



Radar Data Assimilation with WRFDA

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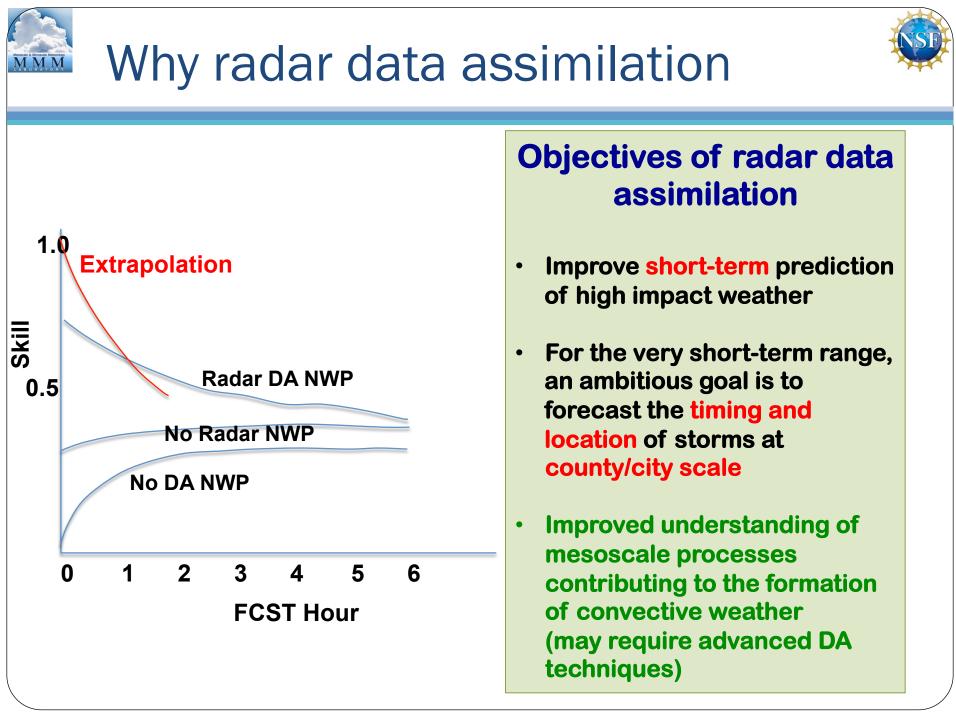
July **2**4–26, 2017, Boulder, CO



OUTLINE



- Motivation, milestones, and current capabilities
- Radar DA method in WRFDA
- Configure and run WRFDA-radar
- Applications and ongoing research





Radar data: the good and the bad

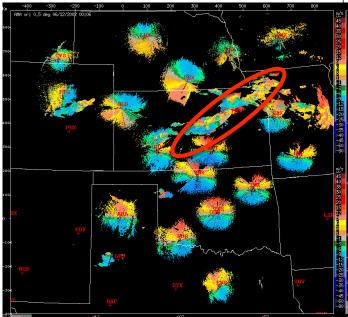


- High spatial and temporal resolutions at the convective-scale
- Observes wind (radial velocity) as well as microphysics (reflectivity)
- Accurate observations
- Observations are mostly in the lower atmosphere

Not so good

- Indirect observations need observation operators
- Incomplete coverage limited range and limited detection ability in clear air
- Only radial velocity and reflectivity
- Nontrivial for QC
- Locally available

Radial velocities from 20 WSR-88D radars





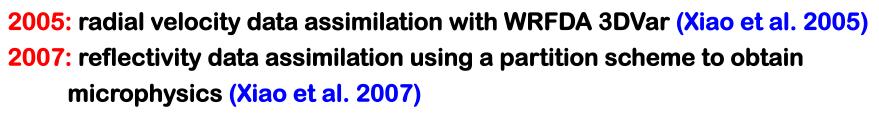






- Impact of radar data assimilation on convective forecasting (Xiao et al. 2005, 2007)
- Validity of the linearization of the observation operator for reflectivity (Wang et al. 2013)
- Validity of the tangent linear model approximation for highly convective weather (Wang et al. 2014)
- How well does 4DVar perform in comparison with 3DVar (Sun and Wang 2014)
- Impact of the choice of momentum control variables (Sun et al 2016)
- Controlling noise in high-resolution analysis with multi-scale balance (Vendrasco et al. 2016, Tong et al. 2017)





