

利用局地系集卡爾曼濾波系統同化 熱動力變數與雷達資料：敏感度實驗測試



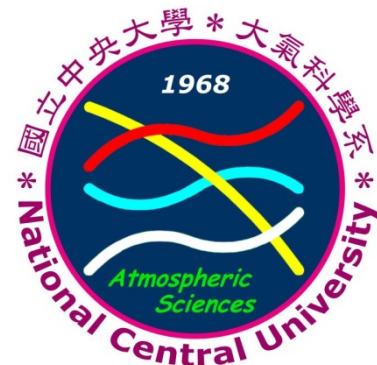
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Acknowledgement: 廖宇慶教授、楊舒芝教授

**2017 Conference on
Weather Analysis and Forecasting
Sep. 12th, 2017**



Outline

1. Motivation

2. Examination of background errors (covariance)
at convective scale

3. Results of OSSEs

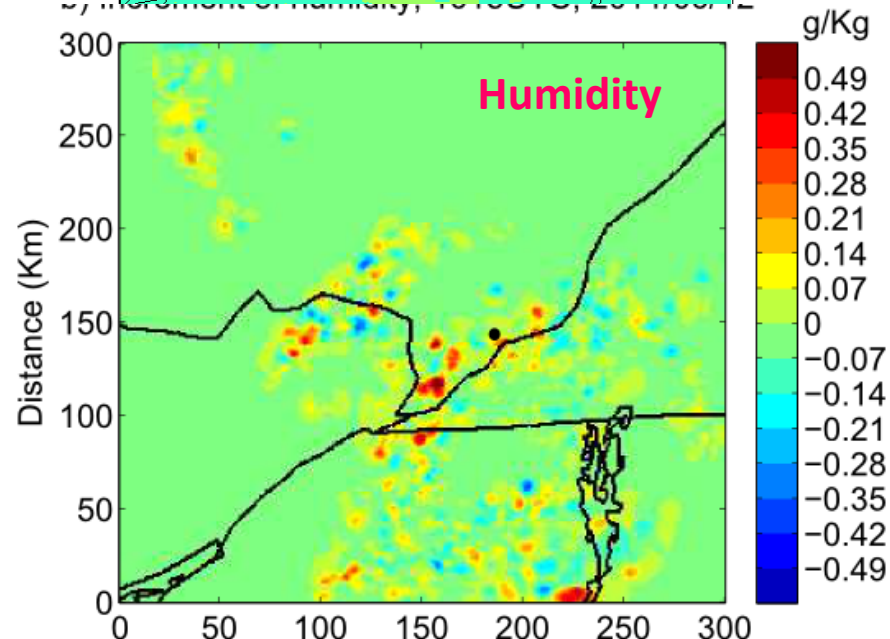
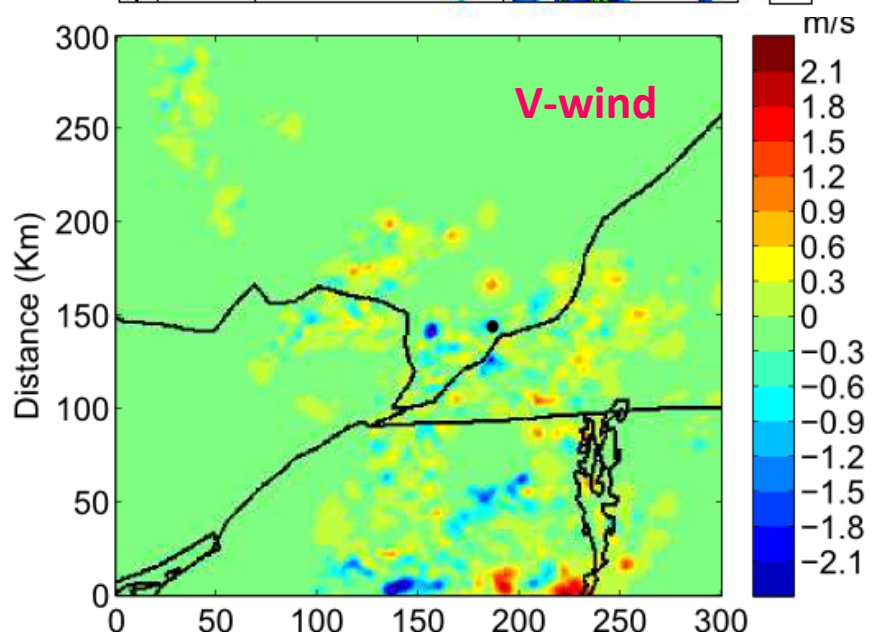
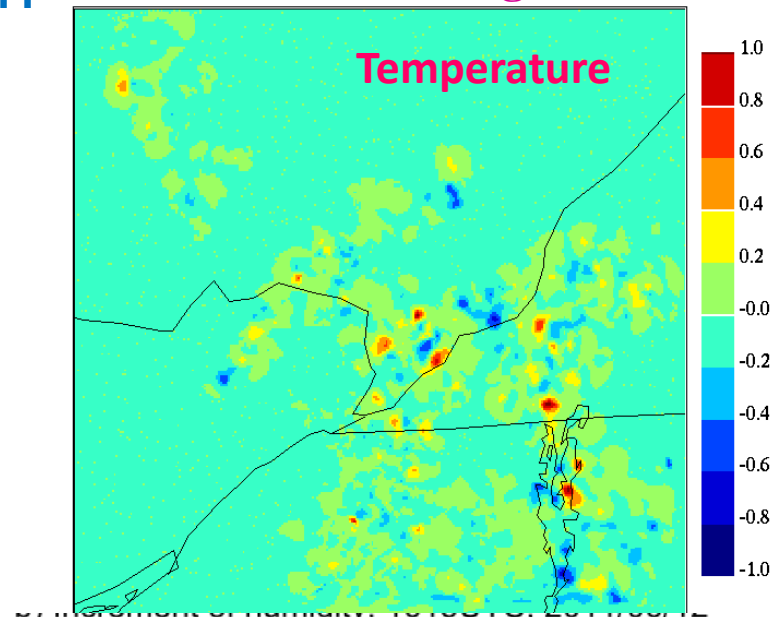
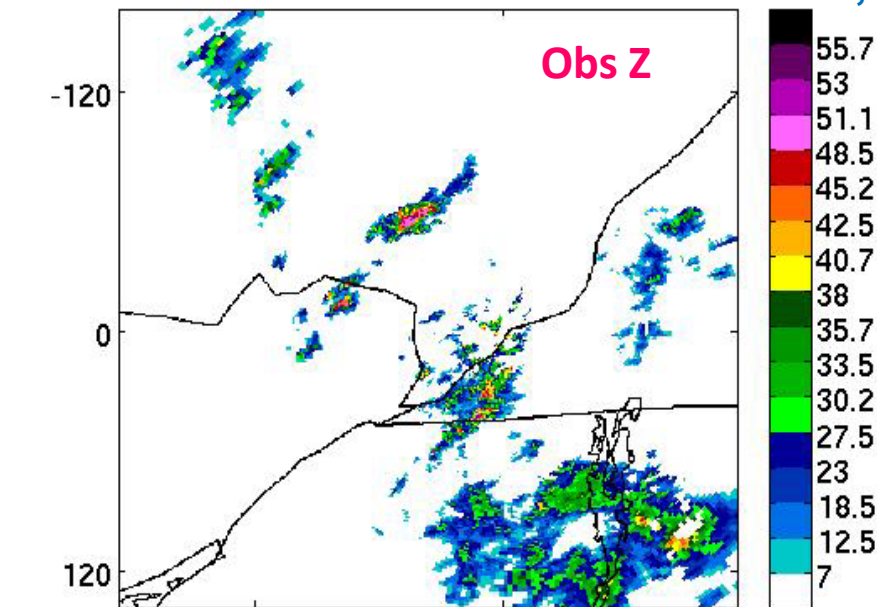
(a) Analysis (B) Short-term forecast

4. Summary

By using Canadian HREnKF and assimilating Vr from **single Doppler radar**
It is able to propagate information to other control variables.

Localized convection on June 12, 2011

(Chang et al. 2014 M.W.R.)

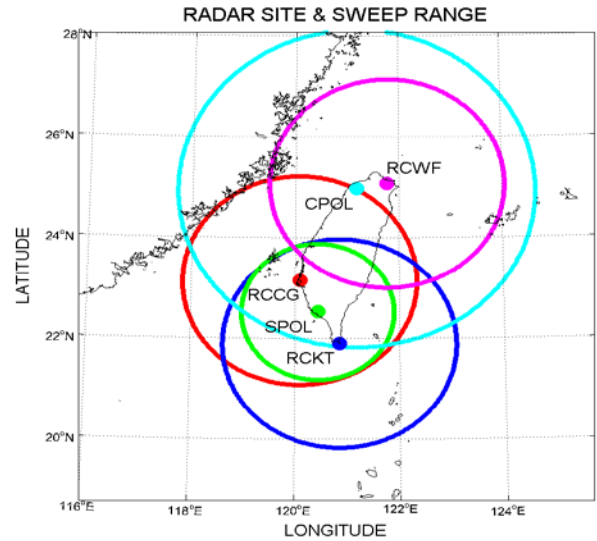


➤ Assimilating Vr and Z of radar network in Taiwan [3-km] (Prof. Yu-Cheing Liou and Shu-Chih Yang)

WRF-LETKF Radar Assimilation System
WLRAS (Tsai et al 2014)

Assimilating both radial wind and reflectivity
Case study on 2008/06/14-15 SoWMEX
IOP8 (MCS)

QPF(5-hr accumulated precipitation)



◆ observation

◆ No DA

◆ 1-hr DA

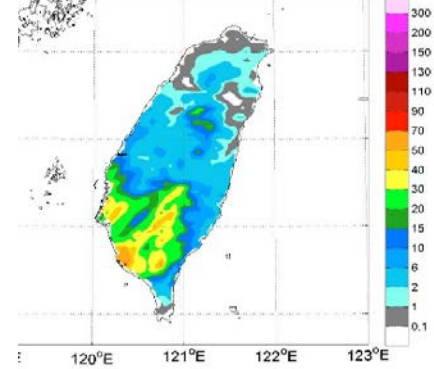
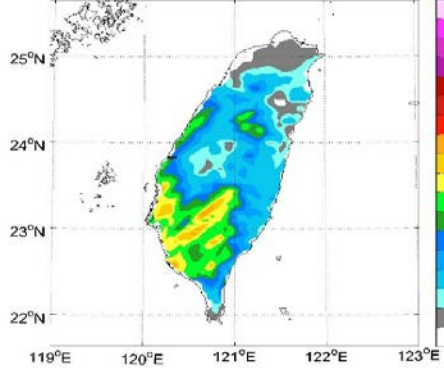
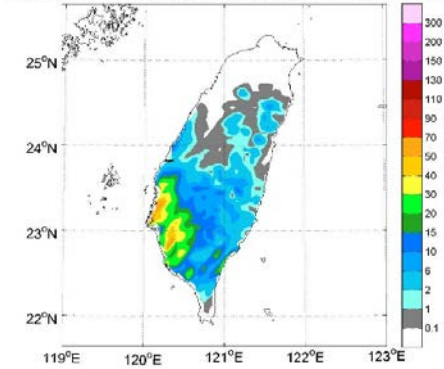
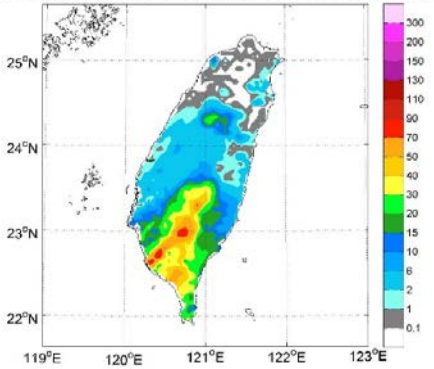
◆ 2-hr DA

