

Numerical weather forecast and Digital forecast at KMA

Young-Youn Park
Numerical Prediction Center/
Korea Meteorological Administration,
currently visiting ECMWF

With contributions from

SangWon Joo, SunOk Mun, Young-Kyoung Seo

Outline

- Overview of KMA's NWP Systems
- Medium-range forecast
 - Recent Changes in KMA's EPS
 - TIGGE : Multi-model test
- Digital Forecast System



KMA Digital Forecast

KMA Digital Forecast

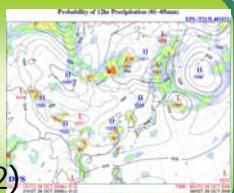
Data Acquisition & Q.C

NWP system

TIGGE

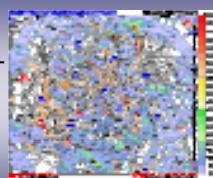
EPS

(T213L40,
16members,
10days,00/12)

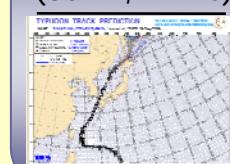


GDAPS

(3dVar, FGAT
T426/L40,
10days)



Typhoon Model (30km, 72hrs)



GWAM

(wave model,
1.25°,
10days)

GDLM

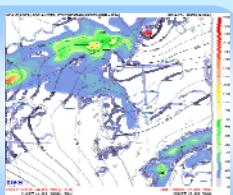
(statistical
model
,10days)

Digital Forecast

TMN
TMX
PoP
PTY
SKY

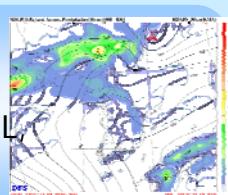
KWRF

(3dVar
10km/30L,
60hrs)



RDAPS

(Downscale,
FDDA, 30km/33L
66hrs)



RWAM (0.25°,66hrs)



RTSM

(1/12°,48hrs)

Temp, POP (statistical model)

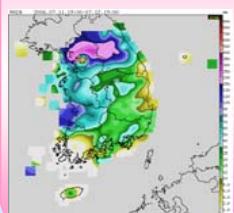
Digital Forecast

MIN/MAX T
PoP
3Hr T
RH

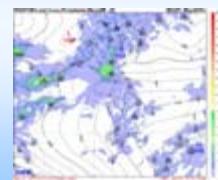
Wave Height
Wind S/D
Prep. Typ.
Sky Cond.
Rain/Snow

On-Going Project

Storm-Scale Model



RDAPS (3dVar,1day 10km/33L)



KWAM

(1/12°, 66hrs)

KLAPS

(15/5km, 18hrs)

RDAPS

(Downscale,
5km/33L, 1day)

- 1~5Km grid model
Radar assimilation
Intensify model physics

EPS (*Ensemble Prediction System*)

	GBEPS 1.1.1 ~ GBEPS 1.2.1	GBEPS 2.1.1 ~ GBEPS 2.3.1	GBEPS 3.3.1
Operation period	2001.3.1 ~	2003.11.1 ~	From 2006.7.
Data assimilation	2dOI → 3dOI	3dOI → 3dVar	3dVar
Model	GDAPS T106L21	GDAPS T106L30	GDAPS T213L40
Vertical resolution	21 levels	30 levels	40 levels
Perturbation method	Breeding	Breeding + Factor Rotation	Breeding + Factor Rotation
Target area (BV)	Global	Northern Hemisphere	Northern Hemisphere
Run per days	1 (12UTC)	1 (12UTC)	2 (00, 12UTC)
Lead time	10 days	8 days	10 days
Ensemble members	16 (16 members + 1 control)		(16+1)*2 members

Recent Changes

Increasing resolution and ensemble size

- T106L30M16 → T213L40M32
- once a day (12Z) → twice a day (00Z, 12Z)

Statistical Post-Processing : Bias Correction

- Decaying averaging bias estimate (from Bo cui)

EPS does not produce enough spread

- Bred vectors seem to have similarity

→ Introduce factor rotation

- Growth rate of perturbation is small

→ Test stochastic perturbation

Factor Rotation

■ What is the Factor analysis?

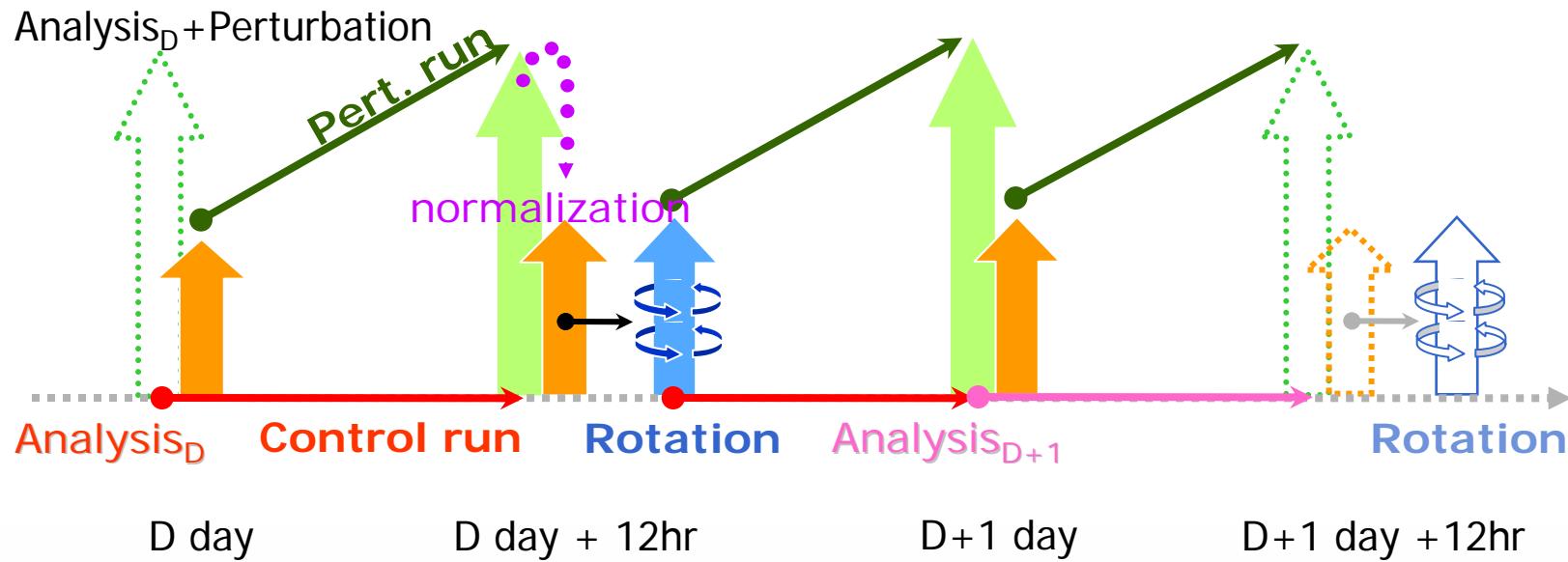
- Use relationship among variables
- Find the structure of the relationship and
- Tie to similar variables → Factors

■ How to produce perturbation with factors?

- Rotate the factors obtained from factor analysis
- With keeping orthogonal feature among factors
- ⇒ New perturbations

Breeding+Factor Rotation

Breeding + Rotation



Bias Corr. & Stochastic Pert.

Bias Correction

- Motivation: strong + bias over east Asia
- Method: Decaying averaging bias estimation (Bo Cui)

Stochastic Perturbation

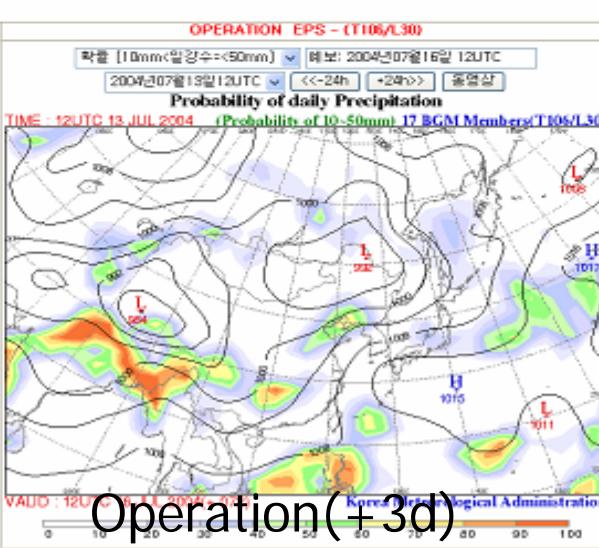
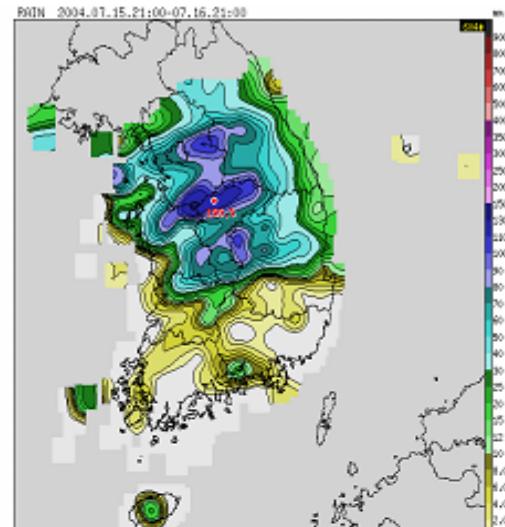
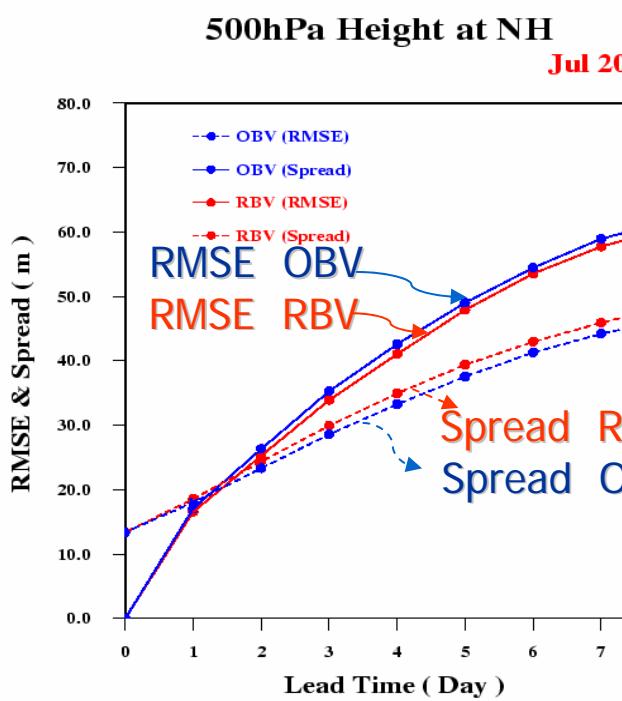
- Estimate error of model physics from background error covariance used in 3dVar.
 - random (stochastic) forcing << model error
- Apply stochastic perturbation to tendency every hour



KMA

Digital Forecast

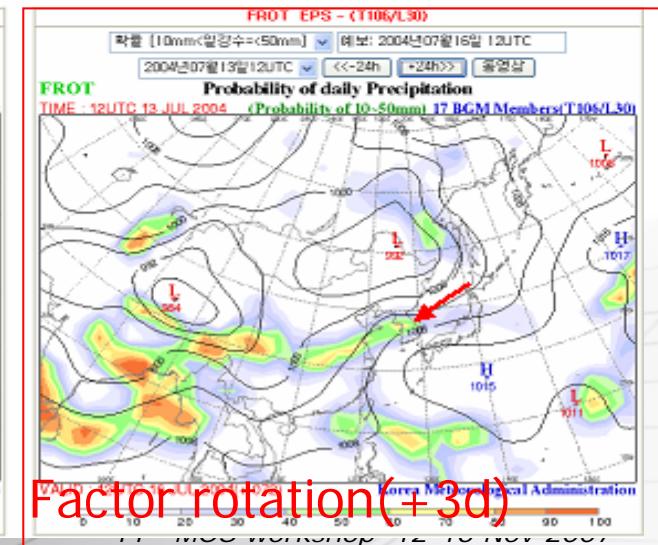
Impact of Factor rotation



Operation(+3d)

