



Assessing the Impact of Surface Data and Profiler Data on Fog Forecast Using WRF-3DVar: A Case Study on a Dense Fog Event over North China

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Fog: vis < 1000 m



234 flights cancelled

500 flights delayed

30 000 travelers stranded



"Dense fog raided Beijing "





Background

- Experimental design
- ➢Results
- ≻Analysis
- ≻Summary



➢Background

➢Experimental design

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≻Analysis



•Fog: horizontal visibility below 1 km due to water or ice droplets suspending in the atmosphere near the surface (Glickman 2000).

- •Fog is high impact weather on aviation, marine and land transportation (Gultepe et al. 2007).
 - → great need for accurate fog prediction.
- Fog is very sensitive to initial conditions (Hu et al. 2014; Zhou and Du 2010)
 making the full use of obs to improve model initial conditions is one of the main concerns for accurate fog prediction.

•one of main methods to improve initial conditions of fog predictions: DA

• Studies of DA on fog:

DA techniques

- ✓ I-DVAR (Bergot 2005; Muller et al. 2007), EnKF(Remy and Bergot 2010),
- ✓ 3DVAR(Gao et al. 2010), cycling 3DVAR (Gao et al. 2015) easy to implement → widely used in the operational system

> Obs used by DA

Review of previous studies on fog DA

Studies of DA on fog:

DA techniques

Obs used by DA:

- conventional obs (GTS) (Liang et al. 2009)
- ✓ AMDAR (Liang et al. 2009) >>> obs only in airports
- ✓ satellite obs (Gao et al. 2015) >>> loc and time limited
- ✓ microwave radiometer (Vandenberghe and Ware 2002)
 - >>> only in field obs experiments and mainly on radiation fog

✓ Fog is **grounded** cloud and **PBL** phenomenon (Gultepe et al. 2007).

- surface information
 - PBL structure (T inversion, U/V)

effective assimilation of surface obs and profiler obs significant for improving the skill of operational forecast of fog

Few studies quantitatively explored the impact of surface data and profiler data on the fog forecast.

Objective of this study

AWS (Automated Weather Stations): Rich surface data resource in China surface T, Q, U/V and P with high temporal and spatial resolutions >>>> only be used to do model verification in most studies.

Profiler : U/V profile, high-frequency, at different vertical levels PBL profiler high vertical resolution below 700 hPa

Explore the impact of surface data (AWS) and profiler data (PBL profiler) on fog forecast with aspects to different obs types, obs densities and distributions, DA cycling frequency for this case.



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OSSE (Observing Simulated System Experiments) configuration:

40-member ensemble simulation with 48-h forecast:
 0000 UTC 20 — 0000 UTC 22 Feb 2007

• one good member \rightarrow Truth \rightarrow simulated obs and verification one bad member \rightarrow BG obs = truth + random(0,obs_err)

BE: NMC (cv7) T24 - T12 from 02/01/2007 to 03/12/2007 40 days

•var_scaling= 1.0 len_scaling= 1.0

•**DA**: D02 and D03

WRF V3.3.1

3-doms 27 km, 9 km, 3 km





Surface data

OBS type	SOUND	SYNOP AWS METAR	Profiler	
Variables	U/V, T, Q, P	U/V, T, Q, P	U/V	
Horizontal density (km)	302	39	164	
Vertical resolution	II levels to 100 hPa	surface	45 levels to 700 hPa	
Temporal density	12 UTC 02/20	06, 09, 12 UTC 02/20	06, 09, 12 UTC 02/20	

Experimental design



3-h cycling

Truth



Background





Radiation Fog

Advection Fog



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Fog coverage (shading) (vis < 1 km) at the first model level at 1800 UTC 20 Feb 2007 (6 h forecast).

Results





➢Background

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Analysis RMSE over D03 at I 200 UTC 20 Feb 2007 (the last cycle)

RMSE_V RMSE_T 500 500 600 600 **BIAS_T** 700 700 p (hPa) p (hPa) 850 850 PFL **PFL** 925 925 BOTH SFC SFC /BC 1000 1000 2.0 3.6 1.6 2.4 2.8 3.2 0.3 1.5 0.6 0.9 1.2 1.8 RMSE_V (m/s) RMSE_T (C)

Forecast RMSE at 925 hPa over D03





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- Compared with PFL, SFC can improve the fog forecast with increasing the ETS by 6.6%. This is because that SFC has better near-surface wind and moisture fields and more stable stratification. It indicates that data with higher horizontal density and more variables at the surface (surface obs) has more impact on fog forecast than that of obs with only wind information but high vertical resolution (profiler).
- ✓ Compared with SFC, BOTH made the fog forecast significantly improved with 13.6 % increase of ETS. This is because that the better U/V vertical structure at PBL leads to better dynamically balanced fields of U/V, T and Q. It indicates that the combination of both surface data and profiler data are extremely necessary to improve the skill of fog forecast for this case.



- Conduct experiments withholding T and Qv to examine the impact of wind vs T and Qv in surface obs on fog forecast
- Examine the impact of obs densities, obs distributions and the DA cycling frequency on fog forecast
- Conduct real data experiments

Thank You



Extra Material

40-member ensemble simulation with 48-h forecast:
 0000 UTC 20 — 0000 UTC 22 Feb 2007

 initial ensemble generated with WRF-3DVar randomly sampling the background error covariance from a fixed covariance model

• The standard deviations of the initial ensemble

0.3 g/kg for water vapor mixing ratio

3 m/s for wind

I.2 K for air temperature

								PRO	JFL_U	UHF:	
Obs = truth + random(0,obs_err)						(PB	(PBL profiler)				
							h(l	h(I)= elev + 100 m			
 Obs_err: Surface obs: SYNOP, METAR, AVVS h(x)=h(1) + k * 60 						* 60 m					
u/v= l r	n/s, T		egree,	Q =	l g/kg	5, P	= I Pa	h(x) < 30 (*	000 n ~700 Ի	n 1Pa)
Pressure (hPa)	1000	925	850	700	500	400	300	250	200	150	100
Wind (m/s)											
SOUND	1.1	1.1	1.1	1.4	2.3	2.8	3.3	3.3	3.3	3.0	2.7
PROFL	1.0	1.0	1.2	1.2	/	/	/	/	/	/	/
Т(К)	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
Q(g/kg)	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0
P(Pa)	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0

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Experimental design

Dom	D01: 159*153 (27 km) D02: 232*214 (9 km) D03: 448*343 (3 km)		
e_vert	40 levels (7 levels below 1 km)		
p_top	50 hPa		
Microphysics	WSM 6-class graupel		
Longwave	RRTM		
Shortwave	Dudhia		
Surface layer	QNSE surface layer		
Land surface	Thermal diffusion scheme		
Boundary layer	QNSE		
Cumulus	Kain-Fritsch (only for D01 and D02)		



- LANDUSE data: Beijing_30s (D03)
- NCEP FNL data: 1deg*1deg
 WRF V3.3.1