- **2013:** indirect assimilation of reflectivity using q_r and q_c as control variables (Wang et al. 2013)
- 2014: adjoint of Kessler scheme for 4DVar radar data assimilation (Wang et al. 2014)
- **2015:** new momentum control variables (u/v) for radar data assimilation (Sun et al. 2016)
- **2015:** a large-scale analysis constraint to maintain large-scale balance (Vendrasco et al. 2016)
- 2017: a strategy for hourly update cycles (Tong et al. 2017)
- 2017: a scheme for assimilating "no rain" reflectivity data (Gao et al., submitted)



Current capabilities



3DVar

- Assimilate both radial velocity and reflectivity
- Direct and Indirect assimilation of reflectivity
- "Warm start" by assimilating estimated humidity within cloud
- Options for choice of momentum control variables
- Constraint terms (Jc) for controlling analysis noise
- Operational 3h/1h cycles since 2014

4DVar

- Use WRF tangent linear model as constraint with multiple outer loops
- Can be run with multi-incremental option
- Adjoint of physics schemes: modified Kessler microphysics, largescale condensation, a simple cumulus scheme, and diffusion scheme
- Assimilate both radar and rainfall data
- 4DVar framework is fully consistent with 3DVar



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- 1. Direct assimilation of reflectivity (Xiao et al. 2007)
 - Requires an observation operator to link the reflectivity with microphysics
 - No cloud control variables
 - Vertical velocity is diagnosed using the Richardson equation
 - Microphysics are diagnosed using a warm rain partition scheme
 - 2. Indirect assimilation of reflectivity (Wang et al. 2013, preferred)
 - Diagnose microphysics (q_r, q_s, q_g) and humidity from reflectivity
 - Assimilate the diagnosed quantities
 - Cloud control variables and vertical velocity control variable



Cost Function



Indirect method with cloud control variables

$$J = J_b + J_o + J_{v_r} + J_{q_r} + J_{q_v}$$
For radar DA

Control variables :

u/v (or $\psi/\chi_{u)}$, T (or T_u), Ps (or Ps_u), RHs, q_c, q_r, and w

- 3DVar critically depends on a cloud analysis scheme that assimilates estimated in-cloud humidity
- A modified Kessler scheme along with its adjoint produces analyses of microphysics in 4DVar



Radial velocity observation operators

$$J_{vr} = \frac{1}{2} \sum (V_r - V_r^{ob})^2 / \sigma_{vr}$$

- V_r Radial velocity from the model V_r^{ob} Radial velocity observations
- $\sigma_{_{_{\it V\!F}}}$ Observation error variance

v_r - (u,v,w,q_r) Relation:

$$v_{r} = \frac{x - x_{r}}{r}u + \frac{y - y_{r}}{r}v + \frac{z - z_{r}}{r}(w - V_{T}(q_{r}))$$



$$J_{qr} = \frac{1}{2} \sum (q_r - q_r^{ob})^2 / \sigma_{qr}$$



- q_r^{ob} Estimated rainwater mixing ratio from reflectivity
- $\sigma_{\it qr}$ Observation error variance

Z-q_r Relation (assume Marshal-Palmer DSD)

$$Z = 43.1 + 17.5 \log_{10}(
ho q_r)$$
 Sun and Crook (1997)

Z-q_s and Z-q_h follow Gao and Stensrud (2012)



Pseudo humidity assimilation

$$J_{qv} = \frac{1}{2} \sum (q_v - q_v^{ob})^2 / \sigma_{qv}$$



- q_v^{ob} Estimated water vapor mixing ratio by assuming near saturation above LCL within cloud
- σ_{qv} Observation error variance

Namelist options: cloudbase_calc_opt

Option for calculating cloud-base height: below this height retrieved humidity will not be assimilated for the use_radar_rqv option 0: fixed value of 1500 meters 1 (default): KNU scheme 2: NCAR scheme radar_saturated_rf rf value (dBz) used to indicate precipitation for rqv (default 25.0)



