

中央氣象局全球數值預報之資料同化系統的初步測試

2013/05/13 天氣分析研討會

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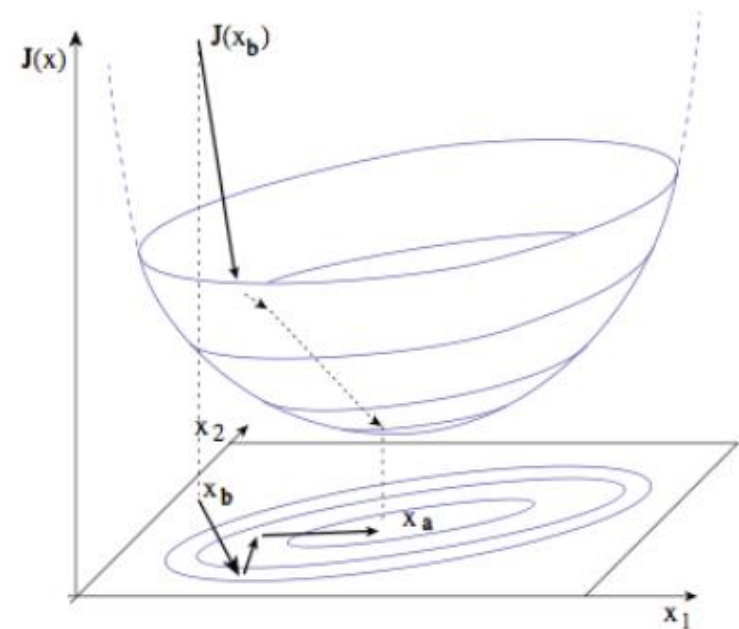
中央氣象局氣象科技研究中心¹ 中央氣象局氣象資訊中心²
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3D VARIATIONAL DATA ASSIMILATION

$$J(\mathbf{x}) = \frac{1}{2}(\mathbf{x} - \mathbf{x}_b)^T \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}_b) + \frac{1}{2}(\mathbf{y}_o - \mathbf{H}(\mathbf{x}))^T \mathbf{R}^{-1}(\mathbf{y}_o - \mathbf{H}(\mathbf{x})) = \mathbf{J}_b + \mathbf{J}_o$$

$$\nabla J(\mathbf{x}) = \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}_b) - \mathbf{H}^T \mathbf{R}^{-1}(\mathbf{y}_o - \mathbf{H}(\mathbf{x}))$$

- Pros:
 1. Easy to add various constraints to cost function
 2. Easy to implement and maintain
- Cons:
 1. Static background error covariance
 2. All observations are treated as instantaneous



From Bouttier and Courtier (1999)

4D VARIATIONAL DATA ASSIMILATION

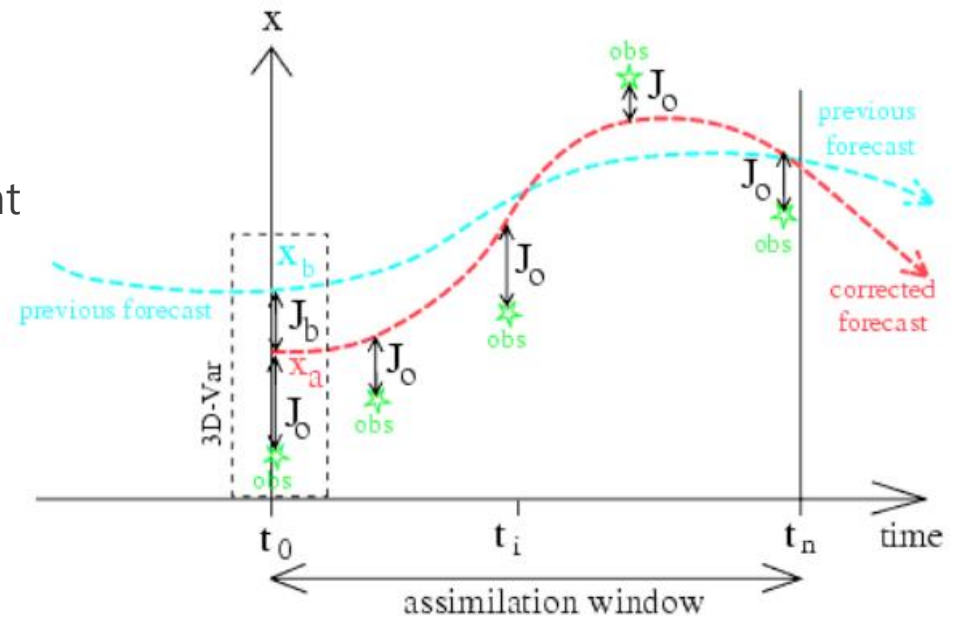
$$J = \frac{1}{2}(\mathbf{x}_0 - \mathbf{x}_b)^T \mathbf{B}^{-1}(\mathbf{x}_0 - \mathbf{x}_b) + \frac{1}{2} \sum_{i=0}^n (\mathbf{H}_i \mathbf{M}_i[\mathbf{x}_0] - \mathbf{y}_i)^T \mathbf{R}_i^{-1} (\mathbf{H}_i \mathbf{M}_i[\mathbf{x}_0] - \mathbf{y}_i)$$

