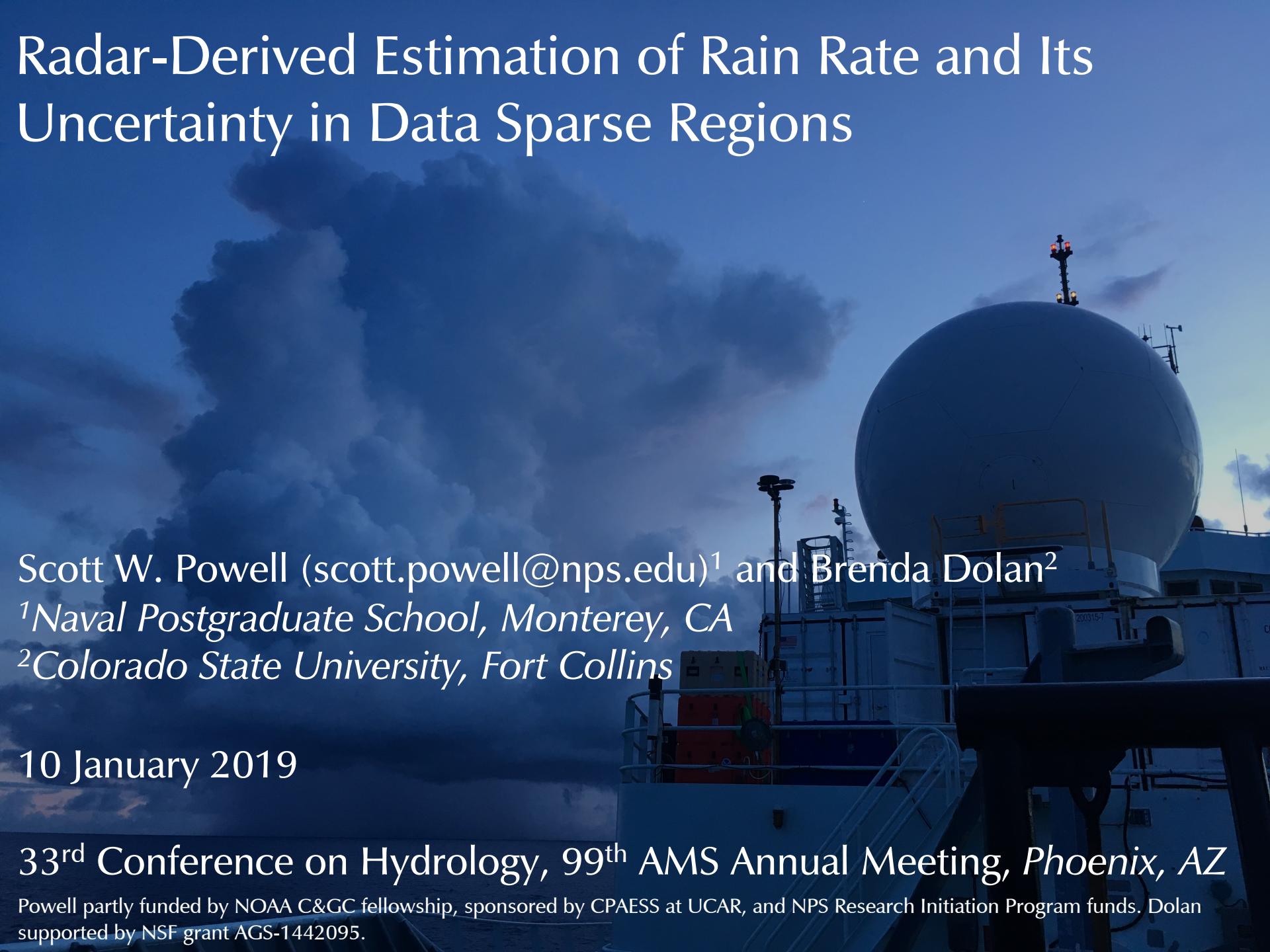


# Radar-Derived Estimation of Rain Rate and Its Uncertainty in Data Sparse Regions



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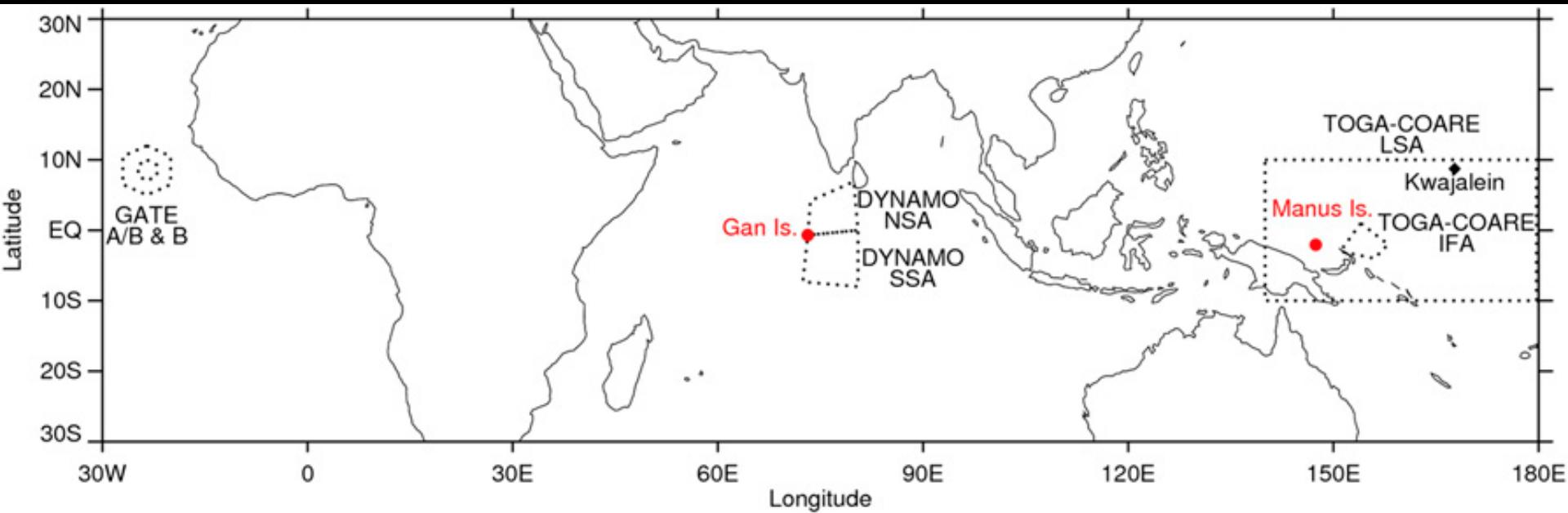
Powell partly funded by NOAA C&GC fellowship, sponsored by CPAESS at UCAR, and NPS Research Initiation Program funds. Dolan supported by NSF grant AGS-1442095.