QPESUMS Accumulated precipitation-5hr 0614-17:00UTC

noda3km025 Accumulated precipitation-5hr 0614 17:00

inn5 Accumulated precipitation-5hr 0614 17:00

Accumulated precipitation-5hr 0614 17:00

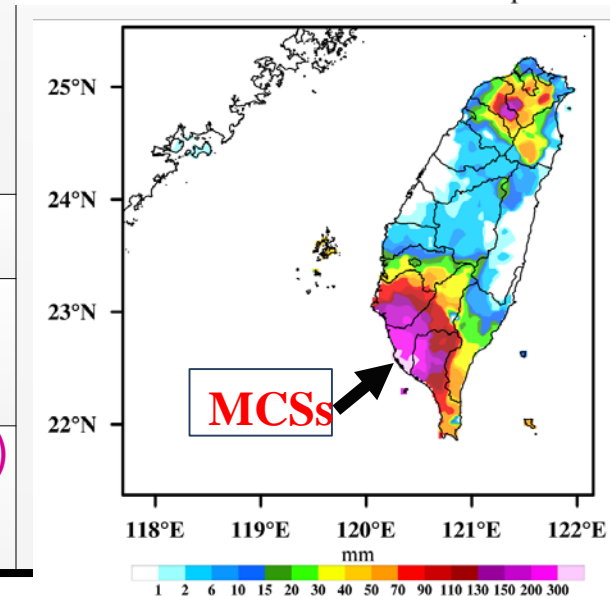
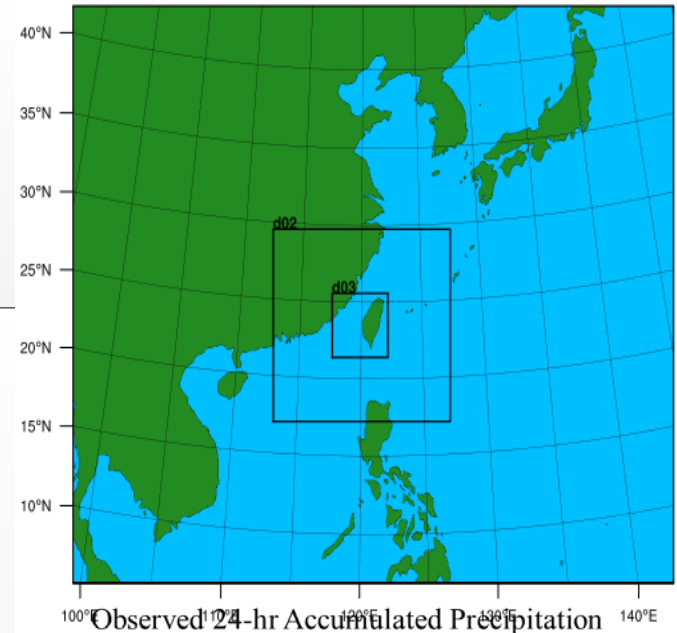


[Motivation]

Q: By assimilating **radial wind and reflectivity** of radar network ,
why we still need a **long assimilation window** to improve QPF?

2. Examination of background (forecast) error

Model Version	WRF 3.2.1
Domains	D01(27-km) 180×150
	D02(9-km) 160×150
	D03(3-km) 150×150
	27 layers, top at 50 hPa
Physical Parameterizations	Longwave Radiation: RRTM scheme
	Shortwave Radiation: Dudhia scheme
	PBL: YSU scheme
	Cumulus: G-D ensemble scheme
	Microphysics: GCE scheme
Initial Time	1200 UTC 15 th Jun., 2008
Initial Condition	WRF-LETKF analysis ensemble (conventional data, AMV, GPS-RO)
Ensemble Members	(Yang et al. 2014 M.W.R.) 36→72



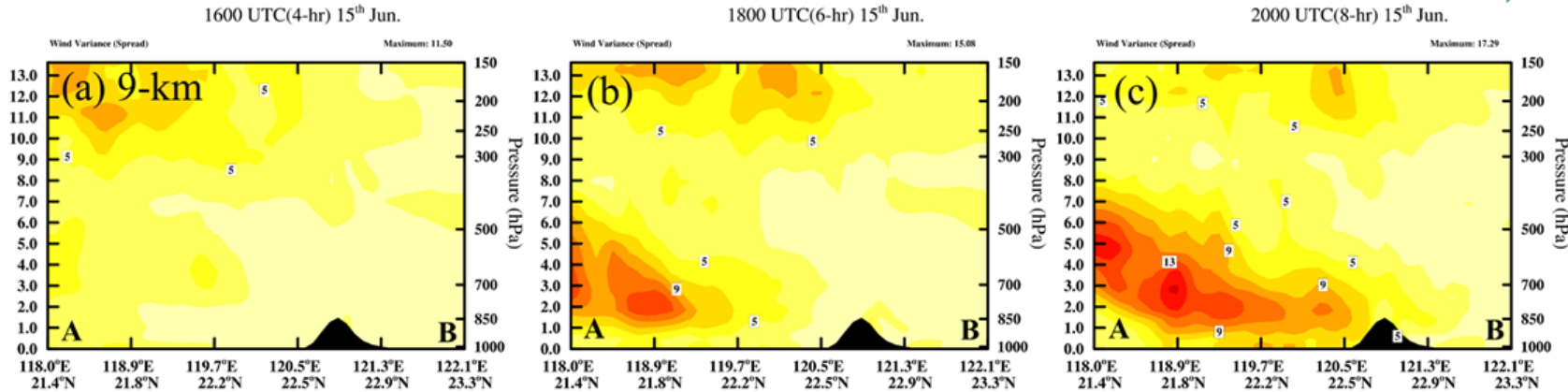
Variance in time

1600

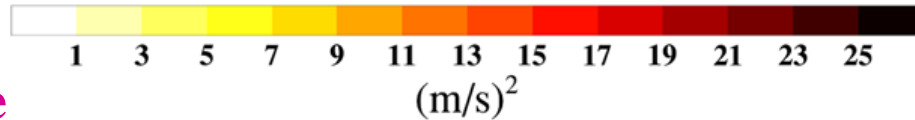
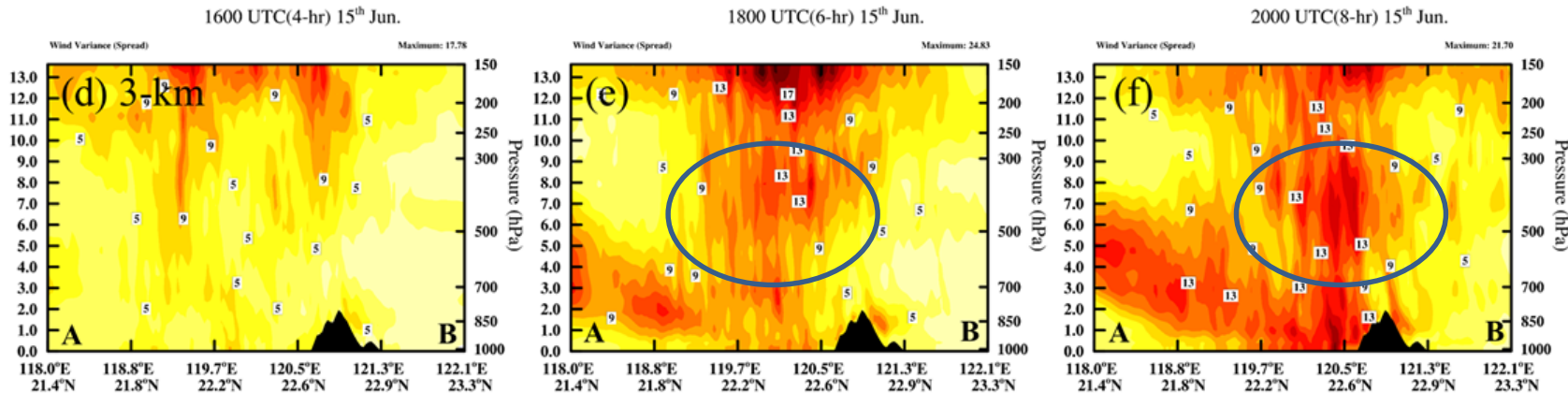
1800

2000 UTC

9-km



3-km



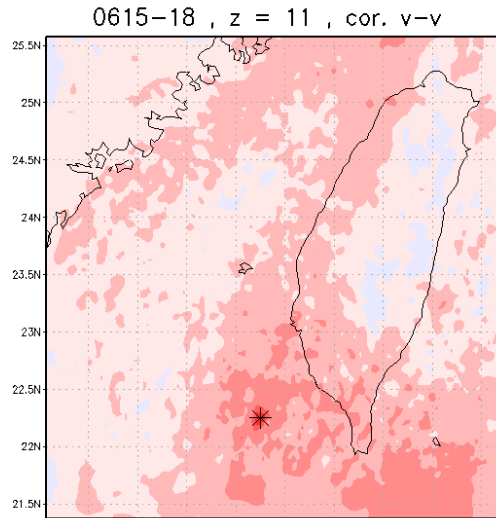
Capture the smaller-scale weather features

➤ multi-scale interactions at 3-km resolution

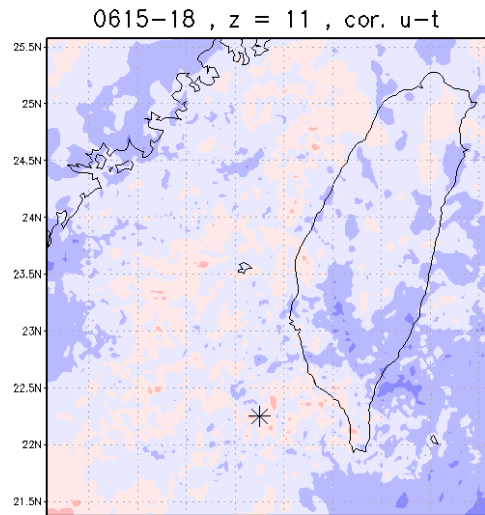
2. Examination of background error at convective scale

◆ Convective scale error correlation (3-km resolution, ~700mb)

