



Preliminary Experimental Results of Polarimetric Radar Data Assimilation in the Case of Typhoon Soudelor (2015)

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Polarimetric variables

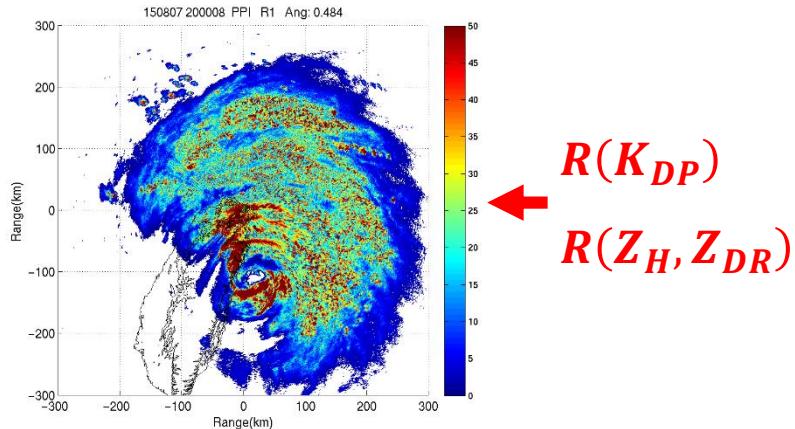
Z_H (dBZ)	Horizontal reflectivity	$Z_H = 10 \log Z_{hh}$
	Rain: 25~60 dBZ Hail: 50~65 dBZ	Snow: -10~35 dBZ Wet hail: 45~80 dBZ
Z_{DR} (dB)	Differential reflectivity	$Z_{DR} = 10 \log \frac{Z_{hh}}{Z_{vv}}$
	Rain: 0.5~4 dB Hail: -1~0.5 dB	Snow: -0.5~1 dB Wet hail: -1~6 dB
LDR (dB)	Linear depolarization ratio	$LDR = 10 \log \frac{Z_{hv}}{Z_{hh}}$ or $10 \log \frac{Z_{vh}}{Z_{vv}}$
	Mixed phase, random orientation, irregular shape → high LDR	

Polarimetric variables

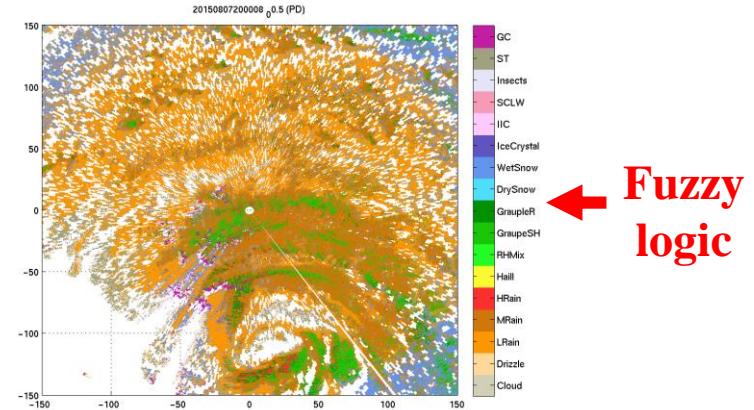
ρ_{HV}	Co-polar correlation coefficient	$\rho_{HV} = \frac{\langle f_a^* f_b \rangle}{\sqrt{\langle f_a ^2 \rangle \langle f_b ^2 \rangle}}$
	Rain: > 0.95 Hail: 0.9~0.95	Snow: > 0.95 Wet hail: > 0.9
ϕ_{DP} (°)	Differential phase	$\phi_{DP} = \phi_{hh} - \phi_{vv}$
	Increase as the radar beam penetrates oblate hydrometeors	
K_{DP} (°/km)	Specific differential phase	$K_{DP} = \frac{1}{2} \frac{d\phi_{DP}}{dr}$
	Rain: 0~20 °/km Hail: -1~2 °/km	Snow: -1~1 °/km Wet hail: 0~20 °/km