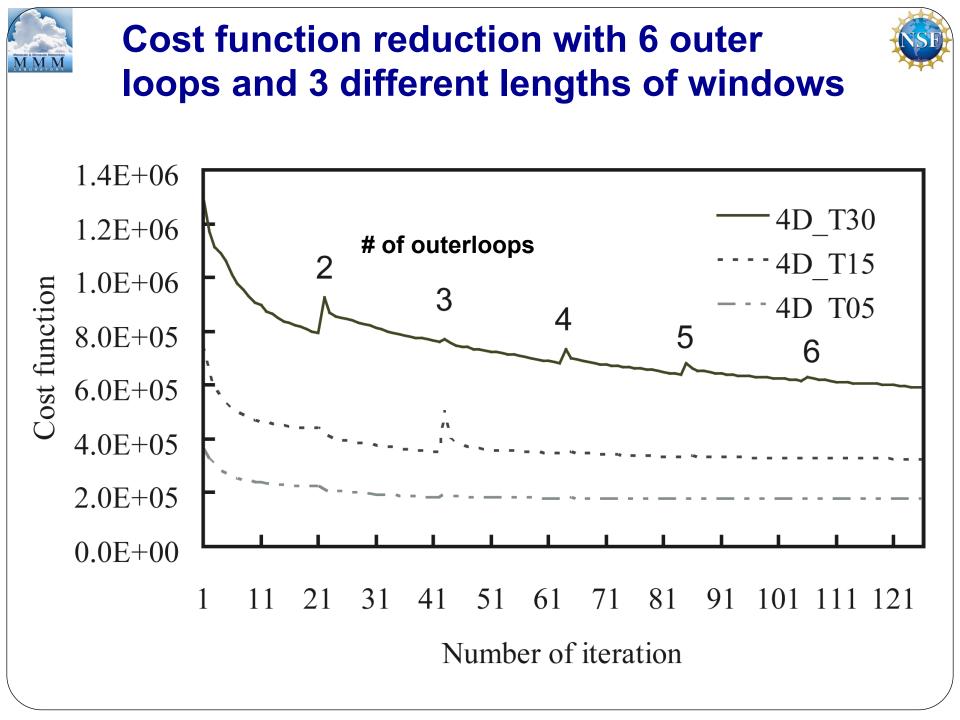
Multi-incremental 4DVar



Following Courtier et al. 1994, WRFDA uses a multi-incremental formulation, which means

- The forward prediction model within the 4DVar window is an approximation of the nonlinear model
- The control variables are increments from the forward model trajectory
- The formulation requires the update of analysis increment in an inner loop but also the update of the nonlinear model in an outer loop
- It makes the cost function better conditioned and allows different spatial resolutions for the inner and outer loops multi-incremental

Details in Huang et al. (2010) and Wang et al (2013)







WRFTL & AD	WRFNL	
mp_physics_ad=98 Large scale condensation	mp_physics can be set to an options for WRF	
Mp_physics_ad=99 Modified Kessler scheme	It can also be set to 98 or 99, same as WRFTL & AD	
cu_physics = 0: no cumulus scheme cu_physics=98: Simplified CU scheme Any other numbers will be	Same as the left column	
	mp_physics_ad=98 Large scale condensation Mp_physics_ad=99 Modified Kessler scheme cu_physics = 0: no cumulus scheme cu_physics=98: Simplified CU scheme	







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Radar DA options in Namelist

- Read radar data use_radarobs = true,
- Assimilate radial velocity use_radar_rv=true,
- Two methods for reflectivity assimilation use_radar_rf =true; (direct method; Xiao et al. 2007) use_radar_rhv=true; (indirect method; Wang et al. 2013)
- Assimilate pseudo humidity use_radar_rqv=true,
- 3DVar (default) or 4DVar var4d=true
- Microphysics control variables cloud_cv_options = 3 BE of regular variables are from gen_be; cloud variables hard coded
- u/v momentum control variables

cv_options = 7 5: psi/chi CV 7: u/v CV (7 is recommended for radar)

 Several options for assimilating weak radar echoes – "no rain" data See the user guide



Basic steps for radar DA

NSI

Step 1: prepare radar data in the correct format and write the data into ob.radar (ob01.radar, ob01.radar, ... for 4DVar)

- Use your own QC software
- (ϕ ,r, θ) => (x,y, θ) => Lat/lon profiles => merge the radars into one file

Step 2: produce 1-3 month WRF forecasts (12h & 24h) over the study domain, and then compute BE using the WRFDA utility **gen_be**

Step 3: modify the namelist.input to make radar DA choices

Step 4: conduct single observation tests to tune the length scale and variance for your specific domain