- Pros:

1. Implicit flow-dependent B
2. Assimilate observations at right time

- Cons:

1. Hard to develop and maintain adjoint model \mathbf{M}^T
2. The assumption of "perfect model"



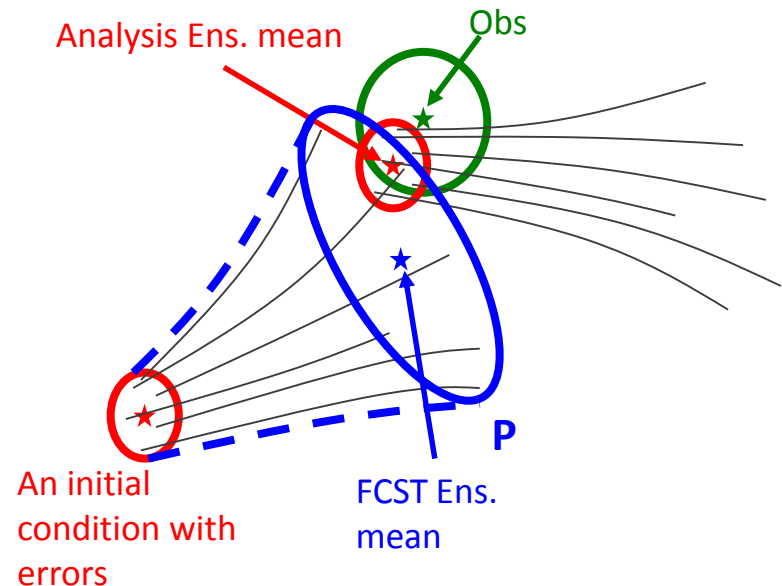
From Tom Auligne

ENSEMBLE KALMAN FILTER

$$\mathbf{x}_a = \mathbf{x}_f + \mathbf{K}(\mathbf{y}_o - \mathbf{H}\mathbf{x}_f) \quad \mathbf{P}^f \mathbf{H}^T = \overline{\mathbf{x}'^f (\mathbf{H}\mathbf{x}'^f)^T} \equiv (1/n - 1) \sum_{i=1}^n \mathbf{x}'_i{}^f (\mathbf{H}\mathbf{x}'_i{}^f)^T$$

$$\mathbf{K} = \mathbf{P}^f \mathbf{H}^T (\mathbf{H}\mathbf{P}^f \mathbf{H}^T + \mathbf{R})^{-1} \quad \mathbf{H}\mathbf{P}^f \mathbf{H}^T = \overline{\mathbf{H}\mathbf{x}'^f (\mathbf{H}\mathbf{x}'^f)^T} \equiv (1/n - 1) \sum_{i=1}^n \mathbf{H}\mathbf{x}'_i{}^f (\mathbf{H}\mathbf{x}'_i{}^f)^T$$

- Pros:
 1. Flow-dependent B
 2. No need to linearize and get adjoint model of \mathbf{H} and \mathbf{M}
- Cons:
 - Under sampling:
 1. Filter divergence: Need to inflate the members
 2. Spurious noise: Need Localization