Capabilities of polarimetric radar data

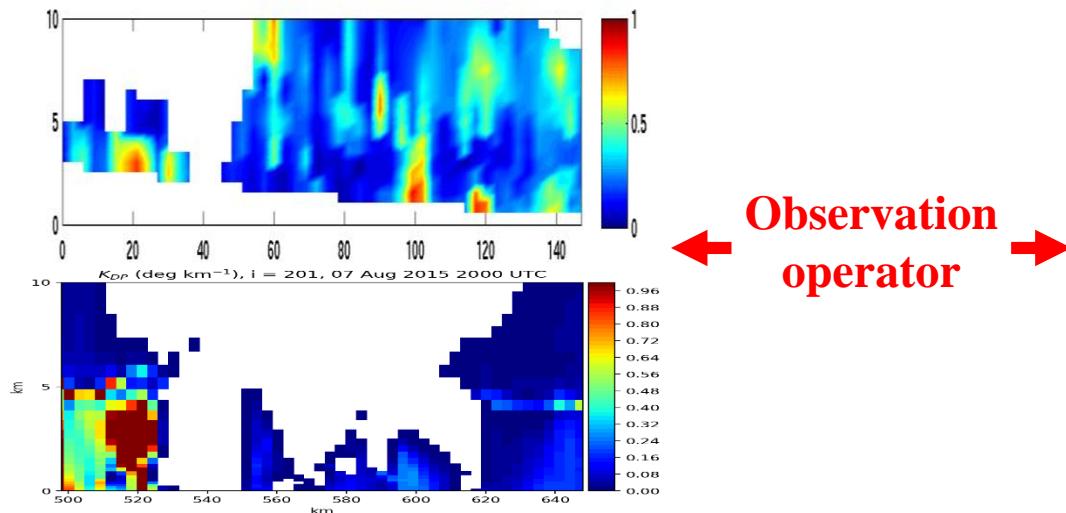
■ Quantitative precipitation estimation



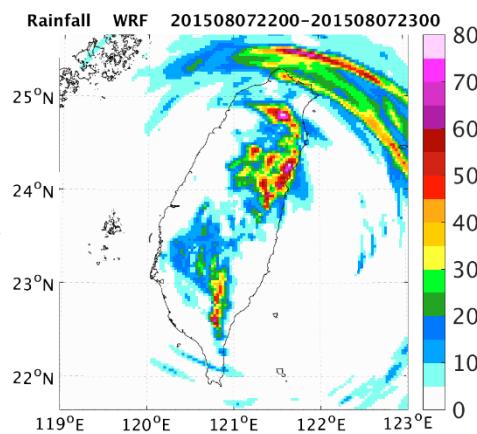
■ Hydrometeor classification



■ Verification of numerical simulation

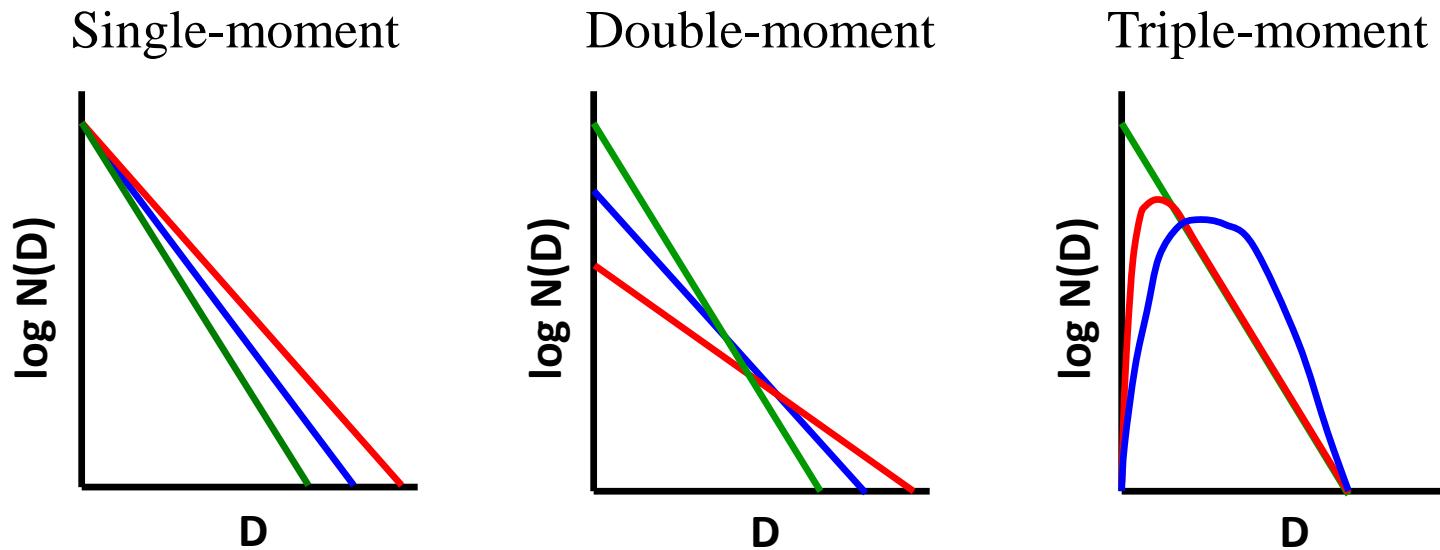


■ Data assimilation



Bulk microphysical schemes

- Gamma drop size distribution: $N(D) = N_0 D^\mu e^{-\Lambda D}$



Fixed parameters: $N_0 \wedge \mu$
 Variable parameters: Λ
 Prognostic variables: q

μ
 $N_0 \wedge \Lambda$
 $q \wedge N_t$

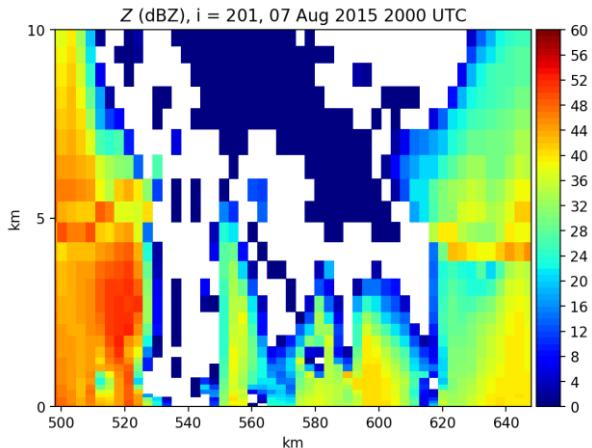
$N_0 \wedge \mu \wedge \Lambda$
 $q \wedge N_t \wedge Z_H$

Examples of bulk microphysical schemes

Scheme	Prognostic q	Prognostic N_t	Constant N_0 (m ⁻⁴)	μ	Constant ρ (kg m ⁻³)
Lin	$q_c \ q_r \ q_i$ $q_s \ q_g(h)$		$N_{0r}: 8e6 \ N_{0s}: 3e6$ $N_{0g}: 4e6$	$\mu_{rsgh}: 0$	$\rho_r: 1000 \ \rho_s: 100$ $\rho_g: 400$
WSM6			$N_{0r}: 8e6 \ N_{0s}: 2e6$ $N_{0g}: 4e6 \ (N_{0h}: 4e4)$		$\rho_r: 1000 \ \rho_s: 100$ $\rho_g: 500 \ \rho_h: 700$
Goddard			$N_{0r}: 8e6 \ N_{0s}: 1.6e7$ $N_{0g}: 4e6 \ (N_{0h}: 2e5)$		$\rho_r: 1000 \ \rho_s: 100$ $\rho_g: 400 \ \rho_h: 917$
WDM6		$N_n \ N_c \ N_r$	$N_{0s}: 2e6 \ N_{0g}: 4e6$ ($N_{0h}: 4e4$)	$\mu_r: 1$ $\mu_{sgh}: 0$	$\rho_r: 1000 \ \rho_s: 100$ $\rho_g: 500 \ \rho_h: 700$
Morrison		$N_r \ N_i \ N_s$ $N_{g(h)}$		$\mu_{rsgh}: 0$	$\rho_r: 997 \ \rho_s: 100$ $\rho_g: 400 \ \rho_h: 900$
Milbrandt -Yau	$q_c \ q_r \ q_i$ $q_s \ q_g \ q_h$	$N_c \ N_r \ N_i$ $N_s \ N_g \ N_h$			$\rho_r: 1000 \ \rho_s: 100$ $\rho_g: 400 \ \rho_h: 900$

Melting ice model

- Rain-snow (rain-graupel, rain-hail) mixture exists when rain and snow (graupel, hail) coexist.
- The fraction of rain or snow in the mixture form: $F = F_{max}[\min(q_s/q_r, q_r/q_s)]^{0.3}$
- The mixing ratio of the rain-snow mixture: $q_{rs} = F(q_r + q_s)$
- The water fraction within the rain-snow mixture: $f_w = q_r/(q_r + q_s)$
- The density of the rain-snow mixture: $\rho_{rs} = \rho_r f_w^2 + \rho_s (1 - f_w^2)$



Prognostic q	$F_{max,s}$	$F_{max,g}$	$F_{max,h}$
$q_r \ q_s \ q_g$	0.5	0.4	
$q_r \ q_s \ q_h$	0.5		0.3
$q_r \ q_s \ q_g \ q_h$	0.35	0.25	0.2

Mixed-phase hydrometeor species q_{rs} , q_{rg} and/or q_{rh} are created.