- 200km for GTS data and 30km for radar data are commonly used

Step 5: configure WRFDA: to invoke the "CLOUD_CV" option, do the following in the configure script setenv CLOUD_CV 1 - for csh Export CLOUD_CV = 1 - for both ksh and bash

STEP 6: link ob.radar and other other observation files and first guess



Data format



```
write(301, '(a14, i3)') 'Total number =', nrad
write(301,'(a)')
'#-----#'
write(301,'(a)') '
do irad = 1, nrad ! nrad: total # of radar
!---Write header
write(301,'(a5,2x,a12,2(f8.3,2x),f8.1,2x,a19,2i6)') 'RADAR', &
radar_name, rlonr(irad), rlatr(irad), raltr(irad)*1000.,
                                                      &
trim(radar_date), np, imdv_nz(irad) write(301,'(a)')
                                                      &
'#-----#'
write(301,*)
!----Write data
do i = 1,np ! np: # of total horizontal data points
write(301,'(a12,3x,a19,2x,2(f12.3,2x),f8.1,2x,i6)') 'FM-128 RADAR', &
trim(radar_date), plat(i), plon(i), raltr(irad)*1000, count_nz(i)
 do m = 1,count_nz(i) ! count_nz(i): # of vertical elevitions for each radar
write(301,'(3x,f12.1,2(f12.3,i4,f12.3,2x))') hgt(i,m), &
rv_data(i,m), rv_qc(i,m), rv_err(i,m),
                                        &
rf_data(i,m), rf_qc(i,m), rf_err(i,m)
enddo
enddo
enddo
```



Data format

Header

nrad: total number of radars
radar_name: name of the ith radar (irad)
rlonr, rlatr, raltr: longitude, latitude, and altitude of irad
radar_date: date of irad observation
np: number of total data points for irad
imdv_nz: number of total elevation angles of irad

Data

plat, plon, raltr: longitude, latitude, and altitude of the ith data point
count_nz: number of data levels at the ith data point
hgt(i,m): height of ith data point at m level
rv_data(i,m), rv_qc(i,m), rv_err(i,m): radial velocity value, qc index, and
obs error. The qc index >= 0 means good data; otherwise bad data.
rf_data(i,m), rf_qc(i,m), rf_err(i,m): reflectivity value, qc index, and obs
error



Example of radar input data

∎OTAL NUMBER = 16 #-----#

RADAR #	KCYS	-104.806	41,152	1887.0	2015-07-07_21 #	:00:00 5497
FM-128		15-07-07_21:				
	5128.9 -	888888.000 - 888888.000 - 888888.000 -	88 -888888	.000	13,029 0	
FM-128	RADAR 20	888888.000 - 15-07-07_21: 888888.000 -	00:00	41,192		•
	5130.6	-7,381	0 1	. 692	13,338 0 8,373 0	0,473
FM-128	RADAR 20		00:00	41.219	-107,189 9,447 0	1887.0
	5133,9	-8,476 888888,000 -	0 1	. 632	12,828 0 8,969 0	0.833
FM-128	RADAR 20	15-07-07_21: 888888.000 -	00:00	41,246	-107,189	
	5139.0 -	888888.000 - 888888.000 -	88 -888888	.000	15,127 0 11,409 0	0,948
FM-128		15-07-07_21: 888888.000 -			-107.153 11.011 0	
	6732.4 -	888888.000 - 888888.000 -	88 -888888	.000	12.650 0 6.896 0	
FM-128	3645.0 -	15-07-07_21: 888888.000 -	88 -888888	.000	11,477 0	0,804
	6732.4 -	-5,278 888888,000 -	88 -888888	.000	9,280 0	0,990 2,035
⊦M-128	KAUAR 20 3646.4	15-07-07_21: -0,267	00:00	41,192 ,448		1887.0 1.225
		-5.217 888888,000 -			14.294 0 10.094 0	0.731 2.072

