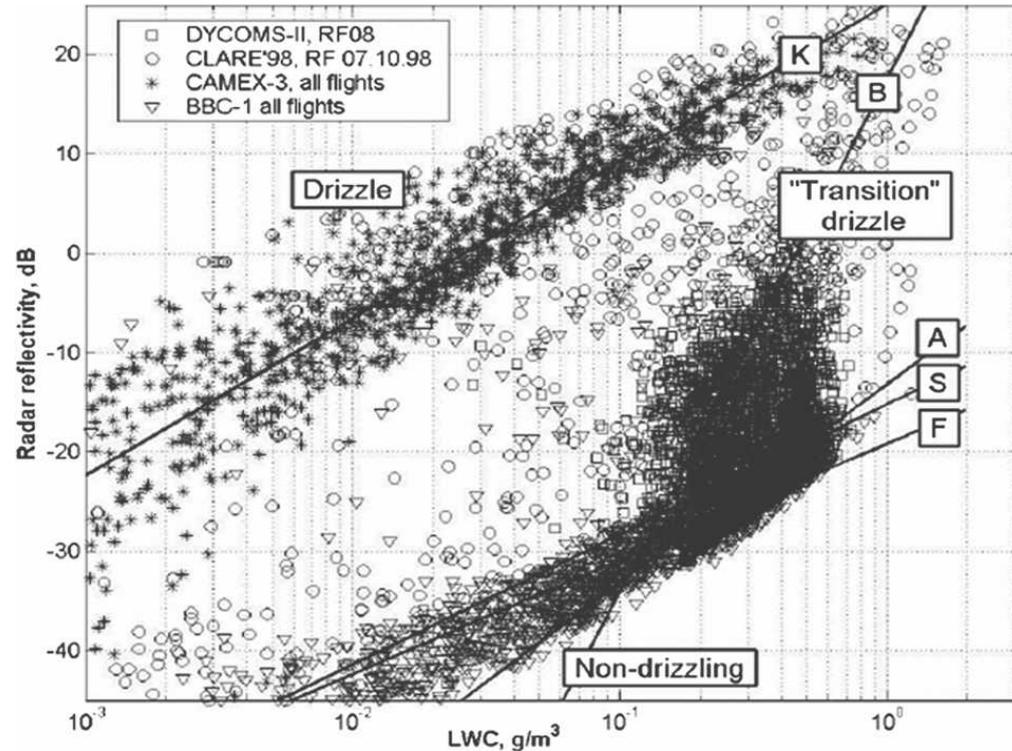
$$M_k = \int_0^{\infty} D^k N(D) dD$$

$$LWC = \frac{\pi \rho_l}{6} M_3$$

$$Z = \int_0^{\infty} D^6 N(D) dD = M_6$$

→  $Z = a LWC^b$

🟢 **problem:**  $N(D)$  results from a number of complex processes and is therefore highly variable even within a cloud!



Khain et al. (2008)

- many Z-LWC relations : Atlas (1954); Sauvageot and Omar (1987); Fox and Illingworth (1997); Khain et al. (2008),...
- large uncertainty, particularly if drizzle drops are present

# LWC from dual-frequency radar

- in Rayleigh regime ( $r \ll \lambda$ ), attenuation of MW radiation is proportional to LWC and increases with frequency  $\rightarrow$  use rate of change with height of  $DWR = Z_{35} - Z_{94}$  to deduce LWC  
*adjustment to account for change in dielectric constants with temperature*

$$\overline{LWC} = \frac{1}{\kappa_{94} - \kappa_{35}} \left( \frac{DWR_2 - DWR_1 - \beta}{2(h_2 - h_1)} - \alpha_{94} + \alpha_{35} \right)$$