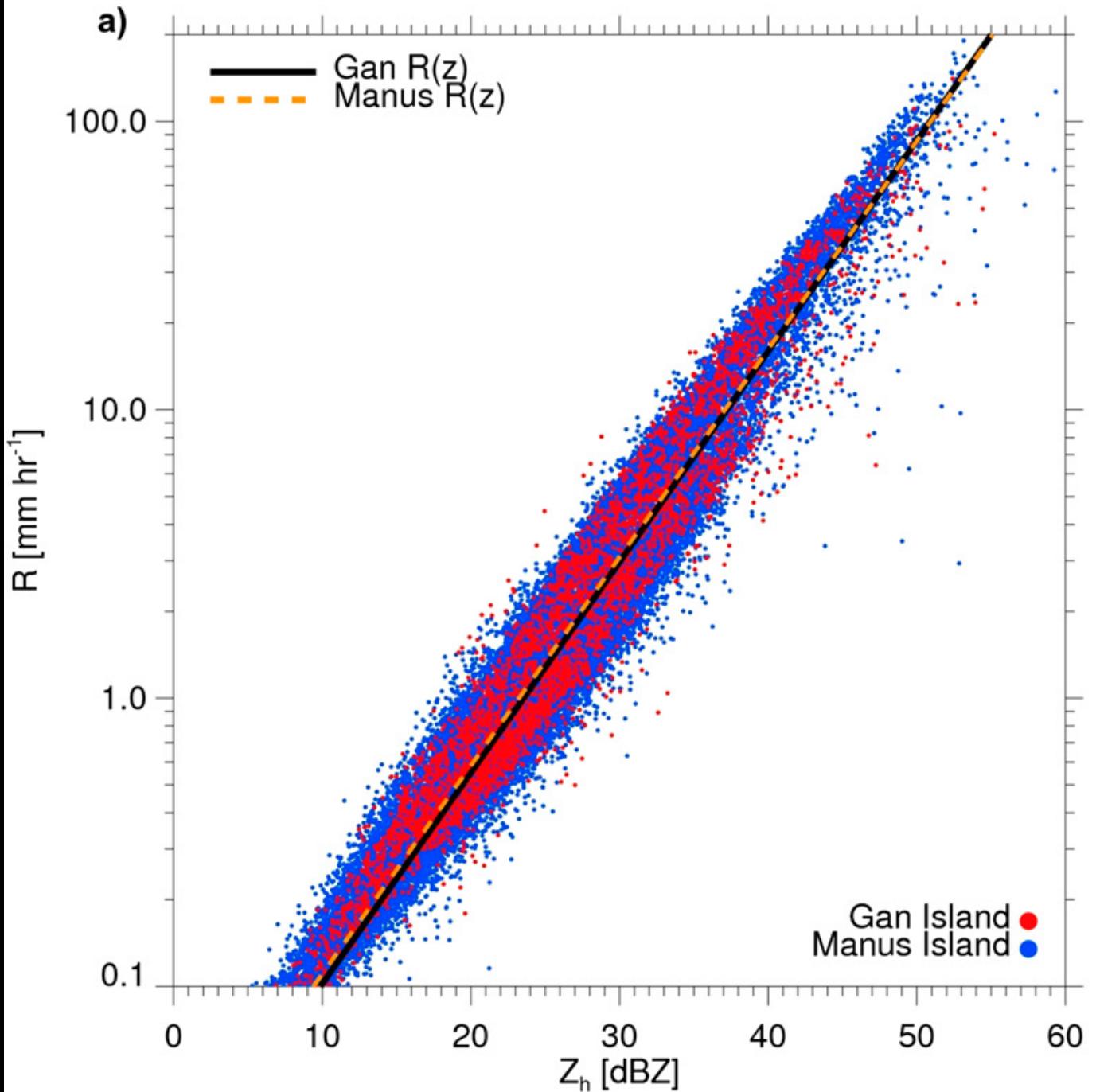
**GOAL:** Describe radar-estimated rain rates as a *range* based on ground-based observations from a disdrometer.

Why?

1. Many tropical oceanic radar datasets exist in data sparse regions without rain gauge data (e.g. radar on remote islands or ship for field campaigns).
2. Uncertainty bounds provide target for validation of other instruments.

Thompson et al. (2015)



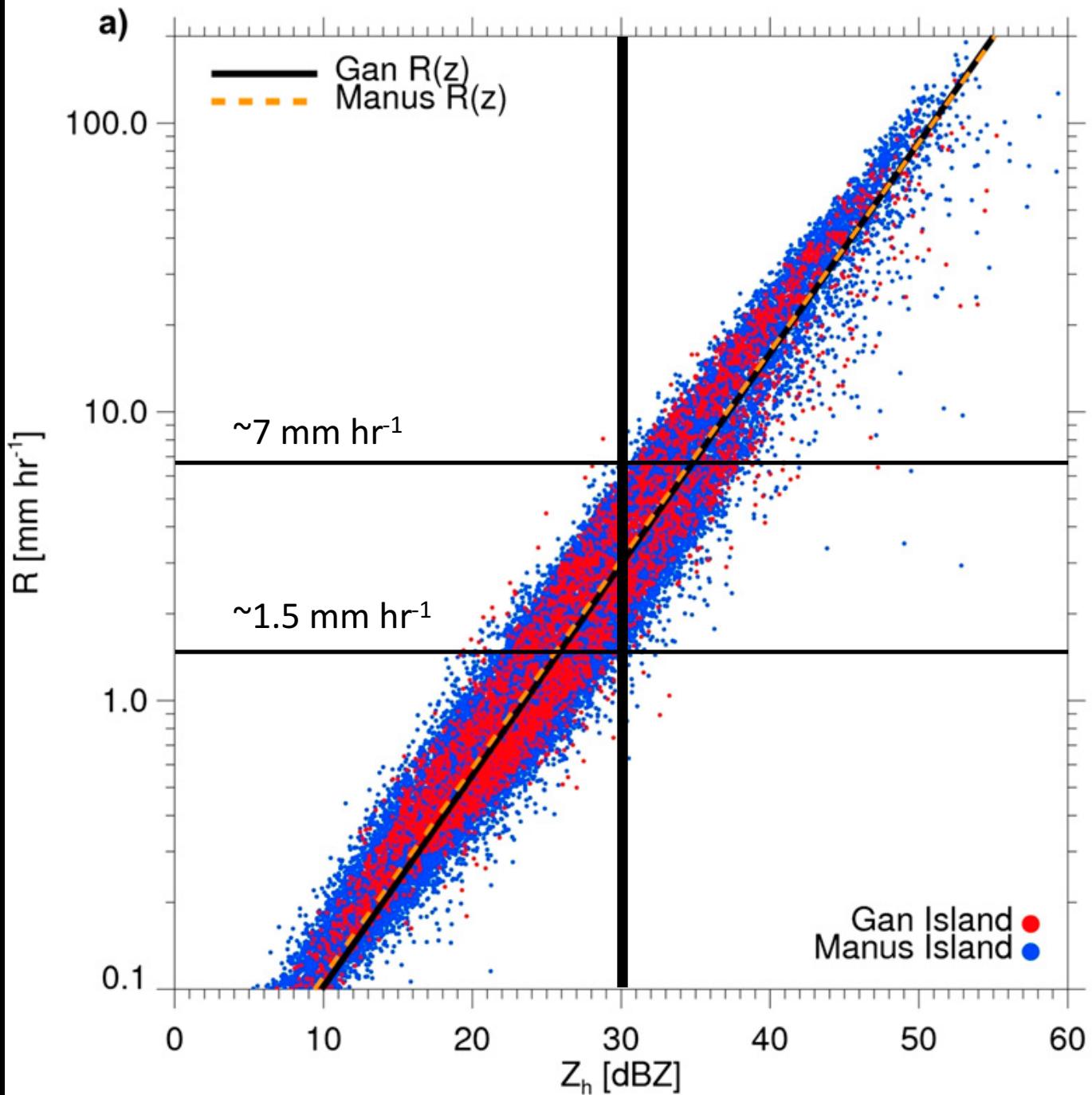
**a)**

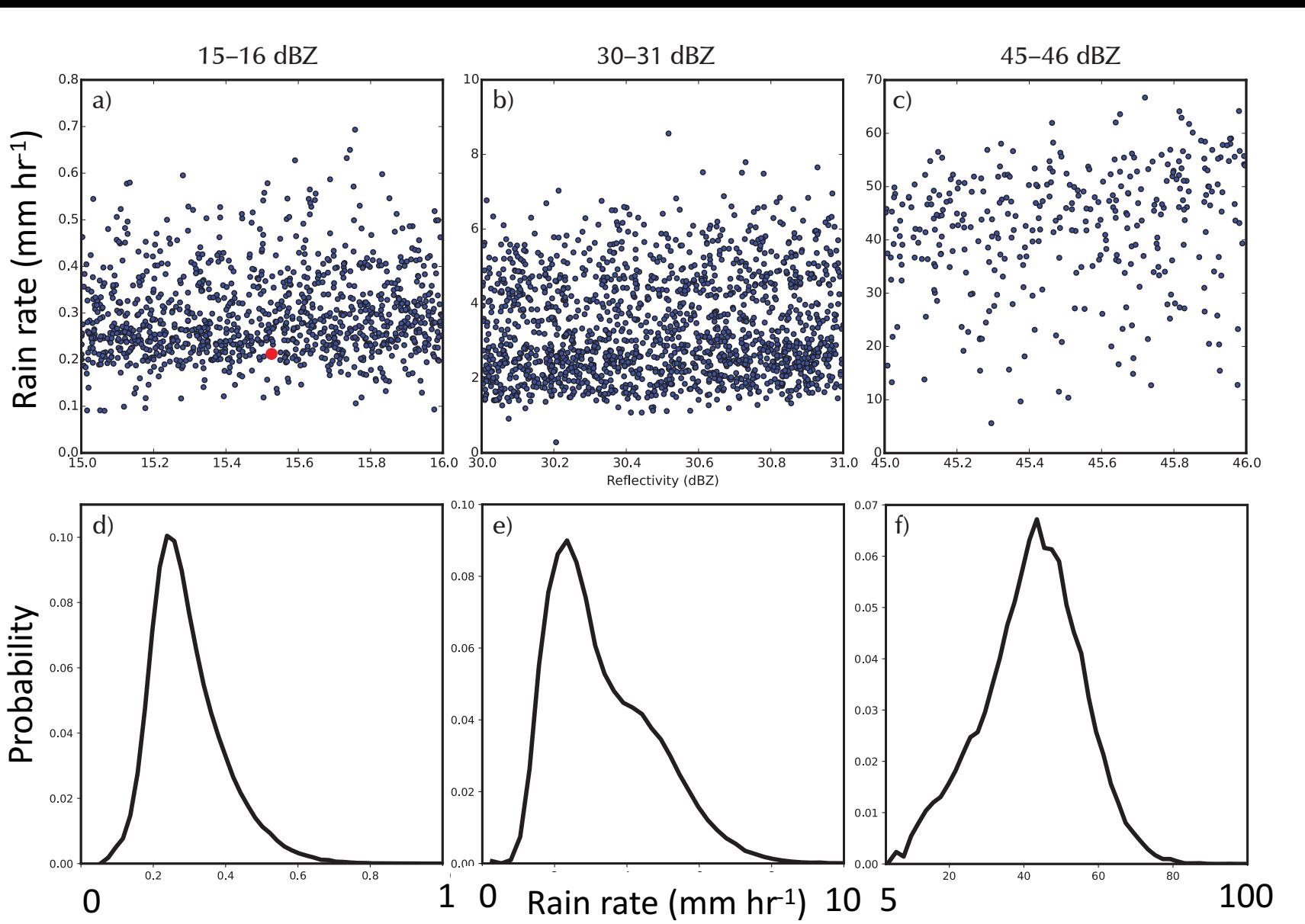
Thompson et al.  
(2015)