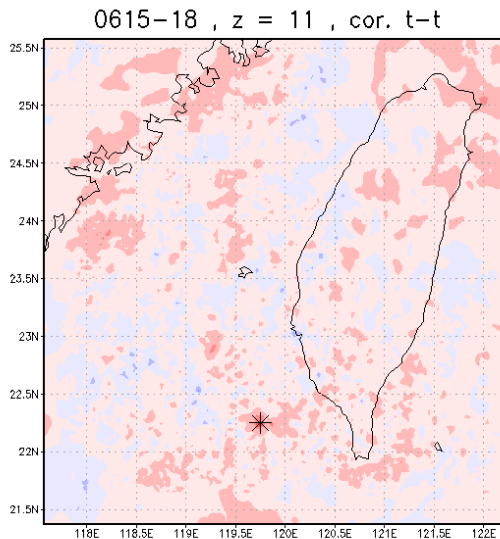
V-V



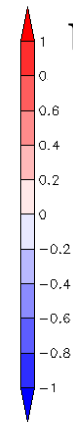
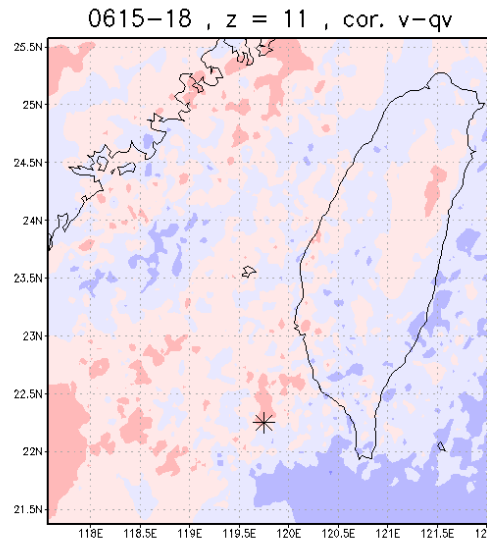
V-qv



T-T

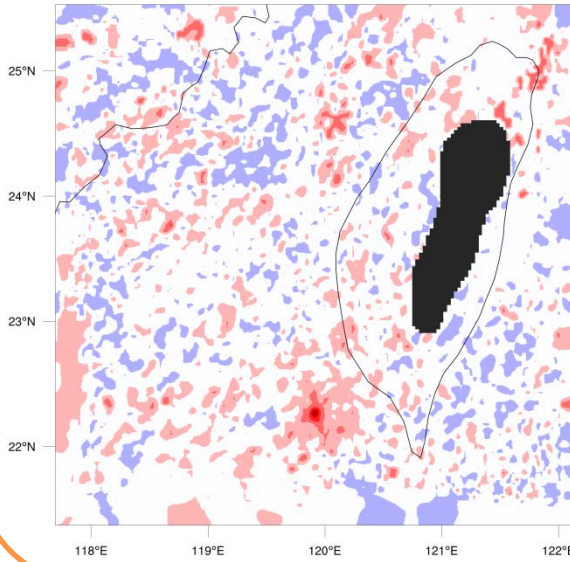


U-T

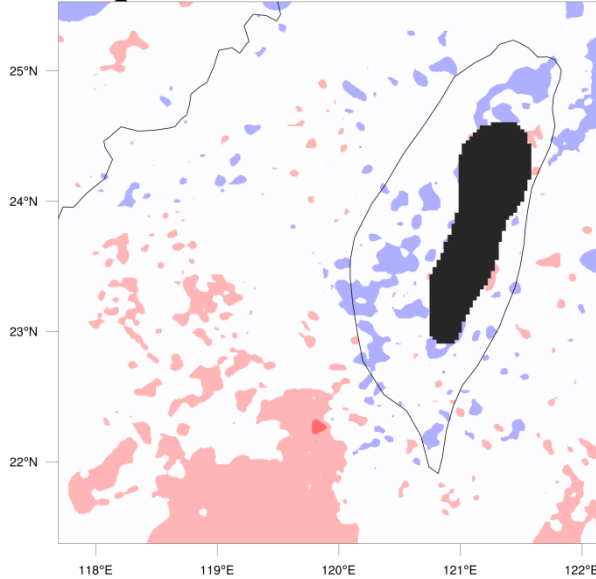


- Compared to Temperature, Wind has longer correlation length.
- Less strong / very local correlations between state variables at convective scale

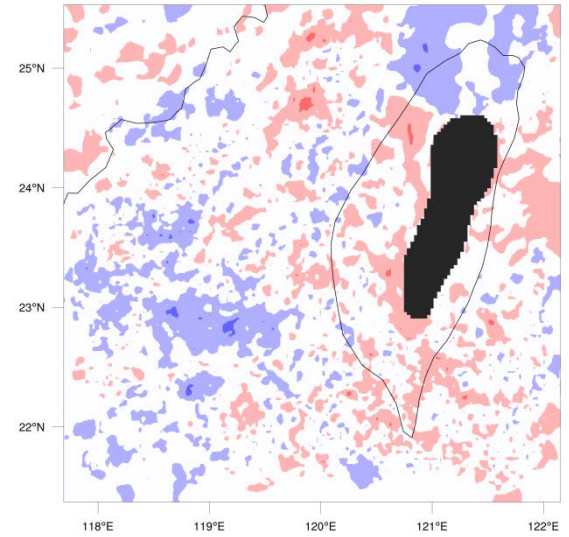
qr-qr auto-correlation



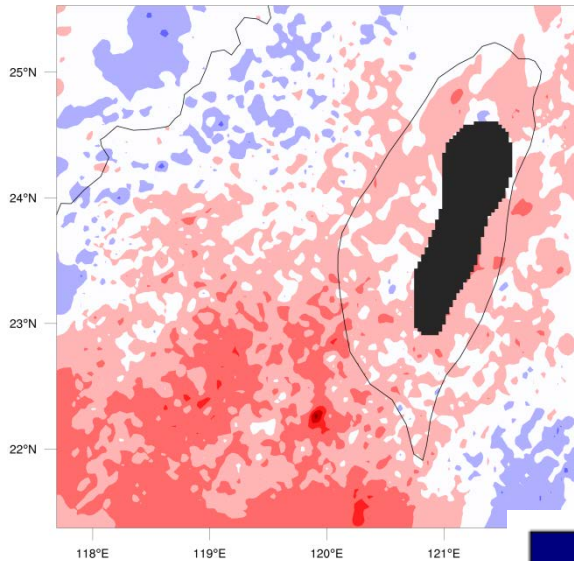
qr-U cross-correlation



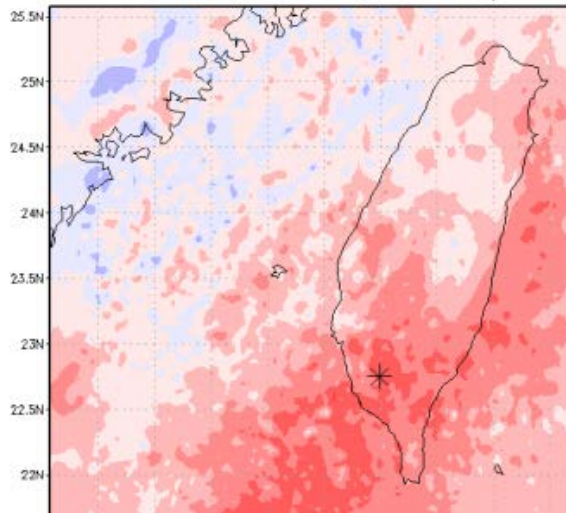
qr-T cross-correlation



qv-qv auto correlation



T-qv cross-correlation



$$Z = 43.1 + 17.5 \log(\rho q_r)$$

$$V_r = u \frac{x}{r} + v \frac{y}{r} + (w + V_t) \frac{z}{r}$$



Explain why we need many cycles in DA even though with Radar network observations



-0.9 -0.7 -0.5 -0.3 -0.1 0.1 0.3 0.5 0.7 0.9

Error Correlation

(cross)

Corr(U, V)

1600

Corr(U, V) 1800

2000 US

Scale analysis in the

previous study

(Yeh and Chen 2002)

$h_m \sim 2 \text{ km}, U \sim 10 \text{ m s}^{-1}, N \sim 10^{-2}$

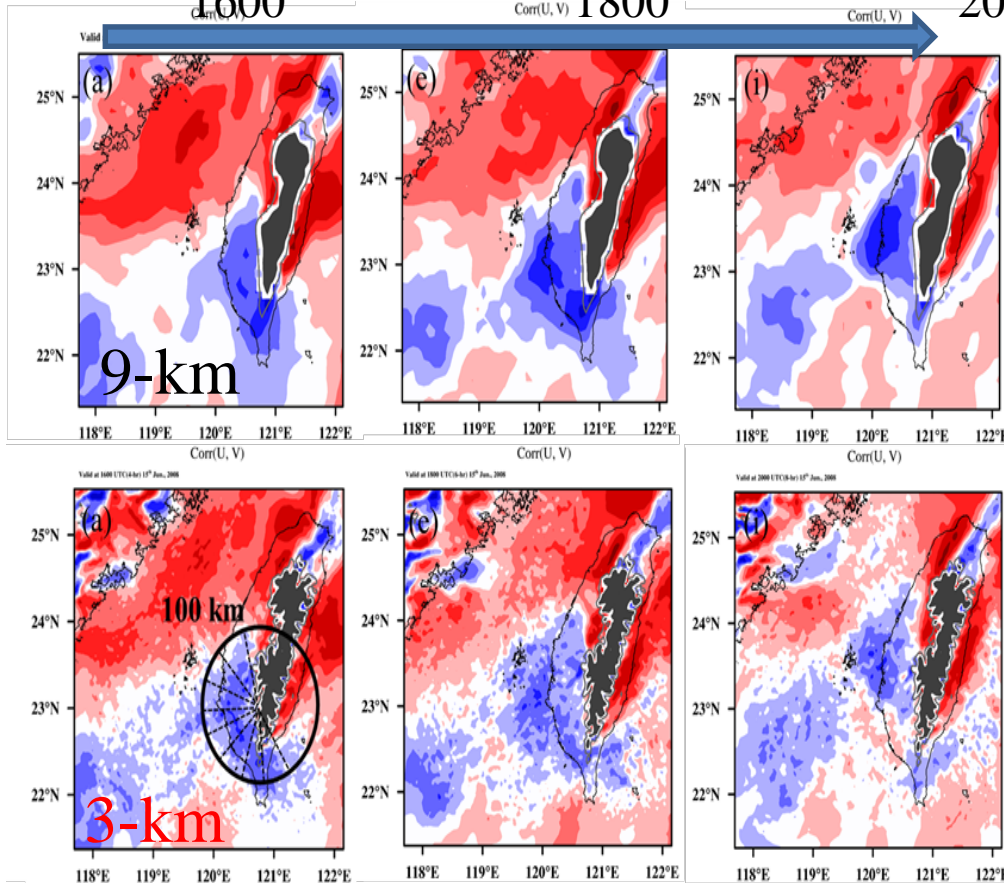
$f \sim 10^{-4} \text{ s}^{-1}$

Froude Number:

$$Fr = \frac{U}{h_m N} = 0.5 < 1$$

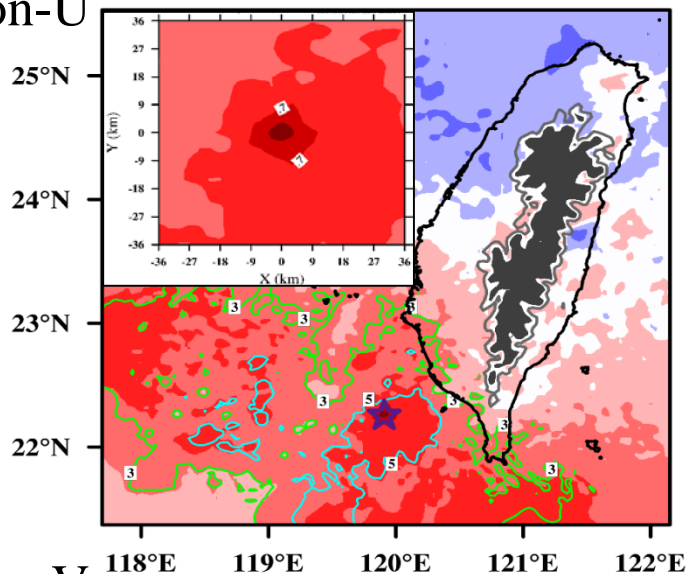
Rossby Radius:

$$L_R = \frac{N h_m}{f} Fr = 100 \text{ km}$$

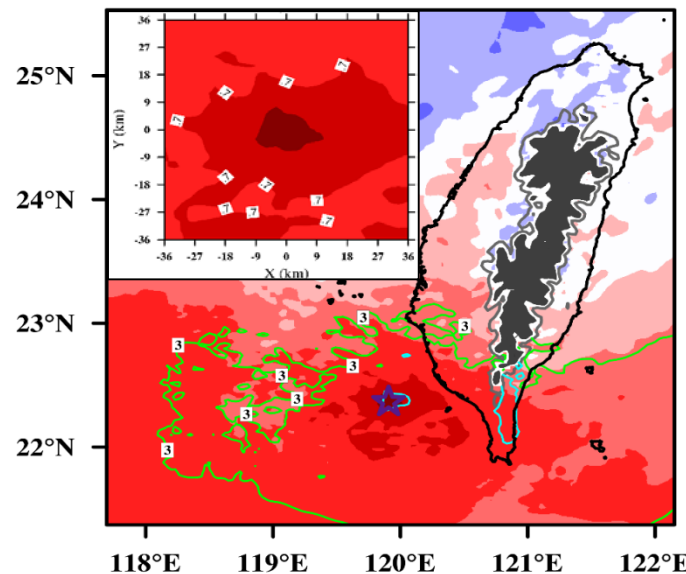


Error covariance at convective scale (I)

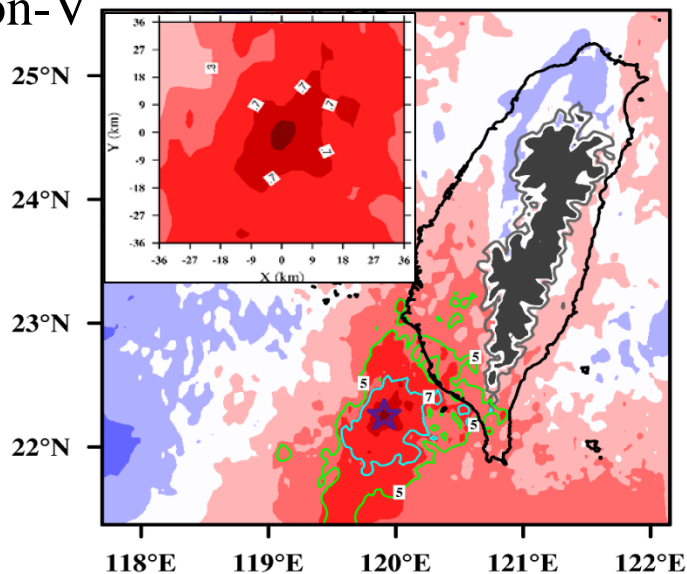
(a) Ucon-U



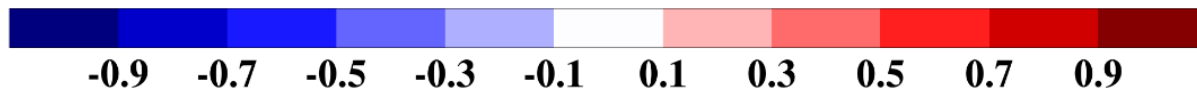
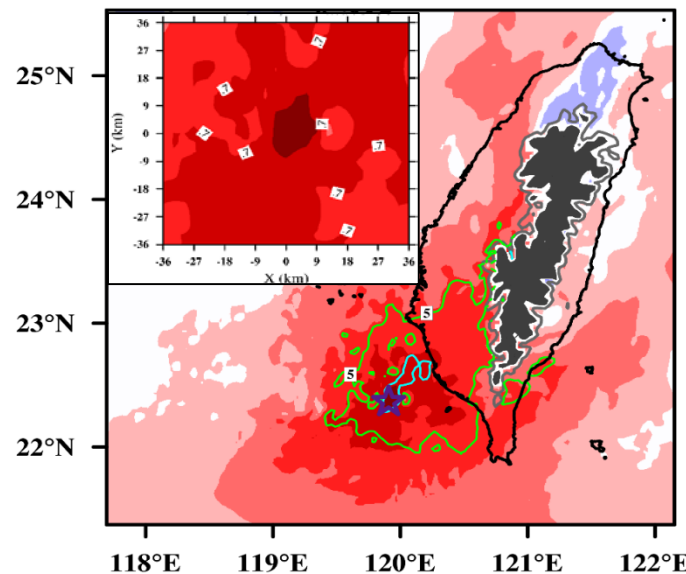
(b) Ustr-U



(c) Vcon-V

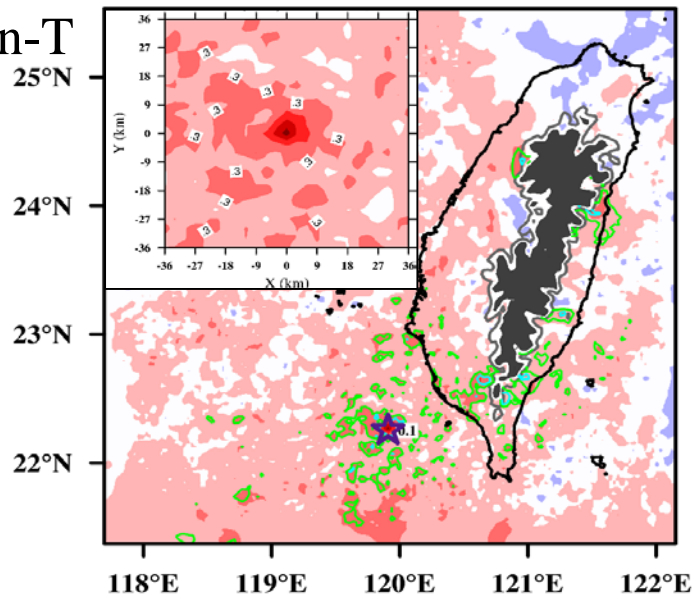


(d) Vstr-V

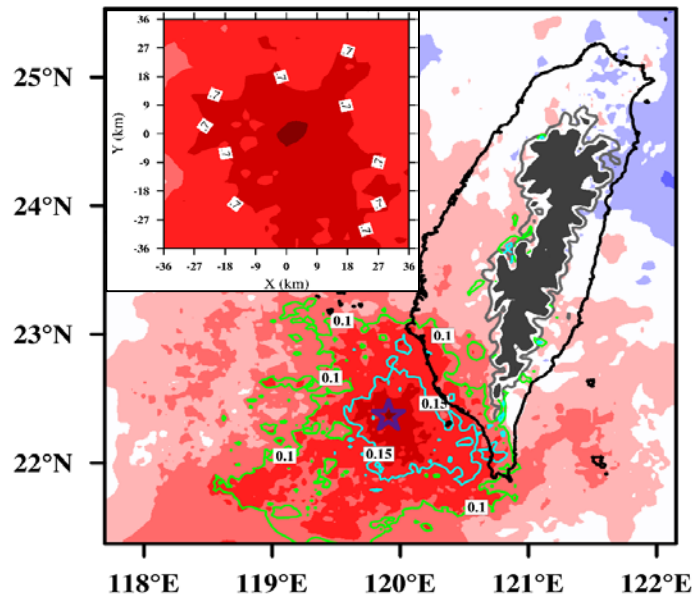


Error covariance at convective scale (II)

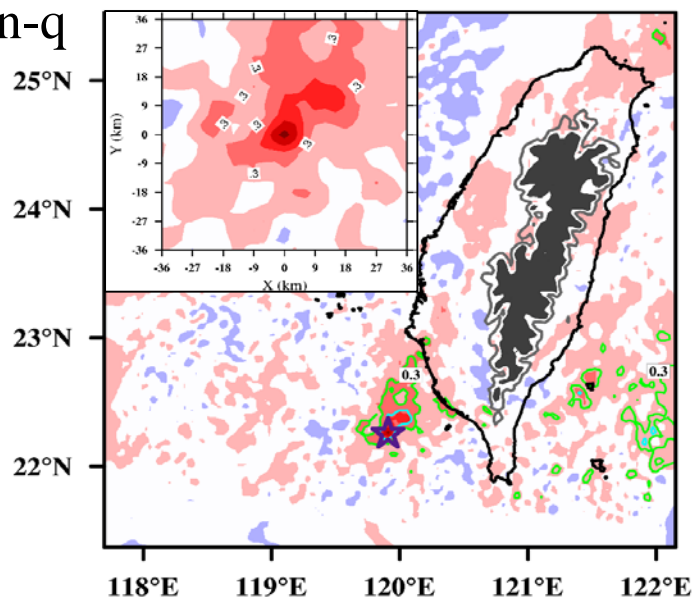
(a) $T_{con}-T$



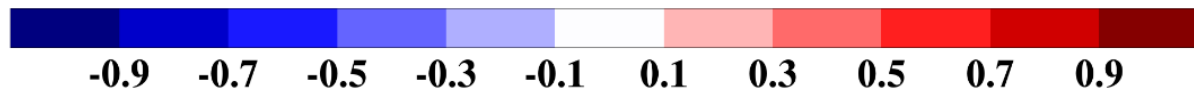
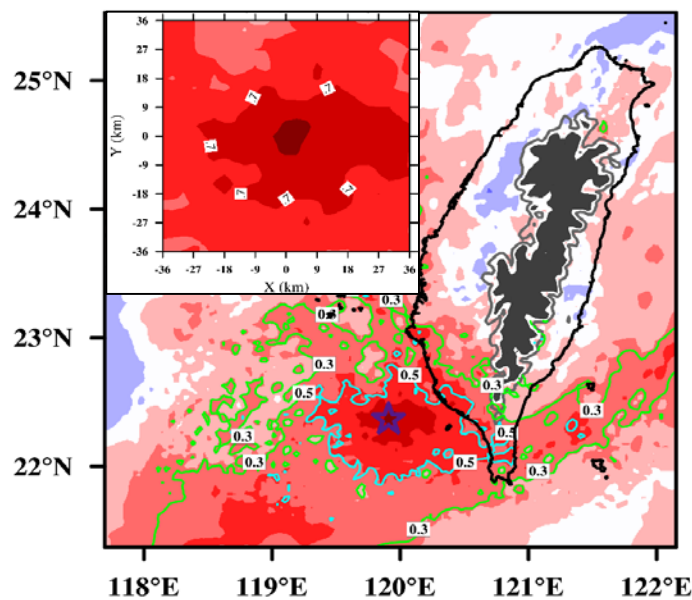
(b) $T_{str}-T$