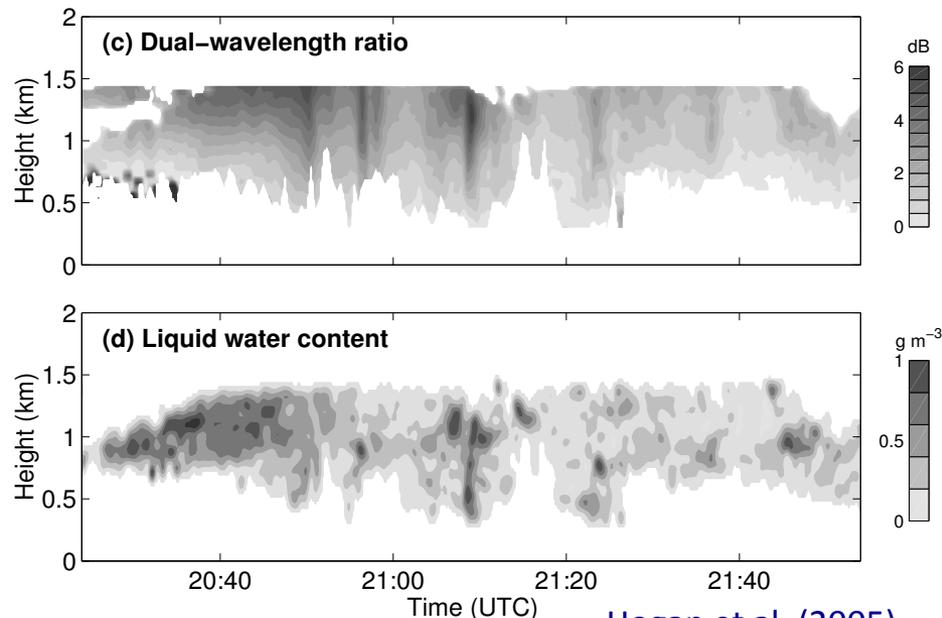
*liquid absorption coeff.*

*atm. absorption coeff.*

- each 1 dB difference corresponds to a LWP of  $\sim 120 \text{ gm}^{-2}$

**advantage:** no assumptions on DSD, insensitive to absolute calibration, presence of drizzle drops does not effect the retrieval

LWC accuracy depends on SNR, limited by number of independent measurements



Hogan et al. (2005)

# Radar + MWR

- ◆ **LWC profile:** MWR LWP is used to scale LWC profile from Z measurements

$$LWC(h) = \frac{Z^{1/2}(h)}{\sum_{i=cb}^{ct} Z_i^{1/2} \Delta h} LWP_{MWR}$$

Frisch et al. (1998)

$\Delta h$ : radar range gate thickness

- ◆  **$r_{eff}$  profile:**  
e.g. Frisch et al. (2002):