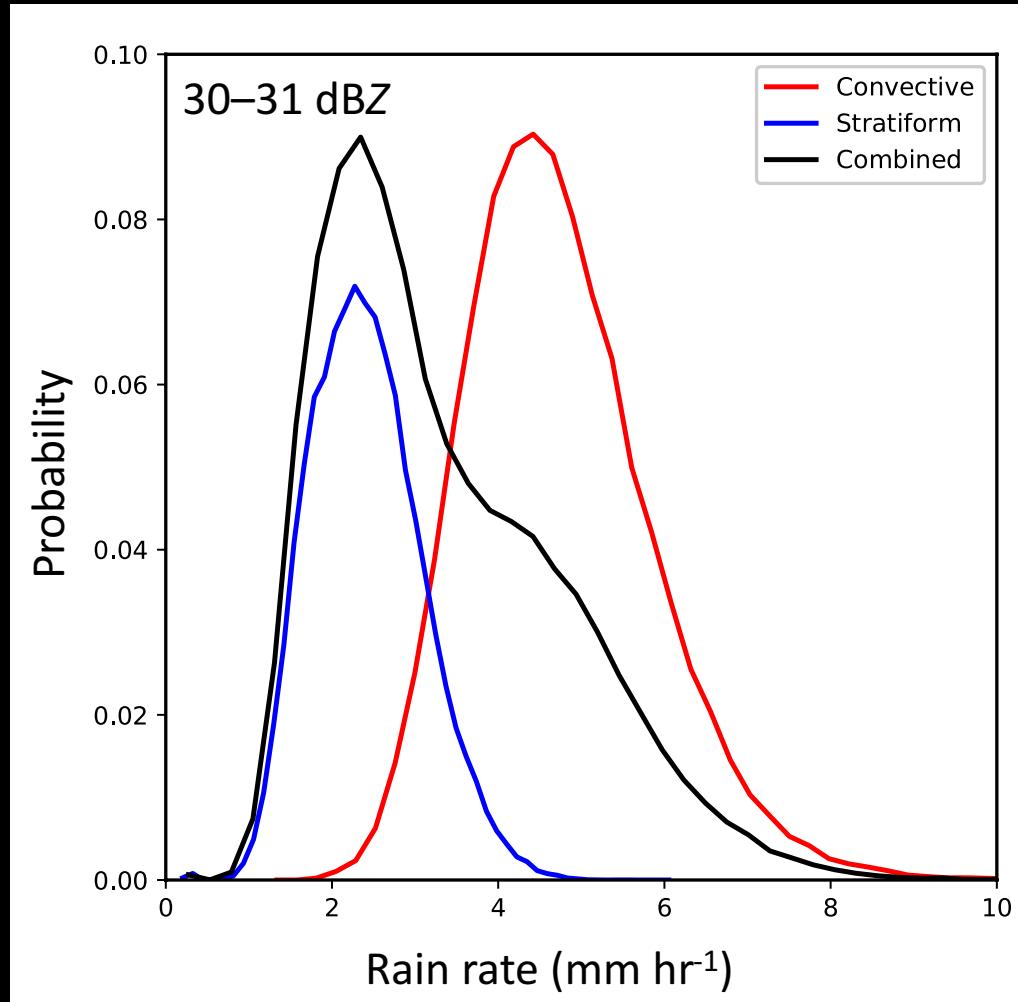
*Z-R* relationships  
derived from  
2DVD data

Thompson et al.  
(2015)

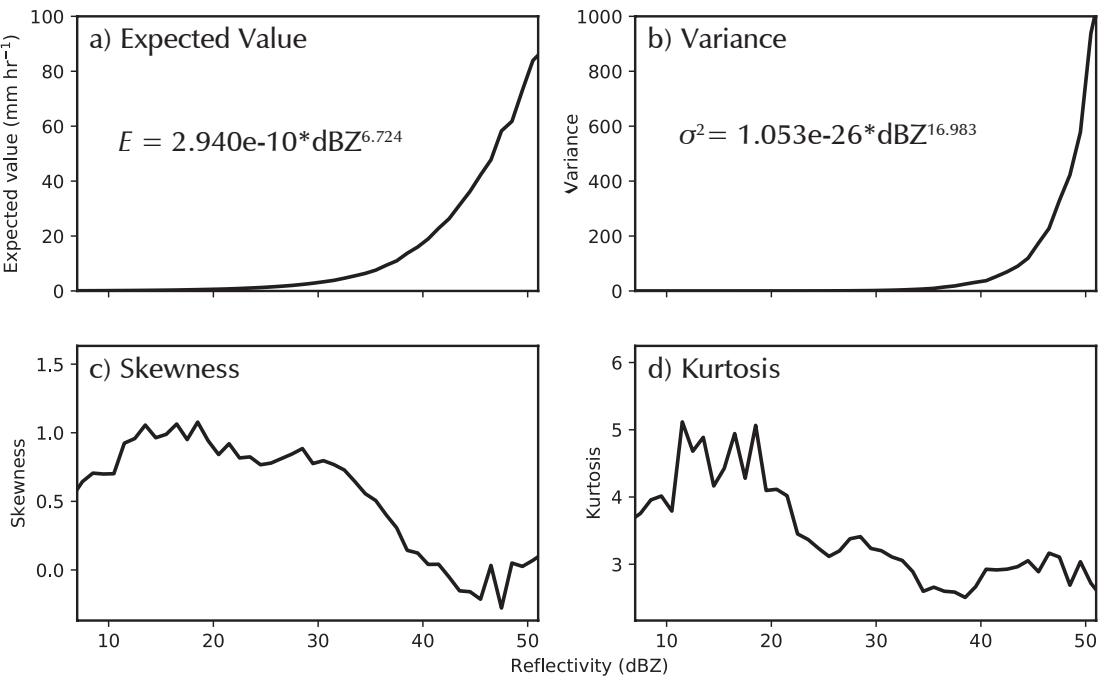
Z-R relationships  
derived from  
2DVD data