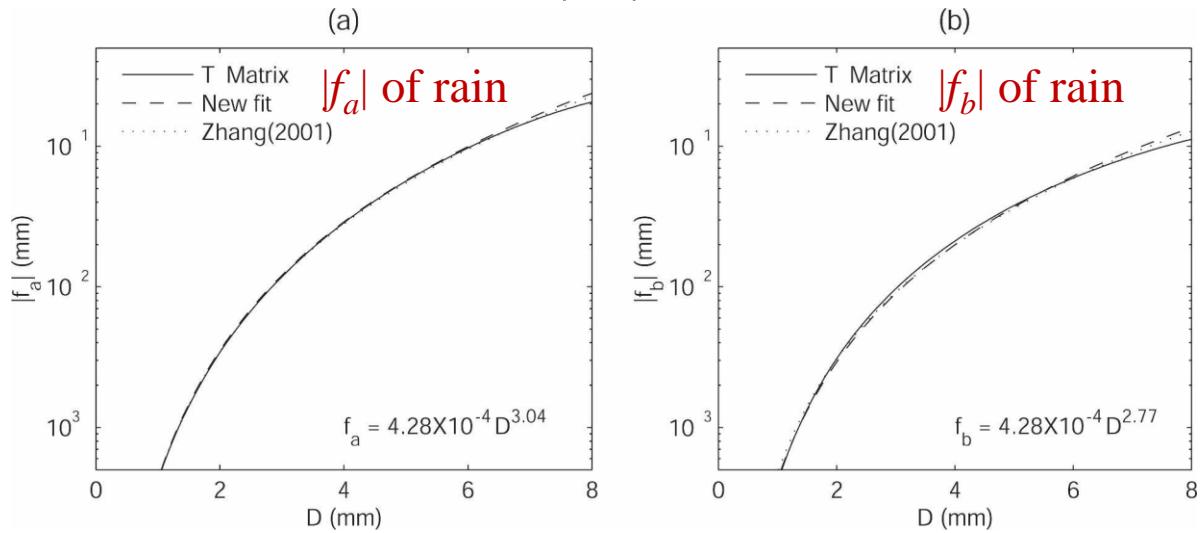
Backscattering amplitudes for H & V polarizations

Horizontal: $|f_a| = \alpha_a D^{\beta_a}$ (mm)

Vertical: $|f_b| = \alpha_b D^{\beta_b}$ (mm)

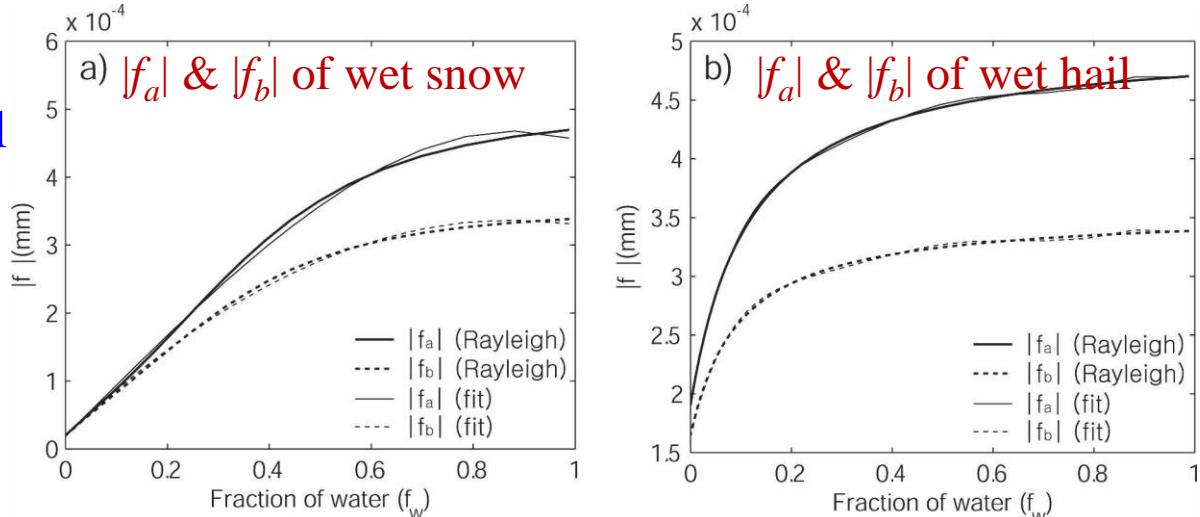
■ Rain:

1. T-matrix method
2. Function of:
 - Wavelength
 - Temperature
 - Drop size
 - Drop axis ratio



■ Dry/wet snow/graupel/hail

1. Rayleigh scattering approximation
2. Drop axis ratio = 0.75



Backscattering amplitudes for H & V polarizations

Horizontal: $|f_a| = \alpha_a D^{\beta_a}$ (mm)

Vertical: $|f_b| = \alpha_b D^{\beta_b}$ (mm)

	α_a	β_a	α_b	β_b
Rain	4.28×10^{-4}	3.04	4.28×10^{-4}	2.77
Snow	1.94×10^{-5}	3	1.91×10^{-5}	3
Wet snow	$(0.194 + 7.094 f_w + 2.135 f_w^2 - 5.225 f_w^3) \times 10^{-4}$	3	$(0.191 + 6.916 f_w - 2.841 f_w^2 - 1.160 f_w^3) \times 10^{-4}$	3
Graupel	8.1×10^{-5}	3	7.6×10^{-5}	3
Wet graupel	$\begin{pmatrix} 0.081 + 2.04 f_w - 7.39 f_w^2 + 18.14 f_w^3 \\ - 26.02 f_w^4 + 19.37 f_w^5 - 5.75 f_w^6 \end{pmatrix} \times 10^{-3}$	3	$\begin{pmatrix} 0.076 + 1.74 f_w - 7.52 f_w^2 + 20.22 f_w^3 \\ - 30.42 f_w^4 + 23.31 f_w^5 - 7.06 f_w^6 \end{pmatrix} \times 10^{-3}$	3
Hail	1.91×10^{-4}	3	1.65×10^{-4}	3
Wet hail	$\begin{pmatrix} 0.191 + 2.39 f_w - 12.57 f_w^2 + 38.71 f_w^3 \\ - 65.53 f_w^4 + 56.16 f_w^5 - 18.98 f_w^6 \end{pmatrix} \times 10^{-3}$	3	$\begin{pmatrix} 0.165 + 1.72 f_w - 9.92 f_w^2 + 32.15 f_w^3 \\ - 56.0 f_w^4 + 48.83 f_w^5 - 16.69 f_w^6 \end{pmatrix} \times 10^{-3}$	3

Consider the cants of hydrometeors

	Mean of canting angles	SD of canting angles
Rain	$\bar{\phi} = 0^\circ$	$\sigma = 0^\circ$
Snow	$\bar{\phi} = 0^\circ$	$\sigma = 20^\circ$
Wet snow	$\bar{\phi} = 0^\circ$	$\sigma = 20^\circ$
Graupel / hail	$\bar{\phi} = 0^\circ$	$\sigma = 60^\circ$
Wet graupel / wet hail	$\bar{\phi} = 0^\circ$	$\sigma = 60^\circ(1 - cf_w)$ $c = 4q_{rg,rh}$ (when $q_{rg,rh} < 0.2 \text{ g kg}^{-1}$) 0.8 (when $q_{rg,rh} \geq 0.2 \text{ g kg}^{-1}$)

$$A = \langle \cos^4 \phi \rangle = \frac{1}{8} (3 + 4 \cos 2\bar{\phi} e^{-2\sigma^2} + \cos 4\bar{\phi} e^{-8\sigma^2})$$

$$B = \langle \sin^4 \phi \rangle = \frac{1}{8} (3 - 4 \cos 2\bar{\phi} e^{-2\sigma^2} + \cos 4\bar{\phi} e^{-8\sigma^2})$$

$$C = \langle \sin^2 \phi \cos^2 \phi \rangle = \frac{1}{8} (1 - \cos 4\bar{\phi} e^{-8\sigma^2})$$

$$C_k = \langle \cos 2\phi \rangle = \cos 2\bar{\phi} e^{-2\sigma^2}$$

Diagnose gamma distribution parameters

- Gamma function: $\int_0^\infty t^b e^{-at} dt = \frac{\Gamma(b+1)}{a^{b+1}}$
- Total number concentration: $N_t = \int_0^\infty N_0 D^\mu e^{-\Lambda D} dD = N_0 \frac{\Gamma(\mu+1)}{\Lambda^{\mu+1}}$

