

# Potential of Higher Moments of the Radar Doppler Spectrum for Studying Ice Clouds

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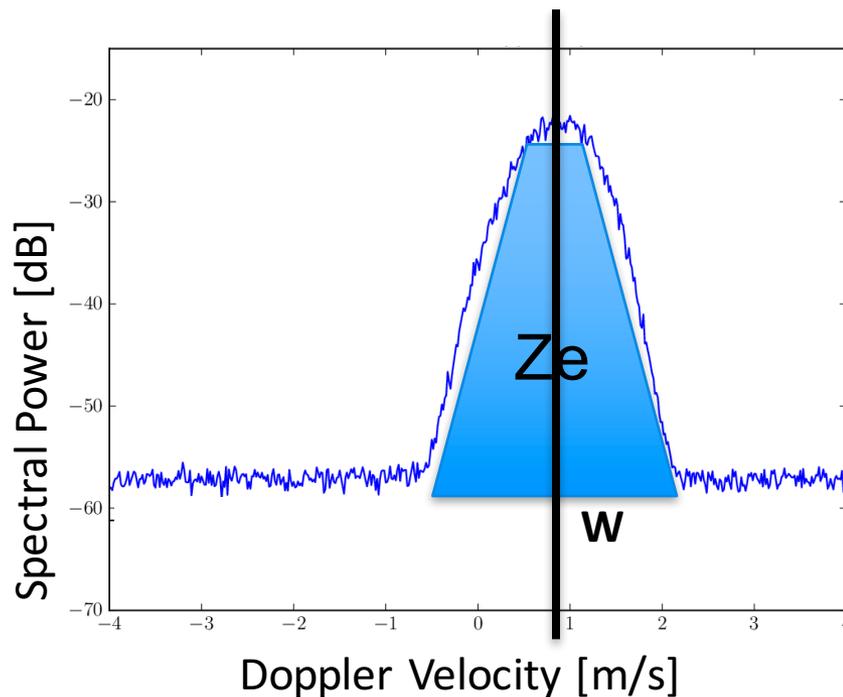
# Vertically pointing Doppler radar

- Give profile information
- High spatiotemporal resolution
- **But: Indirect measurement & high uncertainties!**

**Motivation: We have to increase the number of observables!**



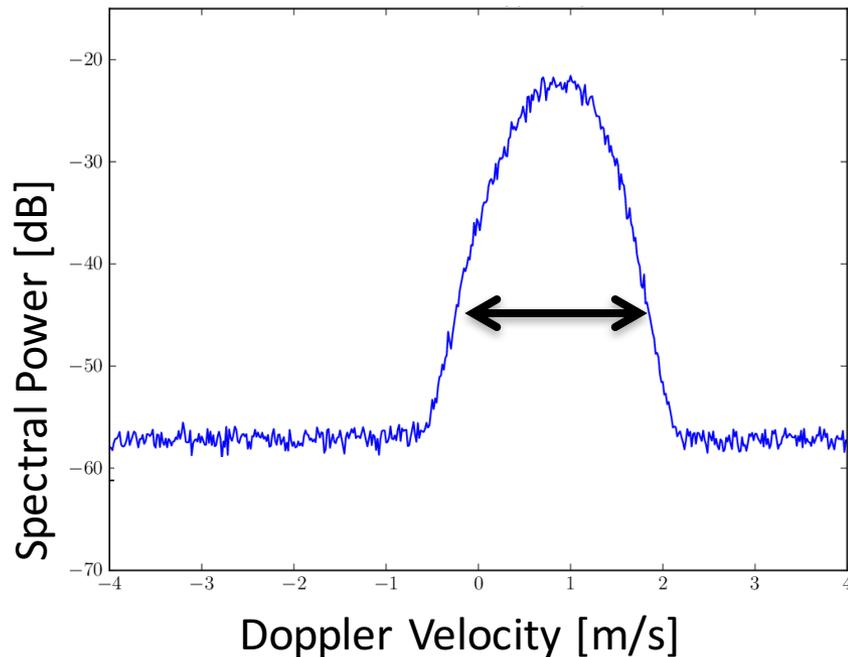
# What does a Doppler radar measure?



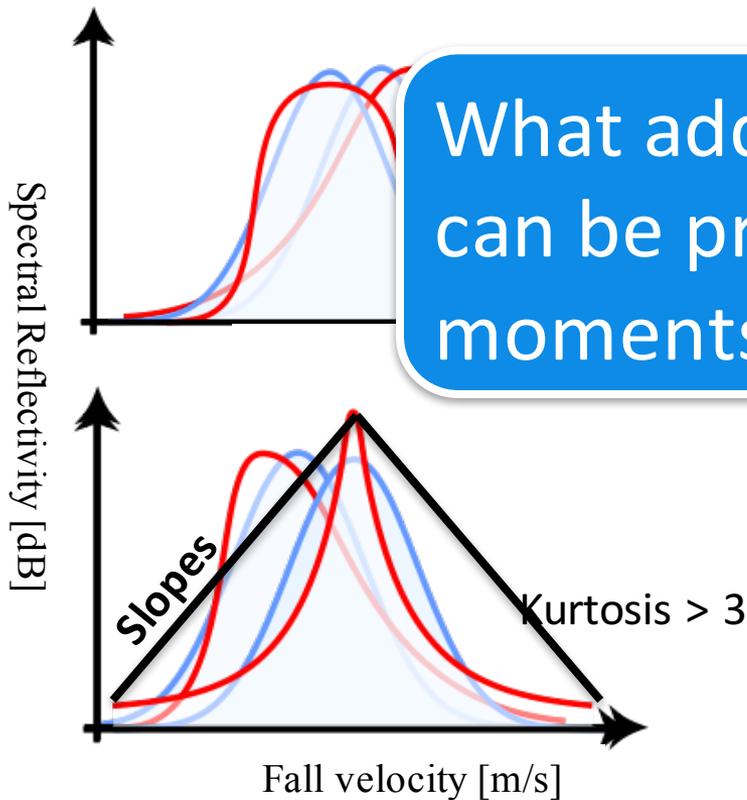
- Reflectivity  $Z_e$  is sensitive to false radar calibration
- Mean Doppler Velocity  $W$  is sensitive to vertical air motion

# Higher order moments

- In addition to Ze and W, use also:
  - Spectrum Width  $\sigma$



# Higher order moments



ents

- In addition to Ze and W, use also:

– Spectrum Width  $\sigma$

at Slope  $S_{l,r}$

H

- Strongly influenced by kinematic broadening  $\sigma_k$  (turbulence, horizontal wind)

# Degrees of freedom for signal

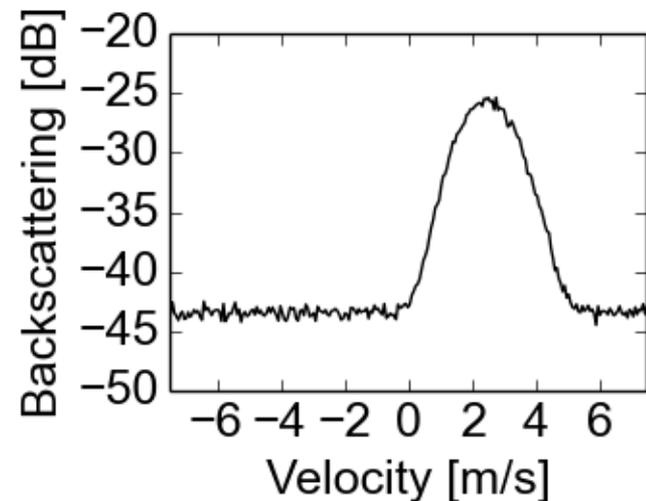
- Estimate the number of independent information pieces
- Use Bayesian retrieval (Optimal Estimation)
  1. Develop forward operator -> PAMTRA
  2. Get a priori data set -> ISDAC + Parameterizations
  3. Apply Retrieval

# Ingredients to simulate a Doppler spectrum with PAMTRA

## Required variables:

- State vector
- Normalized Gamma distribution  $N(D)$ : 3 parameters
  - Mass-size relation  $m(D)$ : 2 p.
  - Cross section area  $A(D)$ : 2 p.
  - Vertical wind  $w$ : 1 p.
  - Turbulent broadening  $\sigma_k$ : 1 p.