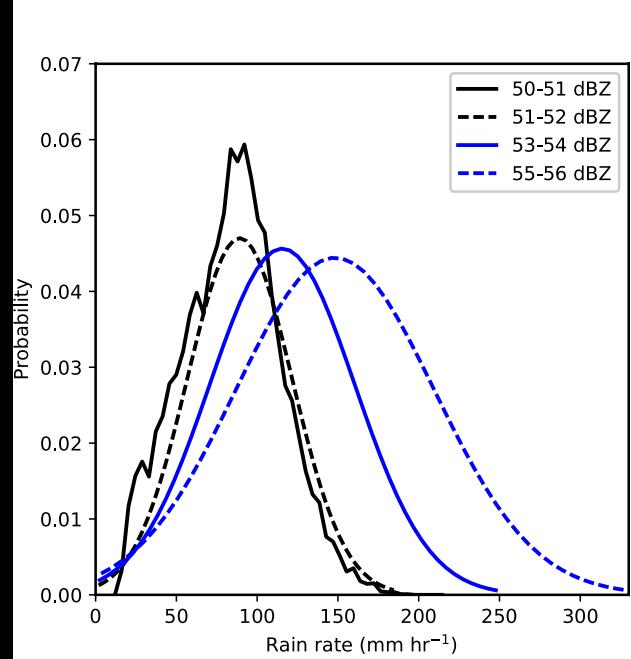


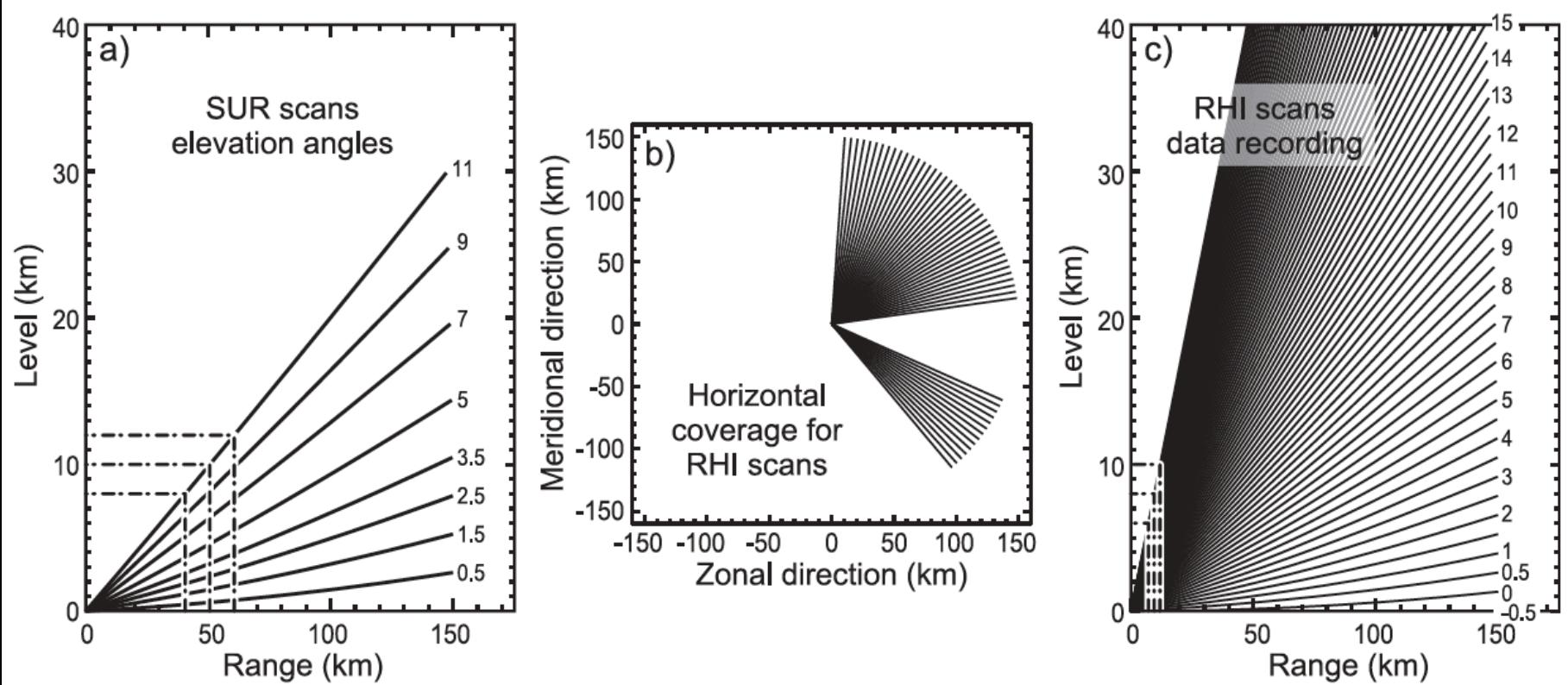
Separate PDFs can be generated for convective and stratiform precipitation.



Generating PDFs for high reflectivity  
 (which is observed too infrequently to do with previously shown method)

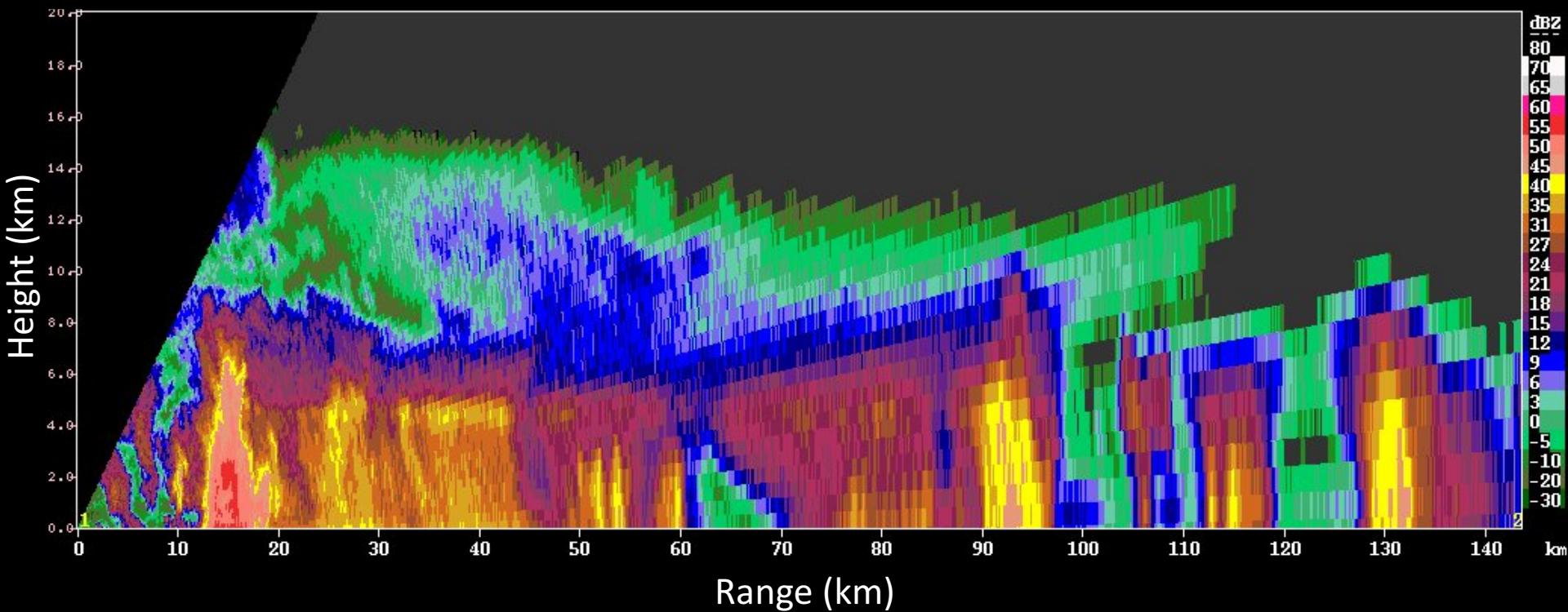
10 Jan 2019





We use data before it is re-gridded to rectilinear coordinates. But most data is above the surface.