From S.-C. Yang

HYBRID DATA ASSIMILATION

$$J(\mathbf{x}'_f, \alpha) = \beta_f \frac{1}{2} (\mathbf{x}'_f)^T \mathbf{B}^{-1} (\mathbf{x}'_f) + \beta_e \frac{1}{2} (\alpha)^T \mathbf{L}^{-1} (\alpha) + \frac{1}{2} (\mathbf{y}'_o - \mathbf{H}\mathbf{x}'_t)^T \mathbf{R}^{-1} (\mathbf{y}'_o - \mathbf{H}\mathbf{x}'_t)$$

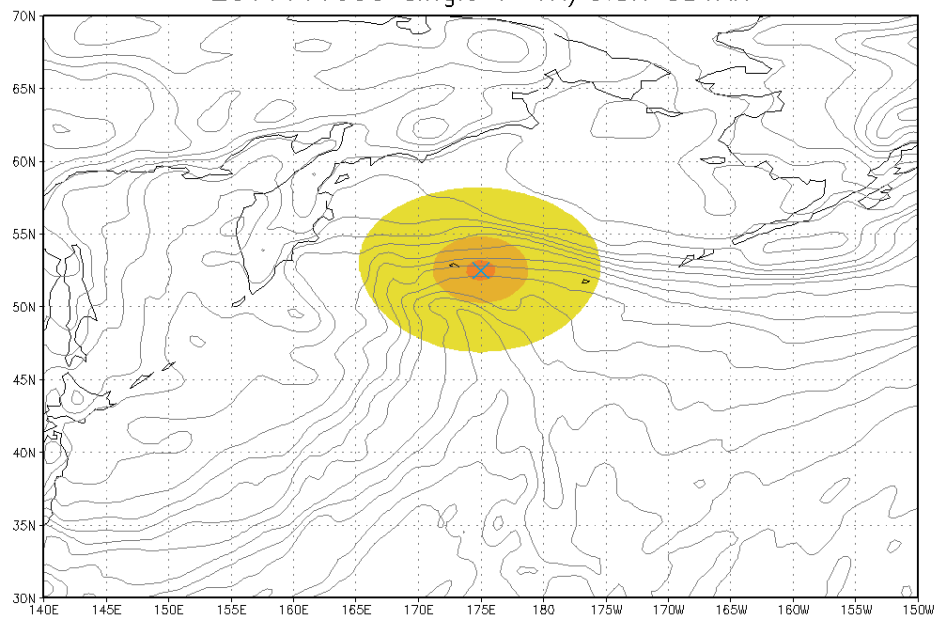
$$\mathbf{x}'_t = \mathbf{x}'_f + \sum_{k=1}^K (\alpha_k \circ \mathbf{x}_k^e) \quad \frac{1}{\beta_f} + \frac{1}{\beta_e} = 1$$

- Benefit from use of flow dependent ensemble covariance instead of static B
 - Avoid “filter divergence” problem in EnKF
 - Less expensive than EnKF (small ensemble)
 - Easy framework to add various constraints (make balance to initial conditions)
 - Use of various existing capabilities in VAR
 - Now hybrid systems are developed or operated ...
 - NCEP, ECMWF, NCAR, CMC, UK ... CWB
- Lorenc, 2003*
Buehner, 2005
Wang et al., 2007, 2008
Kleist et al., 2011
etc.