(c) $q_{con}-q$



(d) $q_{str}-q$



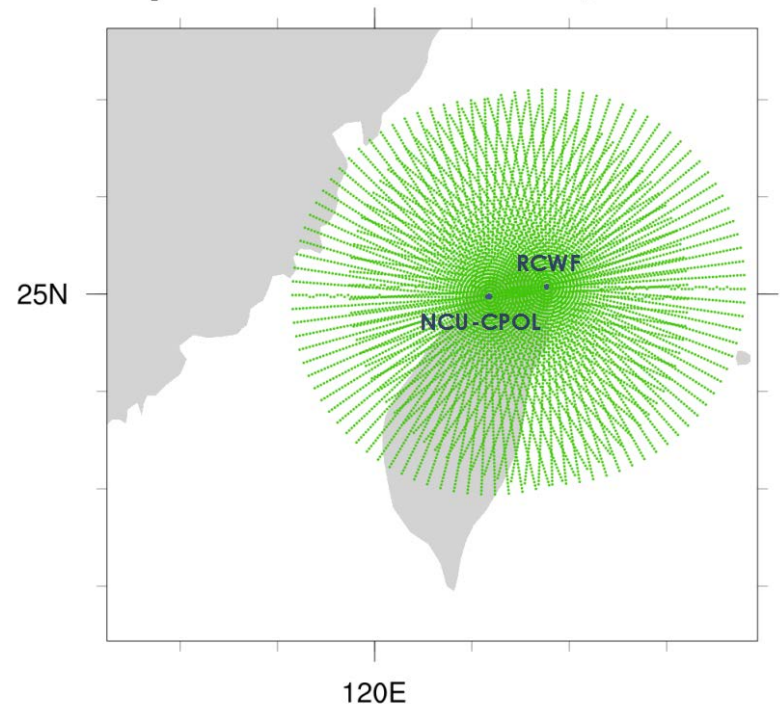
3. Impact Of Assimilating Thermodynamic Variables

Are we able to reduce the cycling procedure for convective scale QPF?

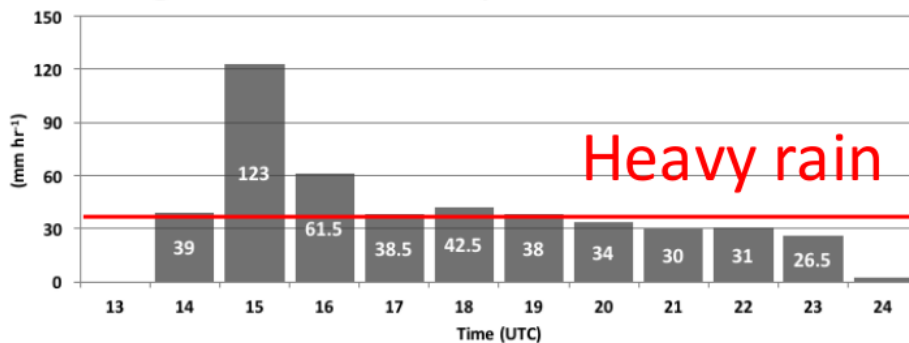
OSSE setup

- WRF-LETKF Radar Assimilation System (WLRAS; Tsai et al., 2014)
- Case study: Mesoscale convective system in Mei-Yu season 11 June 2012
- Radar sites: RCWF and NCU-CPOL

Superobs-RCWF&NCU, ele= 3.4



Yangmei station hourly rainfall



Truth: initial condition: EC-reanalysis (0611 0000UTC)

Exp. : initial condition: NCEP(0611 0000 UTC)

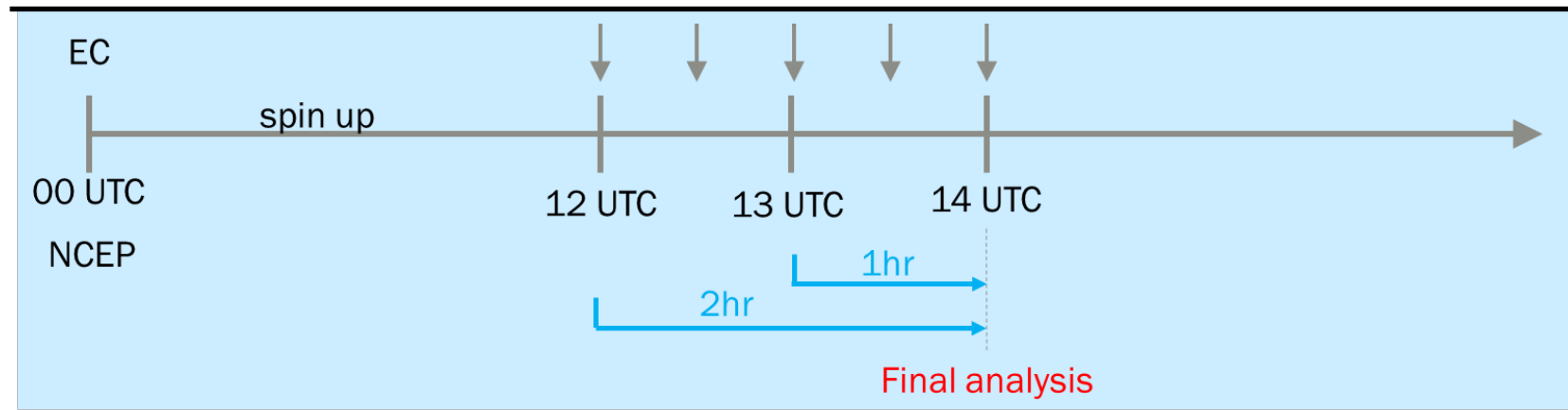
DA period : 1hour /2hours,every 15min

DA variables : Reflectivity, Radial wind, Temperature, Water Vapor

Observation error : Z(3 dBZ), Vr(1 m/s), T(0.5K), Qv(0.5 g/kg)

Member size: 40 members

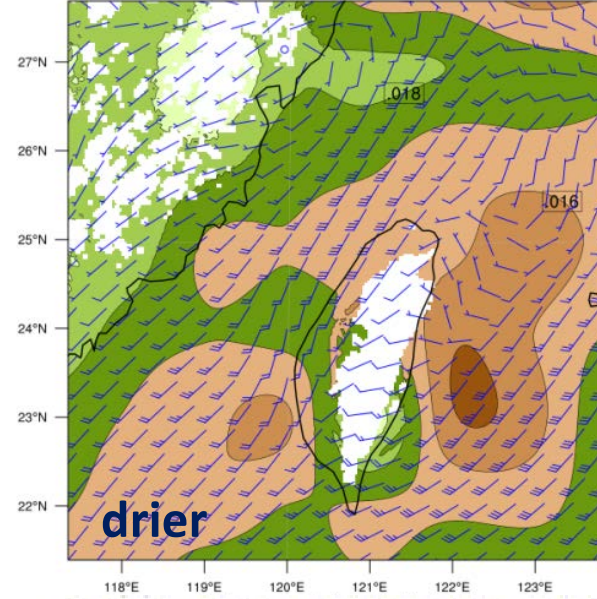
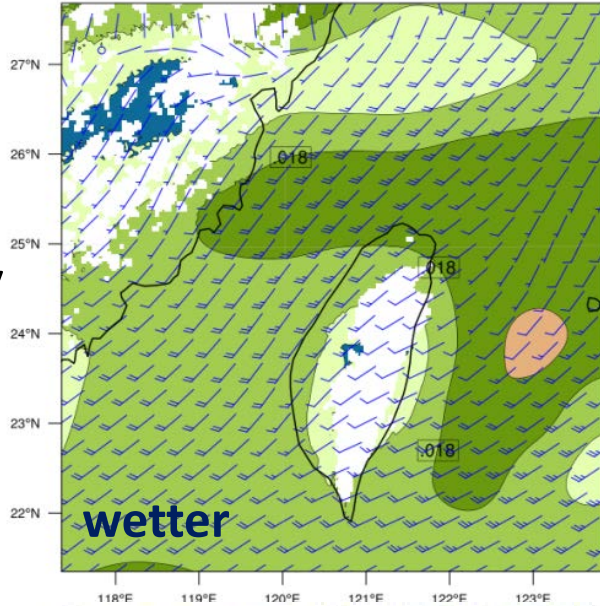
Experiments	Observation	Assimilation period (h)
NoDA	-	-
VrZT	Vr, Z, T	1
VrZQv	Vr, Z, Qv	1
VrZ	Vr, Z	1
VrZ-2	Vr, Z	2



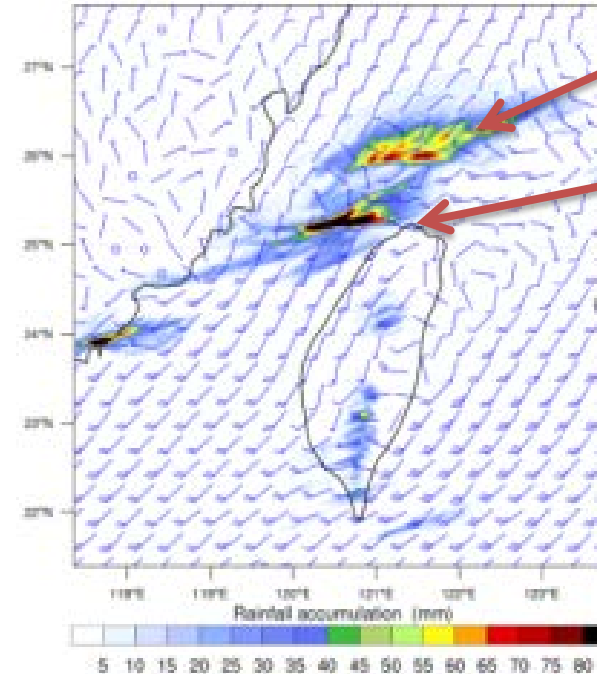
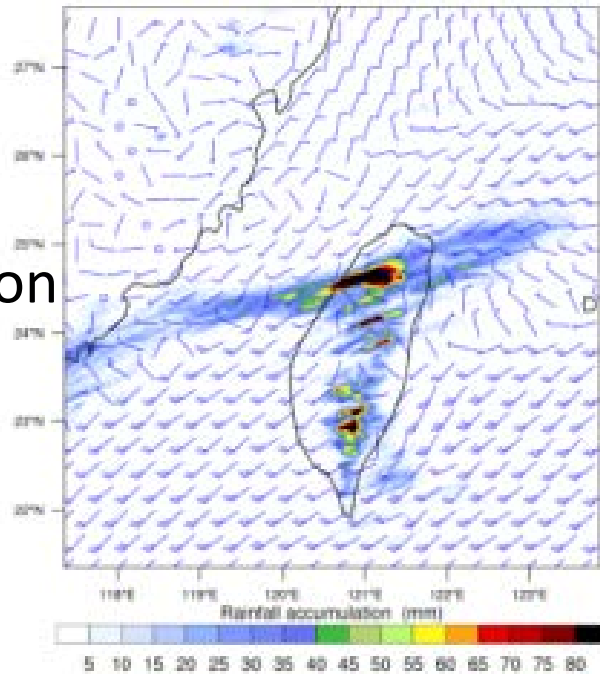
EC (truth)