assuming lognormal DSD  $\rightarrow$

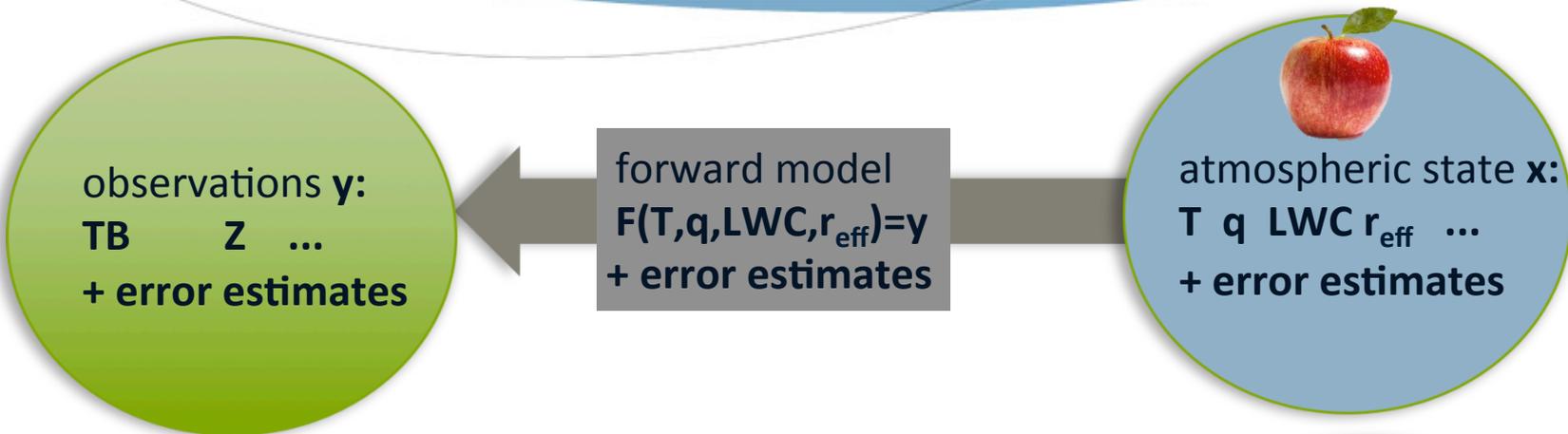
$N(h) = \text{const}$

assuming  $\sigma_x$

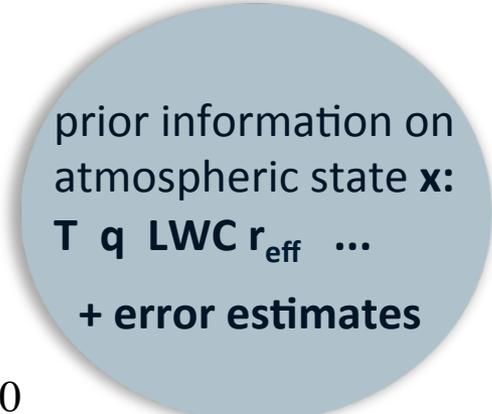
$$n(r) = \frac{N_{tot}}{\sqrt{2\pi}\sigma_x} \exp\left(-\frac{(\ln r - \ln r_0)^2}{2\sigma_x}\right)$$

$$r_{eff}(h) = \frac{Z^{1/6}(h)}{2LWP_{MWR}} \left(\frac{\pi\rho}{6}\right)^{1/3} \left(\sum_{i=cb}^{ct} Z^{1/2}(h_i)\Delta h\right)^{1/3} \exp(-2\sigma_x^2)$$

# Variational-based retrieval methods



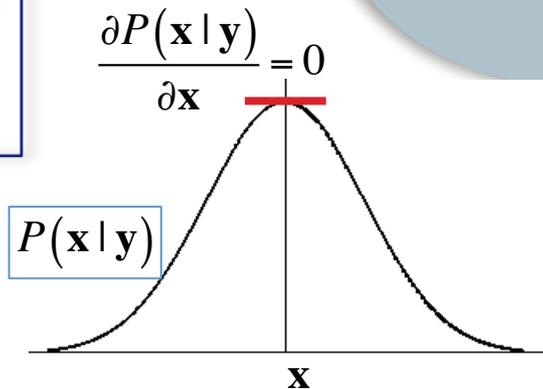
- prior information, observations and forward model have uncertainties  
→ represent probability distributions



Bayes Theorem

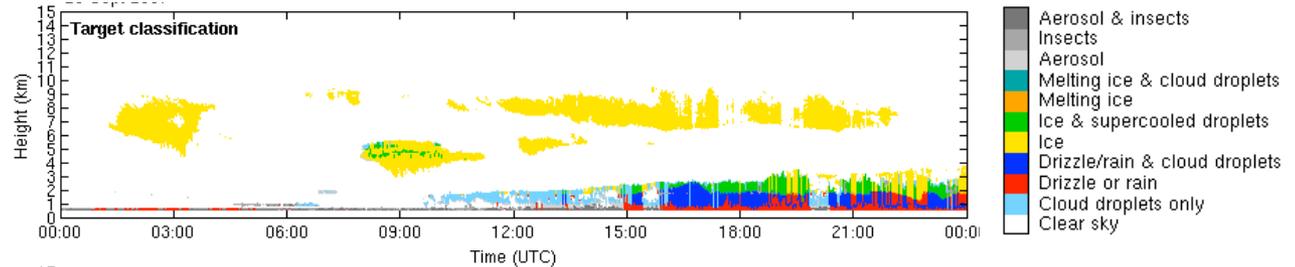
$$P(\mathbf{x} | \mathbf{y}) = \frac{P(\mathbf{y} | \mathbf{x})P(\mathbf{x})}{P(\mathbf{y})}$$

- does provide a class of solutions and assigns a probability density to each



# Integrated profiling technique IPT

Cloudnet Target  
Categorization  
product (Illingworth  
et al., 2007)