➤ 9(!) state variables



# A priori data set

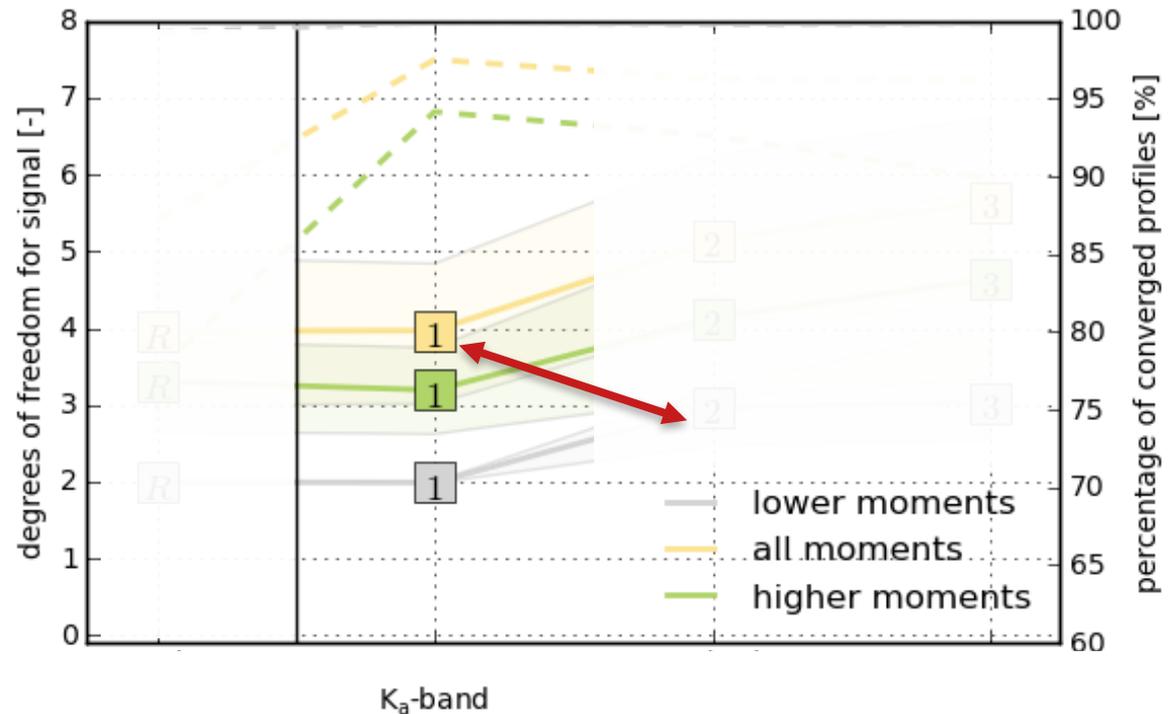
- Indirect and Semi-Direct Aerosol Campaign (ISDAC)
- *low to medium* turbulence Stratocumulus ice clouds
- April 2008 in Alaska
- Convair 580 with in situ instruments



McFarquhar

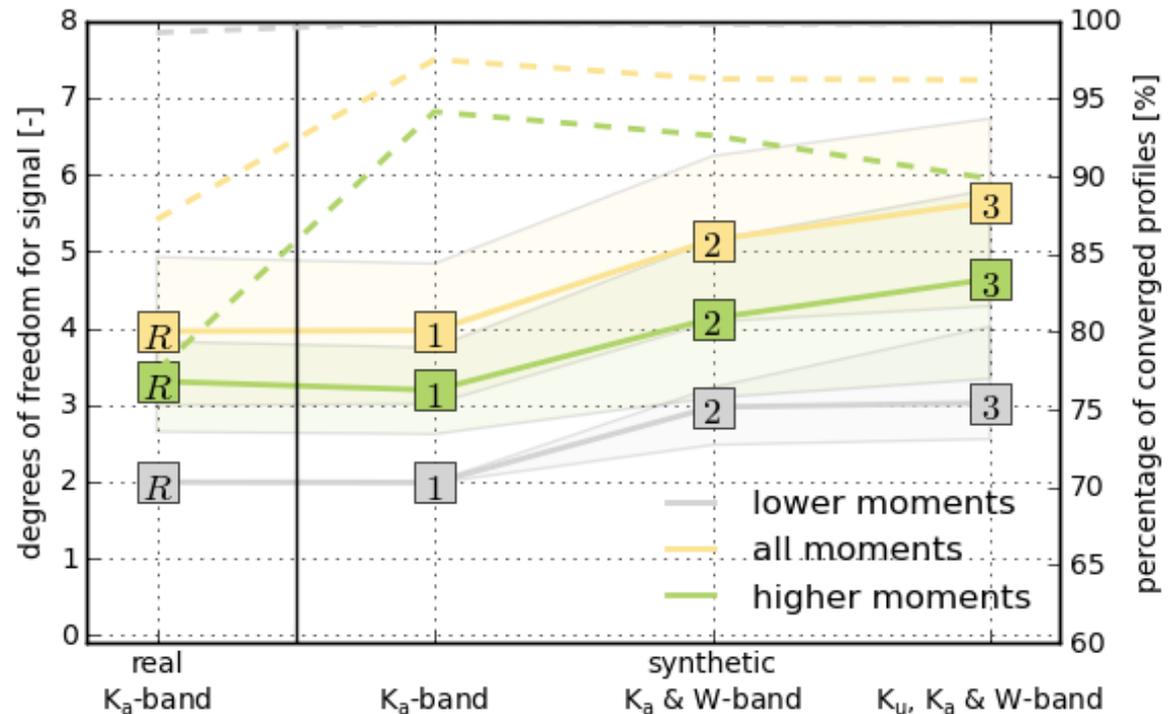
# Degrees of freedom for 1300 profiles

- For  $K_a$ -band observations, higher moments can double the information content.
- Using all moments provides more information for one frequency than lower moments for two frequencies.



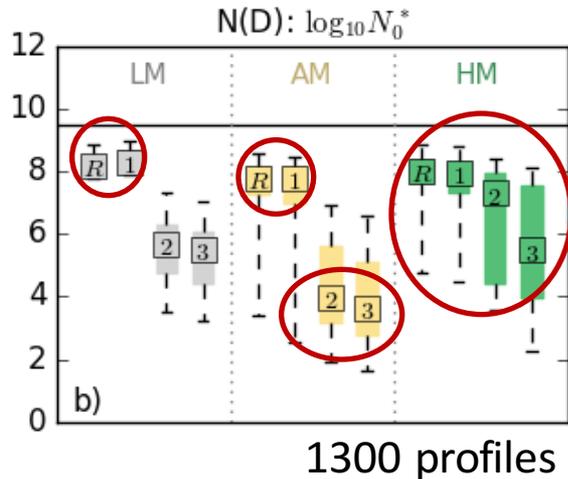
# Degrees of freedom for 1300 profiles

- For  $K_a$ -band observations, higher moments can double the information content.
- Using all moments provides more information for one frequency than lower moments for two frequencies.
- Results for real MMCR observations from Barrow agree well.



# Relative posterior uncertainty [%]

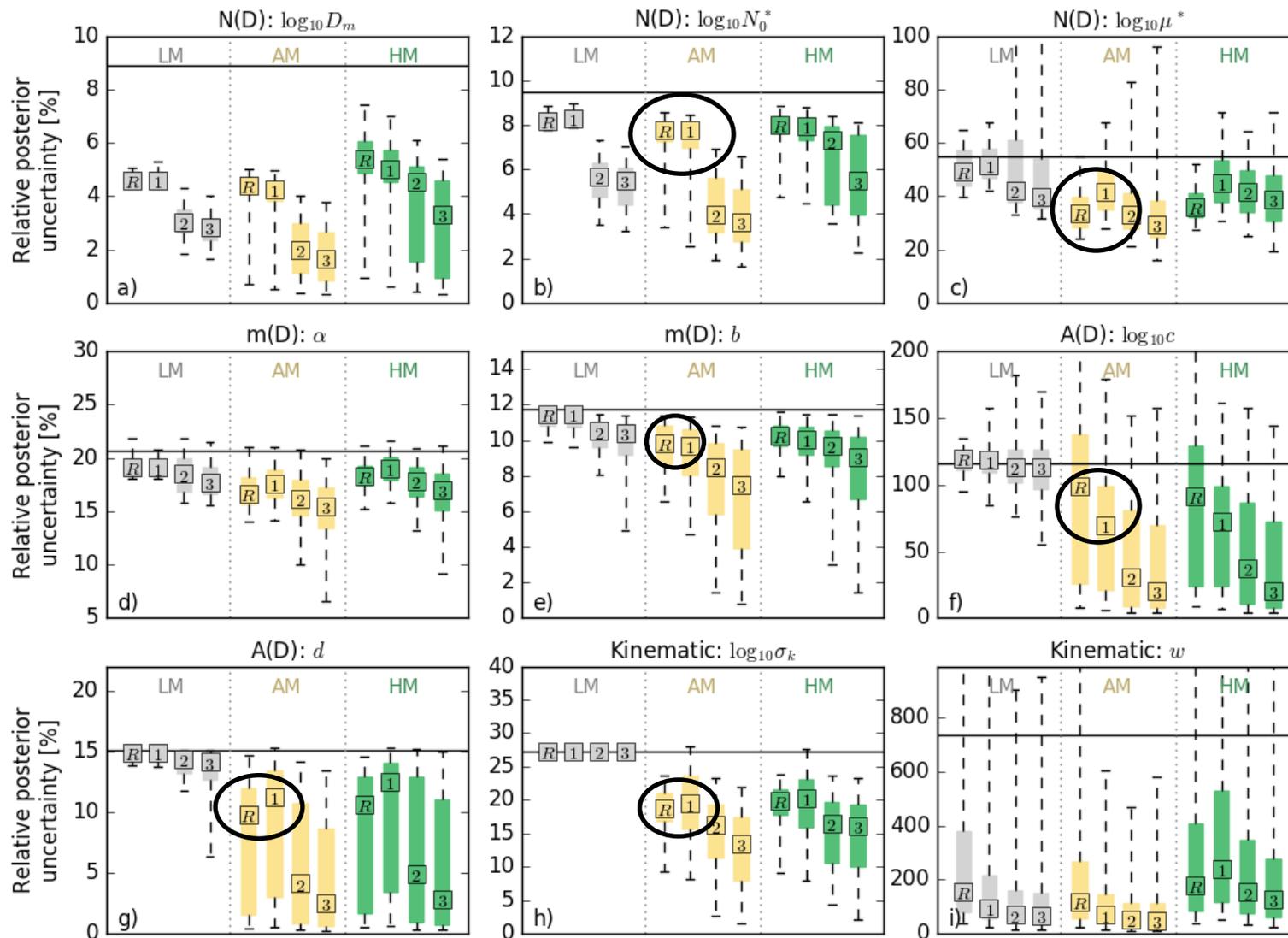
## Particle size distribution $\log_{10}(N_0)$