NCEP (NoDA)

Humidity



Precipitation

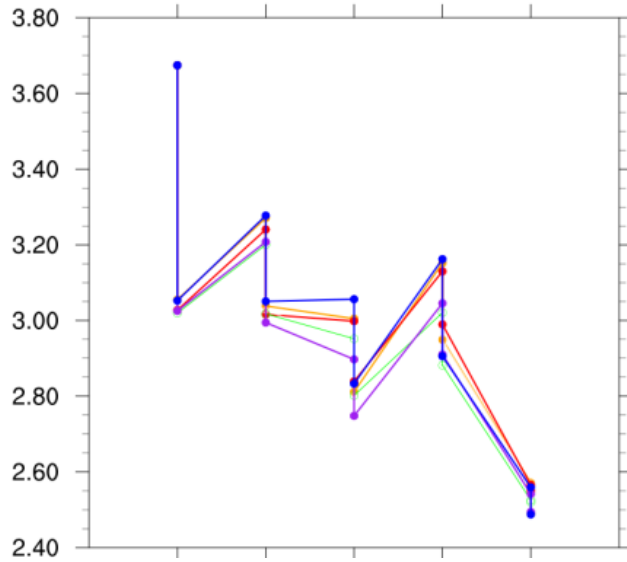


False alarm

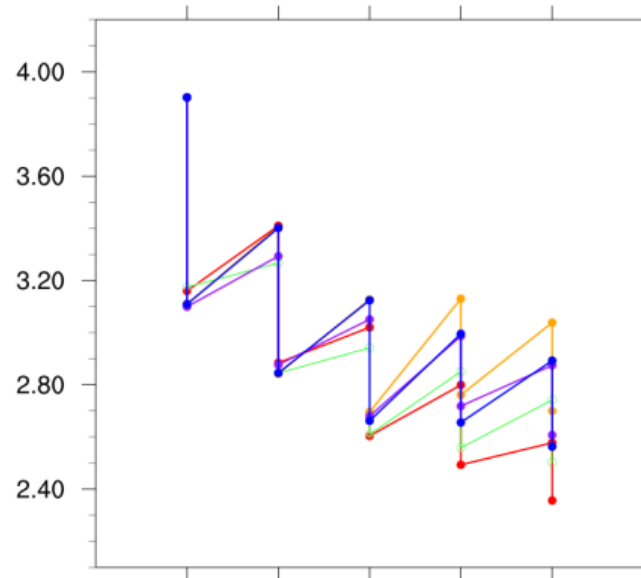
Position error

RMSE of A and P during 1-h assimilation cycles

RMSE U-wind



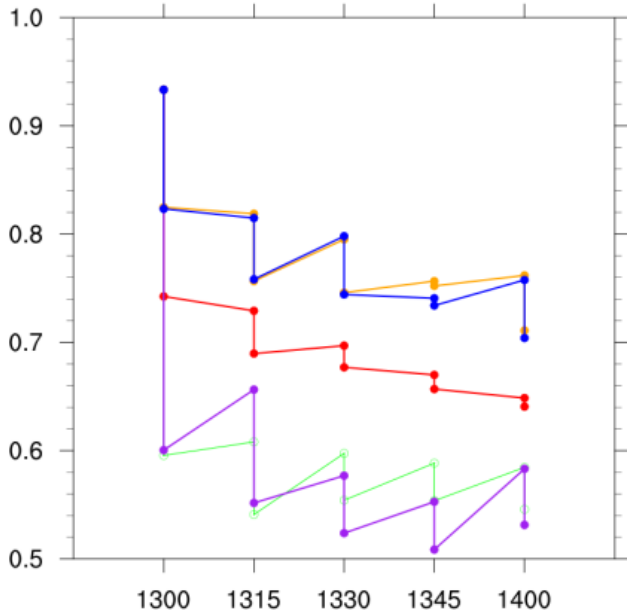
RMSE V-wind



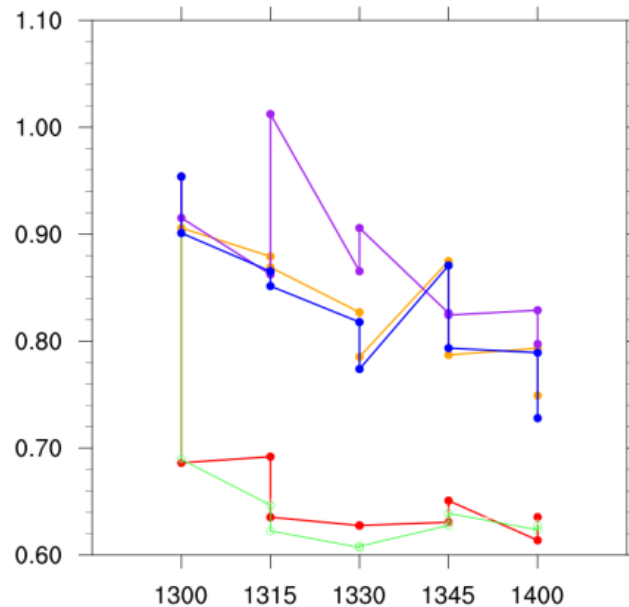
- ZVr
- ZVrT
- ZVrQv
- ZVrTQv

Minor differences:
Assimilating Vr

RMSE Temperature



RMSE Qvapor

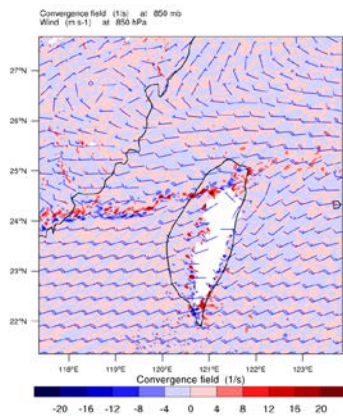


Major differences

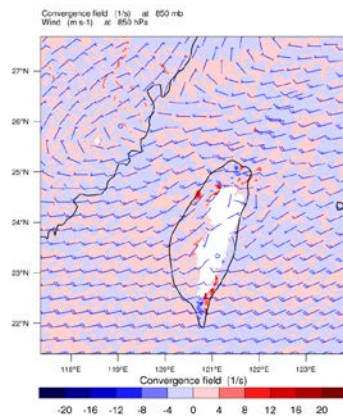
Convergence on 850mb

1400

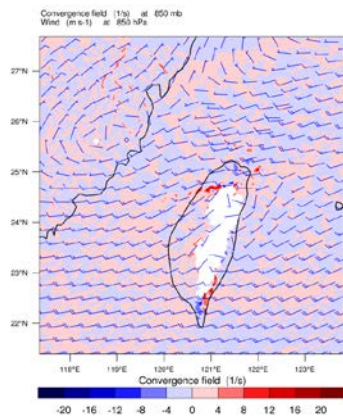
Truth



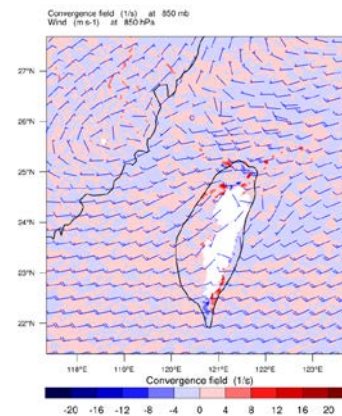
ZVr



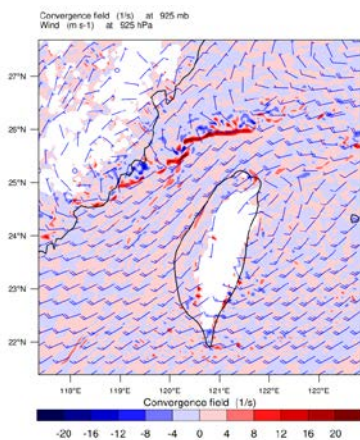
ZVrT



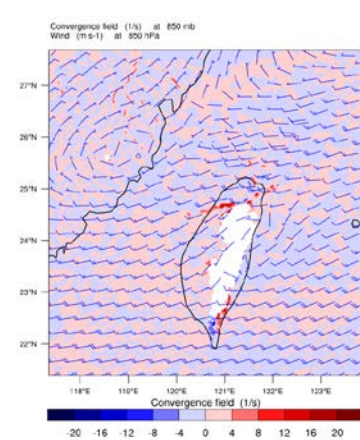
ZVrQv



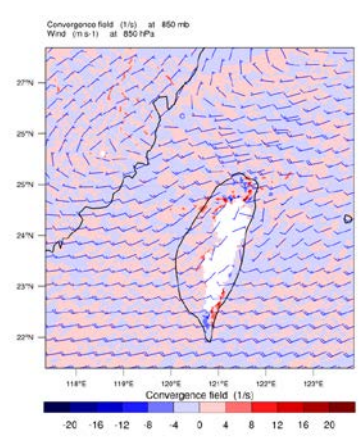
NoDA



ZVrTQv



ZVr2h



Results of assimilation cycles (I)

Vertical structure of temperature & vertical velocity

Cycle2
(1315)

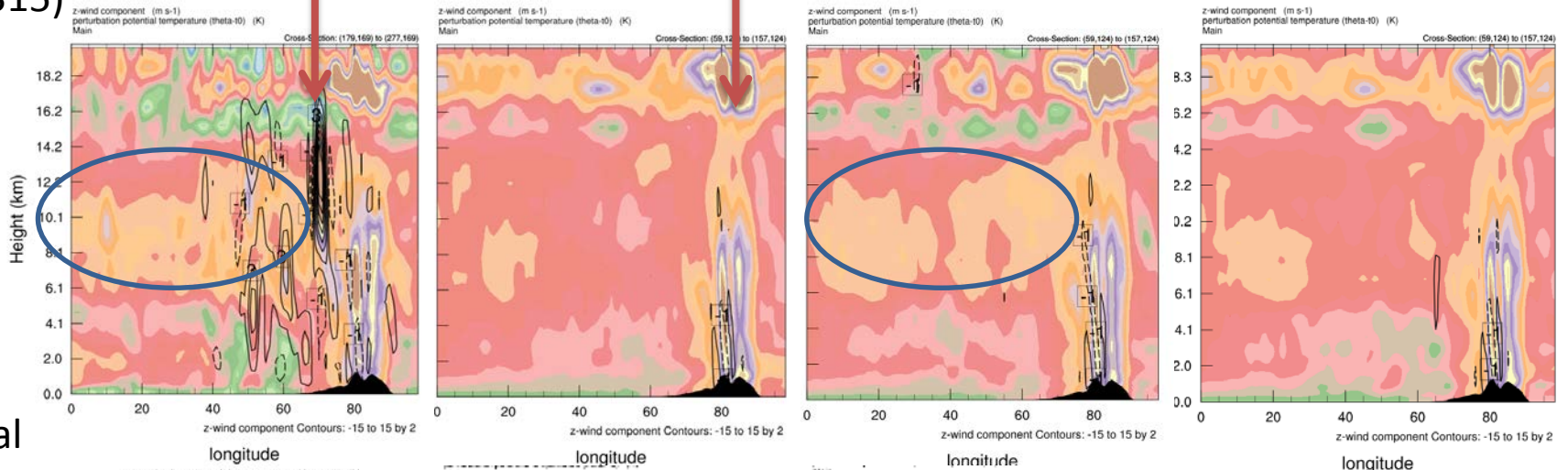
Truth

Conv.