**prior information**  
 $\mathbf{x}_a = [T, q, LWC, \text{reff}]$   
with error covariance  
matrix  $\mathbf{S}_a$

**forward model**  $F(\mathbf{x})=y$   
with error covariance  
matrix  $\mathbf{S}_e$

**measurements**  $y=[TB, Z]$   
with error covariance  
matrix  $\mathbf{S}_e$

1 D variational retrieval algorithm (optimal estimation equation,  
e.g. Rodgers, 2000)

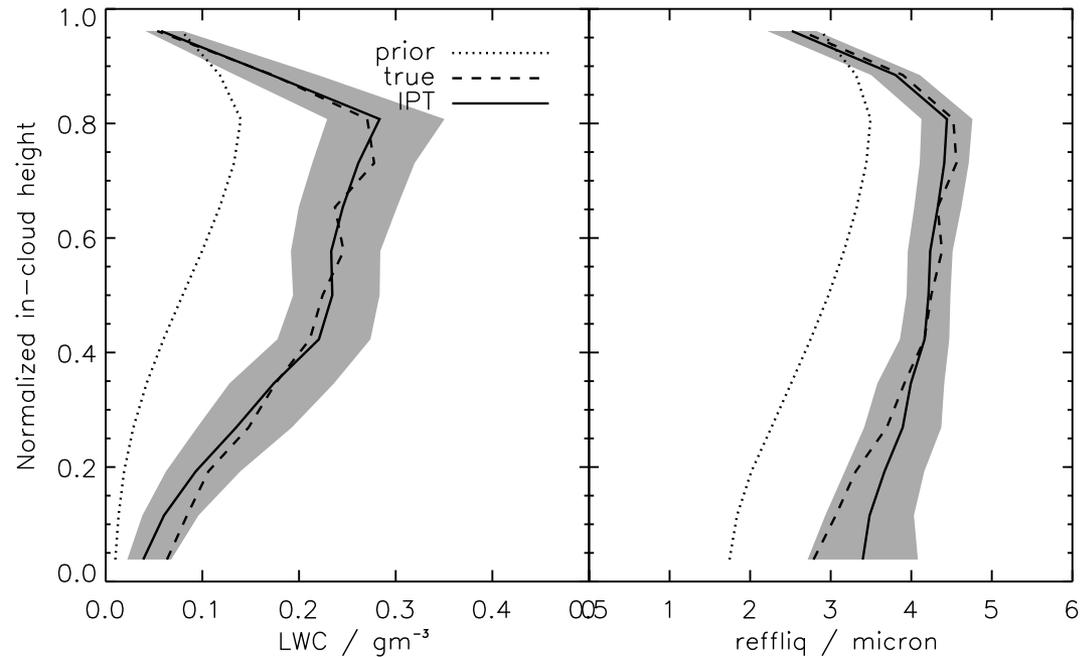
$$\mathbf{x}_{i+1} = \mathbf{x}_i + \left( \frac{\partial \mathbf{F}(\mathbf{x}_i)^T}{\partial \mathbf{x}_i} \mathbf{S}_e^{-1} \frac{\partial \mathbf{F}(\mathbf{x}_i)}{\partial \mathbf{x}_i} + \mathbf{S}_a^{-1} \right)^{-1} \times \left[ \frac{\partial \mathbf{F}(\mathbf{x}_i)^T}{\partial \mathbf{x}_i} \mathbf{S}_e^{-1} (\mathbf{y} - \mathbf{F}(\mathbf{x}_i)) + \mathbf{S}_a^{-1} (\mathbf{x}_a - \mathbf{x}_i) \right]$$

minimizing cost function

$\mathbf{x}_{op} = [T, q, LWC, \text{reff}]$  with  $\mathbf{S}_{op}$

# IPT: synthetic data study

- know the „truth“:  
liquid cloud with  
 $\Delta z=370$  m (13 levels),  
LWP=65  $\text{gm}^{-2}$
- simulate **TB, Z** → IPT
- IPT can very well  
reproduce the truth  
 $\sigma_{\text{reff}} \approx 10\%$ ,  $\sigma_{\text{LWC}} \approx 30\%$



- „Degrees of freedom for signal“: How much information in the retrieved profile comes from the Z and TB measurements?