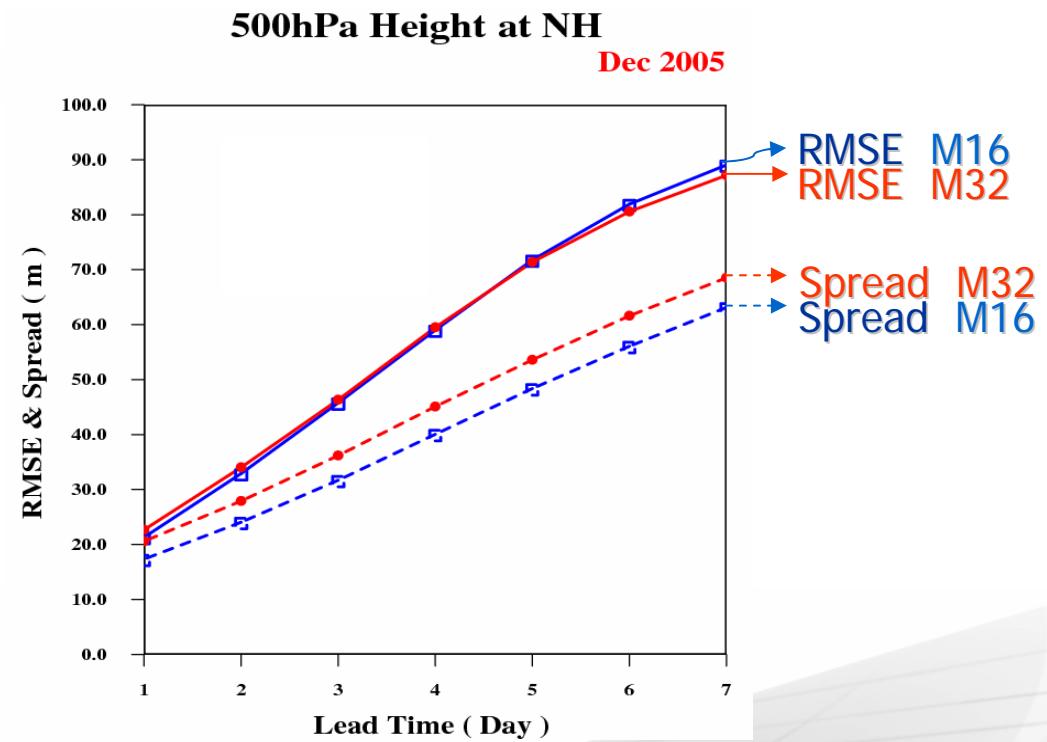
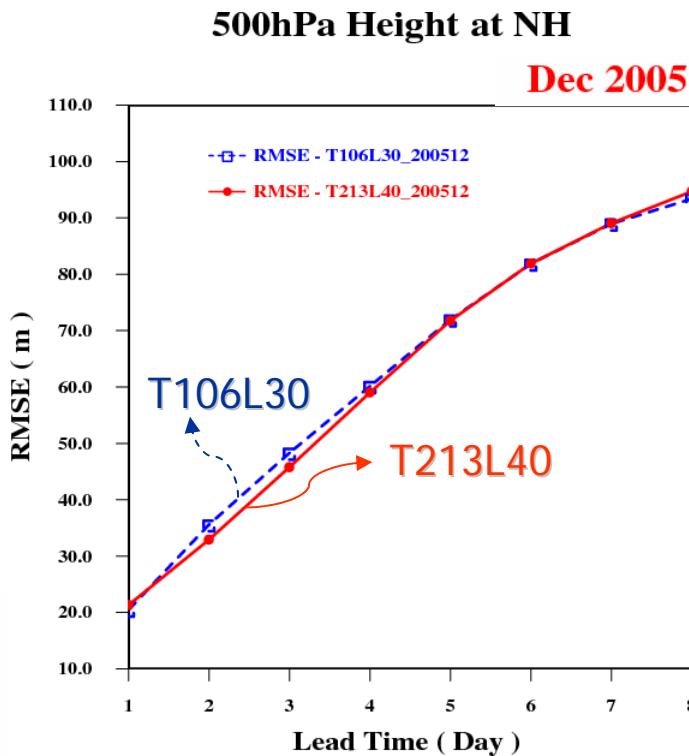
IC: 2004 July 13 12UTC

PoP : July 16 12UTC (+3d)
After factor rotation,
probability of precipitation
successfully reproduces the
heavy rainfall as observed.

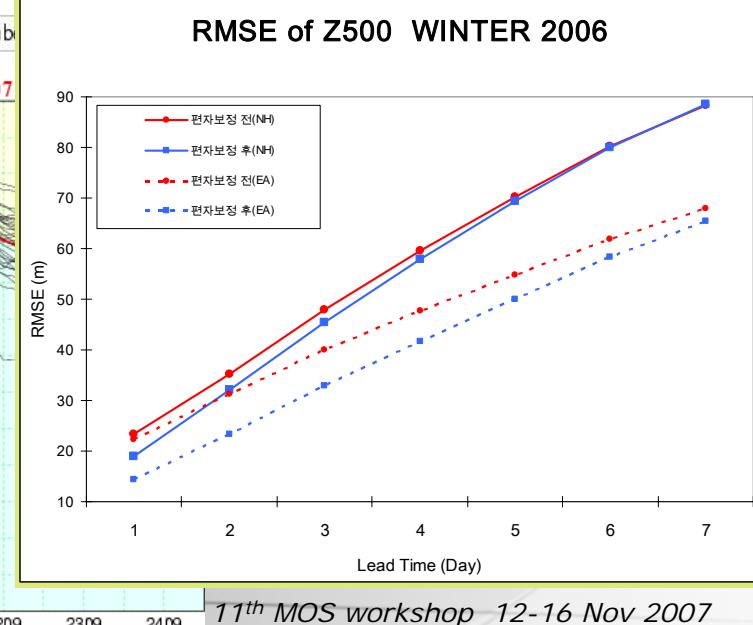
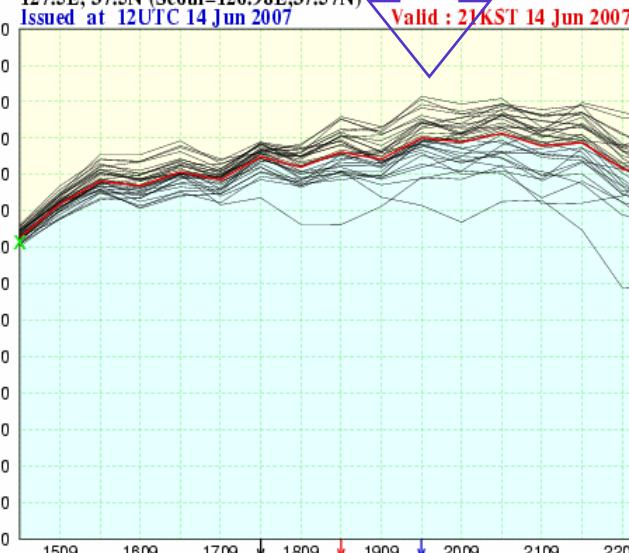
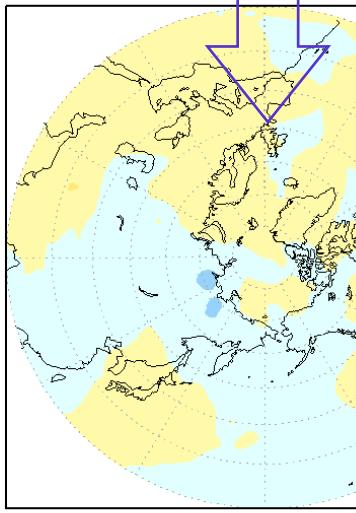
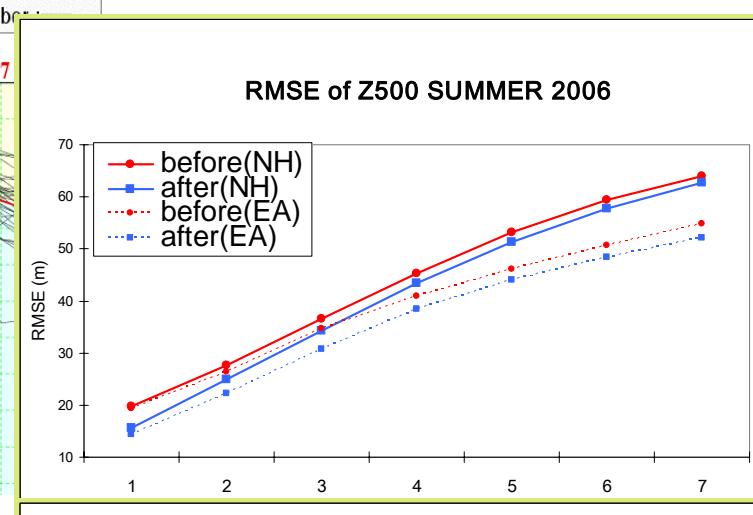
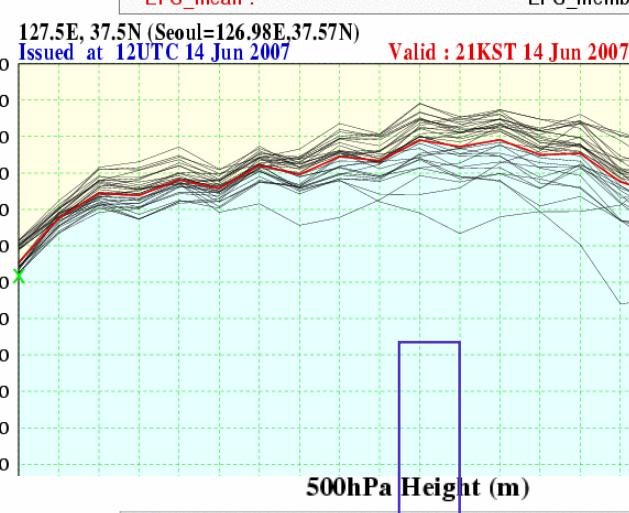
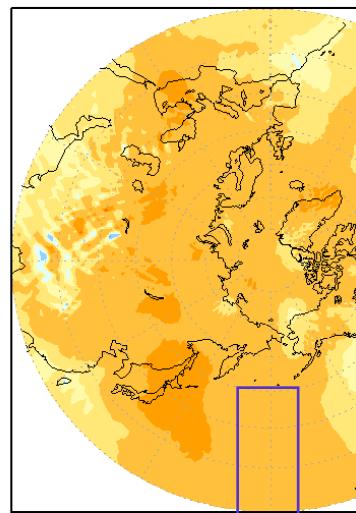


Factor rotation(+3d)