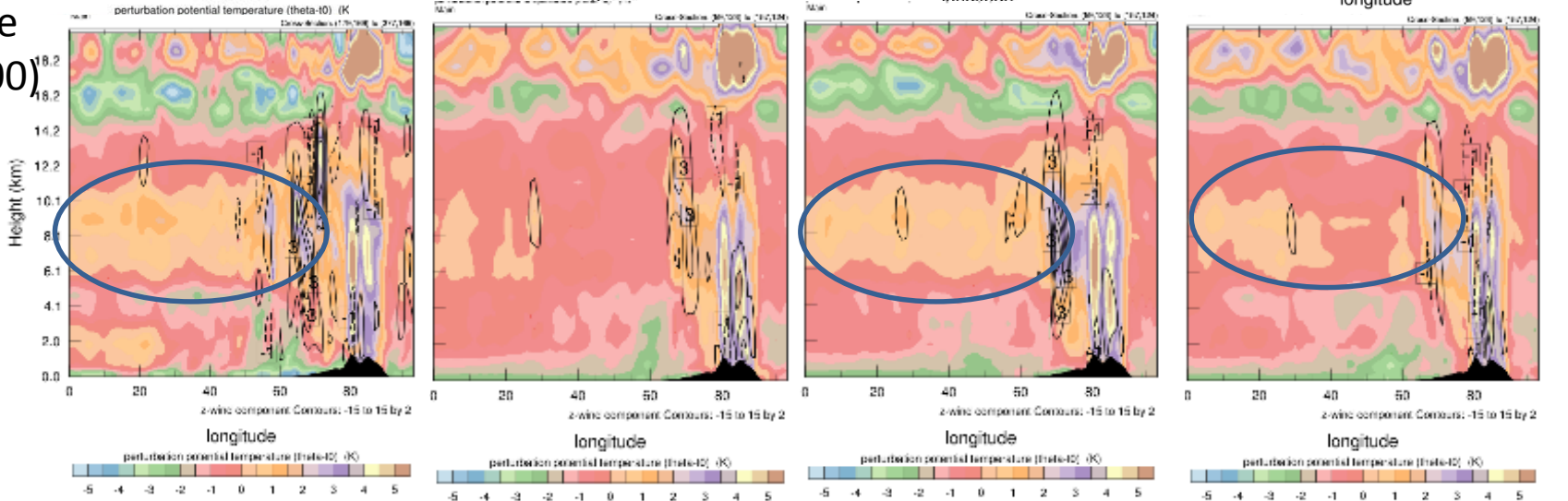
ZVr

ZVrT

ZVrQv



Final
Cycle
(1400)



Results of assimilation cycles (II)

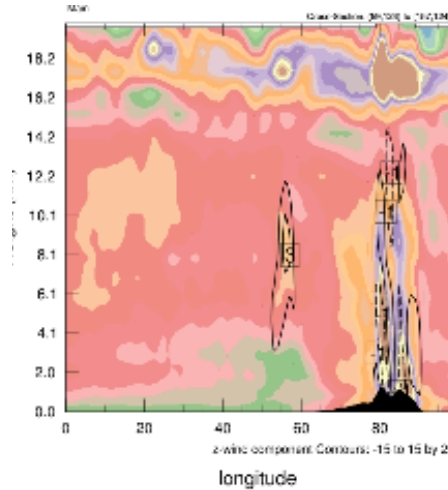
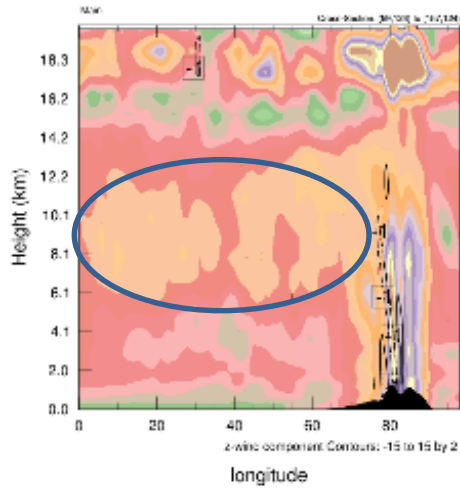
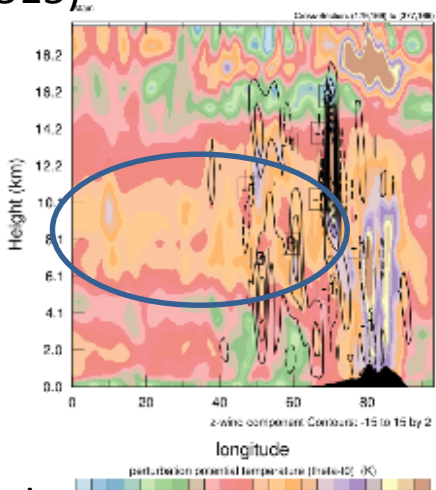
Vertical structure of temperature & vertical velocity

Cycle2
(1315)

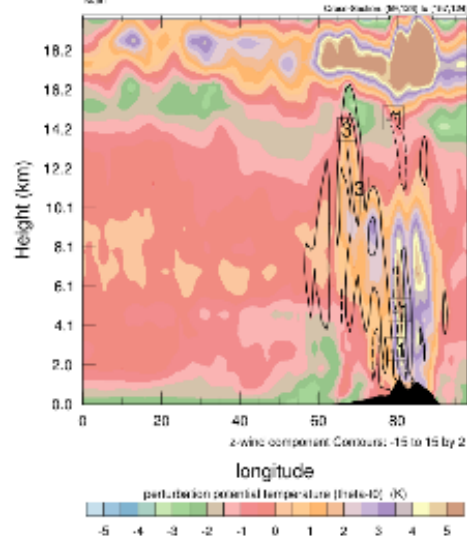
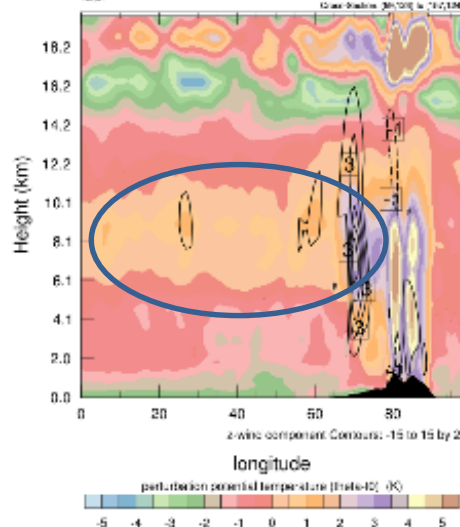
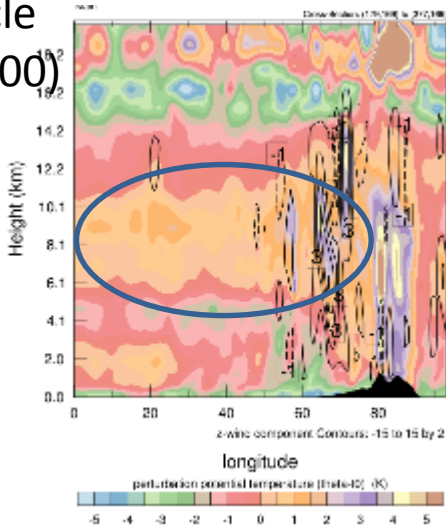
Truth

ZVrT

ZVr2h



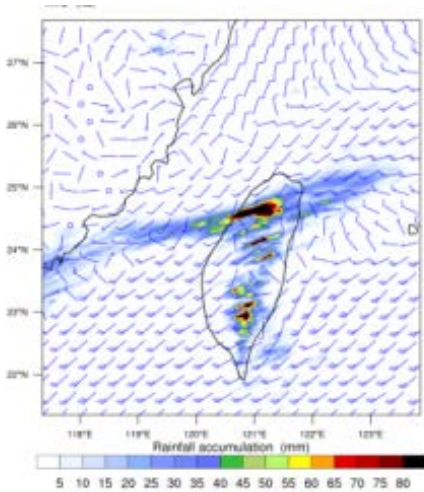
Final
Cycle
(1400)



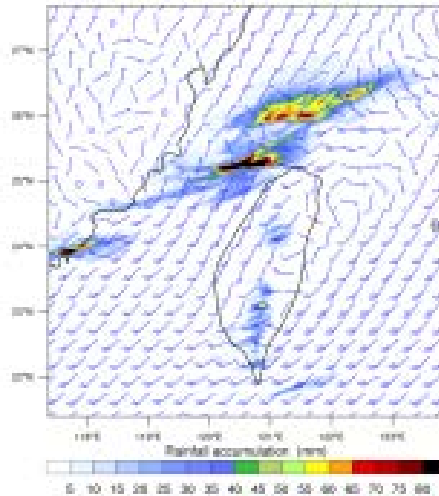
Result of the short-term forecast (I) (3-h accumulated rainfall)

Deterministic forecast from ensemble mean analysis

Truth

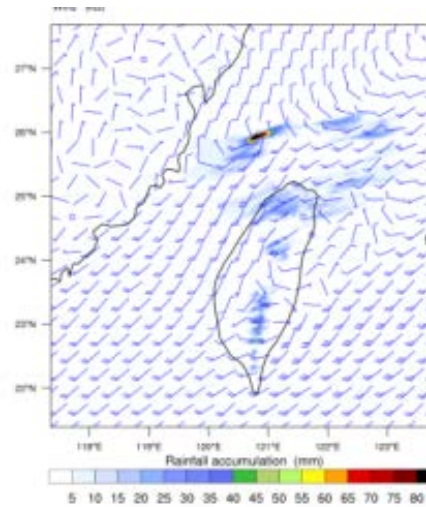


NoDA

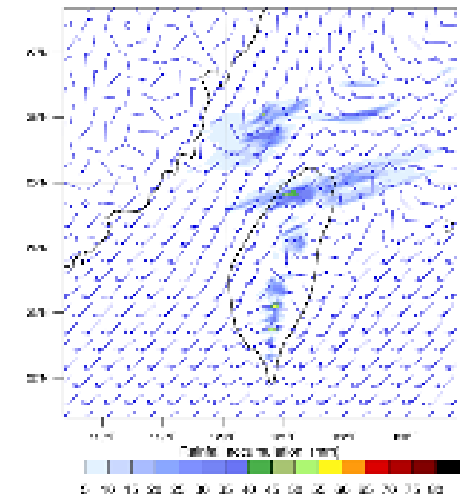


Position improved,
but less accumulated precipitation

ZVr



ZVr2h

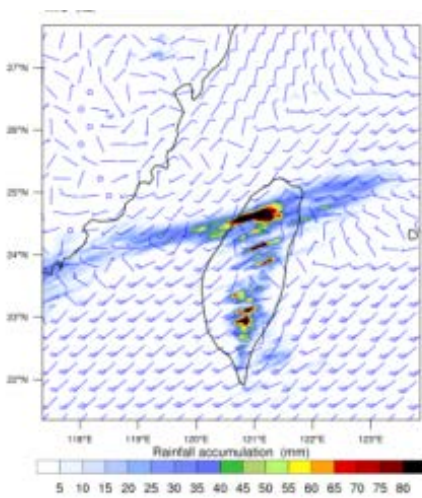


Position and accumulated precip.
slightly improved.