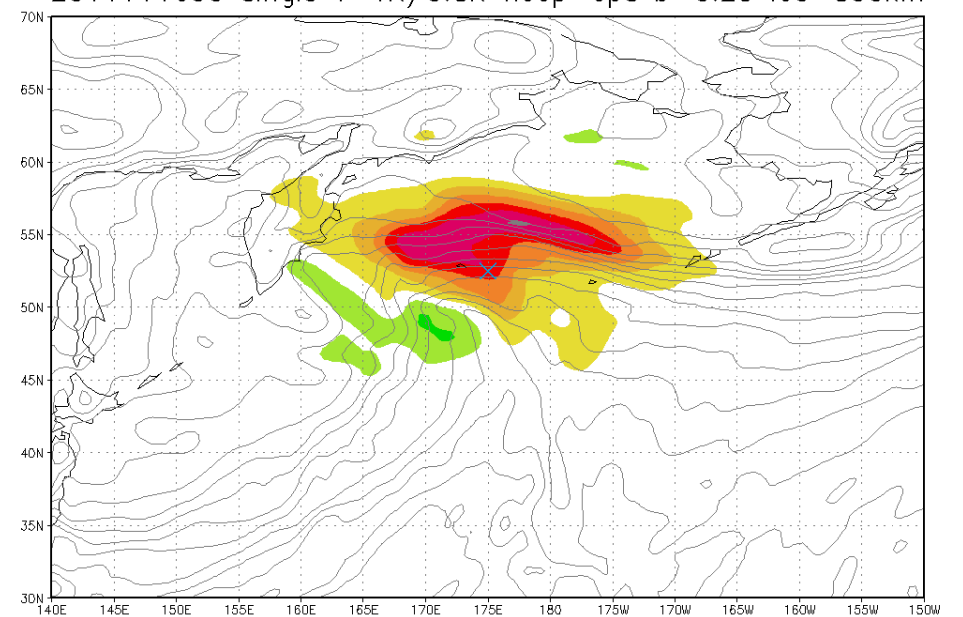
EXPERIMENT DESIGN: SINGLE OBS. TEST

- Data assimilation system: NCEP GSI operational version, may 2012
- Due to lack of EnKF system, the hybrid ensemble members are time lagged members: the forecasts of 6~72HR, per 6hr, total are 12 members
 - 6, 12, 18, 24, 30, 36, 42, 48, 54, 60, 66, 72HR forecasts
- Single observation test 1 on 2011/11/16,00
 - Innovation: 1(K); observation error: 0.8(K)
 - Location: 175W, 52.5N at 850 hPa
 - $\beta_1=0.25$ and $\beta_2=0.75$, localization: 800 km
- Single observation test 2 on 2012/05/26,12
 - Innovation (V): 1(m/s); observation error: 1.0(m/s)
 - Location: 144.2E, 25.5N at 850 hPa
 - $\beta_1=0.25$ and $\beta_2=0.75$, localization: 800 km

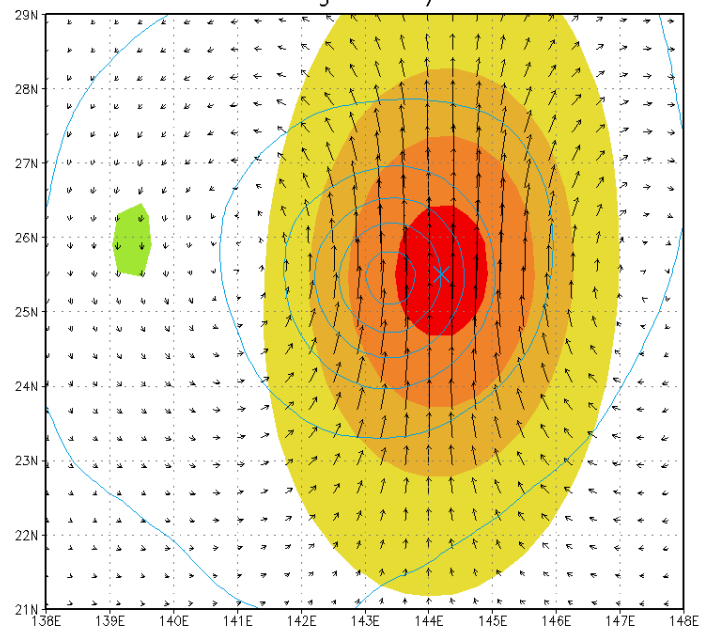
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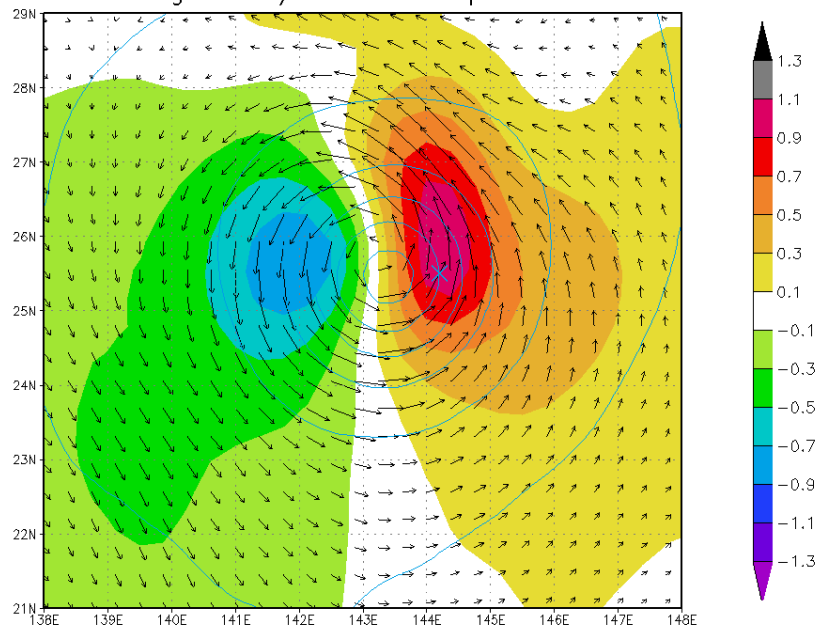
2011111600 single T=1K/0.8K ncep-ops b=0.25 loc=800km