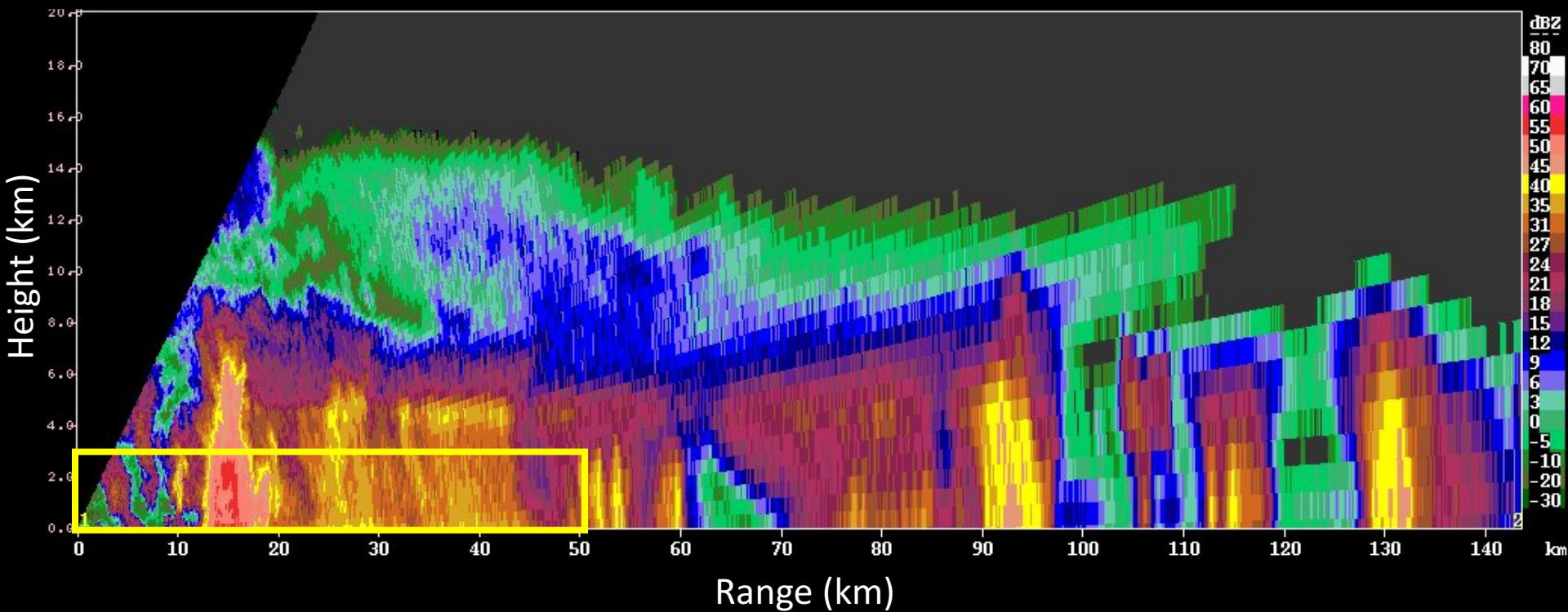
Example of RHI data from S-PolKa during DYNAMO.



Grid this data to 100 meter vertical and horizontal resolution within 50 km of radar.

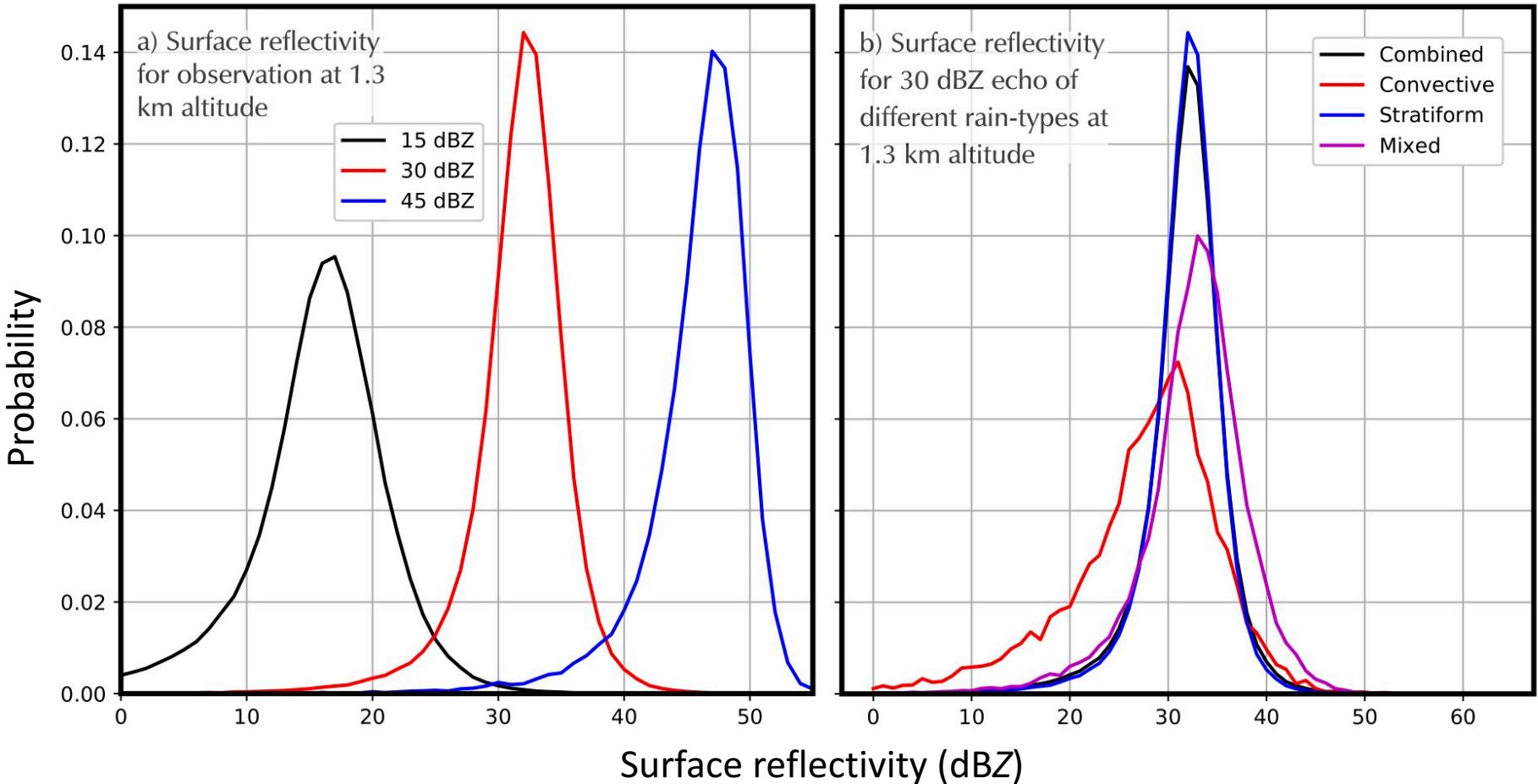
This generates PDFs for downward extrapolation of reflectivity to surface.

Example of RHI data from S-PolKa during DYNAMO.



Grid this data to 100 meter vertical and horizontal resolution within 50 km of radar.

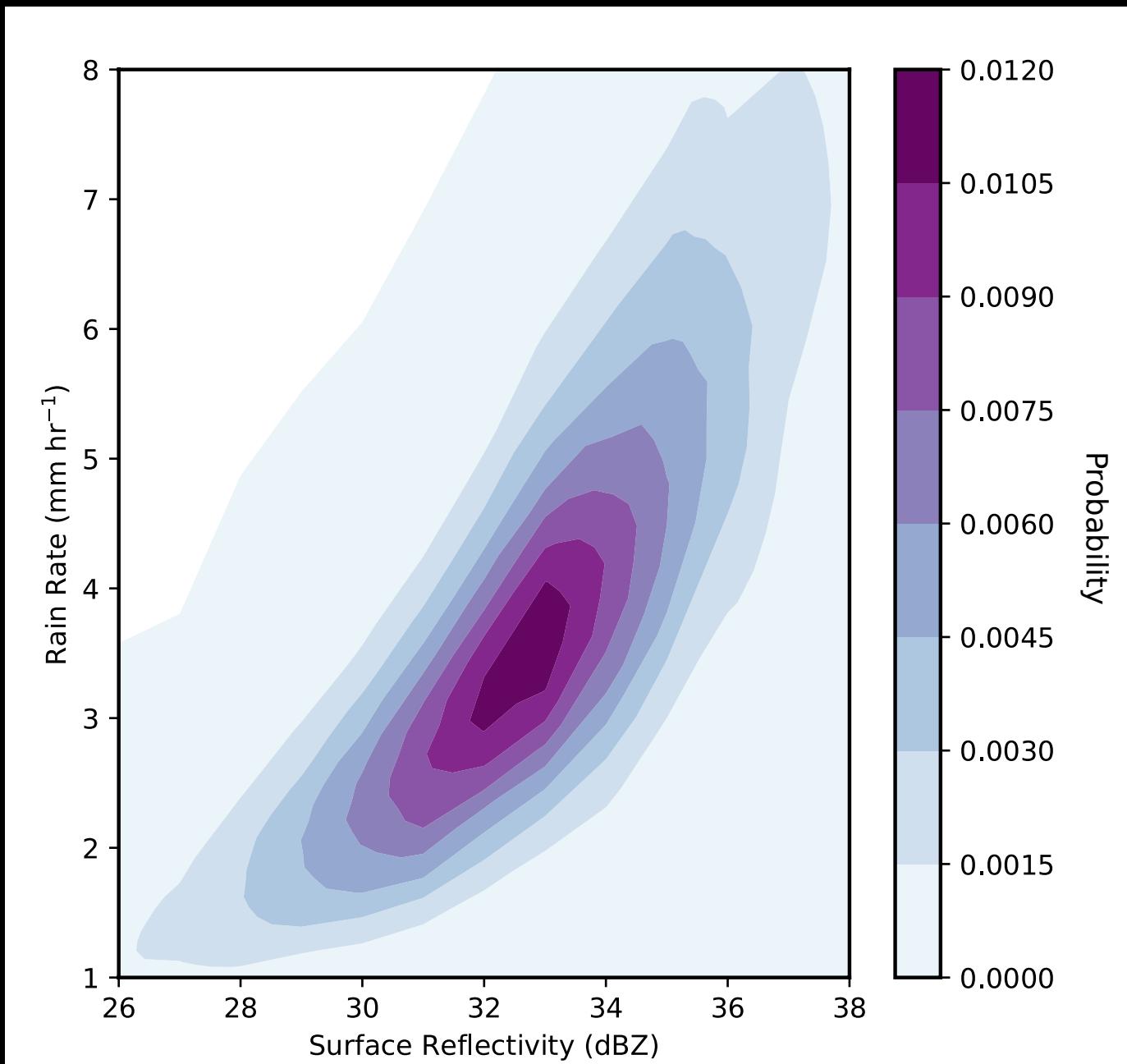
This generates PDFs for downward extrapolation of reflectivity to surface.