**LWC: 38% ,  $r_{\text{eff}}$ : 34%**

- weight of measurements/prior information in the solution determined by their error covariance matrices

y error	1x	2x	3x
<b>LWC</b>			
rel. DOF / %	<b>38</b>	<b>31</b>	<b>27</b>
rel. err. / %	<b>31</b>	<b>38</b>	<b>45</b>
<b><math>r_{\text{eff}}</math></b>			
rel. DOF / %	<b>34</b>	<b>29</b>	<b>26</b>
rel. err. / %	<b>10</b>	<b>13</b>	<b>16</b>

# Variational-based methods

- ◆ provide **physically consistent** solution  
(and if not, something is wrong with your measurements, assumptions,...)
- ◆ directly provide **retrieval uncertainties** based on assumed measurement, forward model and prior uncertainties

Be aware that

- ◆ **prior information is crucial**: constrains solution space
- ◆ **error covariance matrices**, i.e. uncertainties of prior information, forward model, and measurements, need to be **carefully defined**
- ◆ **prior information** and its error often **difficult to determine**  
(e.g. cloud profile, correlation of cloud layers?)
- ◆ variables and errors are **Gaussian distributed**  
→ bias / offset errors can not be handled and need to be removed before

# Liquid clouds: Still a need for research?

**YES!**

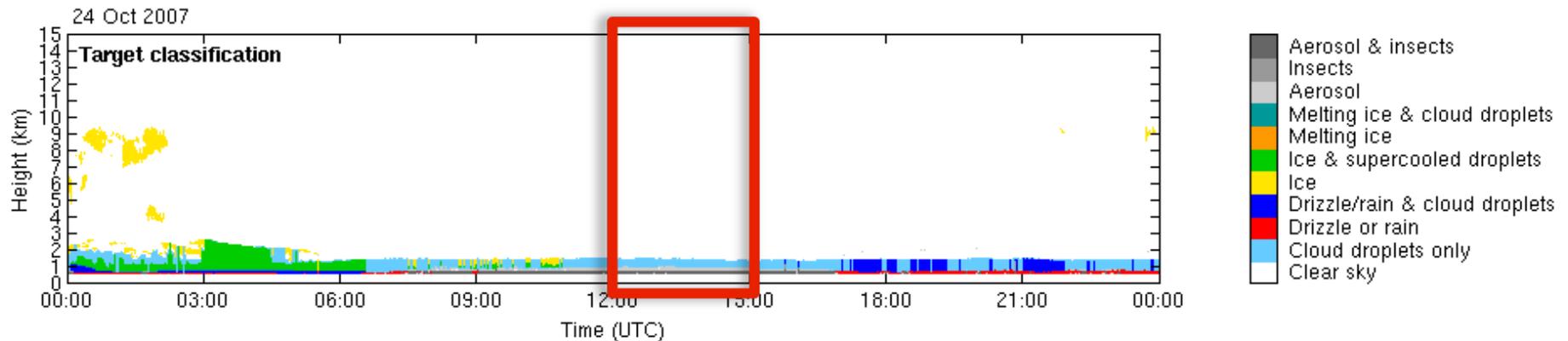
- ◆ many cloud retrievals only applicable in certain atmospheric/cloudy conditions
- ◆ liquid cloud as a result of complex processes  
→ difficult to develop robust „universal“ methods
- ◆ „perfect“ liquid water clouds rarely exist: e.g. drizzle, droplets+ice,...  
→ Doppler spectra analysis  
→ improved phase detection
- ◆ maximize information by integrating as many as observations as possible, e.g. radar, lidar, MWR, IR and solar radiances,...  
→ high demands on measurements: data quality (e.g. bias-free measurements), spatial+temporal matching,...

# Liquid clouds: Still a need for research?

**Thanks for listening!**

# Now, it's up to you...

Let's have a look at this stratus cloud observed by AMF instruments on 24 October 2007 in the Murg Valley, Black Forest (Germany)



I want you to answer a simple (?) question:

What is the **mean liquid water path** of the cloud observed between 12-15 UTC?

→ use data in file: `lwp_exercise.nc`  
(data are already sampled for this time window)