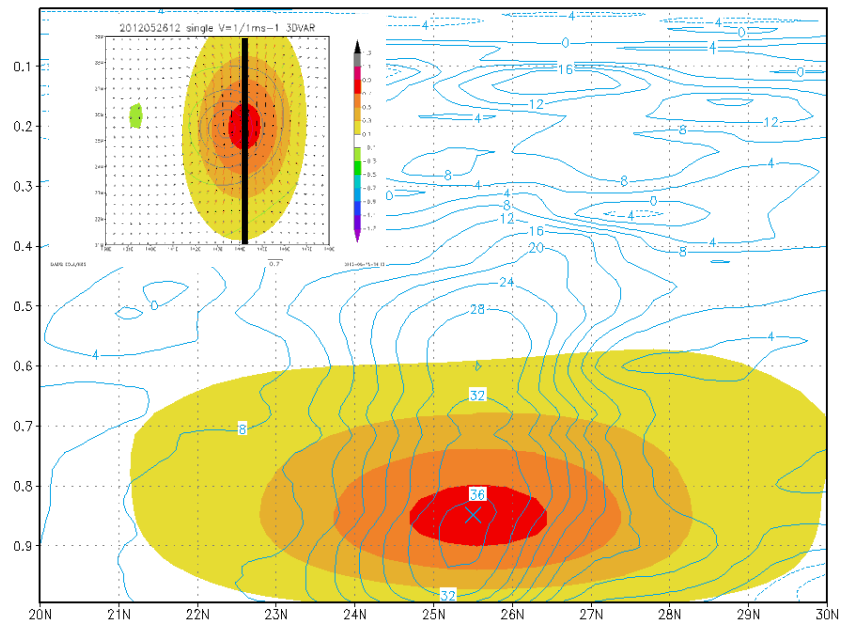
2012052612 single V=1/1ms-1 3DVAR



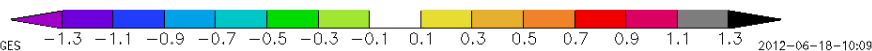
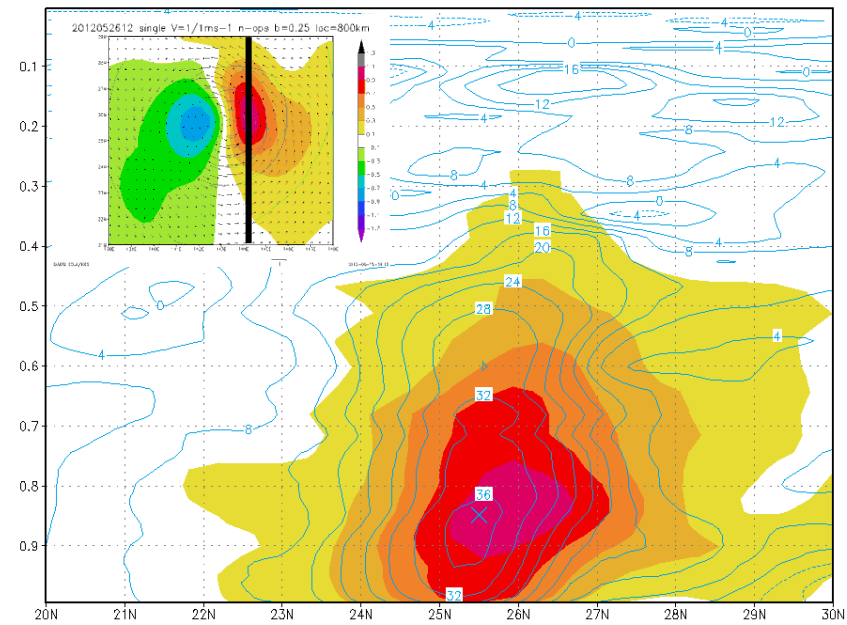
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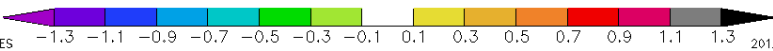