- Mixing ratio: $q = \frac{\rho \int_0^\infty \frac{4}{3} \pi \left(\frac{D}{2}\right)^3 N_0 D^\mu e^{-\Lambda D} dD}{\rho_a}$

$$\frac{\rho_a q}{\rho} = \frac{\pi N_0}{6} \frac{\Gamma(\mu+4)}{\Lambda^{\mu+4}} = \frac{\pi N_t \Gamma(\mu+4)}{6 \Gamma(\mu+1) \Lambda^3}$$

Single-moment scheme: $\Lambda = \left[\frac{\pi \rho N_0 \Gamma(\mu+4)}{6 \rho_a q} \right]^{\frac{1}{\mu+4}}$

Double-moment scheme: $\Lambda = \left[\frac{\pi \rho N_t \Gamma(\mu+4)}{6 \rho_a q \Gamma(\mu+1)} \right]^{\frac{1}{3}}$

Calculate Z_H , Z_{DR} and K_{DP}

$$Z_{hh} = \frac{4\lambda^4}{\pi^4 |K_w|^2} \int_0^\infty (A|f_a|^2 + B|f_b|^2 + 2C|f_a||f_b|) N_0 D^\mu e^{-\Lambda D} dD$$

$$= \frac{4\lambda^4 N_0 \Gamma(2\beta_a + \mu + 1)}{\pi^4 |K_w|^2} \left[\frac{6\rho_a q}{\pi N_0 \Gamma(\mu + 4) \rho} \right]^{\frac{2\beta_a + \mu + 1}{\mu + 4}} (A\alpha_a^2 + B\alpha_b^2 + 2C\alpha_a\alpha_b)$$

$$Z_{vv} = \frac{4\lambda^4}{\pi^4 |K_w|^2} \int_0^\infty (B|f_a|^2 + A|f_b|^2 + 2C|f_a||f_b|) N_0 D^\mu e^{-\Lambda D} dD$$

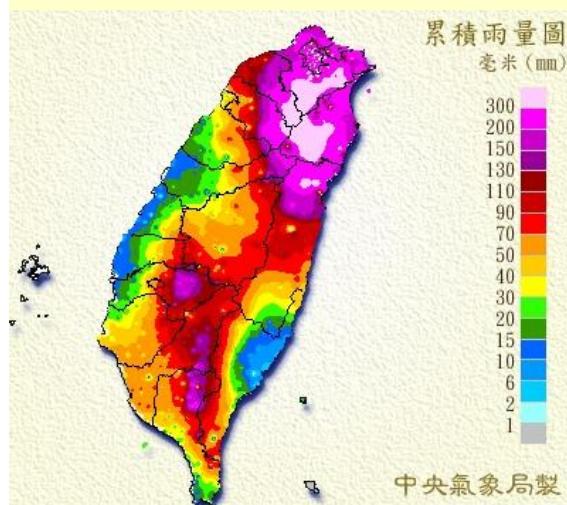
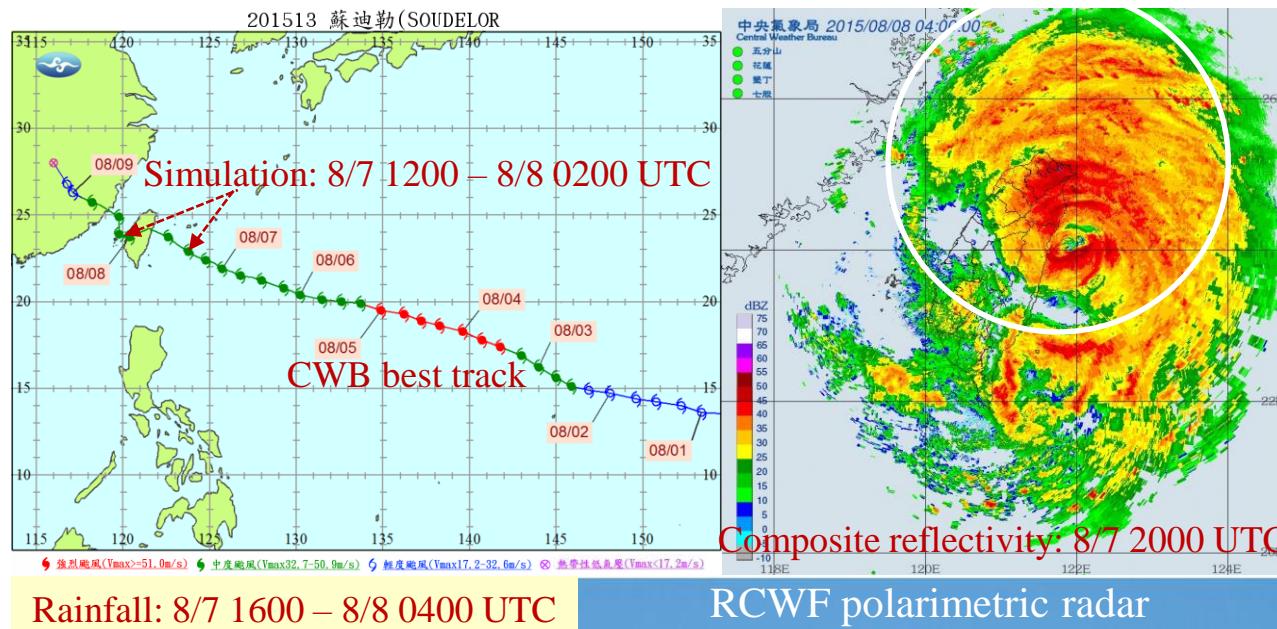
$$= \frac{4\lambda^4 N_0 \Gamma(2\beta_b + \mu + 1)}{\pi^4 |K_w|^2} \left[\frac{6\rho_{air} q}{\pi N_0 \Gamma(\mu + 4) \rho} \right]^{\frac{2\beta_b + \mu + 1}{\mu + 4}} (B\alpha_a^2 + A\alpha_b^2 + 2C\alpha_a\alpha_b)$$

$$K_{dp} = \frac{180\lambda}{\pi} \int_0^\infty C_k \operatorname{Re}(f_a - f_b) N_0 D^\mu e^{-\Lambda D} dD$$