Prior knowledge

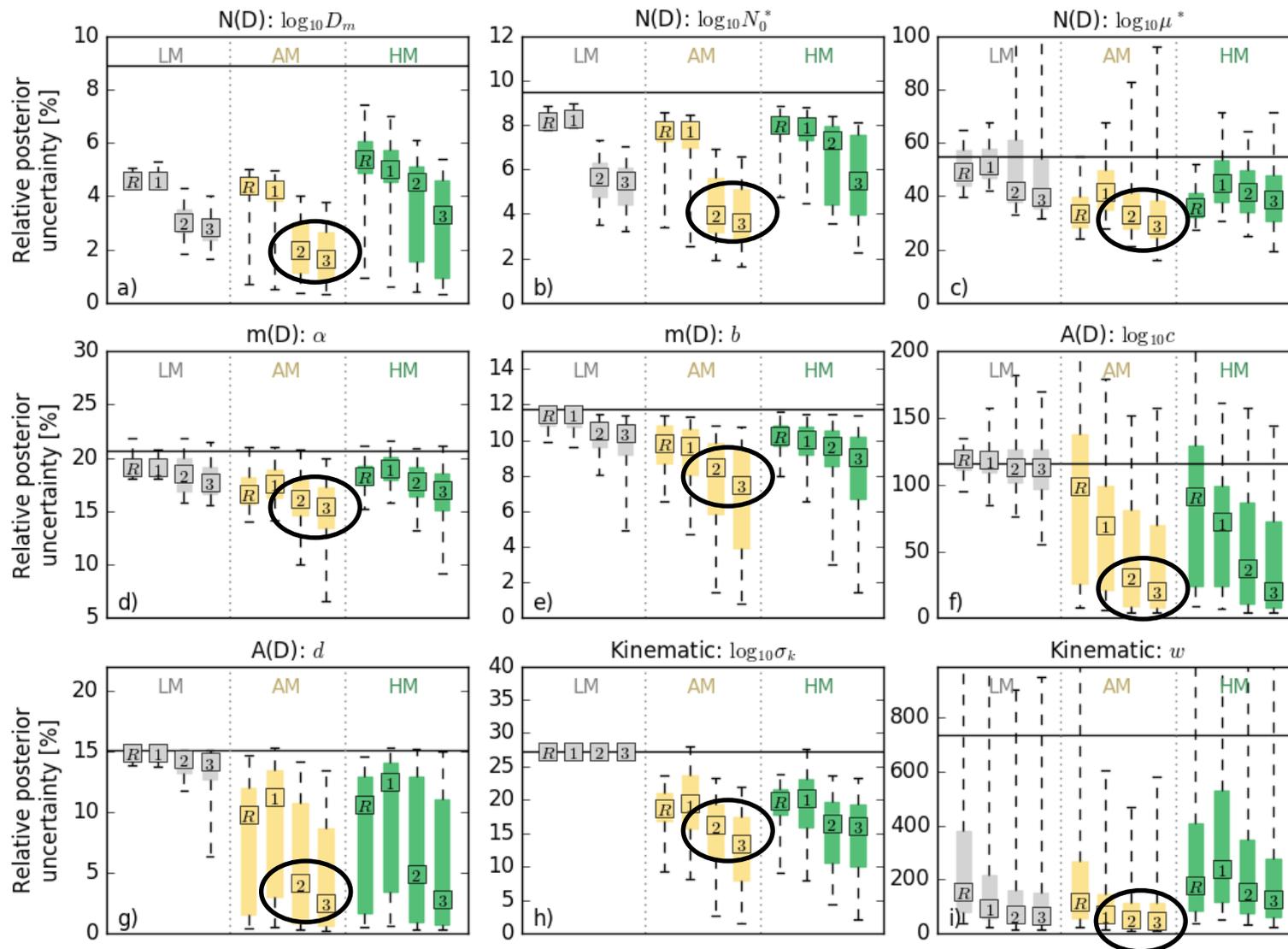
- Single-frequency lower moments retrieval cannot retrieve  $N_0$  well
- Single-frequency *all* moments retrieval reduces  $N_0$  uncertainty
- Dual/triple-frequency all moments retrieval reduces uncertainty by 50%
- Using only higher moments gives less enhancement

R: real measurements  
 1: single frequency  
 2: dual frequency  
 3: triple frequency



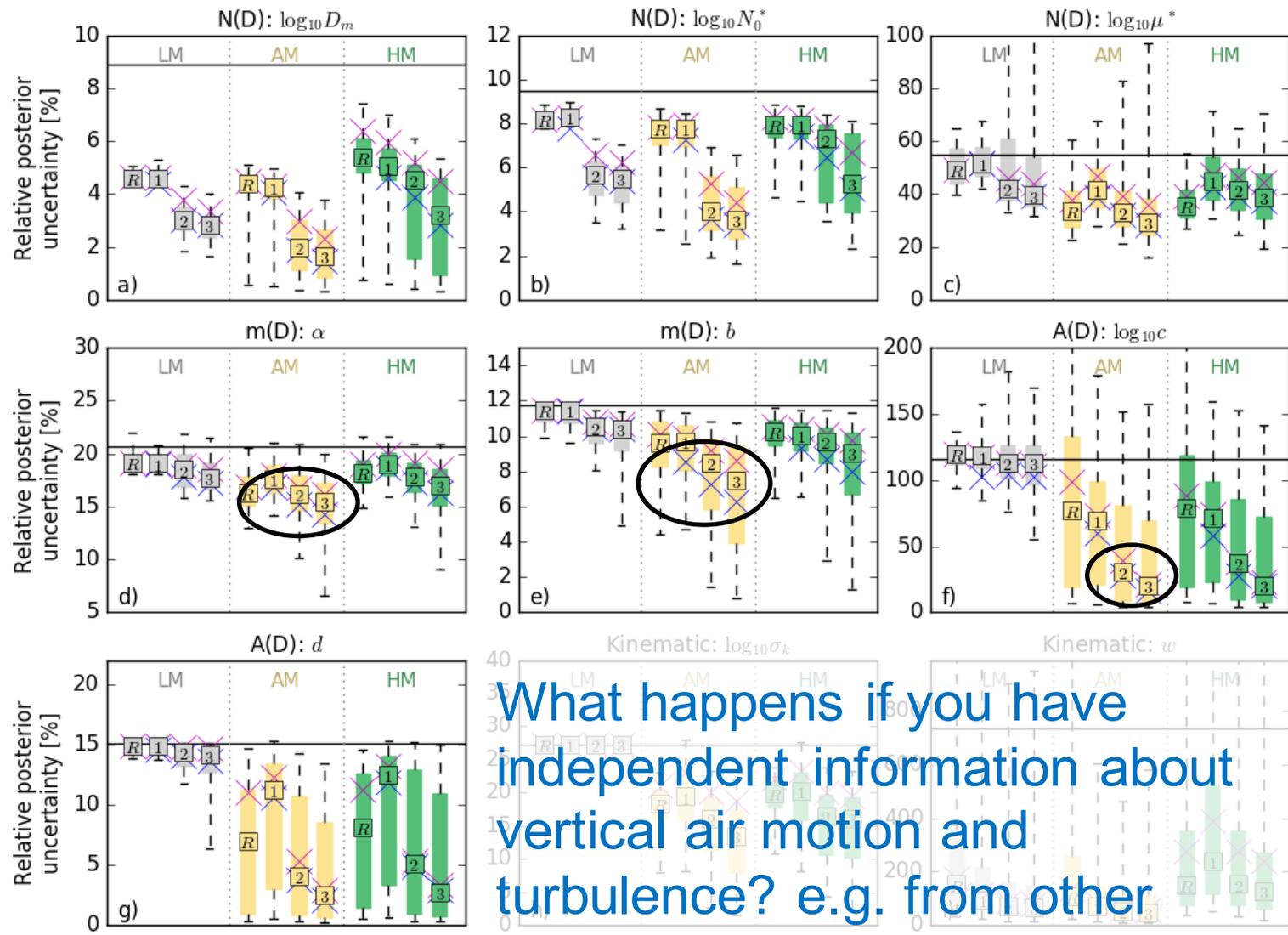
Enhancement for most quantities when using all moments and one frequency