NSF

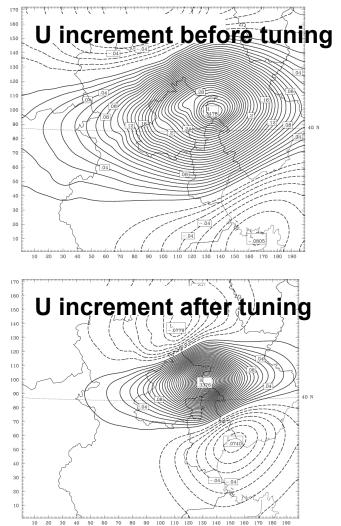


Tuning BES parameters



To change BES variance and length scale, do the following

in your namelist.input: To decrease the weight of the background VAR SCALING1=2.0 VAR SCALING2=2.0 VAR SCALING3=2.0 VAR SCALING4=2.0 VAR SCALING5=2.0 To decrease the length scale LEN SCALING1=0.5 LEN SCALING2=0.5 LEN SCALING3=0.5 LEN SCALING4=0.5 LEN SCALING5=0.5







- The new control variable (CV) option CV7 which uses u/v instead of psi/chi as momentum control variables is added in WRFDA3.7
- CV7 requires the computation of BES of u and v
- In the current version, correlation between variables is not considered
- But will be studied and included in a future release
- Sun et al. (2016, MWR)

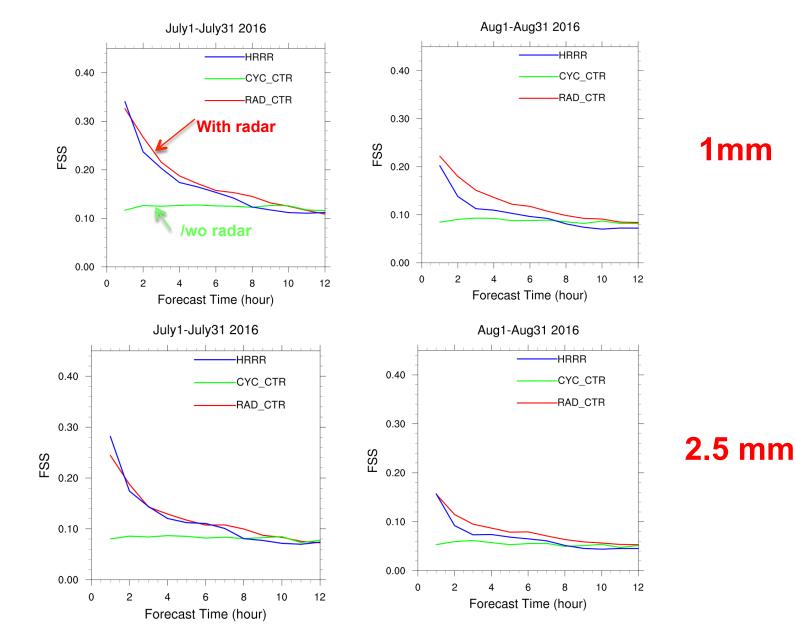




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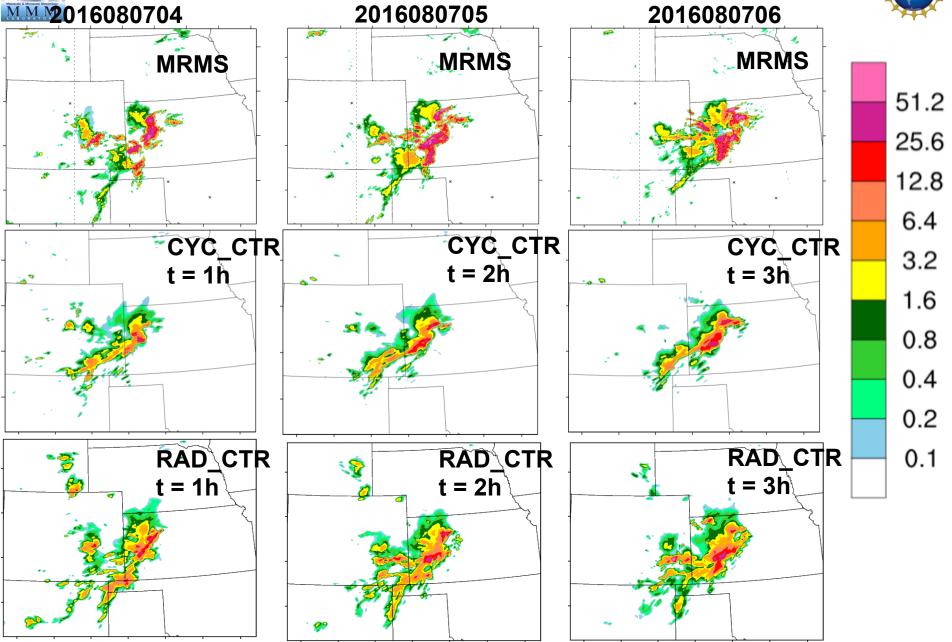