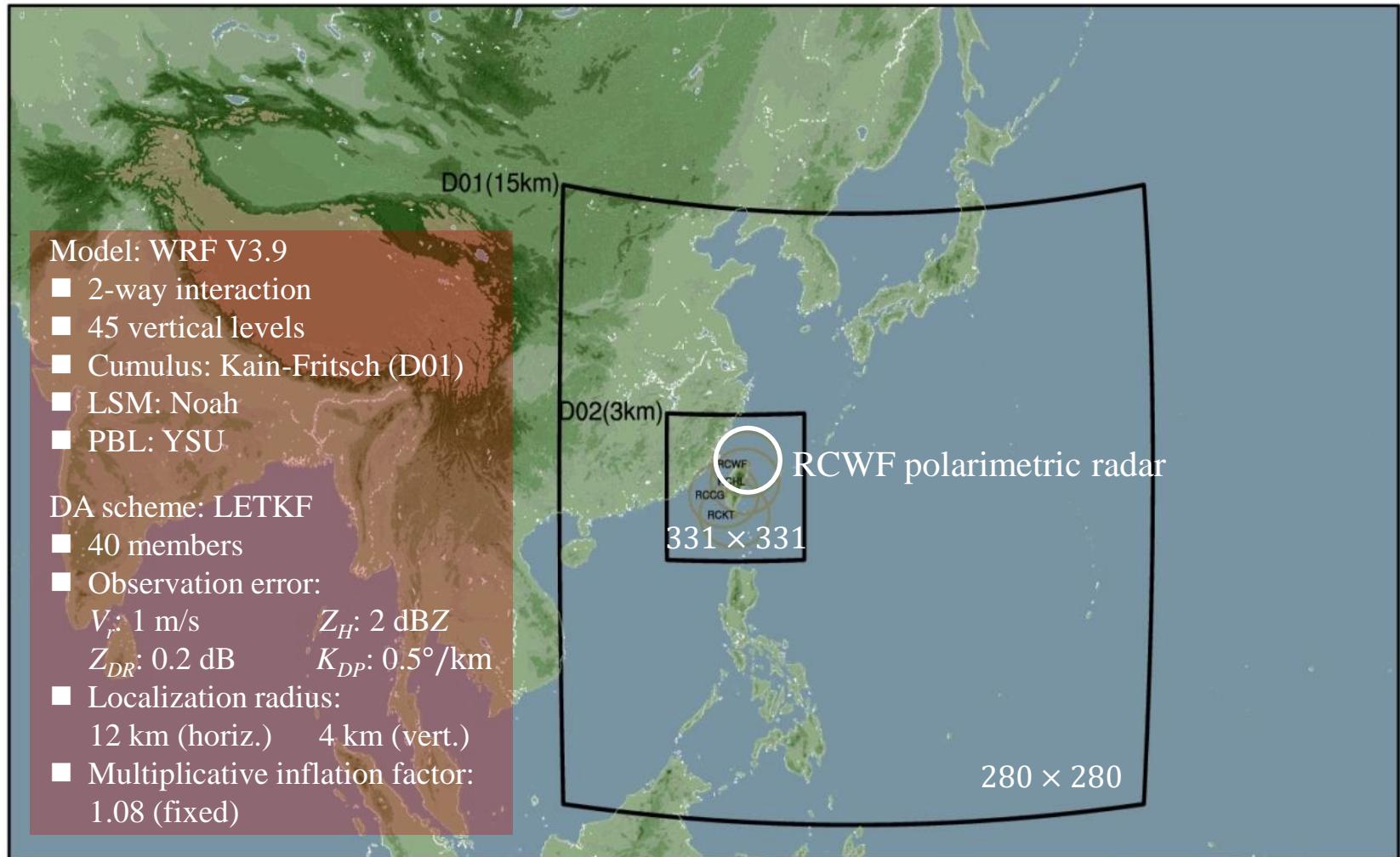
$$= \frac{180\lambda N_0 C_k}{\pi} \left\{ \alpha_a \Gamma(\beta_a + \mu + 1) \left[\frac{6\rho_a q}{\pi N_0 \Gamma(\mu + 4) \rho} \right]^{\frac{\beta_a + \mu + 1}{\mu + 4}} - \alpha_b \Gamma(\beta_b + \mu + 1) \left[\frac{6\rho_a q}{\pi N_0 \Gamma(\mu + 4) \rho} \right]^{\frac{\beta_b + \mu + 1}{\mu + 4}} \right\}$$

$$Z_H = 10 \log \left(\sum_x Z_{hh,x} \right) \quad Z_{DR} = 10 \log \left(\sum_x Z_{hh,x} / \sum_x Z_{vv,x} \right) \quad K_{DP} = \sum_x K_{dp,x}$$

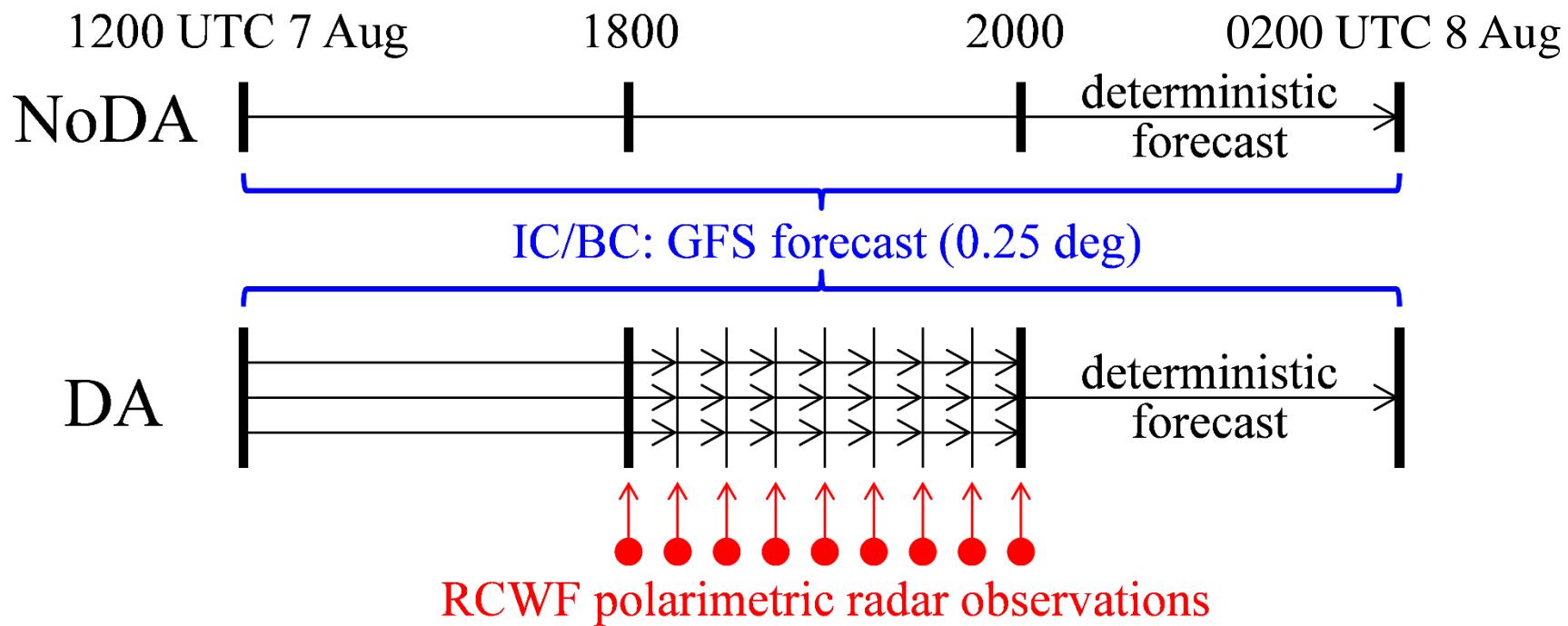
Typhoon Soudelor (2015)