Impact of Resolution & Membership



Impact of Bias Correction



TIGGE: Multi-model test

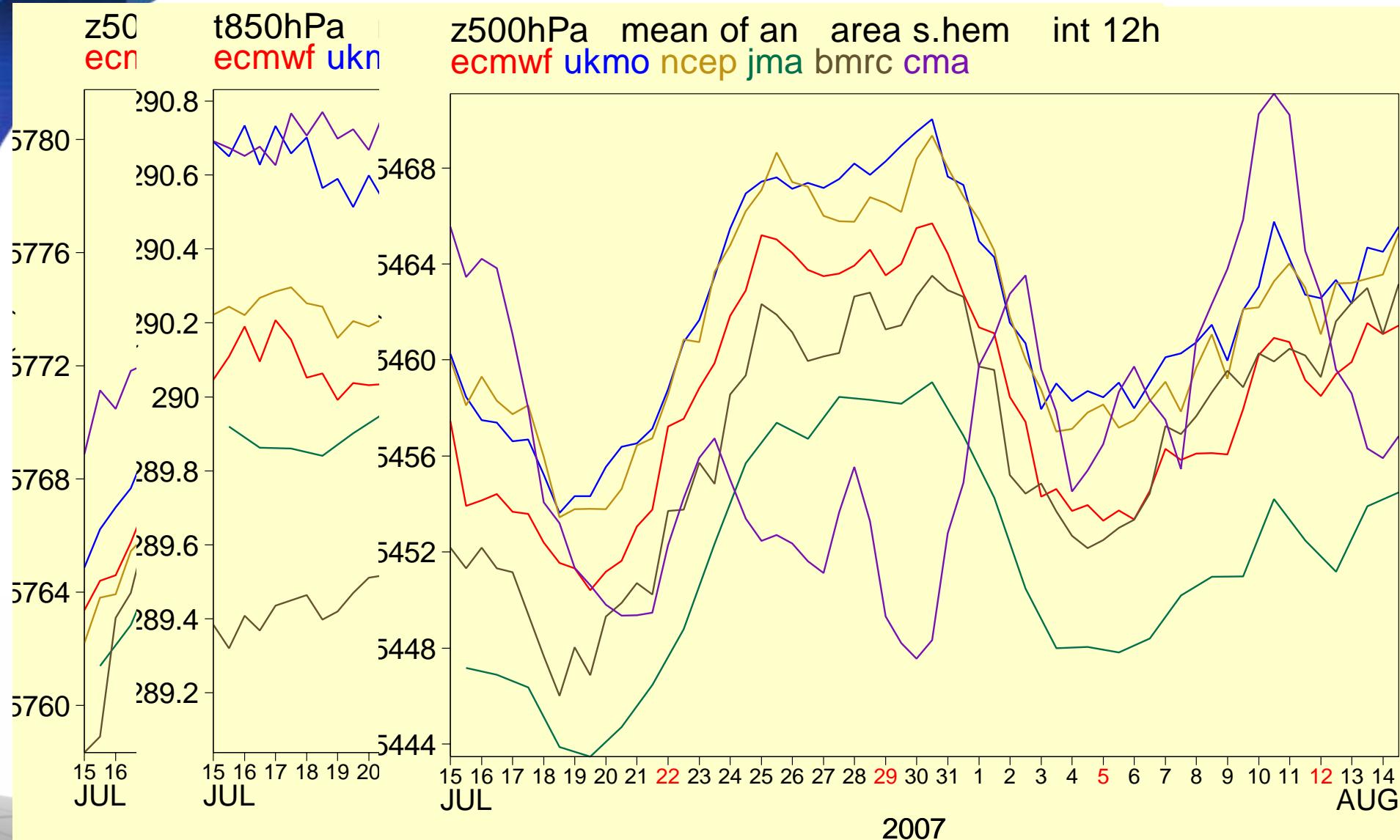
Many Questions/Choices

- Which analysis to use
- How to combine ?
 - Use all the centers/members or only good centers/members?
 - Simple or weighted combine → How to determine the weight ?
 - Using calibration? → How to calibrate?

Multi-model ensemble test

- Combine test period: 8 Jun – 1 Sep (84 cases)
- Centres: ECMWF, UKMO, JMA, CMA
- Method:
 - Simply combine all the members
 - With/without bias-correction (use previous 30 days)
 - Calculate bias for each centres, cf/pf separately

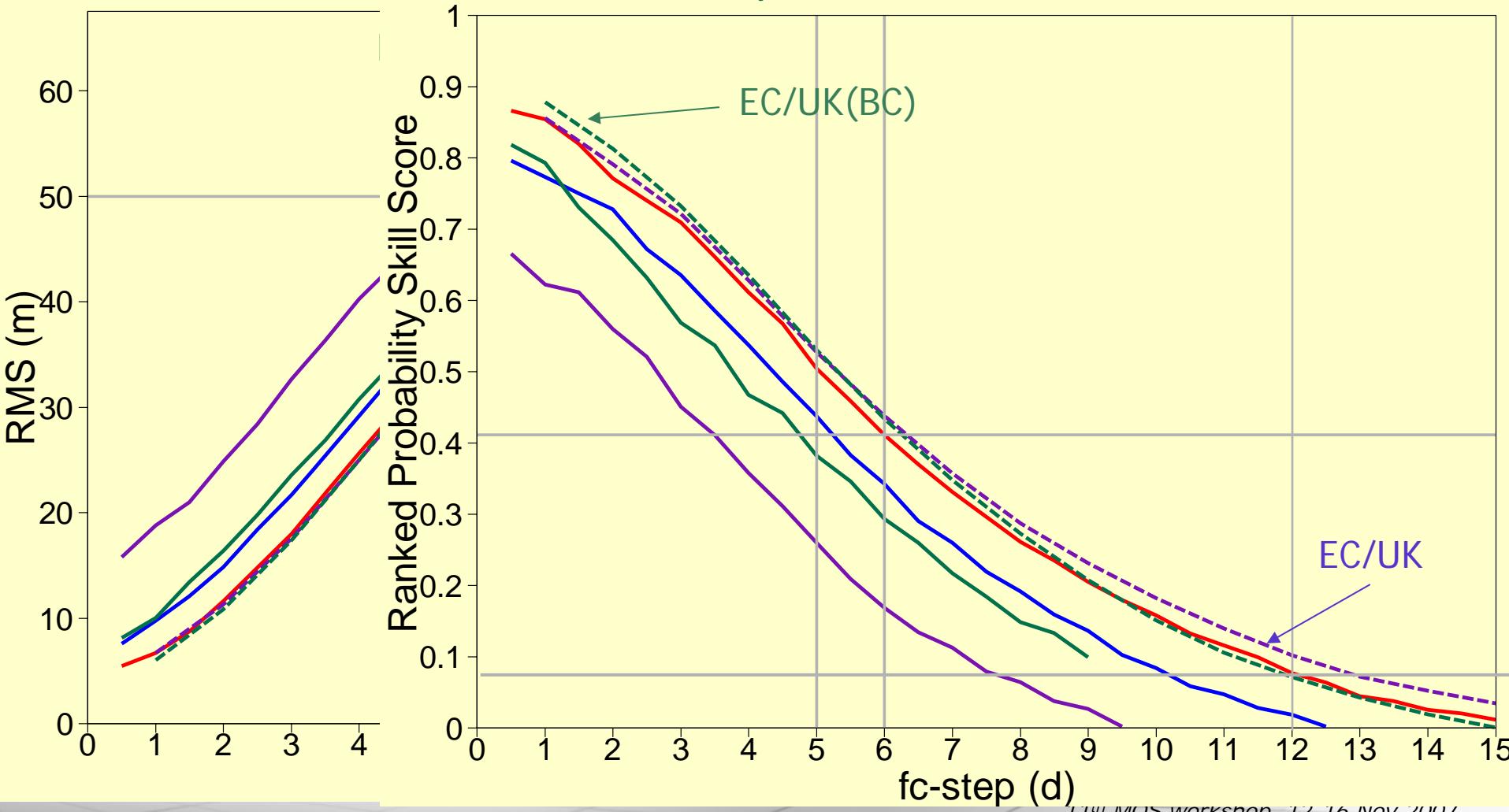
Which an ? (Area mean of an)



Raw/Cal data ? (RMSE/RPSS : Z500 NH)

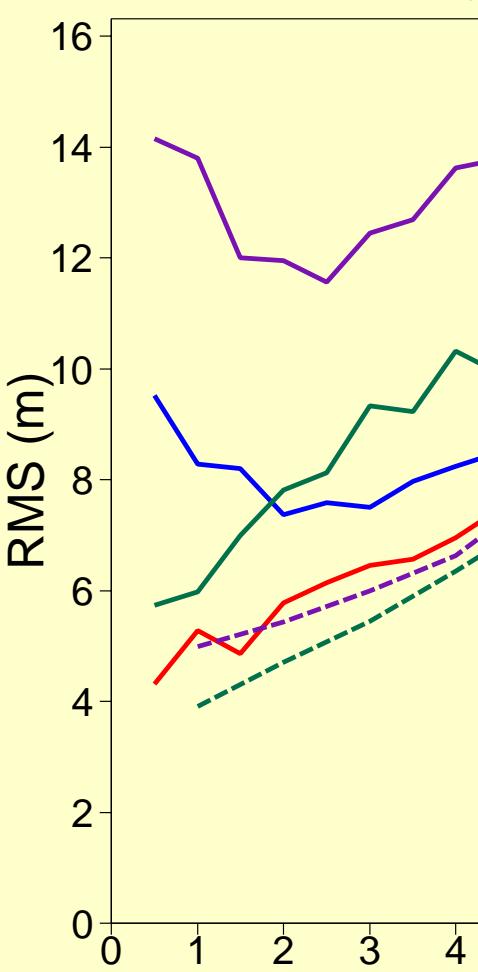
z at 500hPa 12 UTC
cases 20070608-20
ecmwf ukmo cma j

z at 500hPa (ecmwf_as_an)
10 categories, cases 20070608-20070901_N84, area n.hem
ecmwf ukmo cma jma eu eu_bc

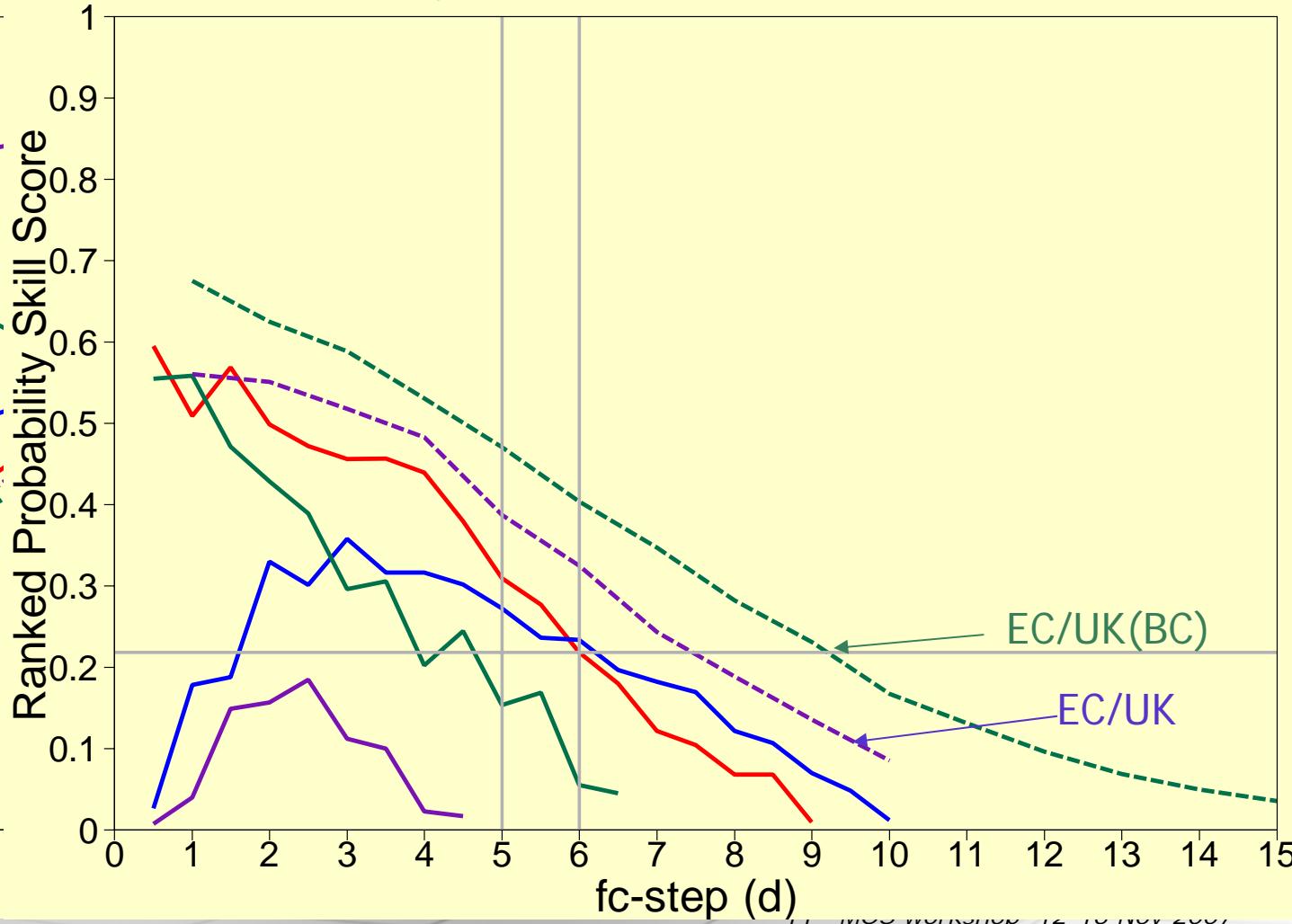


Raw/Cal data ? (RMSE/RPSS : Z500 TR)