Impact of radar data assimilation (FSS)



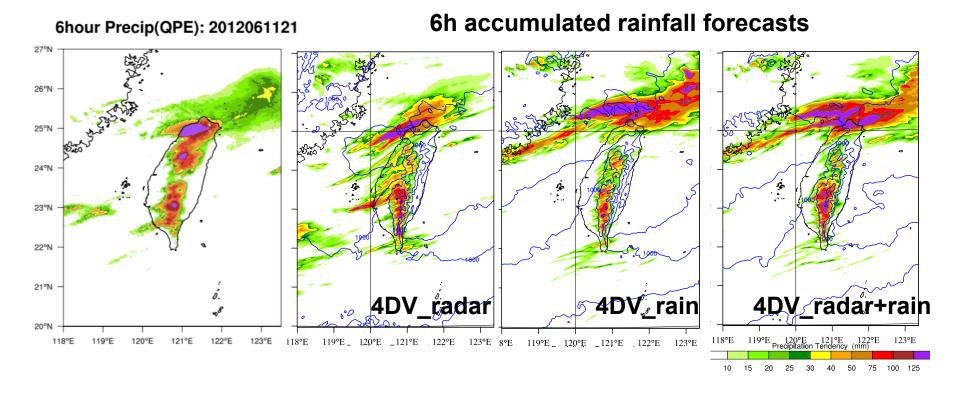
Comparing forecasts with/wo radar







Impact of assimilating both radar and rainfal



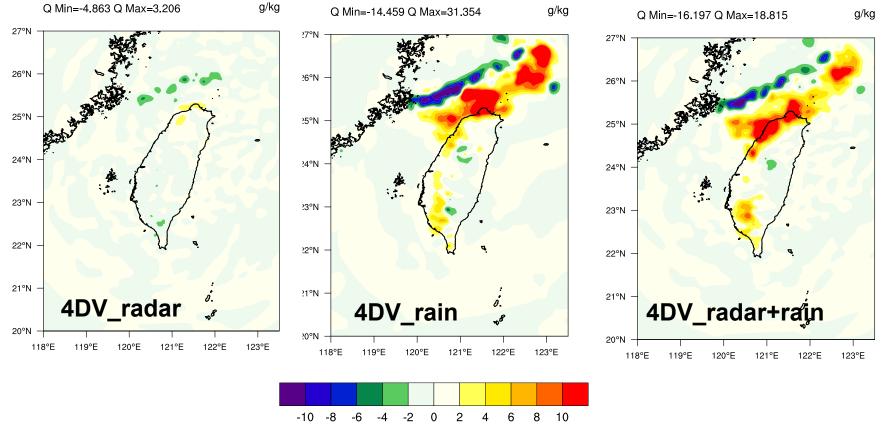
- Radar+rain shifted the northern rain band to the land, agreeing better with QPE
- Radar+rain also improved the intensity of the precipitation on the mountains in southern Taiwan
- Combined assimilation of radar and 10min rainfall data with a 20min 4DVAR window is an innovative study; the result is promising



Comparison of humidity increments



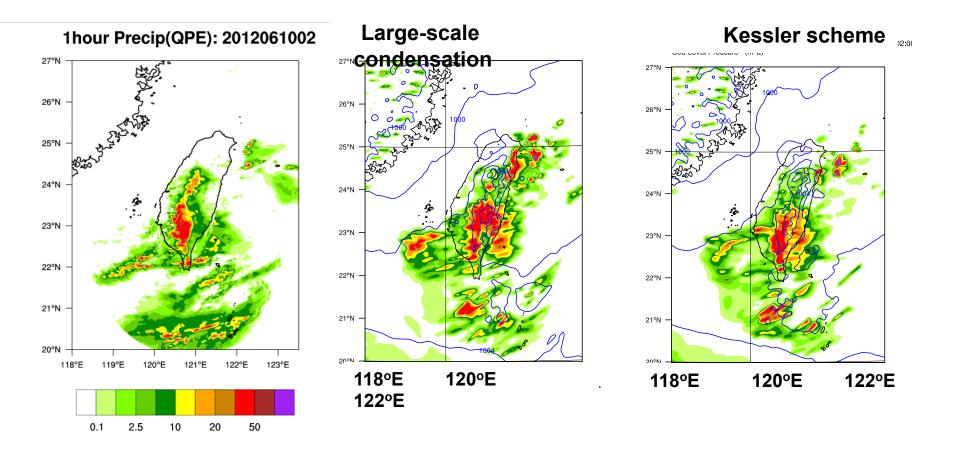
850hpa



- Combined assimilation of radar and rainfall tuned down the overly large humidity increment from that of the rainfall only experiment
- Combined assimilation also moved the humidity increment to the land