R: real measurements  
 1: single frequency  
 2: dual frequency  
 3: triple frequency



Best results for all quantities when using all moments and two or three frequencies

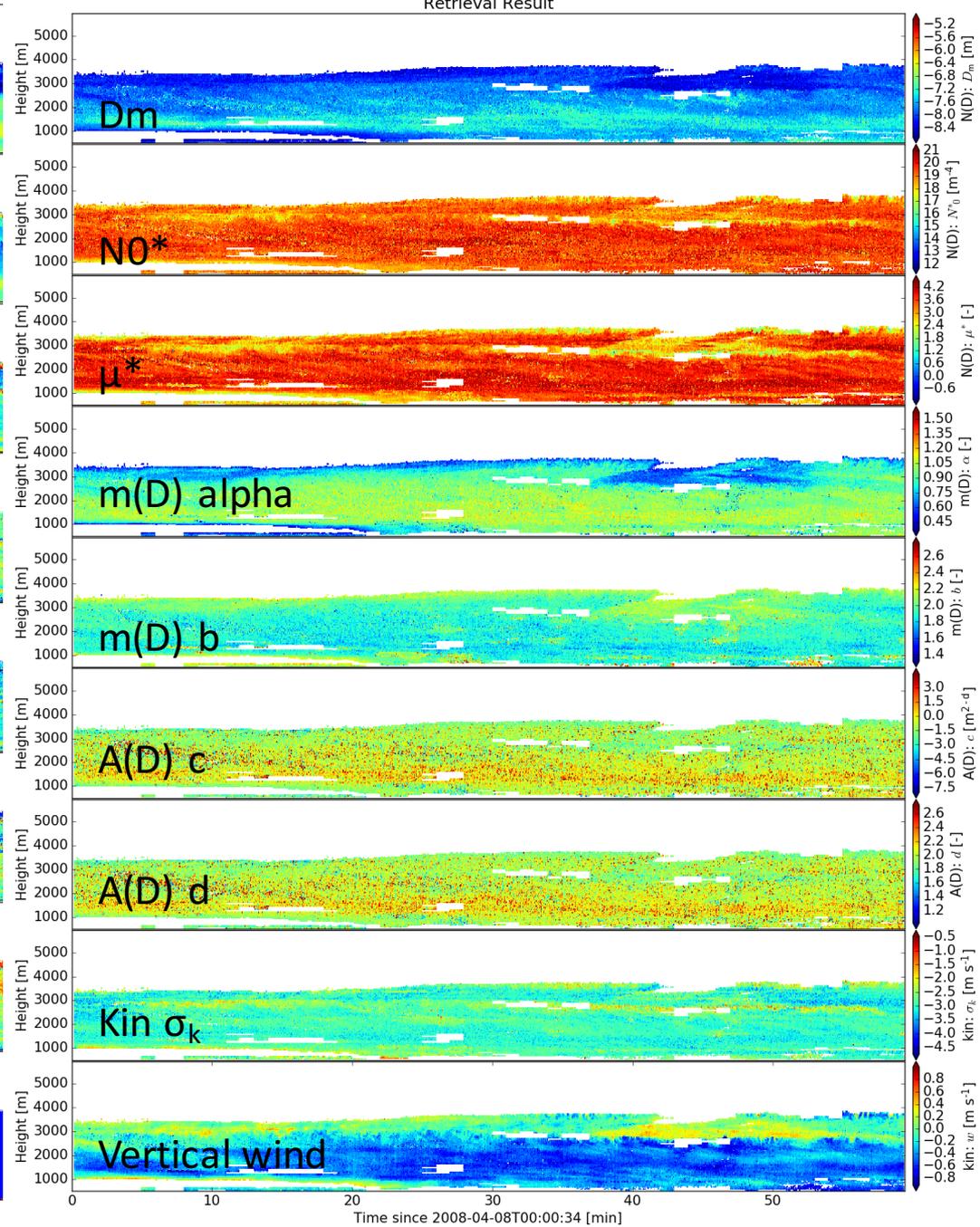
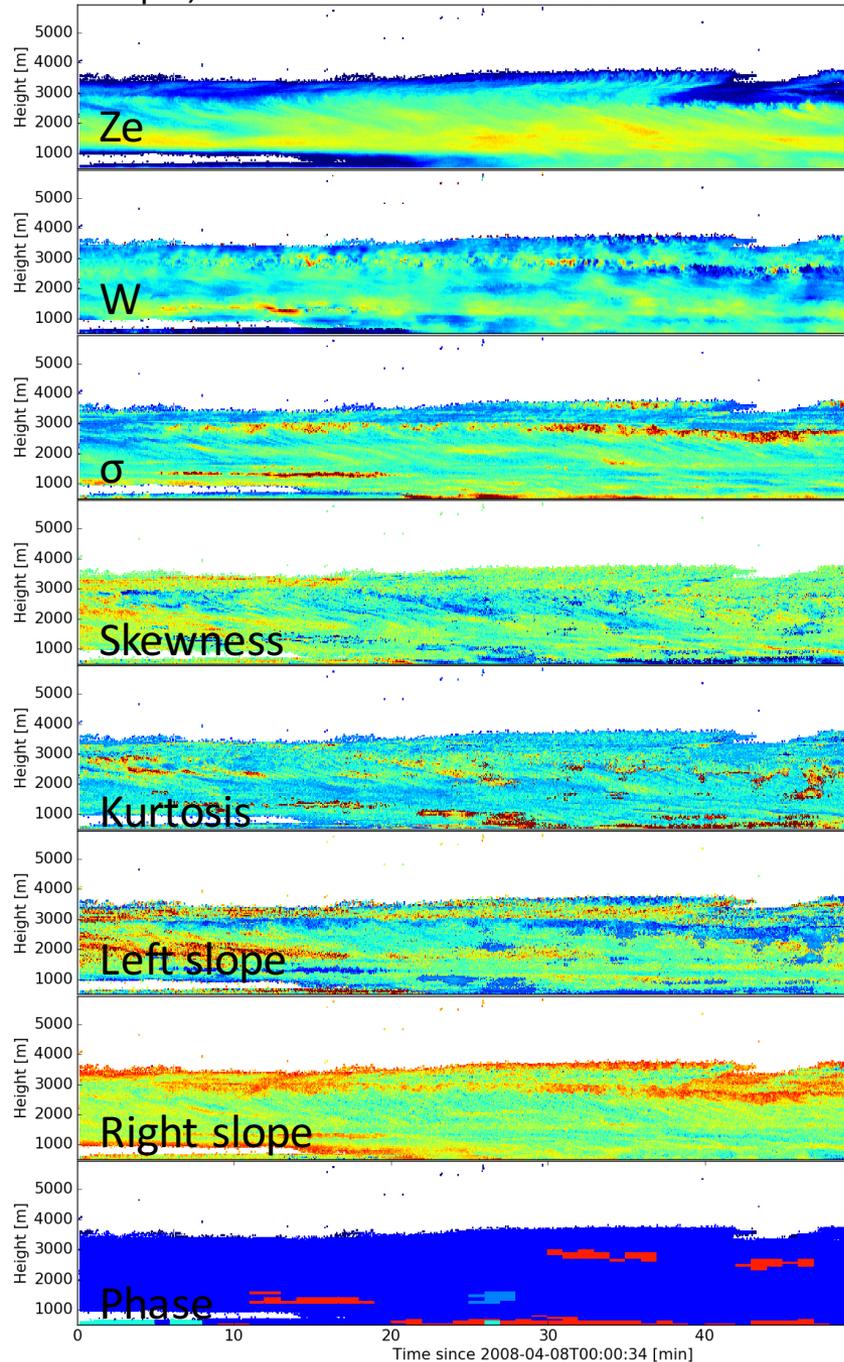
R: real measurements  
 1: single frequency  
 2: dual frequency  
 3: triple frequency



What happens if you have independent information about vertical air motion and turbulence? e.g. from other retrievals?