z at 500hPa 12 UTC
cases 20070608-20
ecmwf ukmo cma j



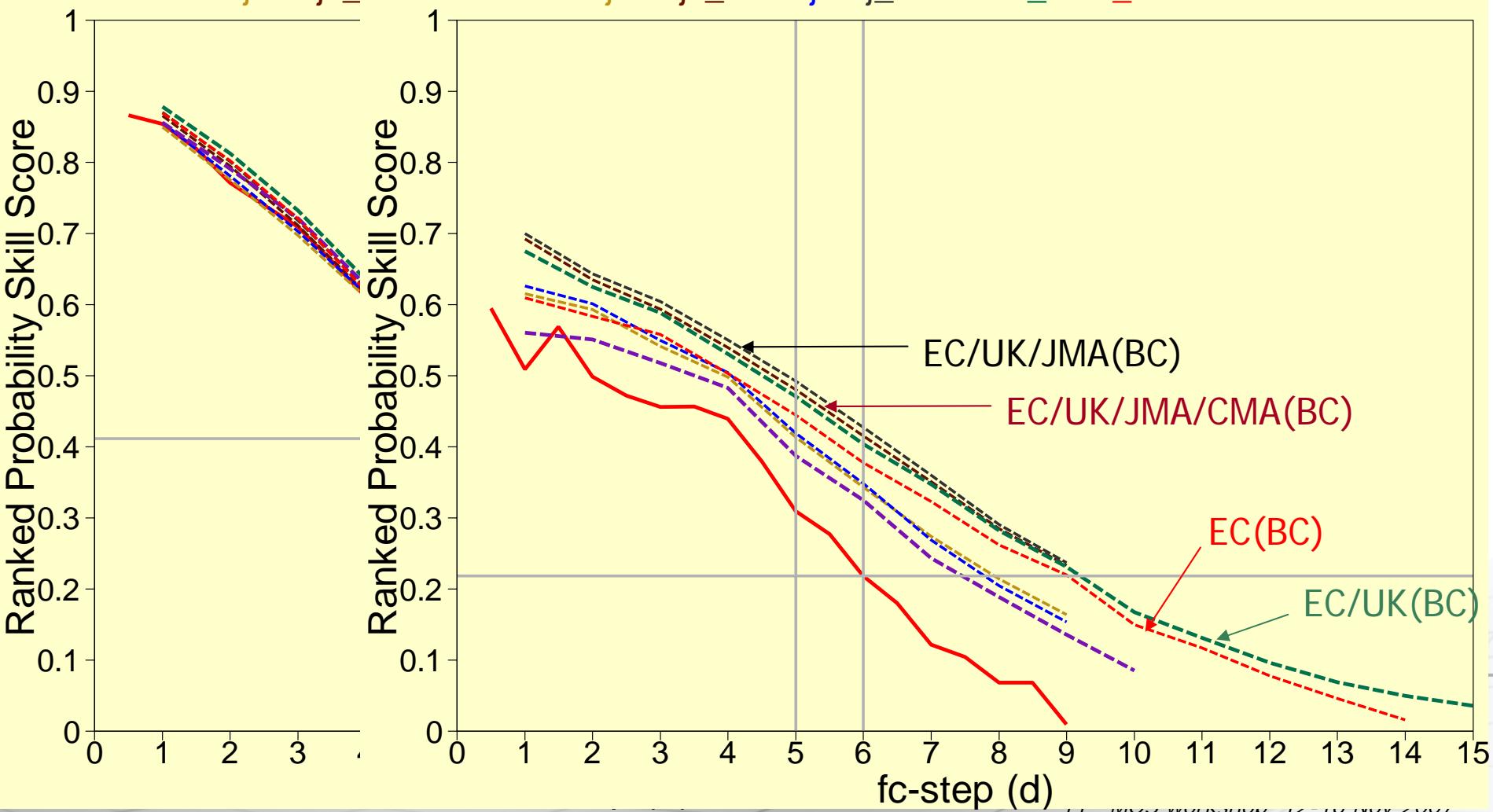
z at 500hPa (ecmwf_as_an)
10 categories, cases 20070608-20070901_N84, area tropics
ecmwf ukmo cma jma eu eu_bc



Which centers ? (RPSS – Z500 NH/TR)

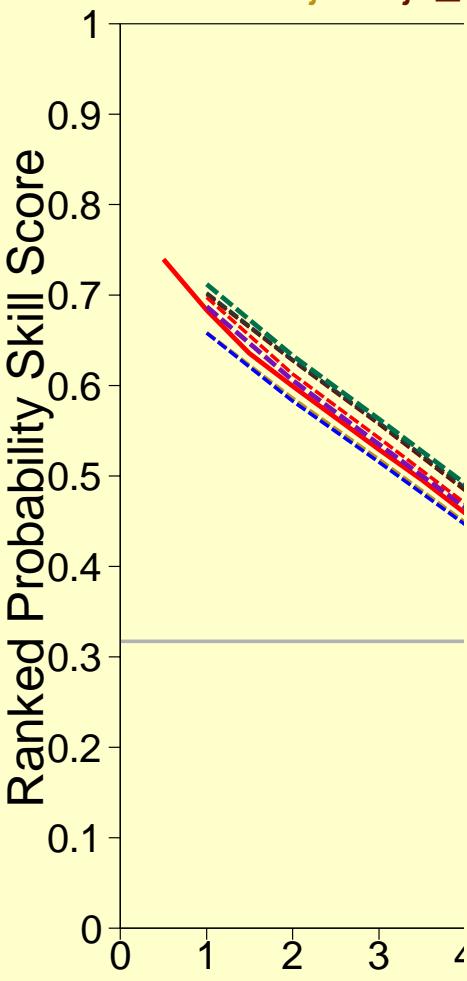
z at 500hPa (ecm
10 categories, cas
ecmwf eujc eujc_

z at 500hPa (ecmwf_as_an)
10 categories, cases 20070608-20070901_N84, area tropics
ecmwf eujc eujc_bc euj euj_bc eu eu_bc e_bc

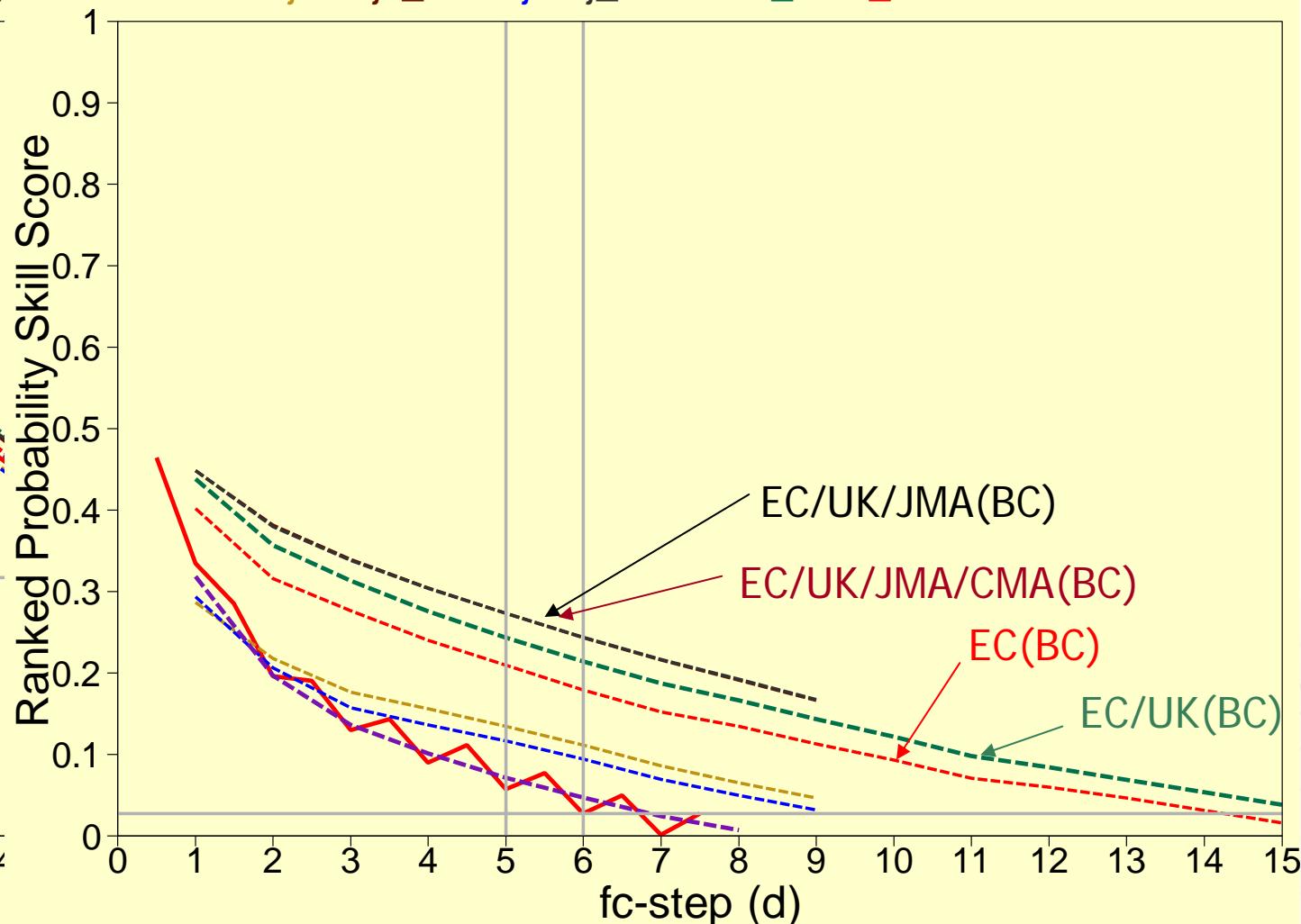


Sensitivity to parameters ?

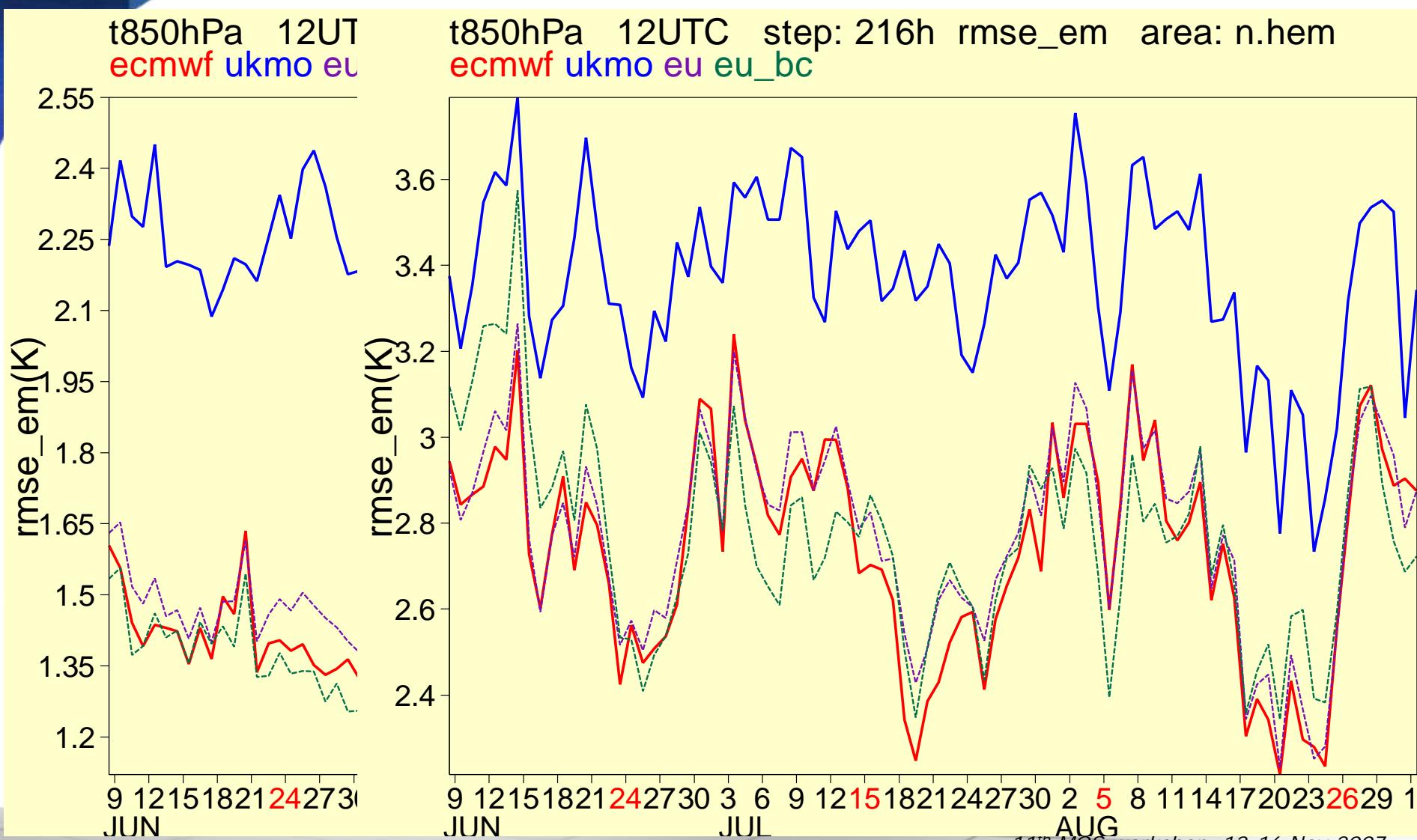
z at 500hPa (ecmwf)
 t at 850hPa (ecmwf)
 10 categories, cases
ecmwf eujc eujc_bc



z at 500hPa (ecmwf_as_an)
 t at 850hPa (ecmwf_as_an)
 10 categories, cases 20070608-20070901_N84, area tropics
ecmwf eujc eujc_bc eu_jc eu_jc_bc eu_bc e_bc



