



Development of assimilation methods for polarimetric radars at storm scales



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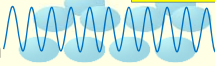
Why Dual-Polarized Radar?

horizontal polarization



Different

vertical polarization



Intensity: $Z_h, Z_v, Z_{DR} (Z_h - Z_v)$
Phase: $\varphi_{DP} (\varphi_h - \varphi_v), K_{DP}$ (spatial derivation of φ_{DP})
Polarimetric radars provide not only intensity of rain but also information on the shape of rain.

1

Two-moment microphysics

Drop size distribution $\phi(D) = N_0 D^\mu e^{-\Lambda D}$

N_0 Intercept, Λ Slope, μ Spectral shape parameters

* Three parameters in the drop size distribution are detected by the two-moment cloud microphysics scheme.

2

Two types of observation operators

Oblateness: $r = 0.9951 + 2.51 \times 10^{-2} D - 3.644 \times 10^{-2} D^2 + 5.303 \times 10^{-3} D^3 - 2.492 \times 10^{-4} D^4$
(Brandes et al. 2005)

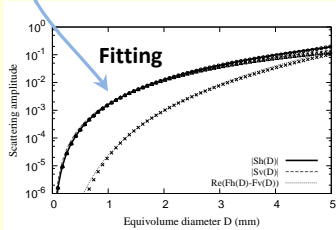
Rain water in the model is considered spheroid as the diameter increases.

1) Model variables -> pol. observation

Fitting (indirect scattering calculation) (Zhang et al. 2001)

$$|S_{h,v}(D)| = \alpha_{h,v} D^{\beta_{h,v}}$$
$$Z_{H,V} = \frac{4\lambda^4}{\pi^4 |K_w|^2} (\alpha_{h,v}^2 N_0 \Lambda^{-(2\beta_{h,v}+1)} \Gamma(2\beta_{h,v} + 1))$$
$$K_{DP} = \frac{180\lambda}{\pi} N_0 \alpha_k \Lambda^{-(\beta_k+1)} \Gamma(\beta_k + 1)$$
$$A_H = \alpha_H K_{DP}^{\beta_H}$$
$$A_{DP} = \alpha_d K_{DP}^{\beta_d}$$

α, β are fitting coefficients.



Scattering amplitude against rainwater size and the fitting functions. Marks represent backscattering amplitudes and the difference between horizontal and vertical polarizations calculated by the T-matrix, and lines are fitted with power-law functions.

2) Pol. Observation -> model variables

Evaluate three converters here.

$$Qr(Z_h, Z_{DR}) = c_1 Z_h^{a_1} 10^{0.1b_1} Z_{DR}$$

$$Qr(K_{DP}) = c_2 \left(\frac{K_{DP}}{f}\right)^{b_2}$$

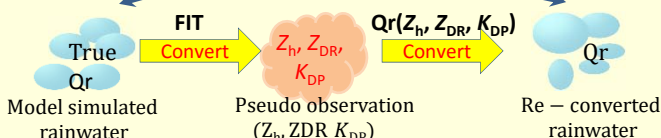
$$Qr(K_{DP}, Z_{DR}) = c_3 K_{DP}^{a_3} 10^{0.1b_3} Z_{DR}$$

(Bringi and Chandrasekar 2001)

$a_{1,3}, b_{1,2,3}$, and $c_{1,2,3}$ are the empirical constants.

How to evaluate these?

Comparison



3

Results

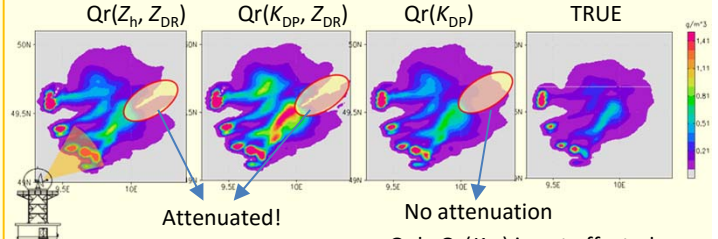
Averages and standard deviations in comparison with actual observations

	Z_h (dBZ)		Z_{DR} (dB)		K_{DP} ($^{\circ} \text{km}^{-1}$)	
	AVG	STD	AVG	STD	AVG	STD
OBS	18.99	11.10	0.93	1.05	1.31	1.54
FIT	19.23	10.37	0.55	0.41	0.25	0.18
FIT - OBS	-0.10	14.68	-0.30	0.99	-0.91	1.47

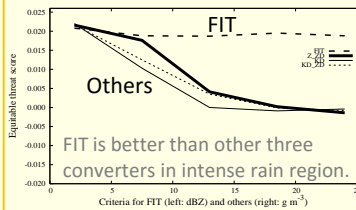
	Bias	STD
$Qr(Z_h, Z_{DR})$	0.033	0.07
$Qr(K_{DP})$	-0.006	0.03
$Qr(K_{DP}, Z_{DR})$	0.044	0.10

$Qr(K_{DP})$ is the most accurate.

Attenuation effect (Simulated Z_h)



Threat scores against actual observations



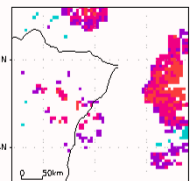
Conclusion
FIT or $Qr(K_{DP})$ are suitable for DA. Since values of the threat scores are quite low, we need to evaluate which one is better through DA.

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Data assimilation

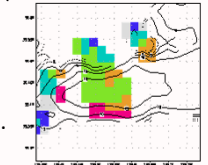
Assimilate Z_h, Z_{DR} and K_{DP} With FIT.

WRF Var at 1-km res.



Analysis increment (shade) and observation (contour) of Z_h .

NHM-4DVar at 2-km res. (Kawabata et al. 2014)



[1] Kawabata, T., T. Kuroda, H. Seko, and K. Saito, 2011: A cloud-resolving 4D-Var assimilation experiment for a local heavy rainfall event in the Tokyo metropolitan area. *Mon. Wea. Rev.*, **139**, 1911-1931.
[2] Kawabata, T., H. Iwai, H. Seko, Y. Shoji, K. Saito, S. Ishii, and K. Mizutani, 2014b: Impact of Doppler Wind Lidar Assimilation on a Heavy Rainfall Forecast at a Meso-gamma Scale. *Mon. Wea. Rev.*, **142**, 4484-4498
[3] Kawabata, T., H.-S. Bauer, T. Schwitalla, V. Wulfmeyer, and A. Adachi, 2016: Evaluation of Forward Operators for Polarimetric Radars Aiming for Data Assimilation. (under review).