Improvements for quantities describing  $m(D)$  and  $A(D)$

- R: real measurements
- 1: single frequency
- 2: dual frequency
- 3: triple frequency



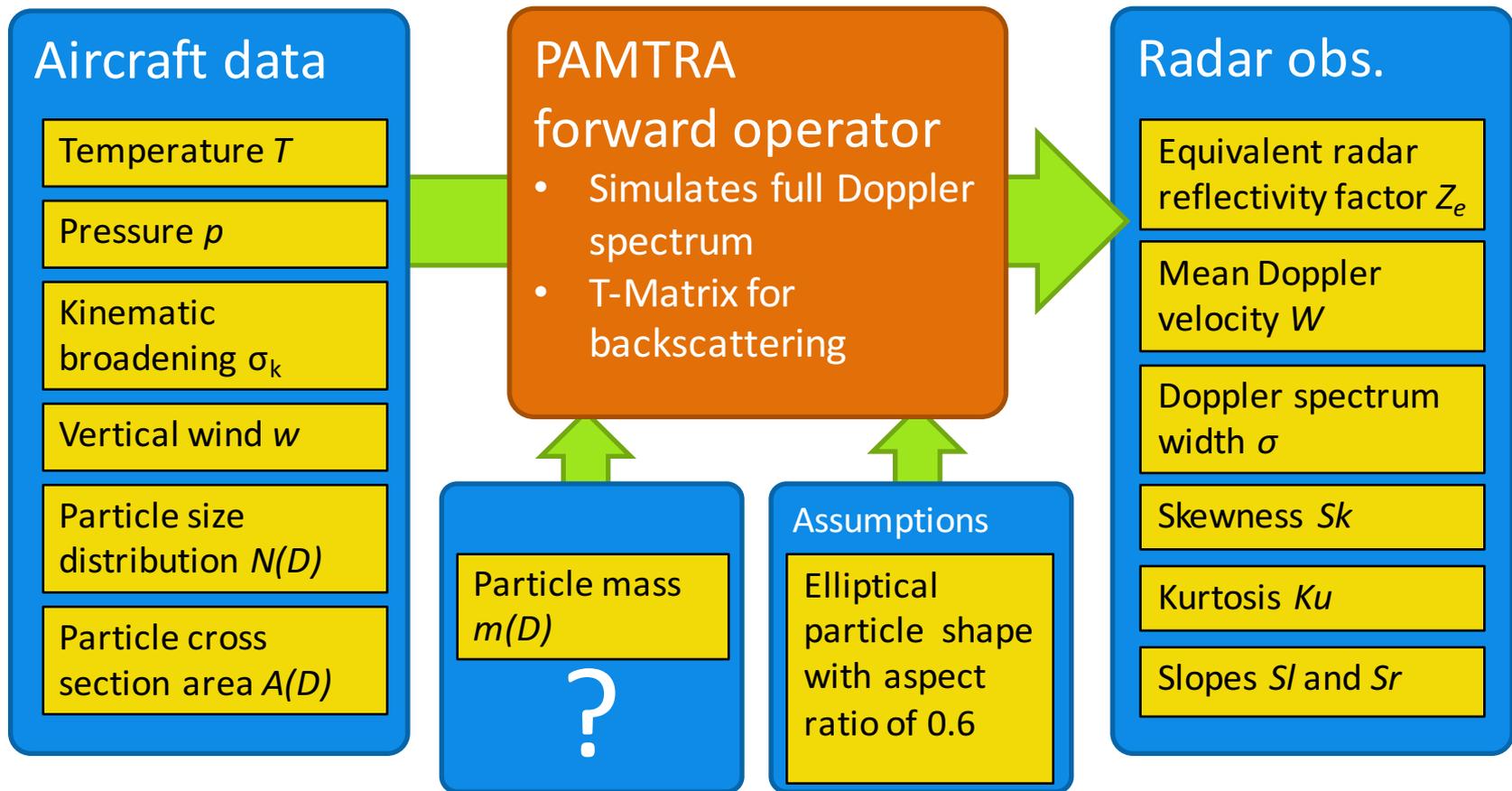
# Summary

- Higher moments of the radar Doppler spectrum can be included into retrievals of polar arctic ice clouds
- Retrievals using also higher moments can enhance retrievals of numerous microphysical *and* kinematic quantities
  - Vertical air motion
  - Kinematic broadening (-> turbulence)
  - Particle distribution
  - Particle cross section Area
  - Mass-size relation when using more than one frequency
- More arctic in situ data sets required (ACME-V!)

Thank you for your attention!

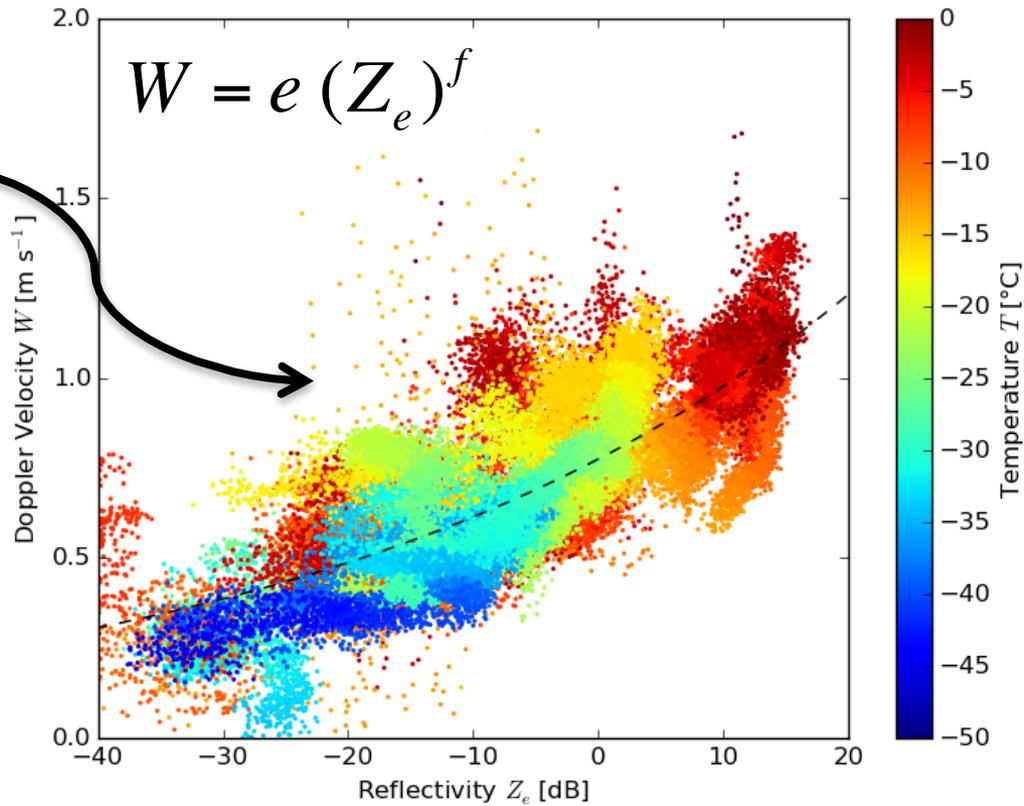


# Apply PAMTRA to ISDAC data



# $Z_e - W$ relation for $m(D)$

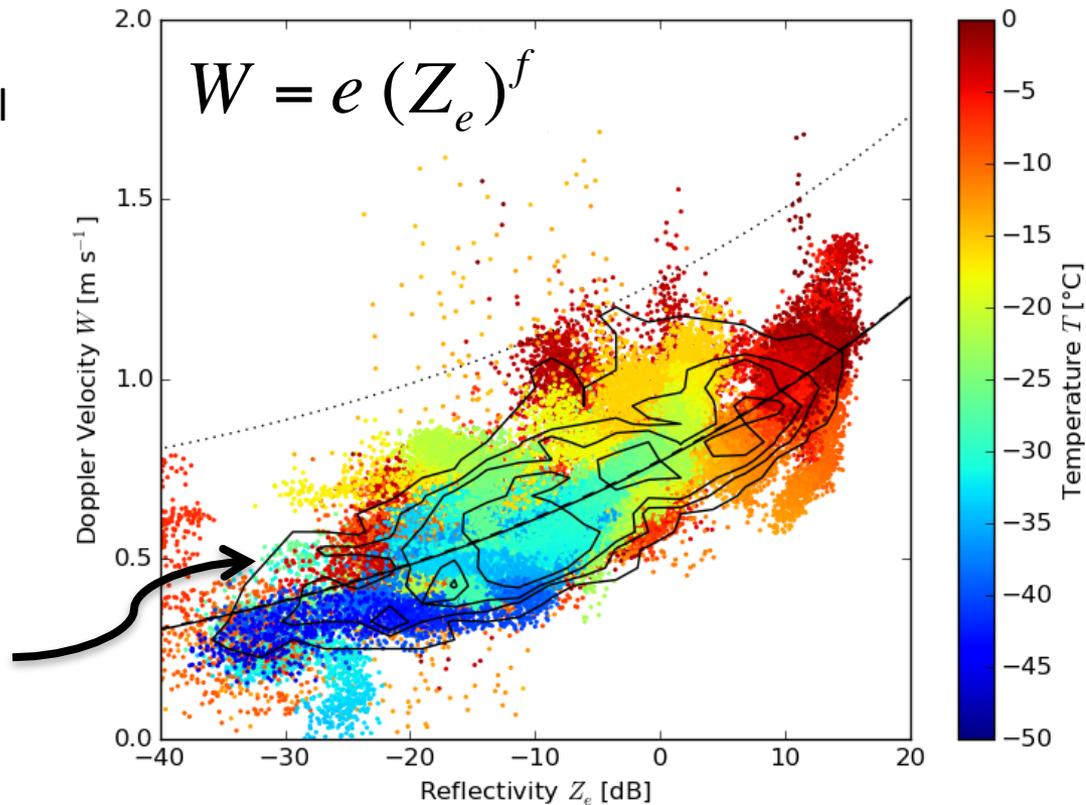
Measured by  
MMCR!