Daily RMSE – t850 +3d/+9d NH

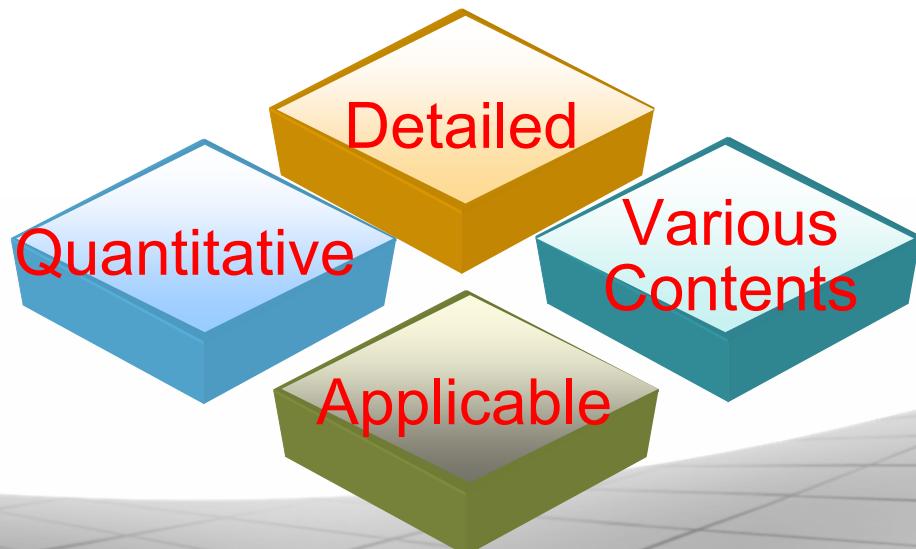


Summary of Preliminary Result

- More gains on probabilistic scores than deterministic scores
- AS, NA, NH: small gains, distinctive gains over tropics (where ecmwf shows under-dispersion)
- Difference btn parameters and areas
- Bc better in early period, but nbc better in the later period of forecast
- Larger benefit of multi-model over strong bias area
- Decision should be made on the bases of areas, parameters of interest
- Lower level parameter (T2m, precipitation) ?
- Advanced calibration ?
- Optimum weight for each centers?

Digital Forecast : Motivation

- Strong demand for detailed, quantitative meteorological services
 - Public does not want weather at station, but at their house
 - The use of meteorological data is too difficult
 - Private sector requires easy to handle data to reproduce new application information
 - Nwp output is used by only very limited area
- ← one of important barrier to develop application information





디지털 예보

KMA Digital Forecast

Current Forecast

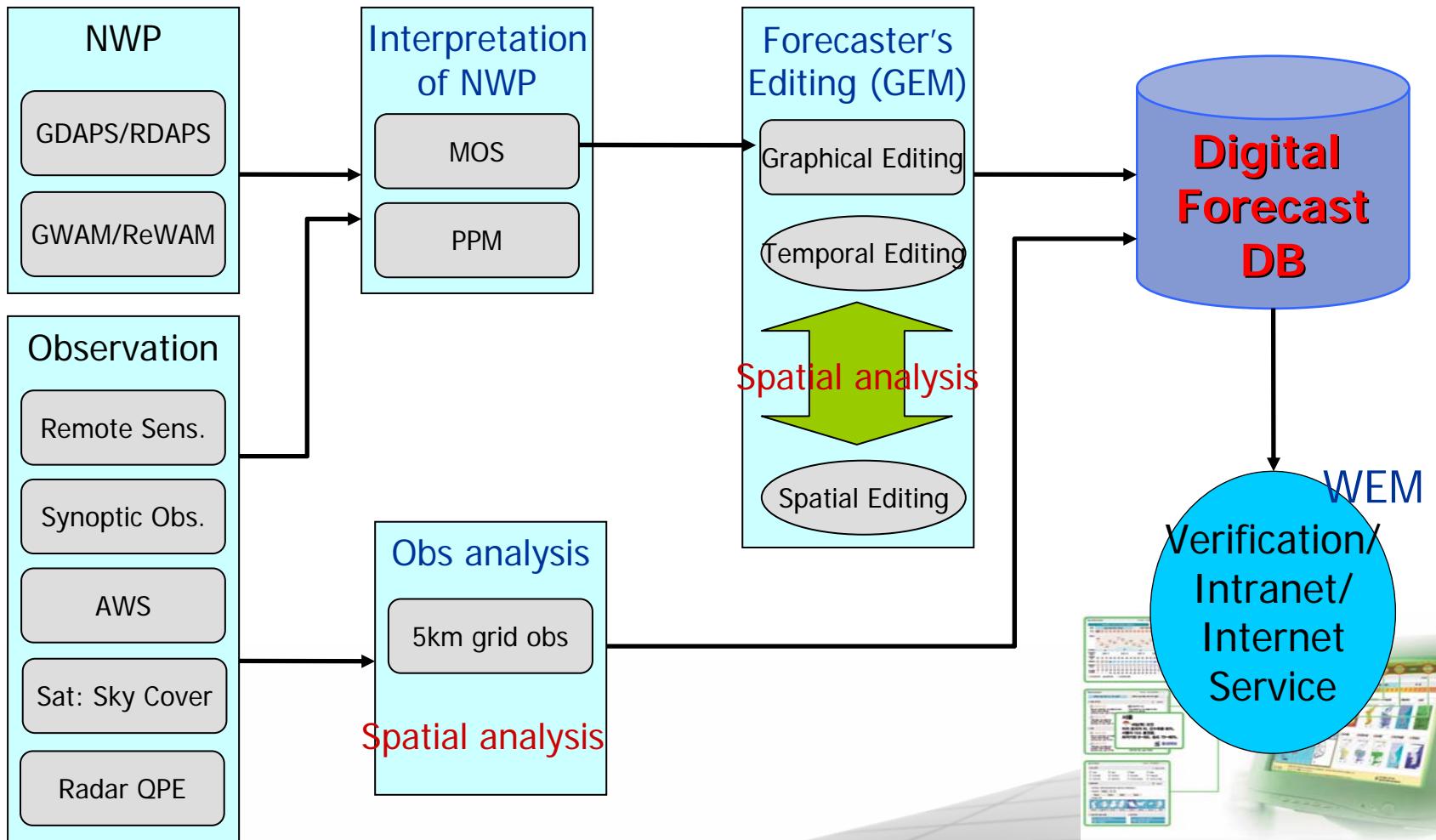


Change

Digital Forecast



Conceptual Structure



Status of Digital Forecast

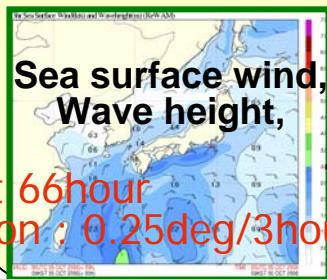
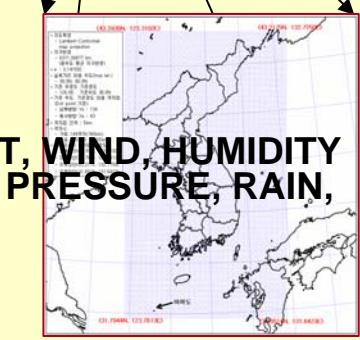
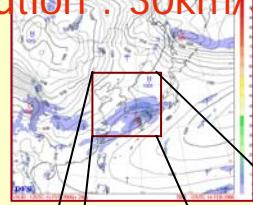
Digital Forecast range	Short range	Medium range
NWP Model	RDAPS	GDAPS
Forecast length/interval	48hours/3hours	10.5day/12hours
Resolution	5km	30km
Frequency (per day)	8	2
Forecast parameter	12	5
Grid/element transform module	11	5
Statistical analysis module (MOS)	MOS 8(/9)	developing
Spatial analysis module (2d-OI)	apply, improvement	not apply
Graphical editing module (GEM)	semi-operation, improvement	
Web service module (WEM)	semi-operation, improvement	
Verification system	operation, improvement	

Interpretation of Model Output : shortDFS

RDAPS:

Forecast 66hour

Resolution : 30km/3hour



ReWAM

Forecast 66hour

Resolution : 0.25deg/3hourly



MOS models
at 103 stations

$$\bar{Y}_t = F_t \left(\frac{\bar{X}_{00h} + \bar{X}_{03h} + \bar{X}_{06h}}{3} \right) + F_{climate}(T_t) + T_{obs}$$

3-hour/Max/Min Temp.,
Probability of Precip., RH,
Wind,
single-station eq.

Precip.

Sky cover, Precip. Type
Snow-depth
Generalized-operator eq.

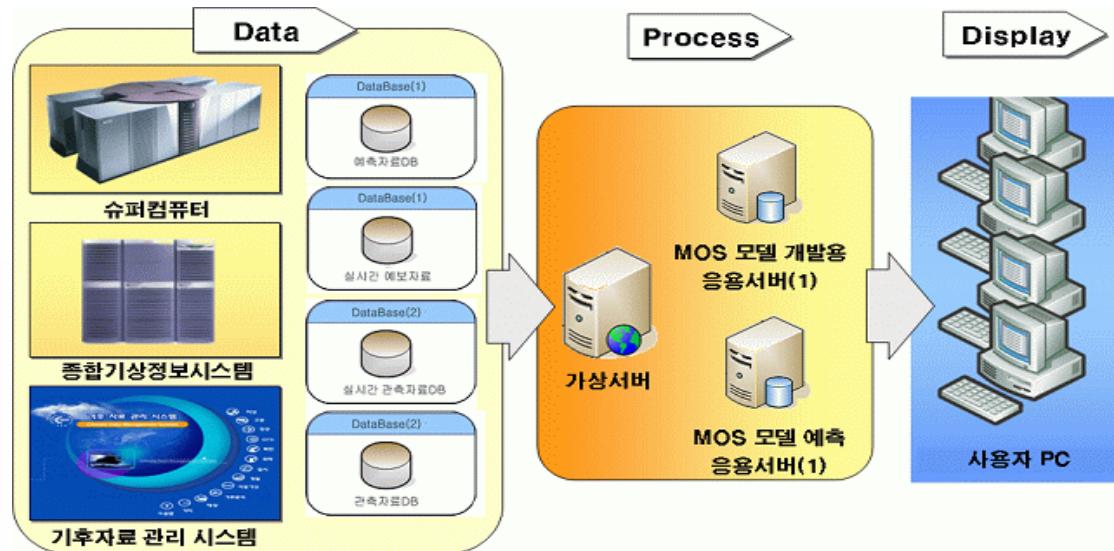
Effective wave height



- Combine NWP Model and statistical methods
- Generate 5 km-grid forecast, Predict every 3 hour up to 48 hours

MOS Tool

- Need tool that mange and produce MOS eq. efficiently according to the changes of NWP models
- Using this tool, make the indicator for MOS model
 - Speed up for developing and improving MOS eq.
 - Standardization of MOS model
 - Total system of development/verification/management of MOS
→ efficiency, accuracy, convenience



1. Sampling Module

표본구성

모델자료선택 KWRF

관측자료선택 SYNOP

관측자료선택 서울

일자선택 구간선택

변수선택

- TMP
- PRES
- MIXR
- RH
- UGRD

피예보인자 - Microsoft Internet Explorer
Mos Option Tools

관측자료
피 예보인자

2. Statistics Module

통계처리방식 - Microsoft Internet Explorer

통계처리 방식선택

회귀분석

Variable1

Forward

Criteria

P-IN P-OUT

MOS Tool

3. Display/GUI Module

MOS 자동 산출 도구 - Microsoft Internet Explorer

eMOS v1.0.0

표본구성

모델자료선택 KWRF
관측자료선택 SYNOP
관측자료선택 서울
일자선택 구간선택 Step1

변수선택

<input checked="" type="checkbox"/> TMP	(K)
<input checked="" type="checkbox"/> PRES	(Pa)
<input type="checkbox"/> MIXR	(Kg/Kg)
<input checked="" type="checkbox"/> RH	(%)
<input checked="" type="checkbox"/> UGRD	(m/s)