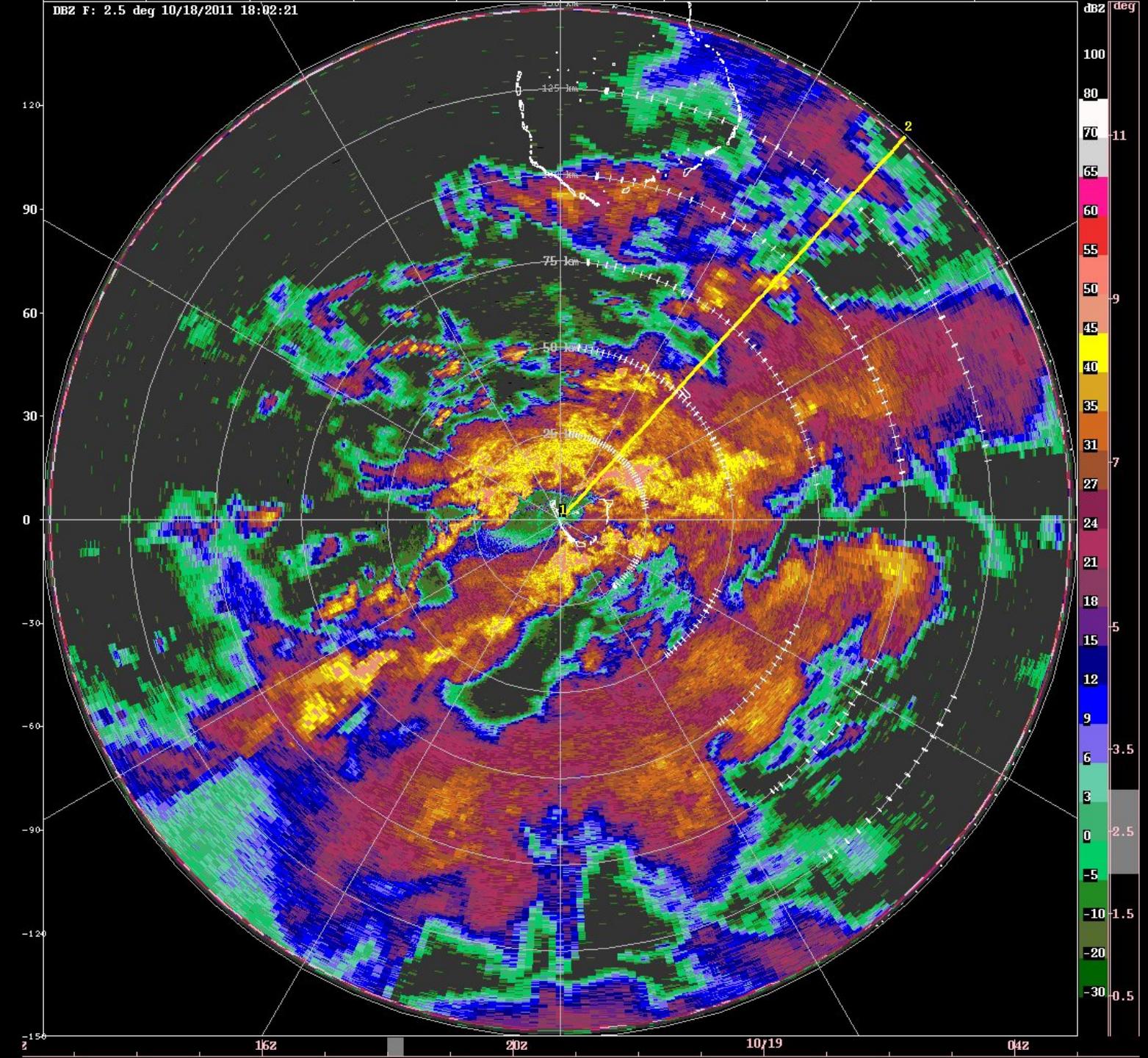
2D PDF  
containing  
spread in  
empirical rain  
rate relation and  
correction for  
extrapolation to  
surface

(This is for a 30  
dBZ stratiform  
echo at 1.3 km.)

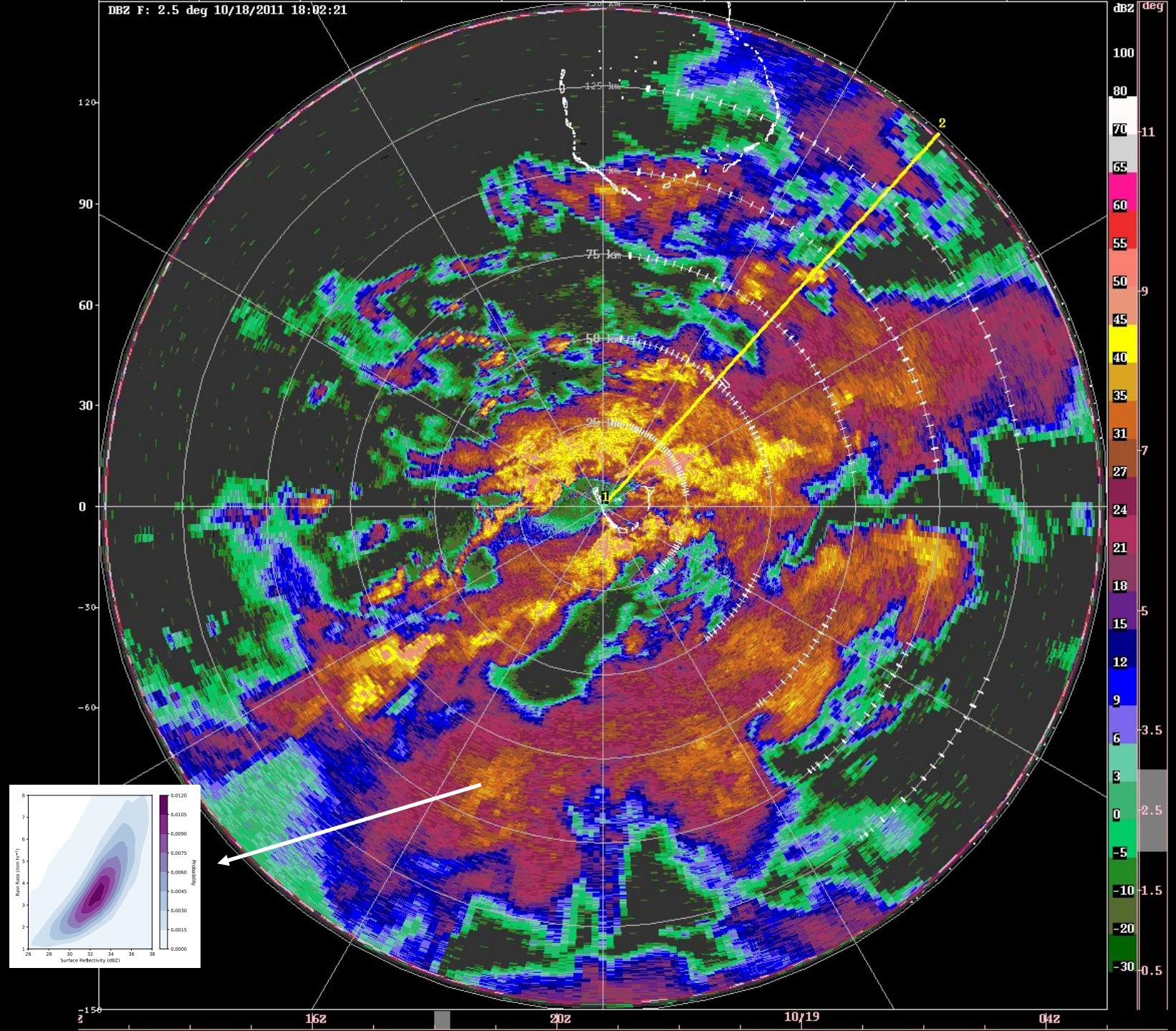
$$P(Z, R) = \frac{P_1(Z)*P_2(Z,R)}{\sum P_1(Z)*P_2(Z,R)}$$