Sensitivity of physics in the TL and AD models

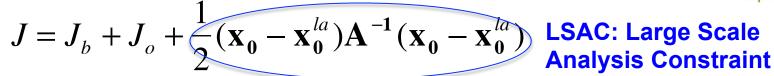
Precipitation forecasts at t = 2h for a Meiyu case occurred in Taiwan





Impact of LSAC on 3DVAR DA (future release)



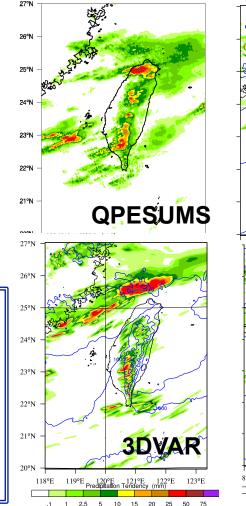


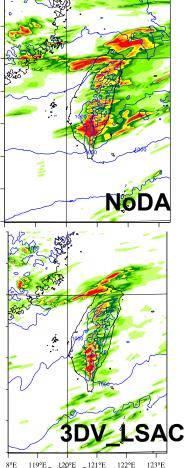
 $\mathbf{X}_{0}^{\text{la}}$: large-scale analysis (GFS)

FSS Score (threshold =5mm) 1.0 0.8 **3DV_LSAC** NODA 0.6 FSS 0.4 **3DVAR** 0.2 0.0 1.0 2.0 3.0 4.0 5.0 6.0 Forecast Time (hour)

3DVAR w/o LSAC eliminates noise in NoDA, but also some signals
3DVAR_LSAC is able to keep the analysis close to high resolution observations while still maintaining large-scale balance, resulting in improved rainfall forecast

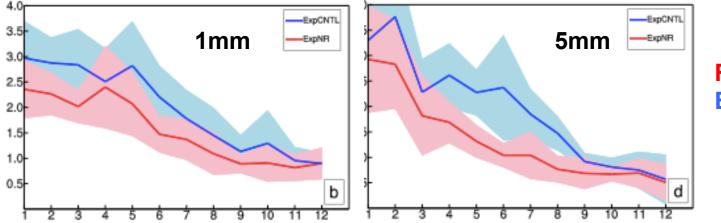
6th hour rainfall forecasts



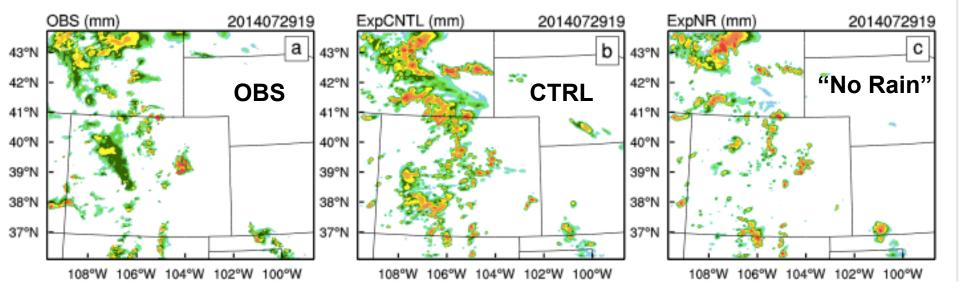




BIAS Julv 1 – August 31 2016



Red: "no rain" DA Blue: Control





Ongoing research



- How to configure 4DVar for radar data assimilation and short-term convective forecasting?
 - physics options
 - cycling strategy
 - multi-incremental method
- Improve flow-dependent BE (error of the day) using the ensemble method
 - Application of WRFDA hybrid-3DVar to radar data assimilation
 - Benefit of using EnKF to update perturbation (En3DVar vs. 3DEnVar)
- Combined assimilation of radar and rainfall
- How to define rainfall error
- Assimilate radar QPE and surface rainfall