현재 예보인자

<input type="checkbox"/> SQ	(g/g)
<input type="checkbox"/> DQ	(g/g)
<input checked="" type="checkbox"/> RH	(%)
<input checked="" type="checkbox"/> RHM	(%)
<input checked="" type="checkbox"/> U	(m/s)

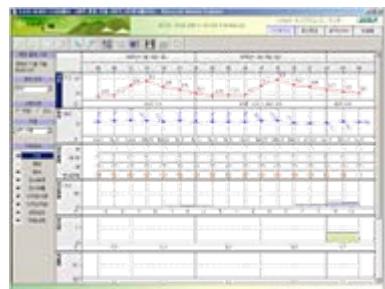
피 예보인자

<input checked="" type="checkbox"/> 3시간 기온	(K)
<input checked="" type="checkbox"/> 최고기온	(K)
<input type="checkbox"/> 최저기온	(K)
<input checked="" type="checkbox"/> 3시간 U	(m/s)
<input type="checkbox"/> 3시간 V	(m/s)

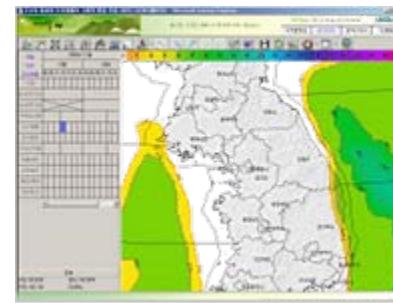
선택요약

- 모델자료 : KWRF
- 관측자료 : SYNOP
- 지점자료 : 서울
- 일자선택 : 20070909:20070910
- 구간선택 : UTC 0 / 3h / 84
- 변수선택 :
- 현재 예보인자
- 피 예보인자
- 통계처리환경

Forecaster's Editing (GEM)



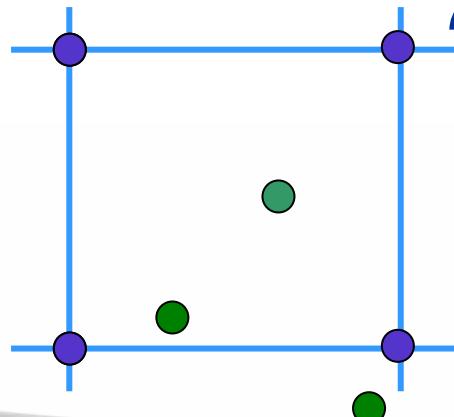
Time Series Editing



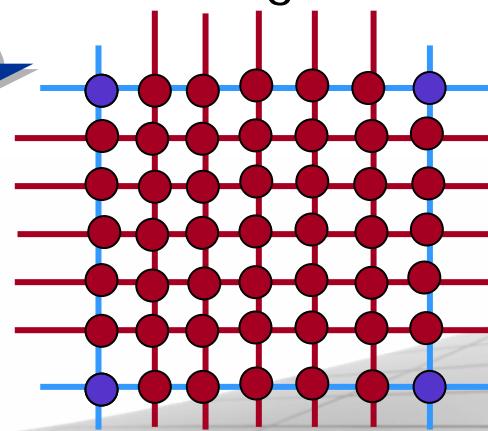
Spatial Editing



Issue Forecast



Objective
Analysis



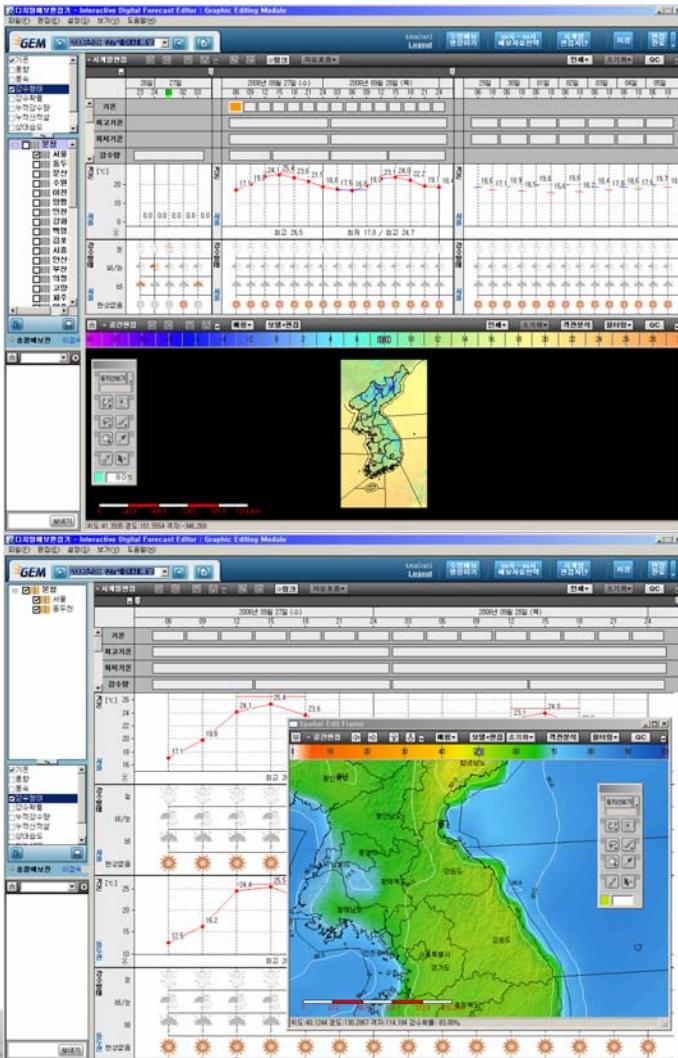


디지털 예보

KMA Digital Forecast

GEM

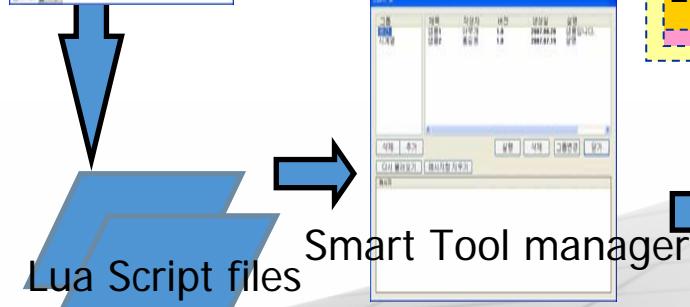
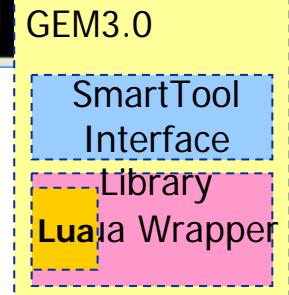
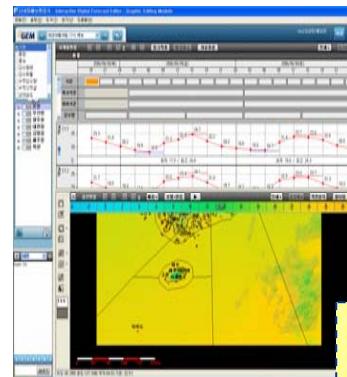
GEM Editing Window



Apply Smart Tool

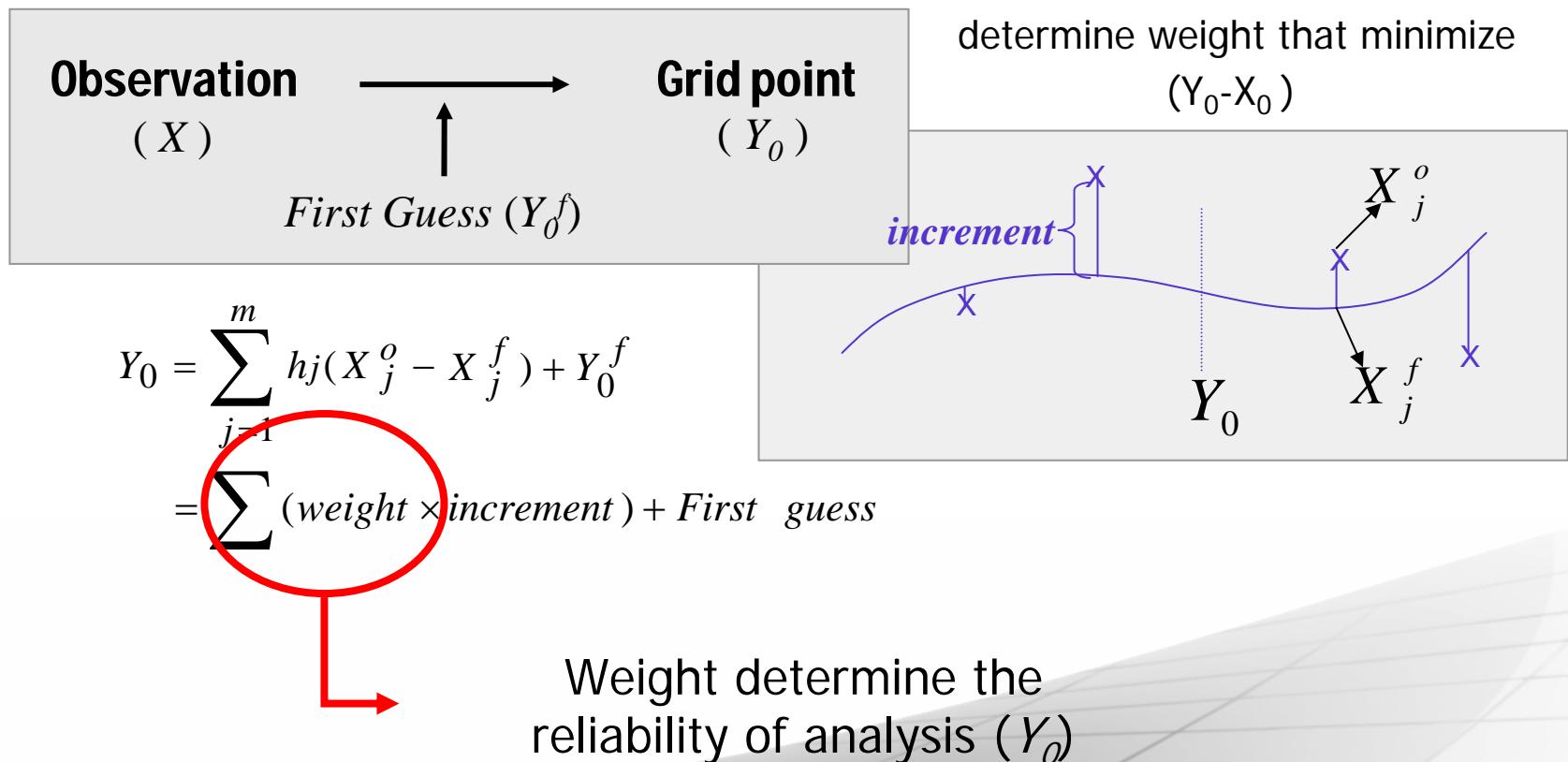


Lua Editor



Spatial analysis module (2d-OI)

2 dimensional-Optimal Interpolation



Weight in 2d-OI

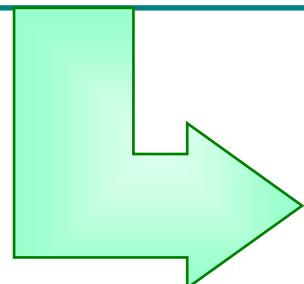
Weight (h) : 2dOI

$$\sum_{j=1}^J (\mu_{ij}^P + \mu_{ij}^0 \lambda_i \lambda_j) H_j = \mu_{ig}^P \quad \lambda = \frac{\sigma_o}{\sigma_b}$$

i,j : station g: digital grid

μ_{ij}^P : BE covariance μ_{ij}^0 : OE covariance

μ_{ig}^P : BE covariance between grid & station



assume OE covariance ~ 0
 (Forecaster editing \sim true)

Weight (h): DFS 2dOI

$$\sum_{j=1}^N (\mu_{ij}^P) H_j = \mu_{ig}^P$$

$$\begin{bmatrix} h_1 \\ h_2 \\ \vdots \\ \vdots \\ h_m \end{bmatrix} = \begin{bmatrix} \sigma_{11}, \sigma_{12}, \dots, \sigma_{1m} \\ \sigma_{21}, \sigma_{22}, \dots, \cdot \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ \sigma_{m1}, \dots, \sigma_{mm} \end{bmatrix}^{-1} \begin{bmatrix} \sigma_{01} \\ \sigma_{02} \\ \cdot \\ \cdot \\ \cdot \\ \sigma_{0m} \end{bmatrix}$$