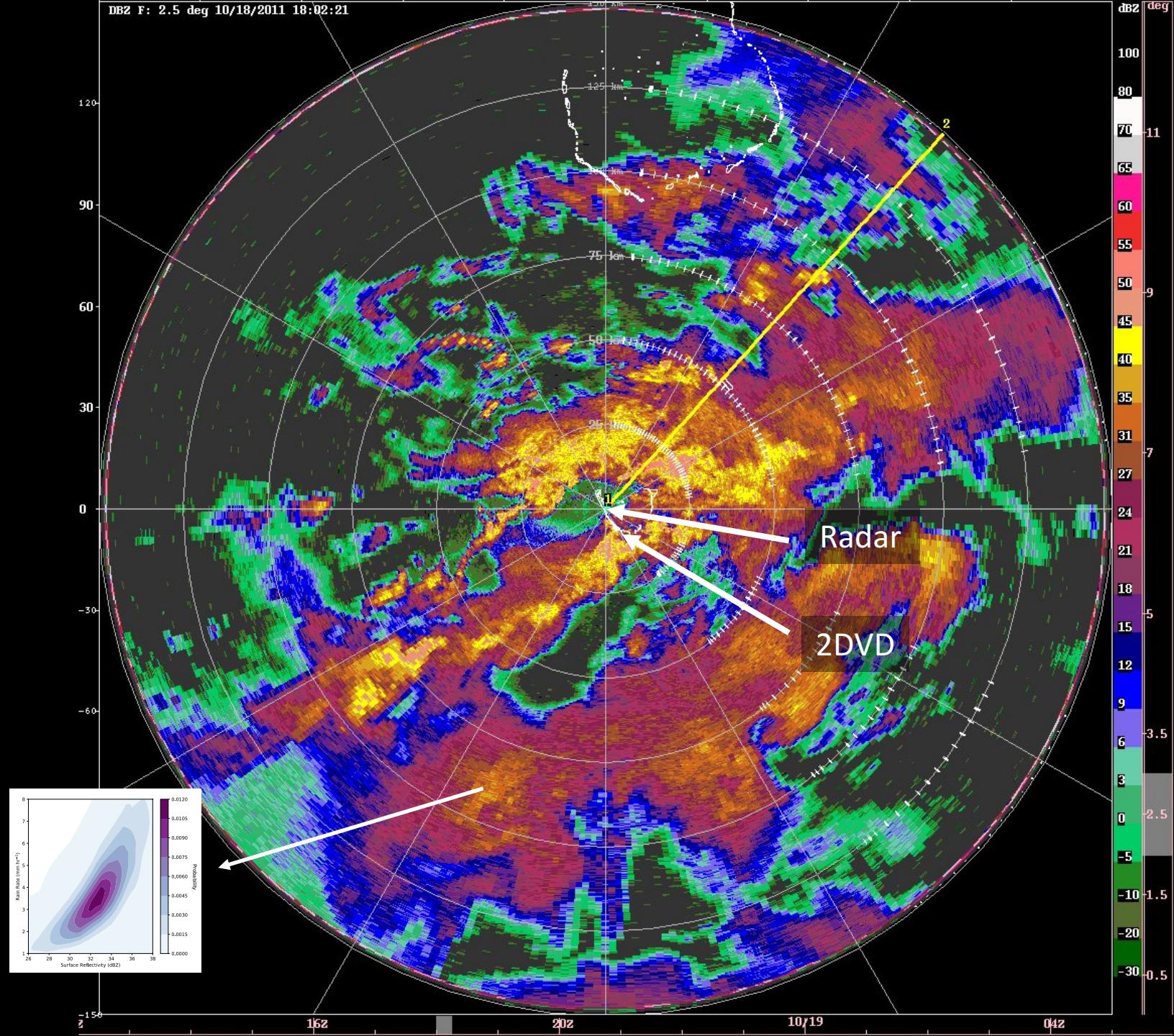
DBZ F: 2.5 deg 10/18/2011 18:02:21

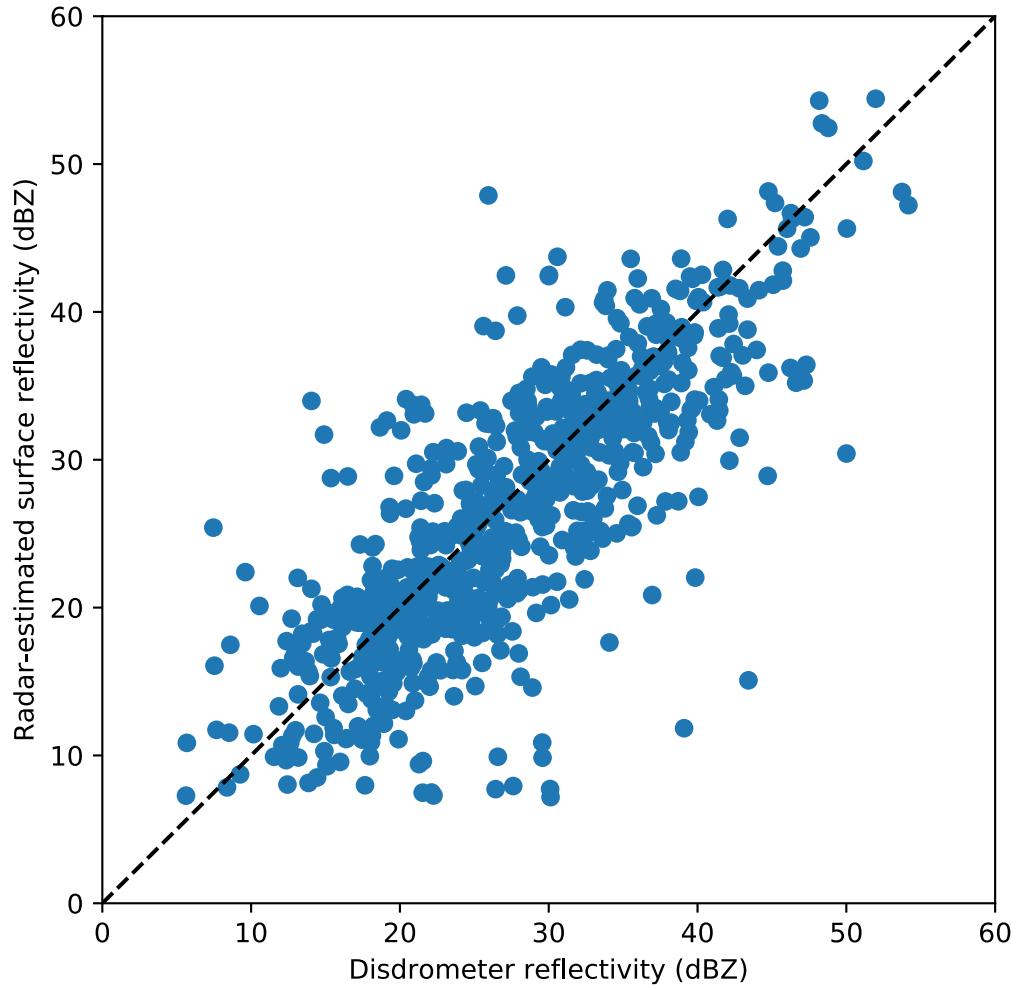


DBZ F: 2.5 deg 10/18/2011 18:02:21



DBZ F: 2.5 deg 10/18/2011 18:02:21

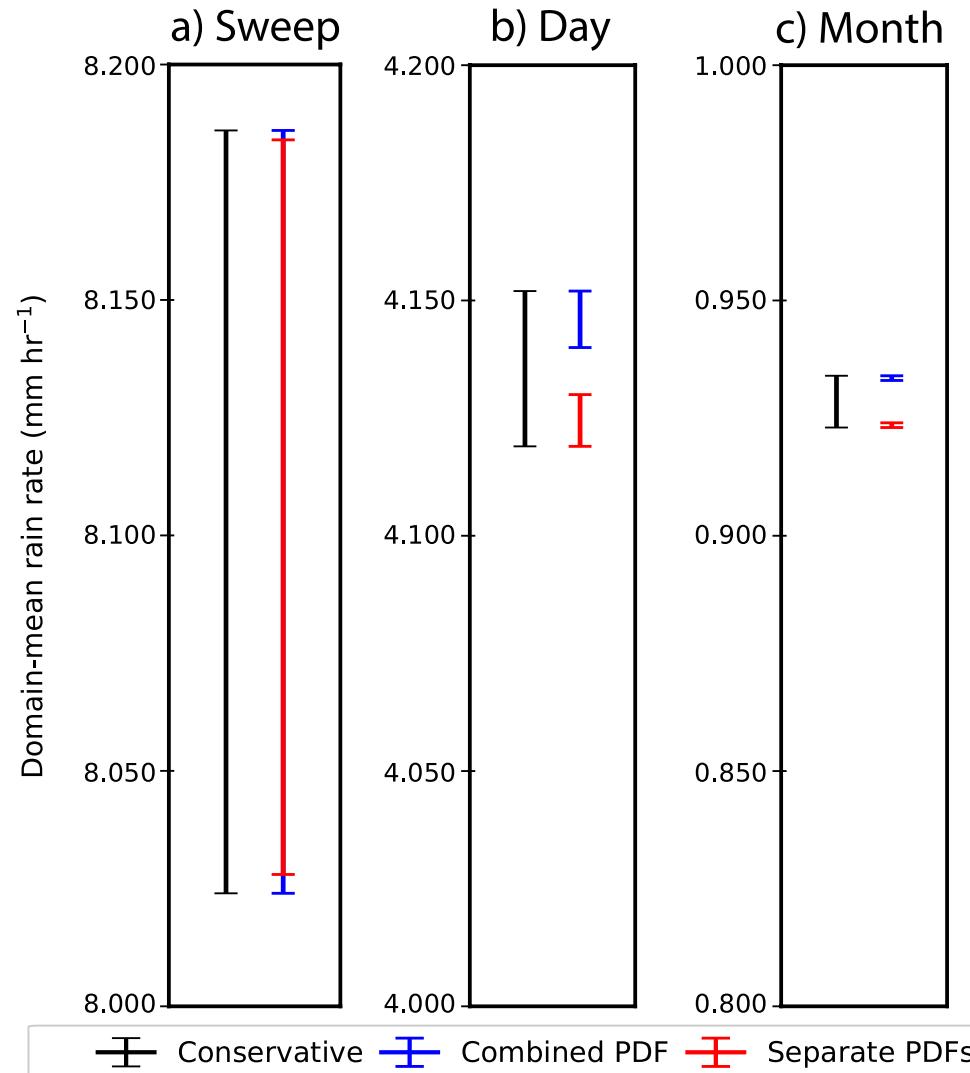




~95% of 2DVD observed rain rates fall within the error bounds determined from the radar-derived rain rate. However, for individual points, the uncertainty is very large!

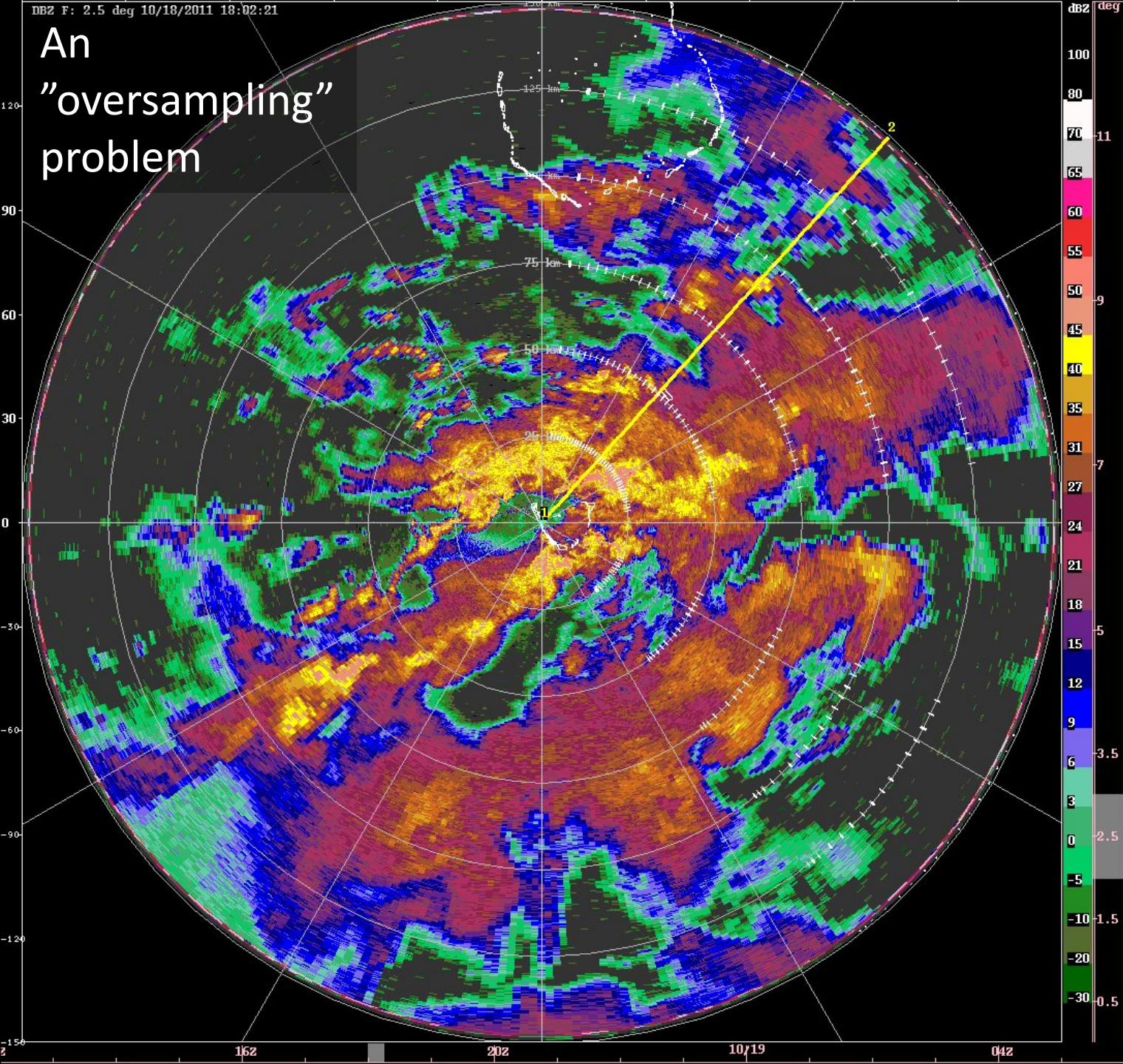
A first attempt at applying this to real radar data.

But...there's a problem.



DBZ F: 2.5 deg 10/18/2011 18:02:21

An  
“oversampling”  
problem



About 50,000  
data points but  
only  $O(10)$  to  
 $O(100)$  degrees  
of freedom?

# Conclusions

- Use 2D video disdrometer data as a climatological dataset of rain rate vs reflectivity (and other dual-polarization variables).
- To express uncertainty in rain rate, describe the spread in empirical (e.g. Z-R) relationships and surface reflectivity as a function of observed reflectivity above the surface. This generates a 2D PDF for surface rain rate per radar observation.
- Comparing radar-derived results to disdrometer observations, the method validates well, but...
- We have an oversampling problem and haven't accounted for spatio-temporal autocorrelation in radar volumes. Rain rate for two adjacent, similar echoes should not be independently estimated! If this is done, the calculated uncertainty goes to zero very quickly.