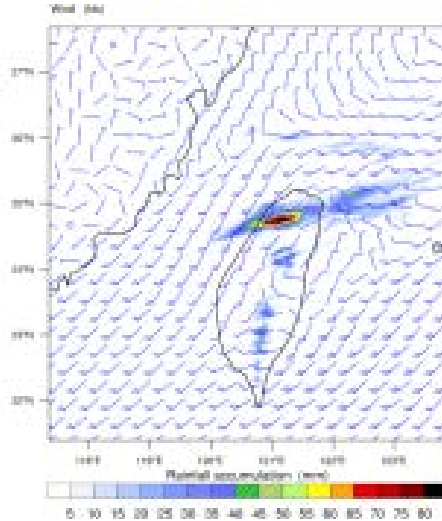
Result of the short-term forecast (II) (3-h accumulated rainfall)

Deterministic forecast from ensemble mean analysis

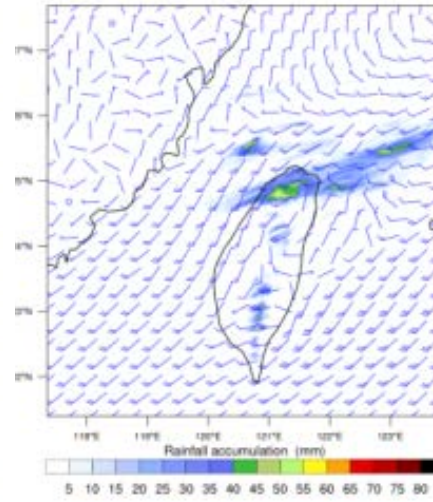
Truth



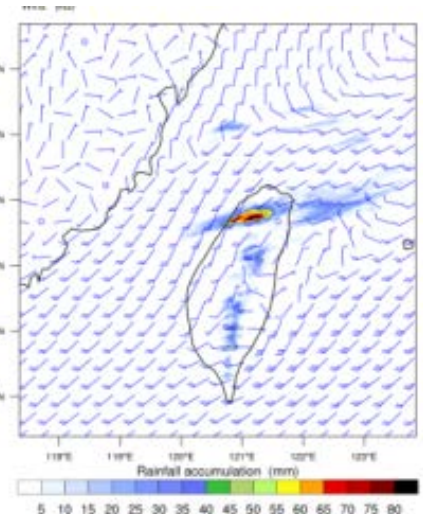
ZVrTQv



ZVrT



ZVrQv



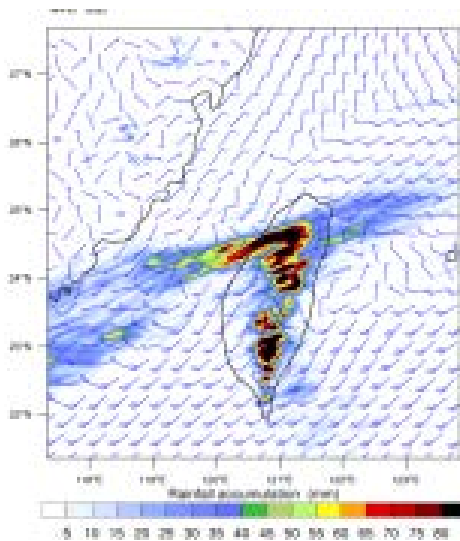
↑
Better (location)

↑
Better (intensity)

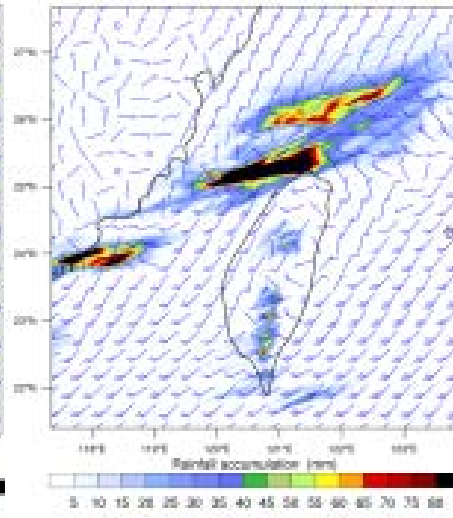
Result of the short-term forecast (II) (6-h accumulated rainfall)

Deterministic forecast from ensemble mean analysis

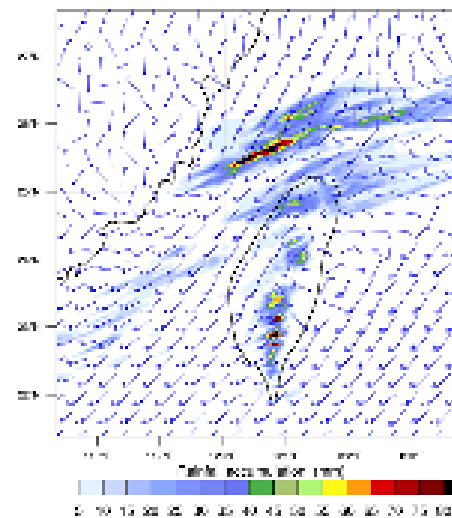
Truth



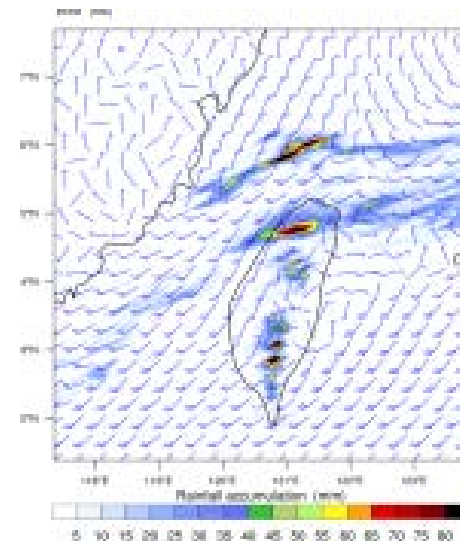
NoDA



ZVr2h



ZVrTQv



Summary

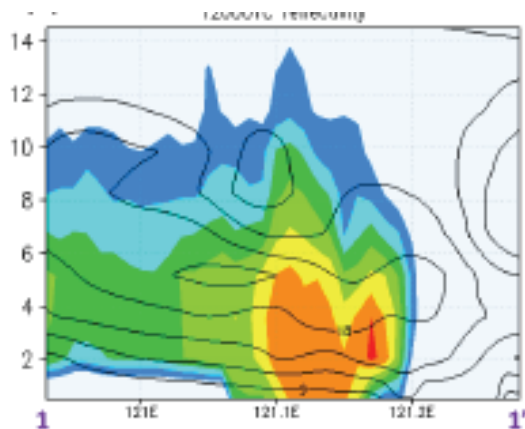
- **The examination of background errors at convective scale shows**
 - 1) model of higher resolution is associated with larger uncertainties;
 - 2) multi-scale situation is both showed in variance and correlation;
 - 3) less strong or very localized correlation between state variables
-> **this explains why we need many cycles with radar network**
 - 4) error covariance has different impact at **convective** and **stratiform** area
- **Results of OSSEs show that**
 - 1) one is able to **reduce the cycling procedure** and obtain better analysis if it is able to assimilate temperature or humidity;
 - 2) To correct position error of precipitation: assimilating radial wind and reflectivity may not be enough.
→ **Info. of temperature or humidity is needed at convective scale;**
 - 3) To improve the QPF for longer time: information of humidity is the key.

Observations lead the way:
Observations of temperature and humidity in 3D
are needed for convective scale

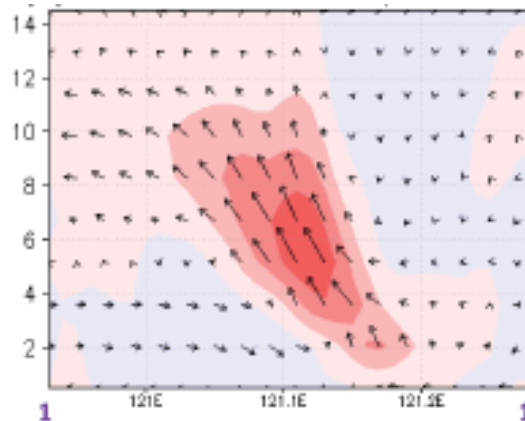
How to obtain the info ?

Retrieval from remote sensing (Himawari 8, GPS-RO, radar) &
mesonet, mesoscale radiometer

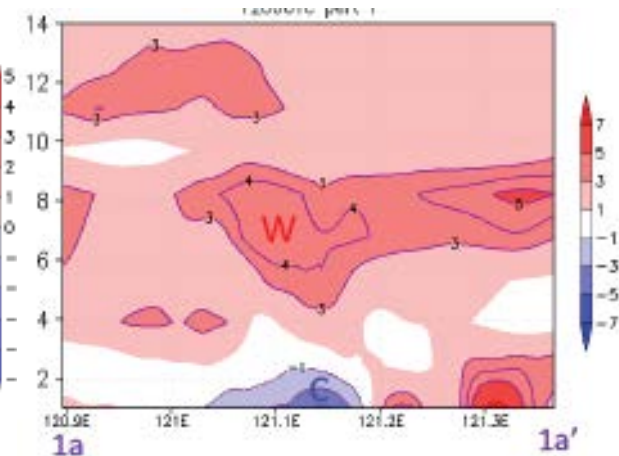
Z-Qr



$V_r \rightarrow u, v, w$



T





Thank you for your attention!

Any questions?