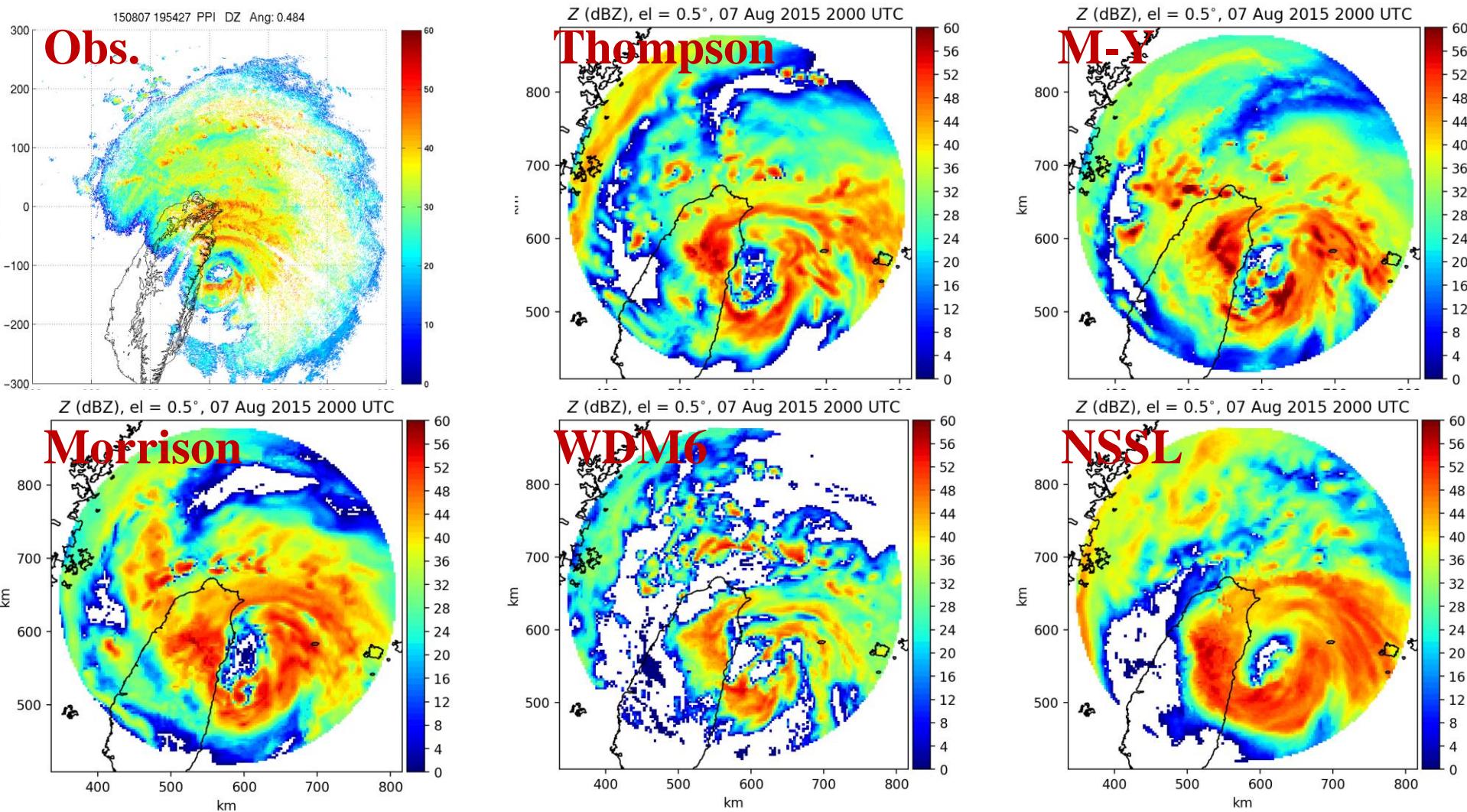
Simulation domains



Simulation period

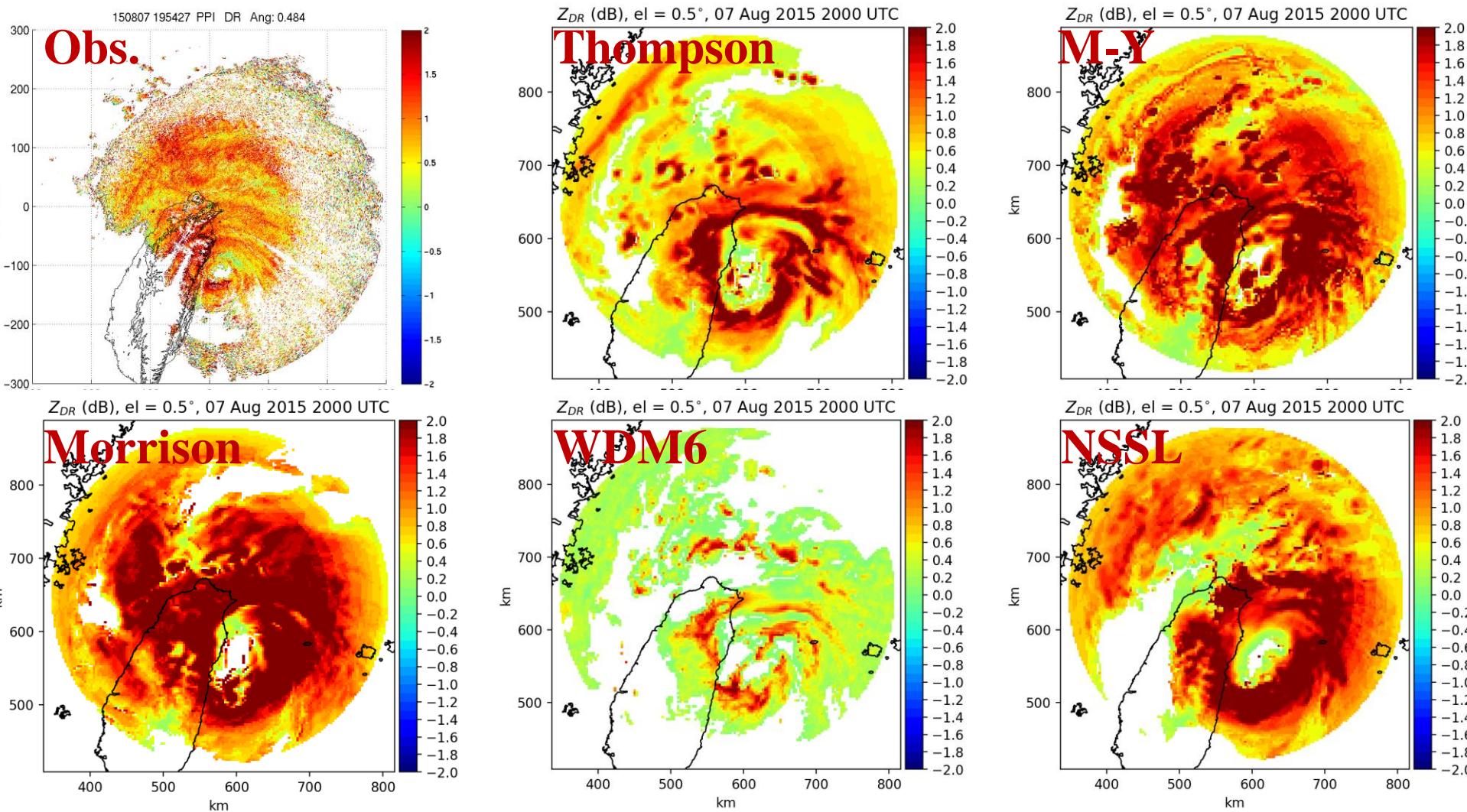


NoDA with various microphysics: Z_H



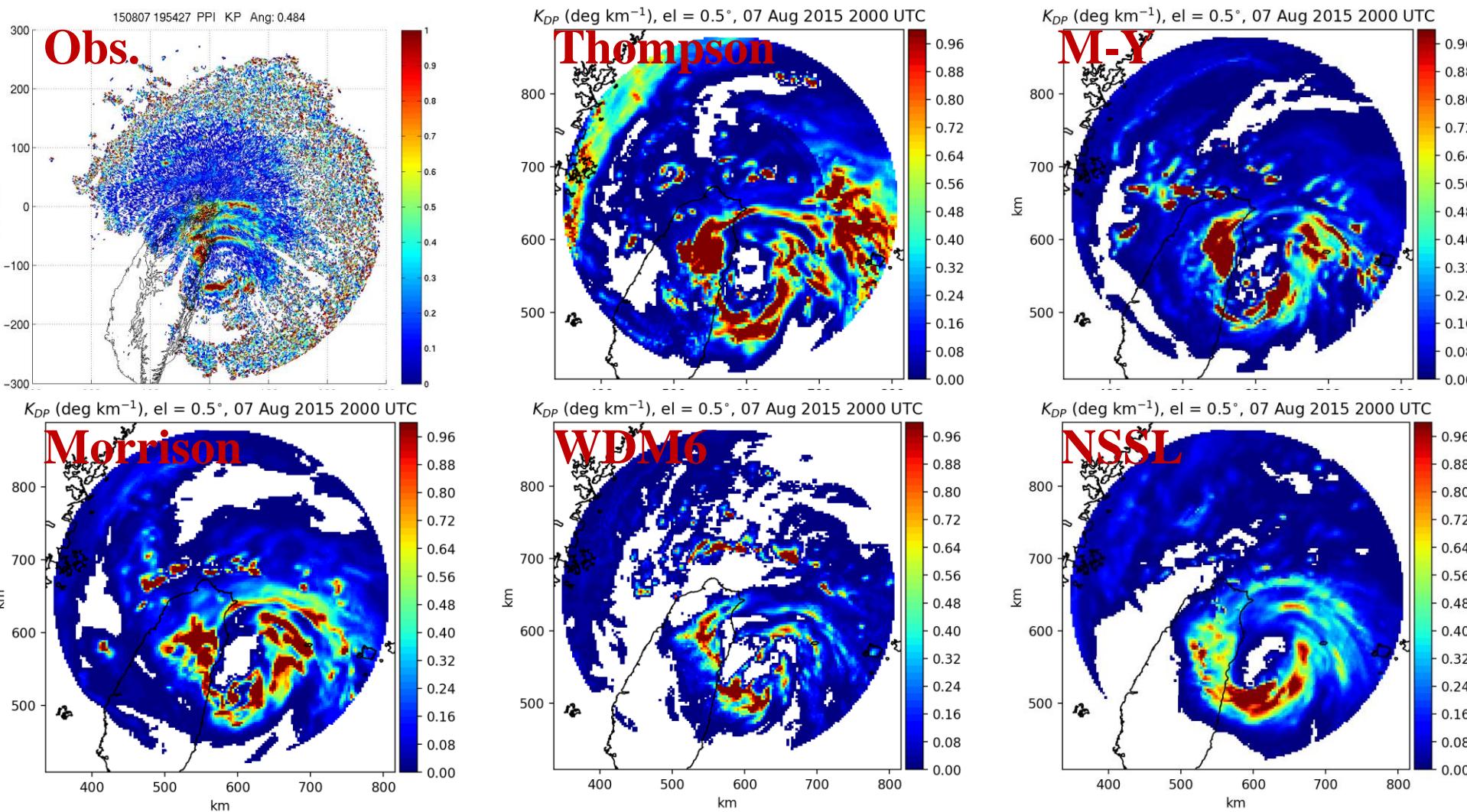
WDM6 gives the least over-forecasting of Z_H with a clear structure of spiral rainbands.

NoDA with various microphysics: Z_{DR}



Similar result with respect to Z_{DR} .