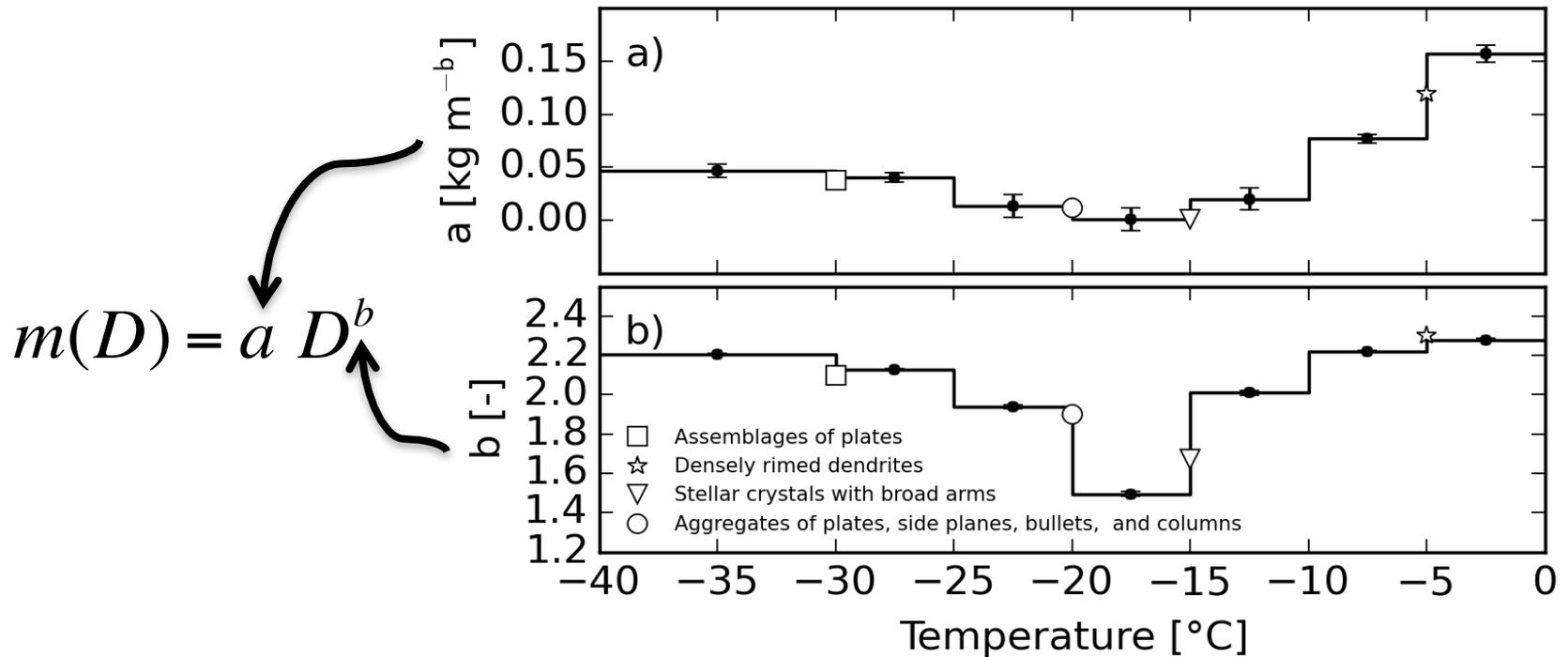
# $Z_e - W$ relation for $m(D)$

Using optimal estimation choose  $m(D)$  such that MMCR observations are matched

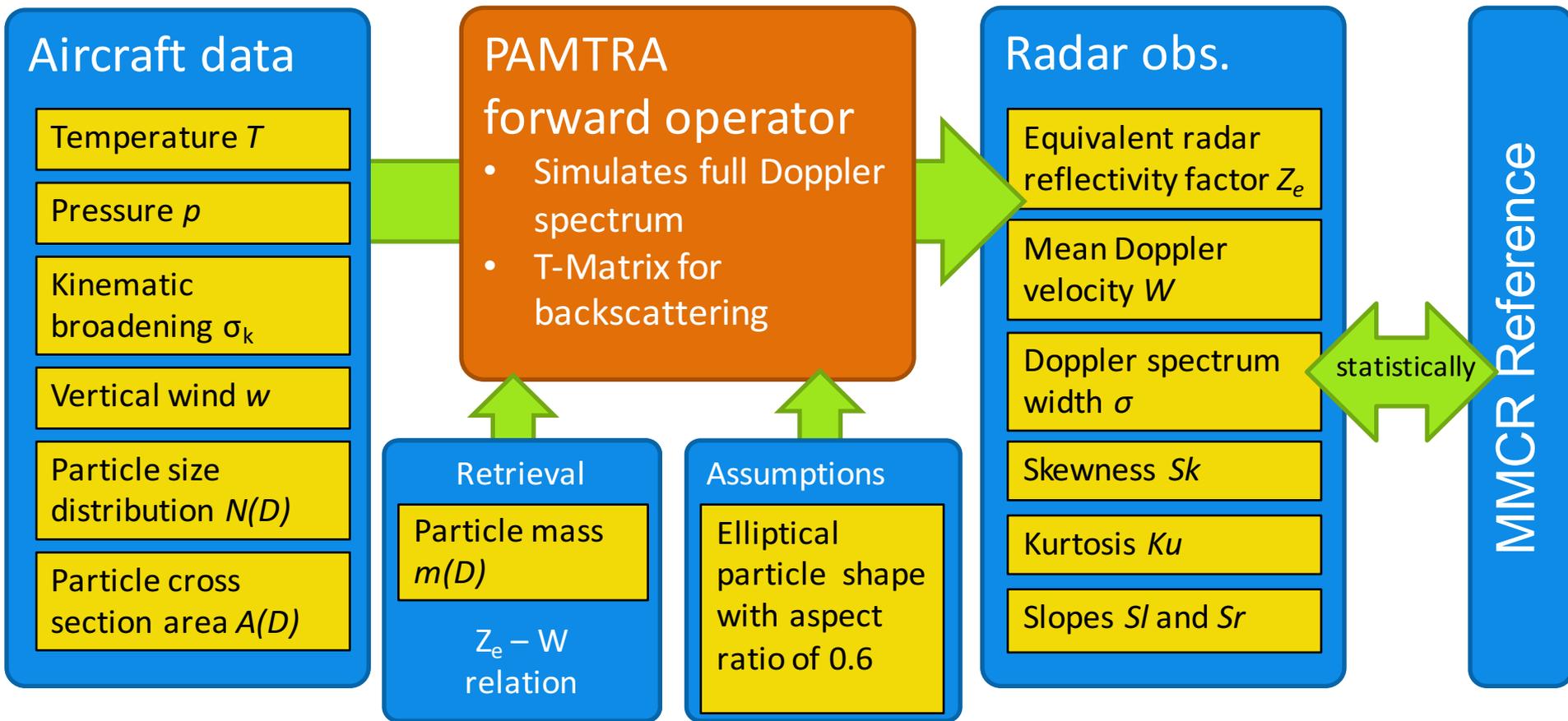
Forward modelled ISDAC data



# Exploit temperature dependence

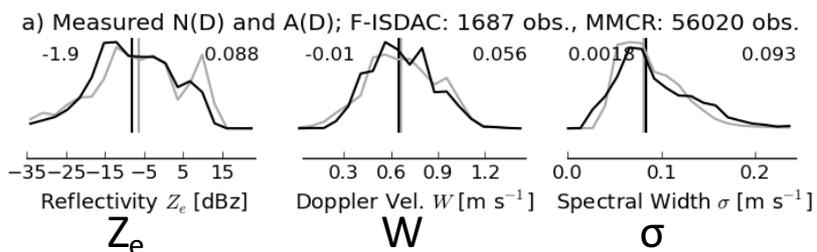


# Validate simulation



# Compare ISDAC and MMCR

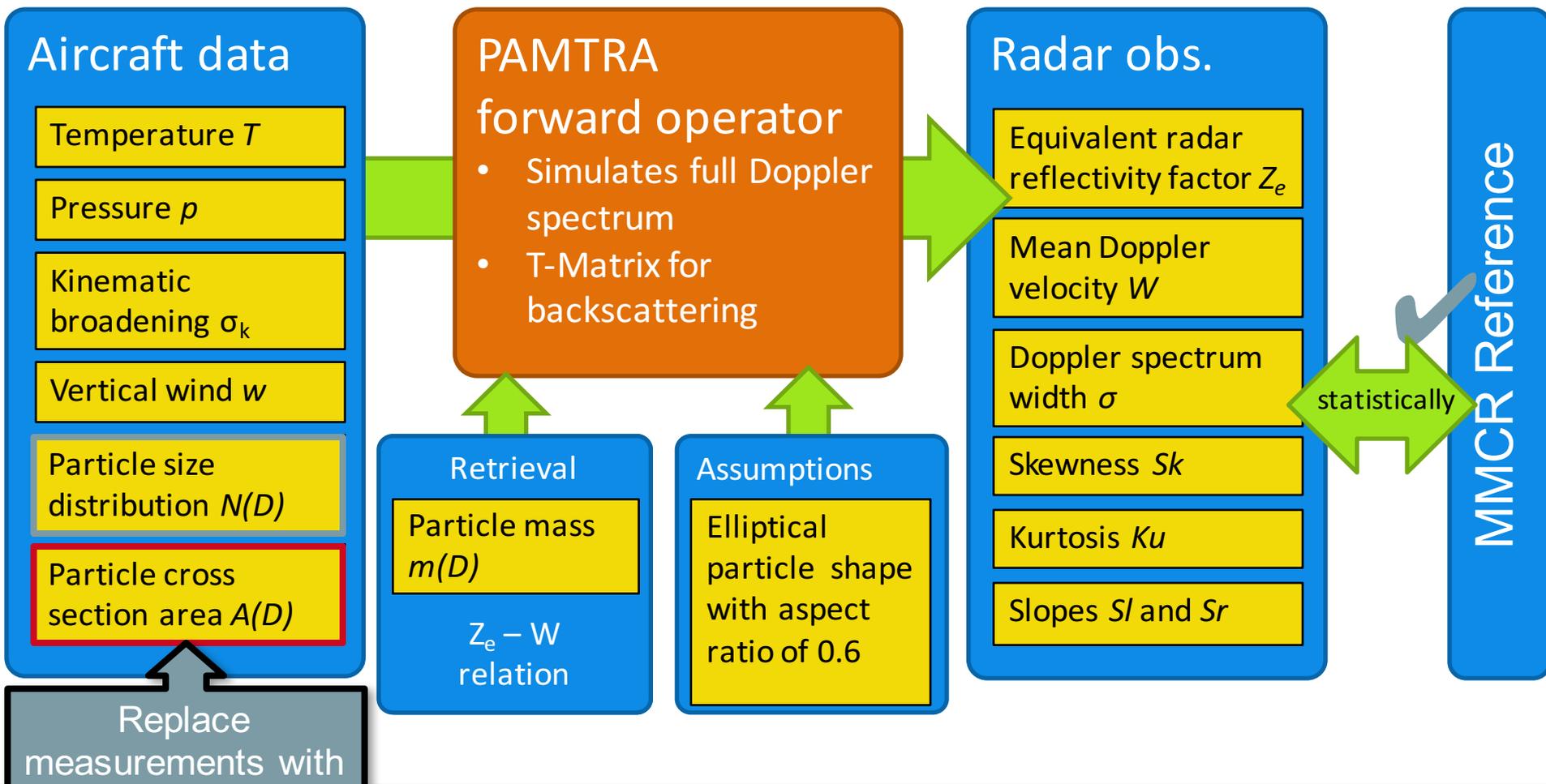
High agreement, because retrieved  $m(D)$  is used



— MMCR — F-ISDAC

Also agreement of higher moments is high, problems with right slope: rare, large particles