Weight co-variance of obs to obs co-variance of obs to grid
 (fixed) (alter)

Correlation between obs stations

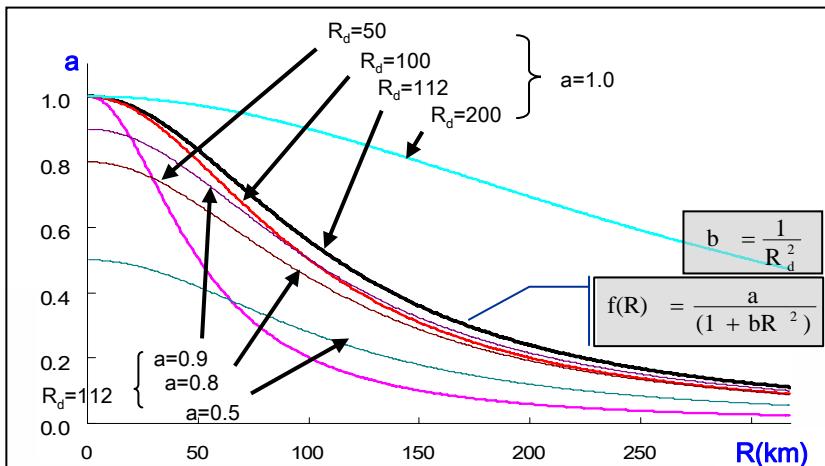
OLD

$$\mu_{ij}^P = \frac{a}{1 + br^2}$$

r : distance

a : error ratio (fixed, 1)

b : error cov wrt r (fixed, 1/1600)



Lorentz's type function fitting

NEW

Use BE cov between stations

$$\rho_{ij} = \frac{\sigma_{ij}}{\sqrt{\sigma_{ii}\sigma_{jj}}} \quad \sigma_{ij} : \text{covariance}$$

- Construct full matrix
 - isotropic \Rightarrow anisotropic
 - homogeneous \Rightarrow inhomogeneous

$$\mu_{ij}^P = \begin{pmatrix} \rho_{11}, \rho_{12}, \dots, \rho_{1m} \\ \rho_{21}, \rho_{22}, \dots, \cdot \\ \vdots & \ddots & \ddots \\ \rho_{m1}, \dots, \dots, \rho_{mm} \end{pmatrix}_{(75 \times 75)}$$

Correlation between grid and station

OLD

$$\mu_{ig}^P = \frac{a}{1 + br^2}$$

r : distance

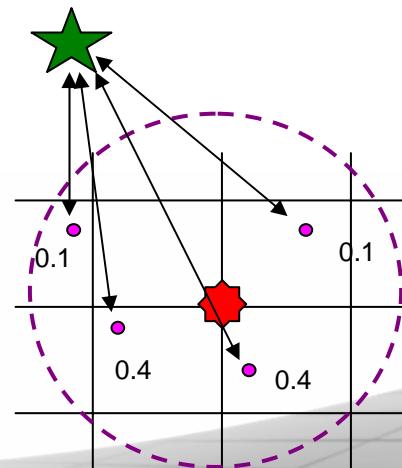
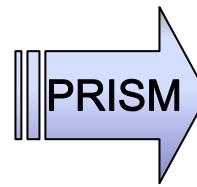
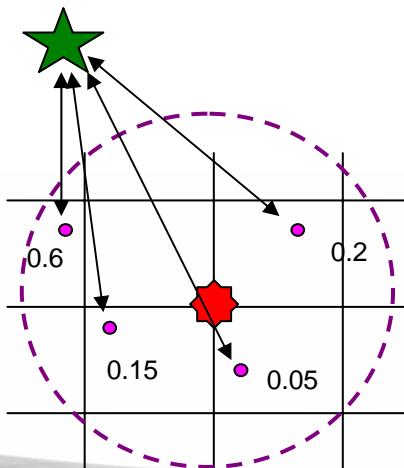
a : error ratio (fixed, 1)

b : error cov wrt r (fixed, 1/1600)

NEW

assume: correlation of spatial error
~ correlation of geographic information

1. Use BEC between stations
2. Use BEC between station and AWS
3. Use grid GIS (PRISM : weight wrt distance, altitude, slope, ,)



$$\mu_{ig}^P = \frac{\sum_{k=1}^K \omega_{gk} \rho_{ki}^P}{\sum_{k=1}^K \omega_{gk}}$$

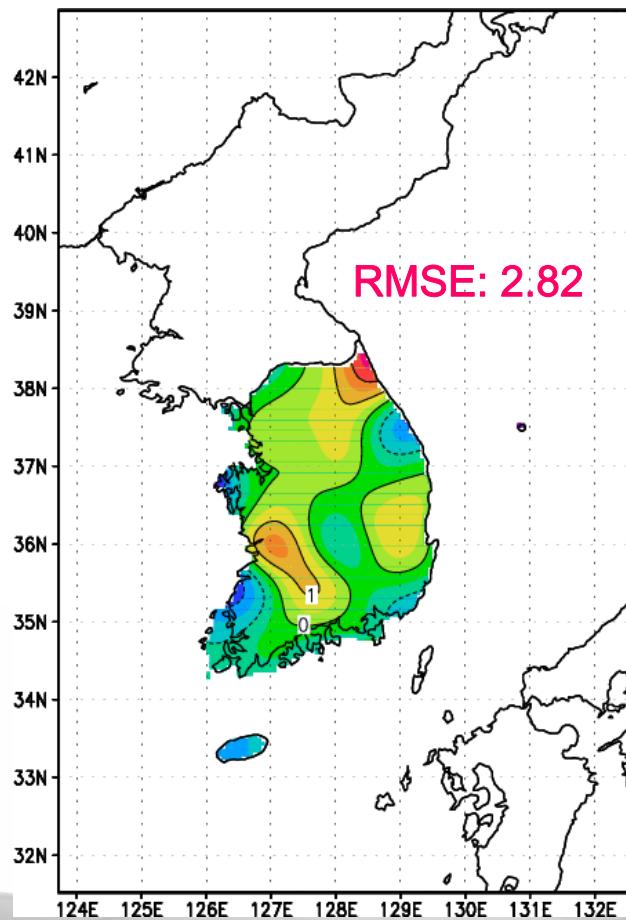
ρ_{ik}^P : BEC of neighboring station

ω_{gk}^P : weight of neighboring stations

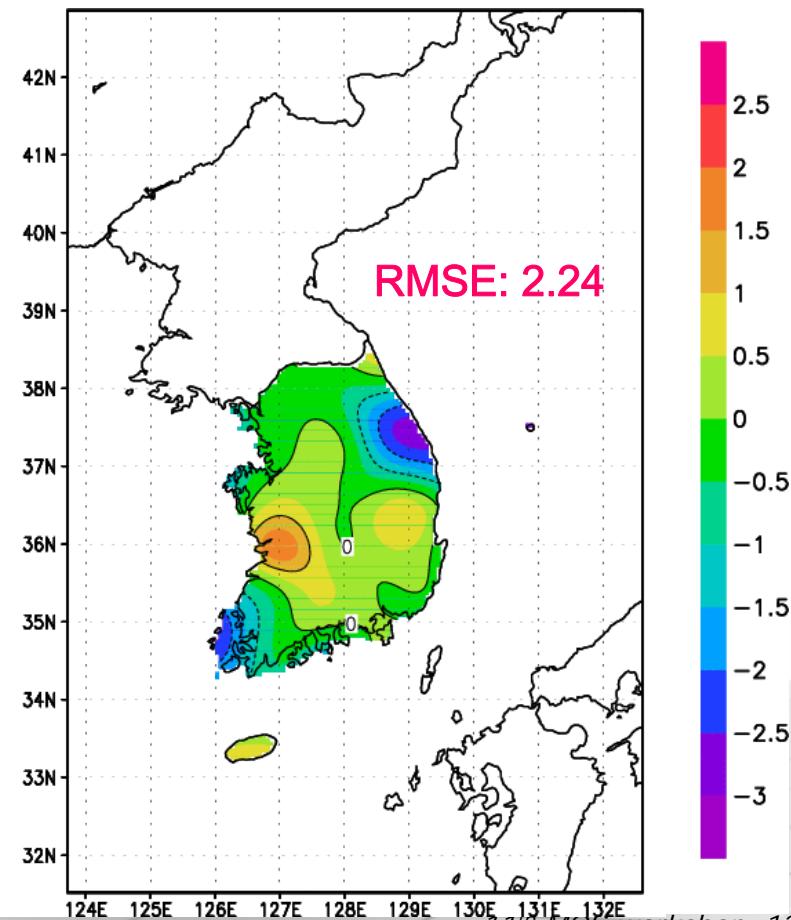
Impact: ERROR

Temperature (06UTC 5 OCT 2007)

OLD - OBS



NEW - OBS

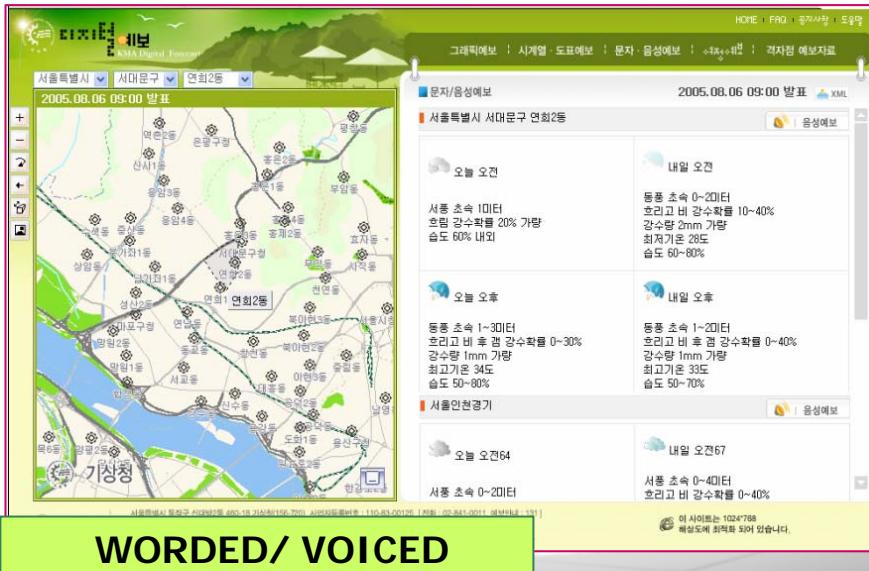
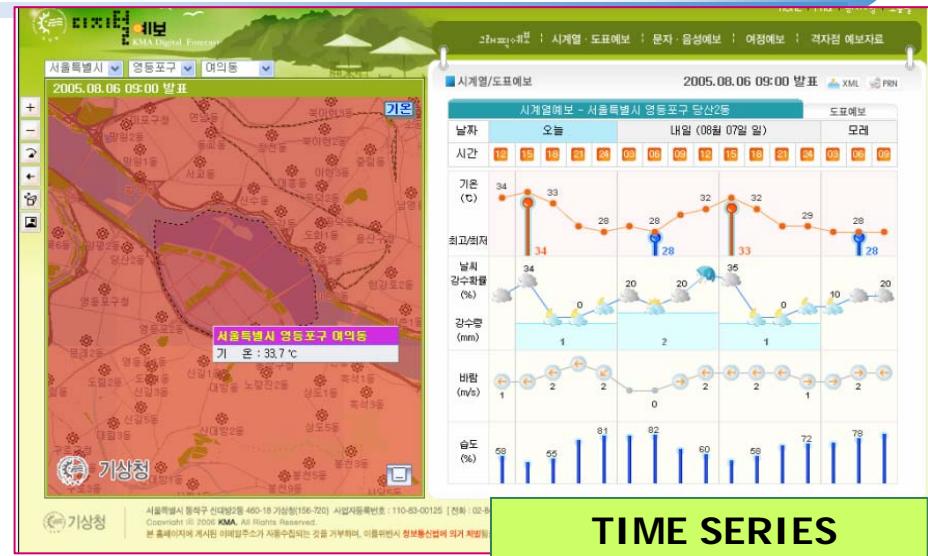
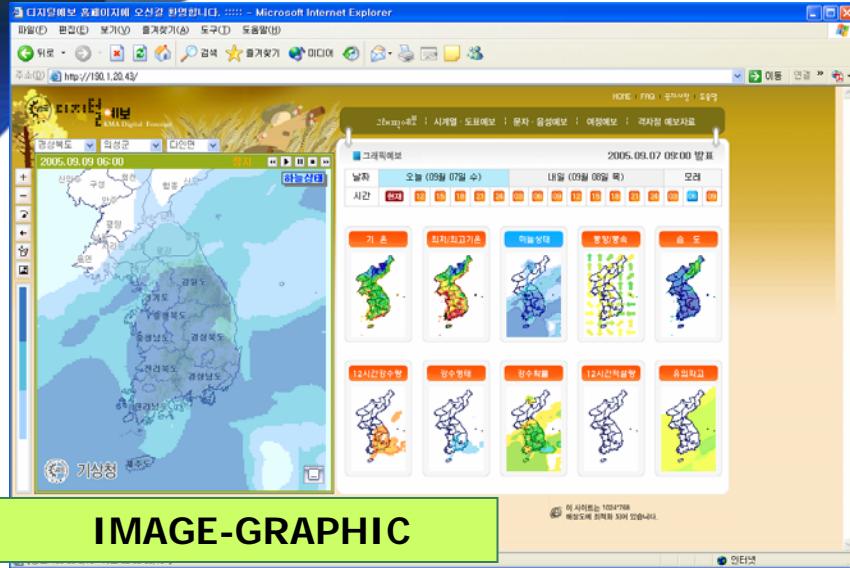