NoDA with various microphysics: K_{DP}

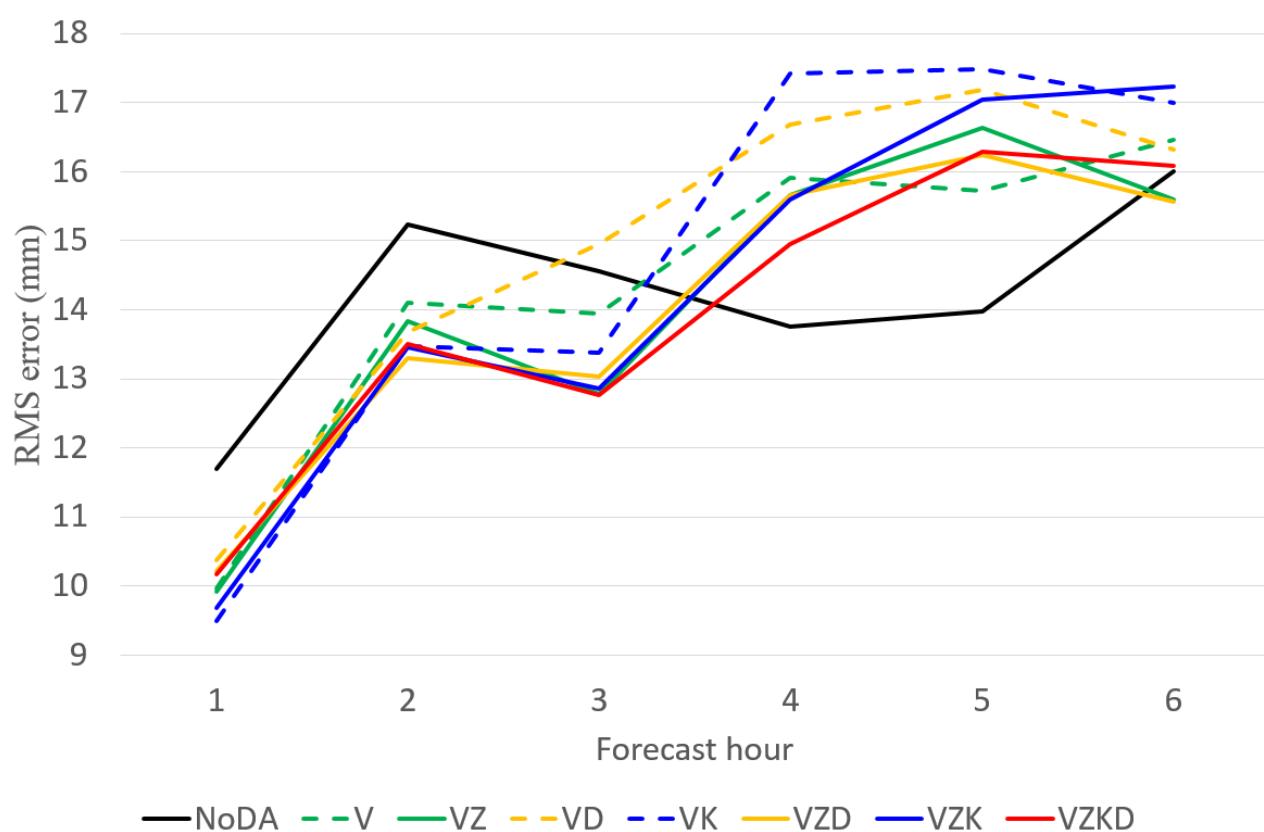


Similar result with respect to K_{DP} . **WDM6** is selected for DA experiments.

DA experiments

	Assimilated radar variables			
NoDA	—			
V	V_r			
VZ	$V_r \ Z_H$			
VD	$V_r \ Z_{DR}$			
VK	$V_r \ K_{DP}$			
VZD	$V_r \ Z_H \ Z_{DR}$			
VZK	$V_r \ Z_H \ K_{DP}$			
VZKD	$V_r \ Z_H \ Z_{DR} \ K_{DP}$			

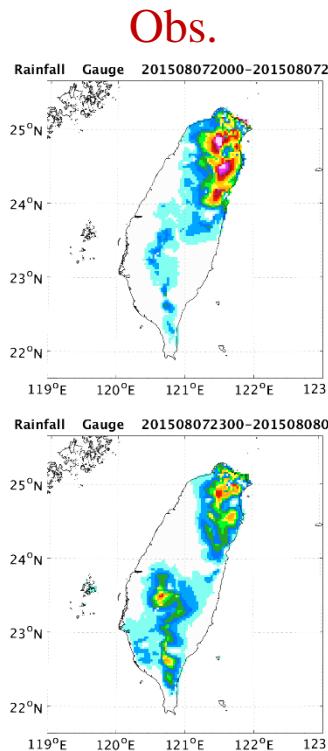
Quantitative precipitation forecasting (QPF)



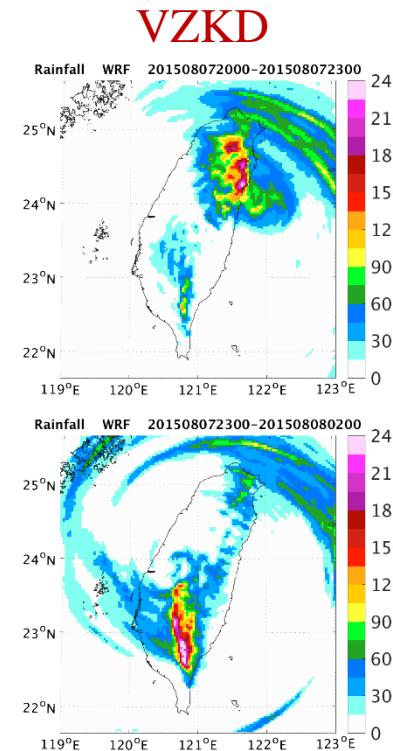
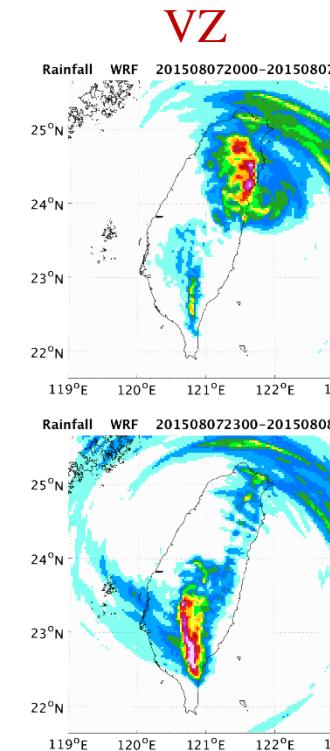
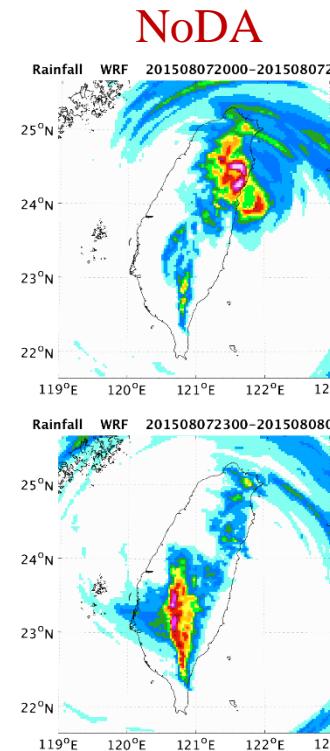
- Generally speaking, radar DA greatly improves QPF in the first 3 hours, but deteriorates that in the following 3 hours.
- Additional assimilation of Z_H is beneficial for any combination of the rest variables.
- K_{DP} plays a more important role than Z_{DR} .
- Assimilating all variables (VZKD) is a good choice.

Quantitative precipitation forecasting (QPF)

8/7 2000 –
8/7 2300 UTC



8/7 2300 –
8/8 0200 UTC



- In the first 3 hours:
Radar DA leads to more accurate intensity and distribution of QPF in northern Taiwan.
- In the following 3 hours:
The deterioration results from the over-forecasting in southern Taiwan using the analysis ensemble mean beyond RCWF's coverage.

Summary

- An observation operator for Z_H , Z_{DR} and K_{DP} is incorporated into a WRF-LETKF system and tested with a typhoon case.
- WDM6 is found to give the least over-forecasting with a clear structure of spiral rainbands in this typhoon case.
- Preliminary results:
 1. With the limited coverage of RCWF, radar DA greatly improves QPF in northern Taiwan for 3 hours.
 2. However, the analysis ensemble mean beyond RCWF's coverage results in worse QPF than NoDA in southern Taiwan.
 3. It is a good choice to assimilate all variables, among which Z_H and K_{DP} are more important than Z_{DR} .
- Sensitivity tests in future prospects:
 1. Assimilation of other observations, including radars in southern Taiwan
 2. Assimilation strategies
 3. Source of the background ensemble

Thank you very much!