# Validate simulation



Replace measurements with parameterizations

IASR PI Meeting

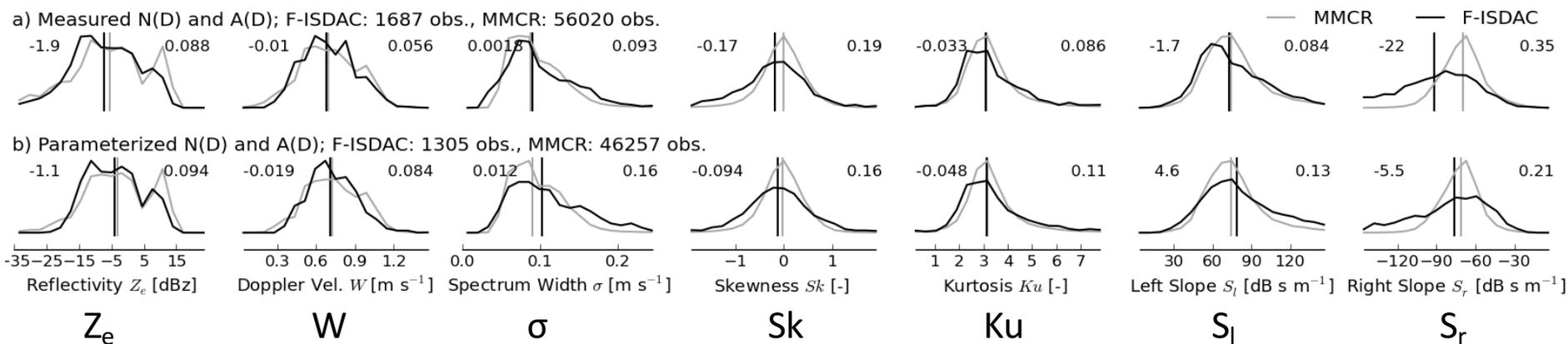
Maximilian Maahn  
May 2 - 6, 2016

Cooperative Institute for Research in Environmental Sciences

UNIVERSITY OF COLORADO BOULDER and NOAA

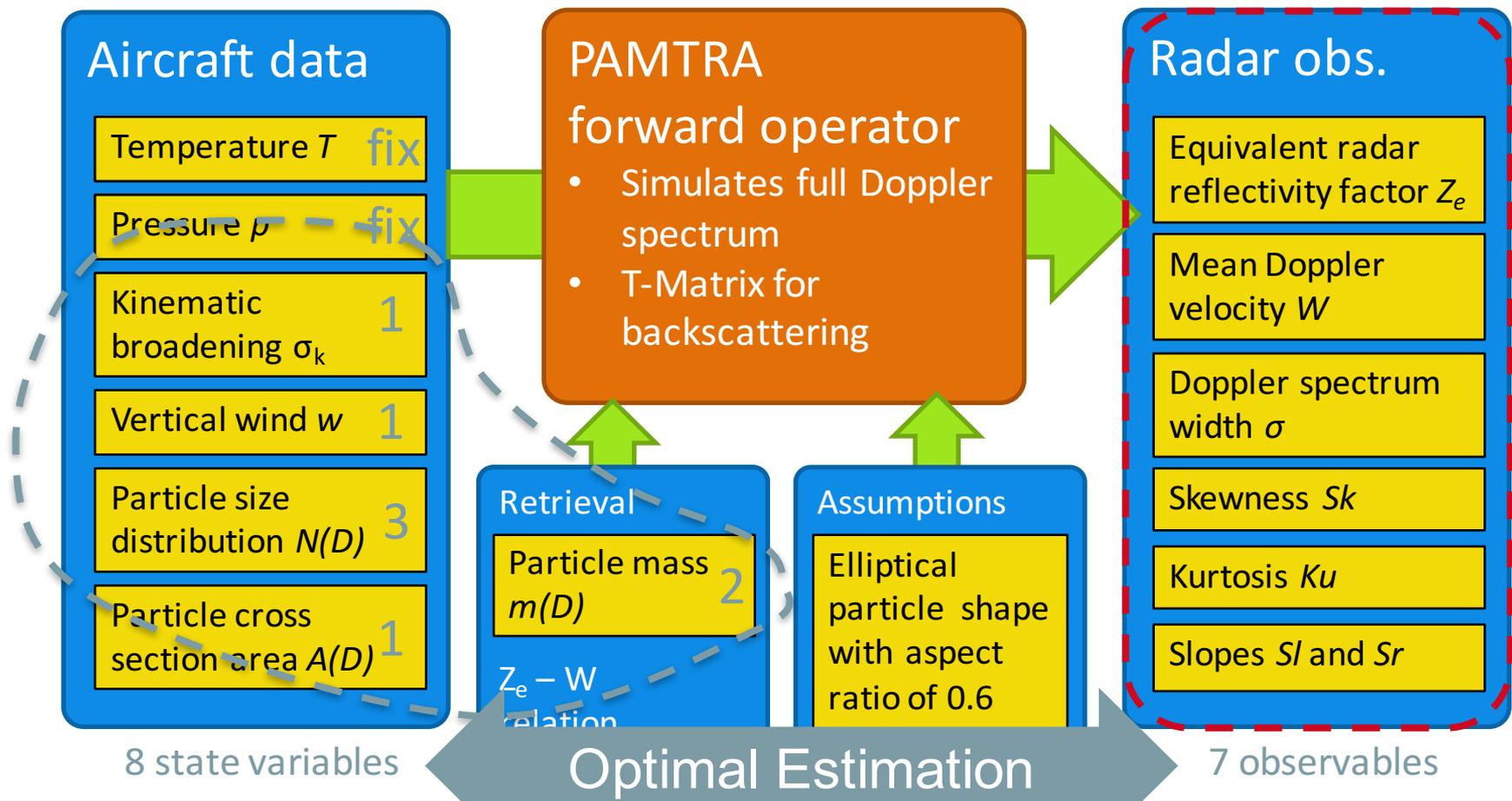


# Compare ISDAC and MMCR

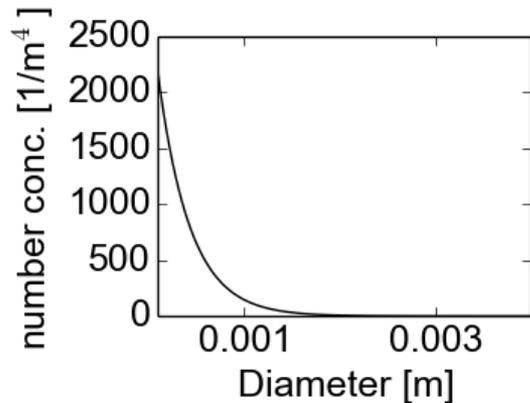


Using the correct parameterizations, agreement is still high

# Set up idealized retrieval



# How to simulate a Doppler spectrum (1)

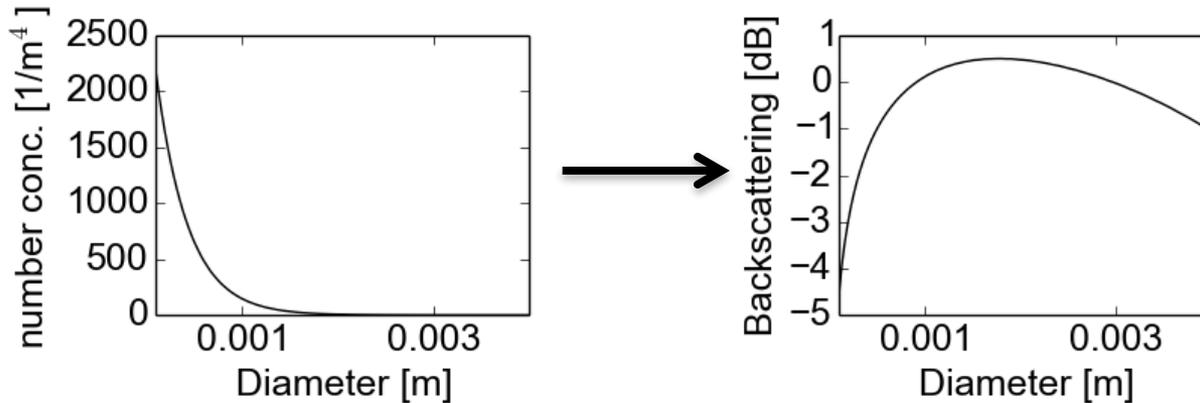


Particle Size  
Distribution  
 $N(D)$ , e.g.  
Gamma  
distribution

Required  
variables:

$N(D)$  (3p)

# How to simulate a Doppler spectrum (2)



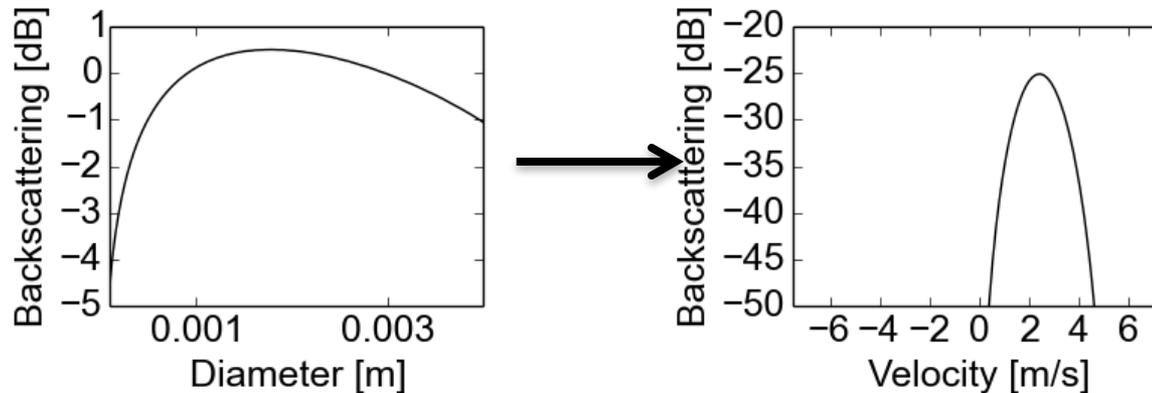
$$N(D) \rightarrow \eta(D)$$

Single Scattering properties, e.g. T-Matrix, depend on temperature  $T$ , particle mass  $m(D)$  and aspect ratio  $AR$

Required variables:

$N(D)$	(3p)
$m(D)$	(2p)
$AR$	(1p)
$T$	(1p)