디지털 예보

KMA Digital Forecast

WEB SERVICE MODULE (WEM)

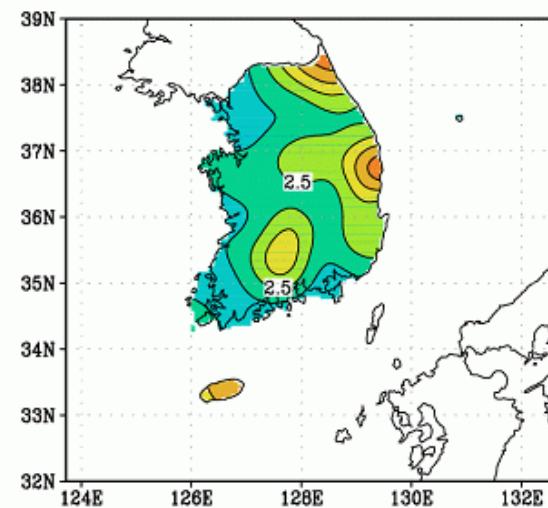
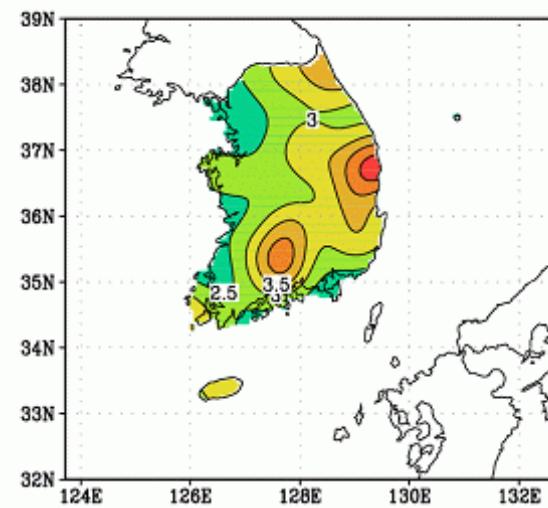
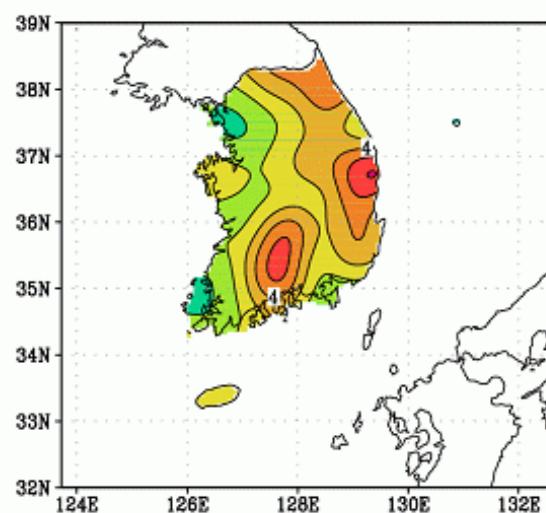


Accuracy : Short range DFS

DEC 2005 - SEP 2006

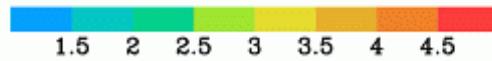
Verification items	FCST	MOS	NWP	IMPROVEMENT % (NWP vs.FCST)	Remarks
RMSE TMP(C)	2.1	2.0	2.7	22.2	
RMSE TMAX(C)	2.4	2.4	2.9	17.2	
RMSE TMIN(C)	2.4	2.4	3.0	20.0	
RMSE PPTN(mm)	10.5	-	11.7	11.1	
ACC PPTN IDF	0.86	-	0.87	1.5	
RMSE WS (m/s)	2.8	-	2.9	3.4	
RF WD (deg)	0.45	-	0.44	2.27	
ACC2 SKY	0.77	-	0.74	4.05	Comparison with PPM
RMSE RH (%)	14.0	12.4	18.0	22.2	MOS since March 2006

Distribution of RMSE



WINTER

Mountainous area large error



SPRING

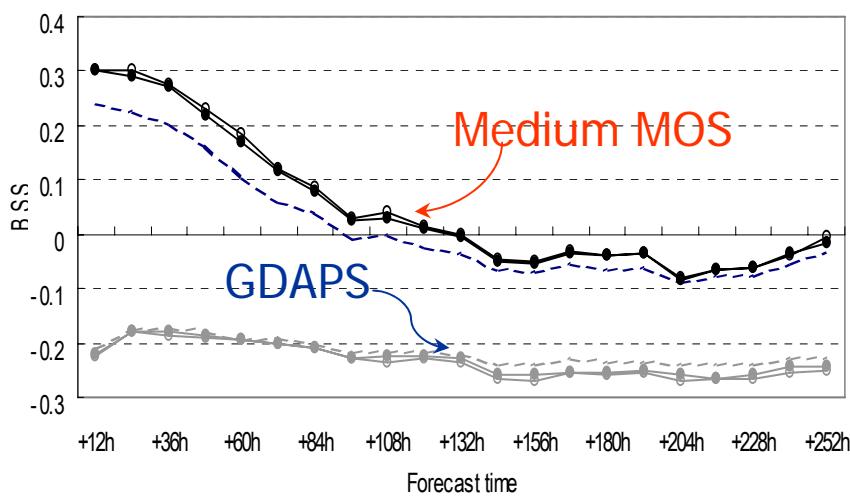
Summer better than winter



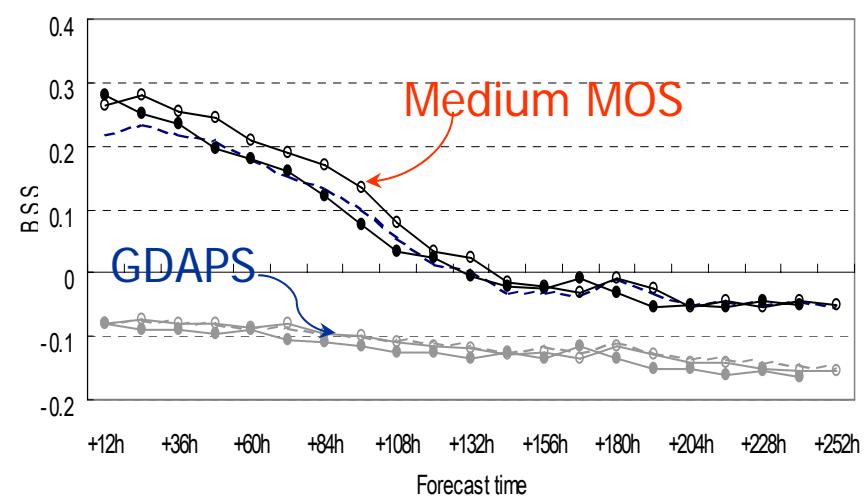
SUMMER

Accuracy: Medium range DFS

BSS (Warm Season)



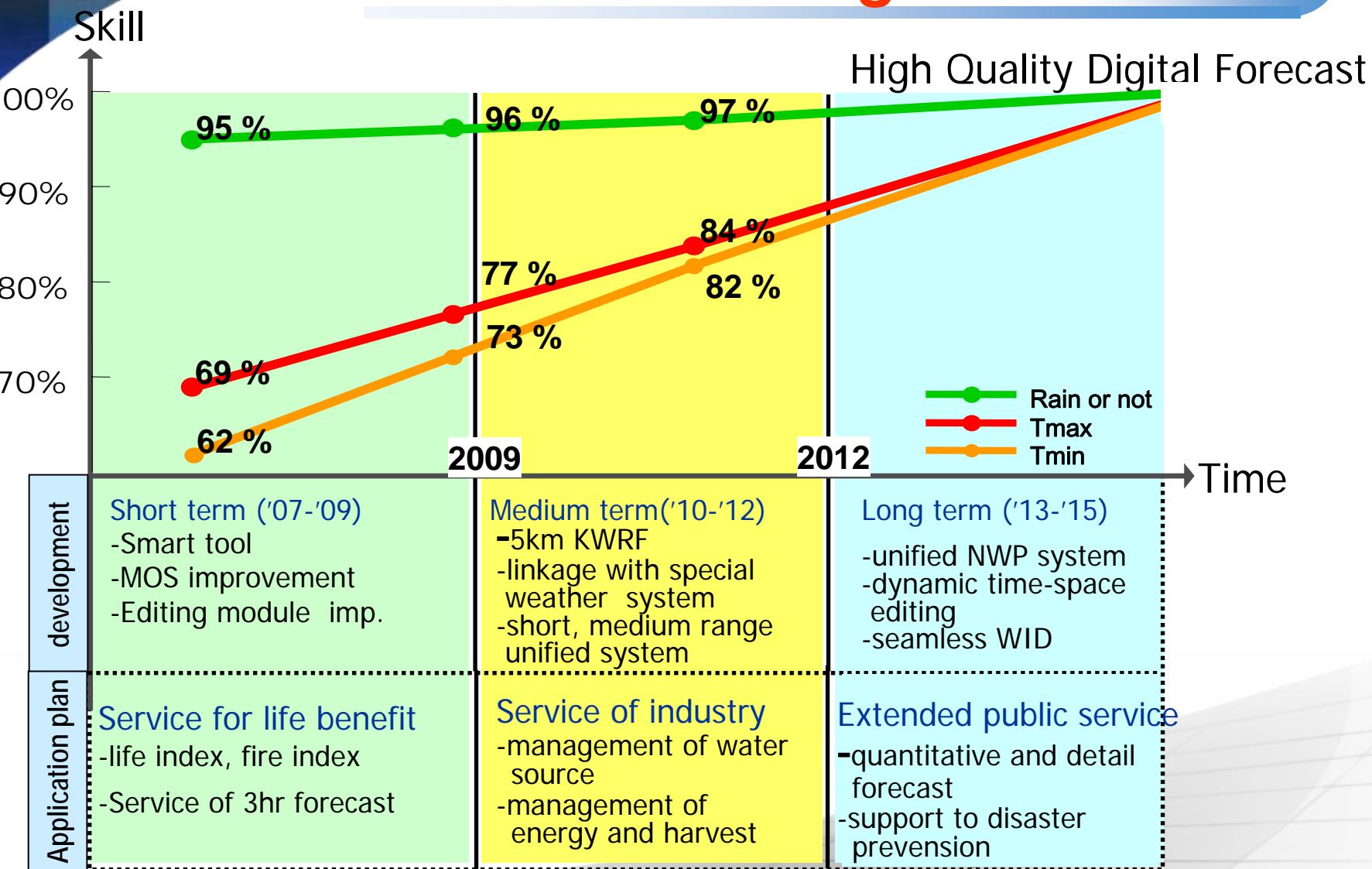
BSS (Cold Season)



- - - 103STN_MOS ○ 76STN_MOS ● 41STN_MOS
 - - - 103STN_GDP ○ 76STN_GDP ● 41STN_GDP

- - - 103STN_MOS ○ 76STN_MOS ● 41STN_MOS
 - - - 103STN_GDP ○ 76STN_GDP ● 41STN_GDP

Plan : Digital Forecast



Thank You ~