# What is the mean liquid water path of the cloud observed between 12-15 UTC?

data in file: lwp\_exercise.nc

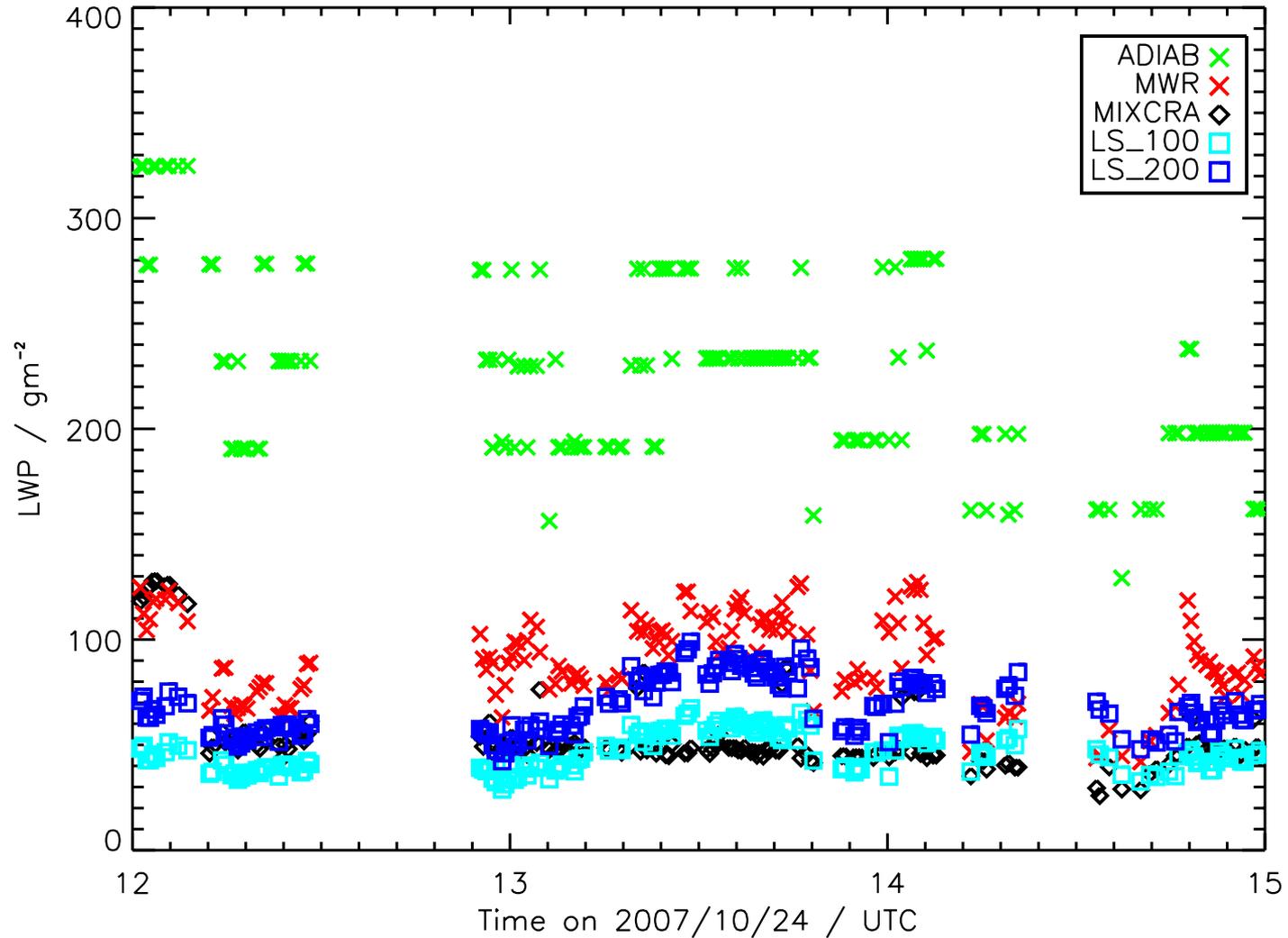
- ◆ **time** [time]
- ◆ **lwp\_ad** [time] ( $\text{gm}^{-2}$ ): adiabatic LWP
- ◆ **lwp\_mixcra** [time] ( $\text{gm}^{-2}$ ): LWP from MIXCRA (variational-based retrieval using MWR and AERI)
- ◆ **lwp\_mwr** [time] ( $\text{gm}^{-2}$ ): LWP from MWR (a two-channel, regression-based retrieval)
- ◆ **Z** [time, cloudlayer] ( $\text{mm}^6/\text{m}^3$ ): radar reflectivity factor (fill values -9999. !)
- ◆ **rg**: radar range gate length (m)  
→ to calculate lwc profile for each time step from Z use the relationship from *Liao and Sassen (1994)*