# How to simulate a Doppler spectrum (3)



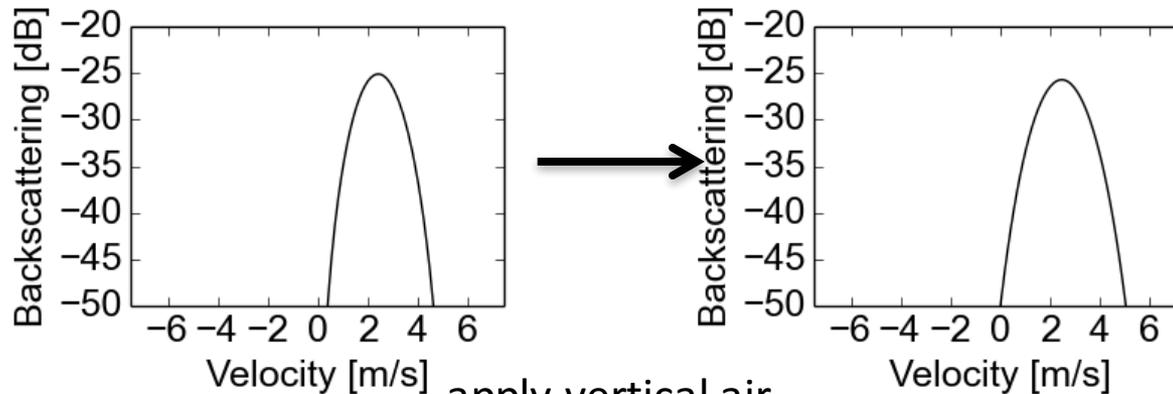
$$\eta(D) \rightarrow \eta(v)$$

Fall velocity relation  $v(D)$ , depends on cross section area  $A(D)$ ,  $m(D)$ , pressure  $p$  and  $T$

Required variables:

$N(D)$	(3p)
$m(D)$	(2p)
AR	(1p)
T	(1p)
$A(D)$	(2p)
$p$	(1p)

# How to simulate a Doppler spectrum (4)



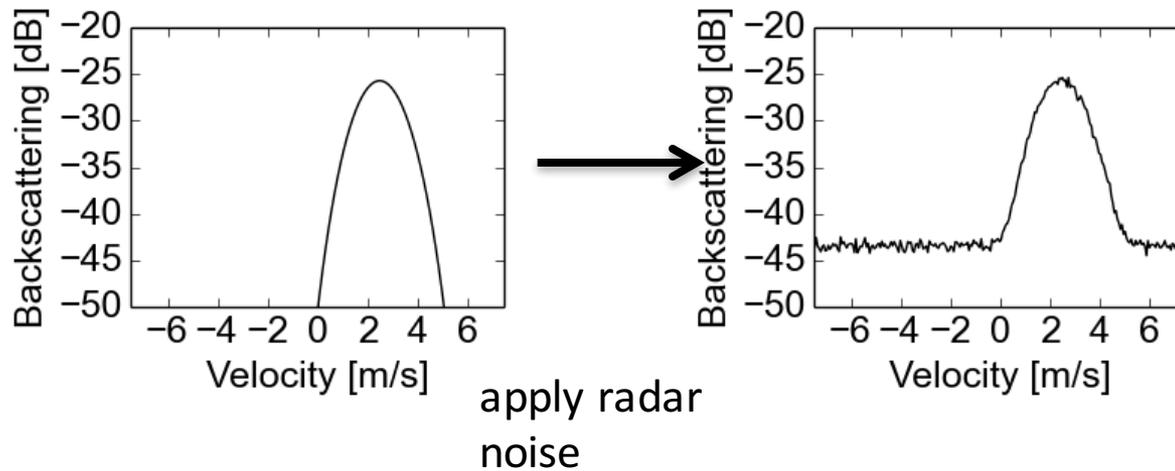
apply vertical air motion and kinematic broadening

depends (among others) on horizontal wind  $U$  and Eddy dissipation rate  $\epsilon$

Required variables:

N(D)	(3p)
m(D)	(2p)
AR	(1p)
T	(1p)
A(D)	(2p)
p	(1p)
U	(1p)
kin	(2p)

# How to simulate a Doppler spectrum (5)



depends on radar specifications  
(receiver, integration time etc.)

Required variables:

N(D)	(3p)
m(D)	(2p)
AR	(1p)
T	(1p)
A(D)	(2p)
p	(1p)
U	(1p)
kin	(2p)

# Required data

In total 13 required quantities!  
Not all of them equally important

Required variables:

N(D)	(3p)
m(D)	(2p)
AR	(1p)
T	(1p)
A(D)	(2p)
p	(1p)
U	(1p)
kin	(2p)