$$LWC = \left( \frac{ZN_d}{3.6} \right)^{1/1.8}$$

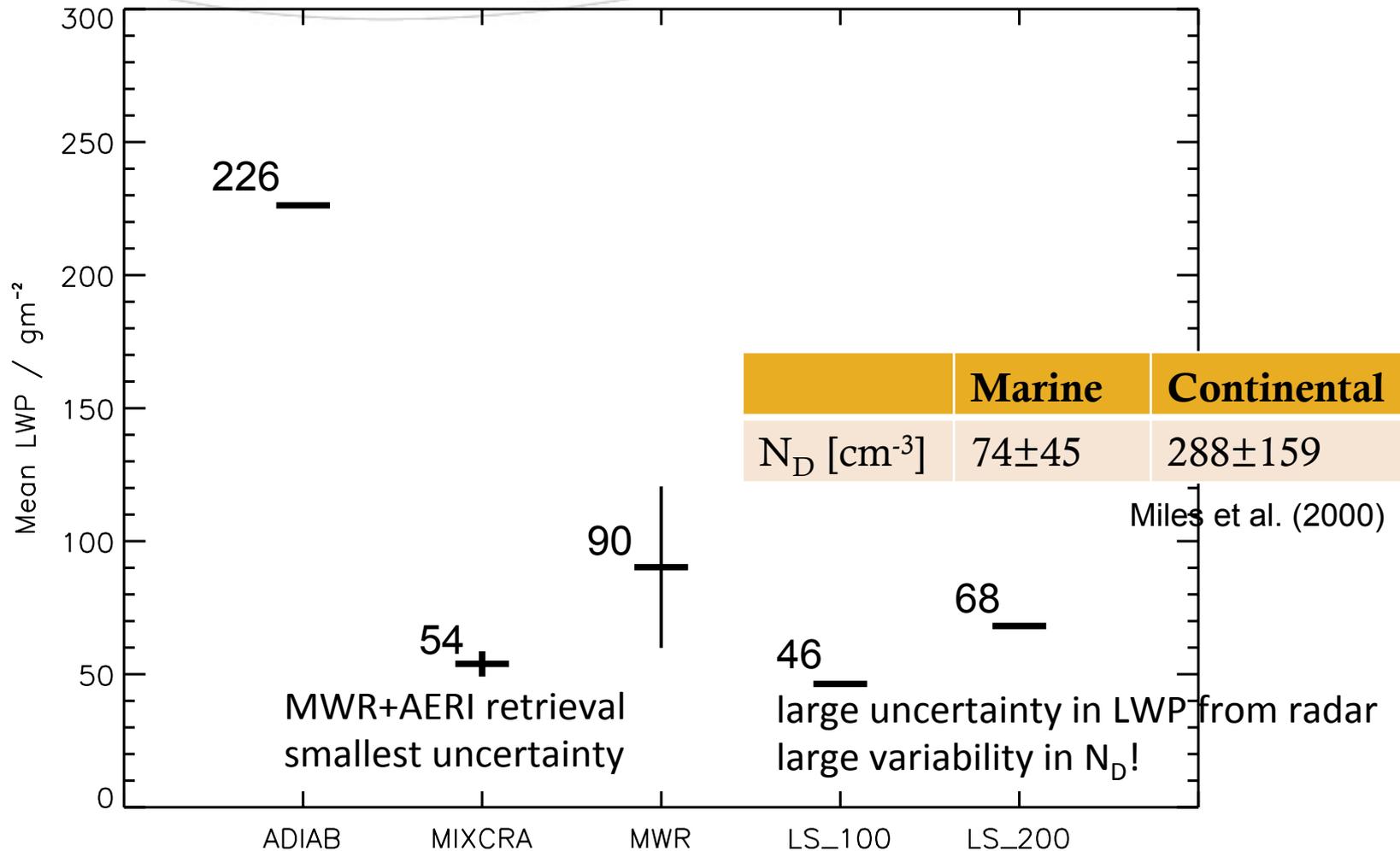
with LWC in  $\text{gm}^{-3}$ , Z in  $\text{mm}^6/\text{m}^3$ ,  $N_d$  droplet number concentration in  $\text{cm}^{-3}$   
set  $N_d=100 \text{ cm}^{-3}$  ?  $200 \text{ cm}^{-3}$  ?

→ **lwp\_radar** for each time step is then **TOTAL(lwc\_radar\*rg)**

# LWP on 24 October 2007



# Mean LWP on 24 October 2007, 12-15 UTC



all methods reveal subadiabtic clouds

ARM → could perform radiative closure study (SW/LW) to further check with method is best