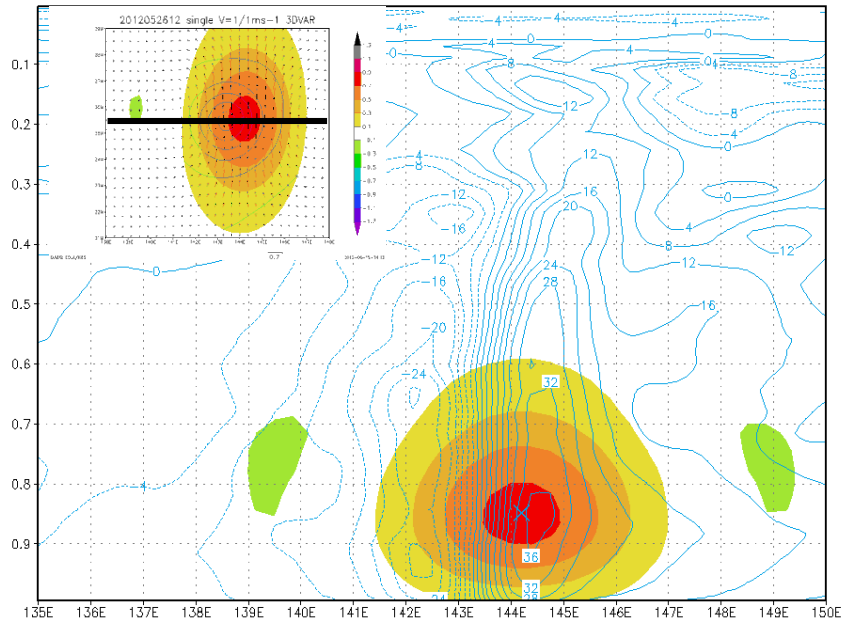
2012052612 single V=1/1ms-1 3DVAR



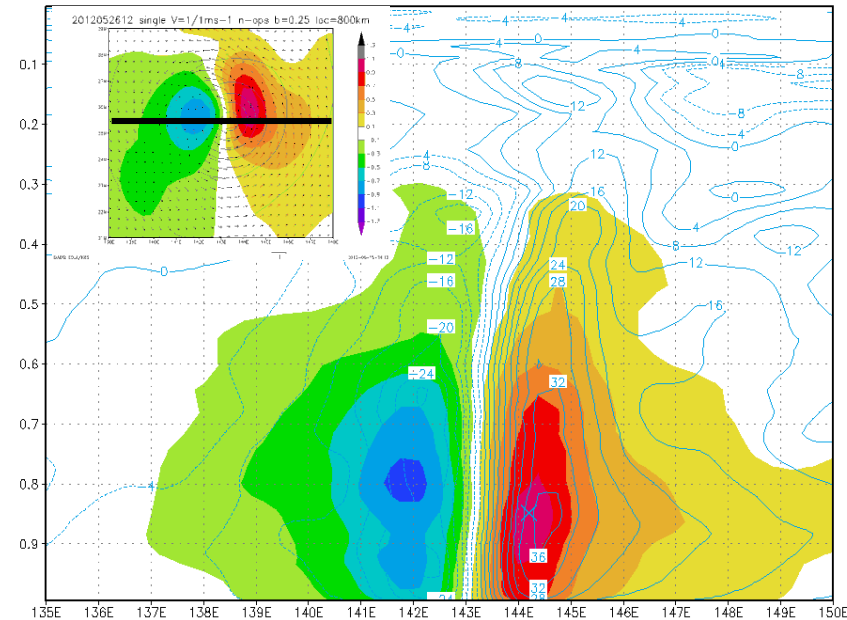
2012052612 single V=1/1ms-1 n-ops b=0.25 loc=800km



2012052612 single V=1/1ms-1 3DVAR



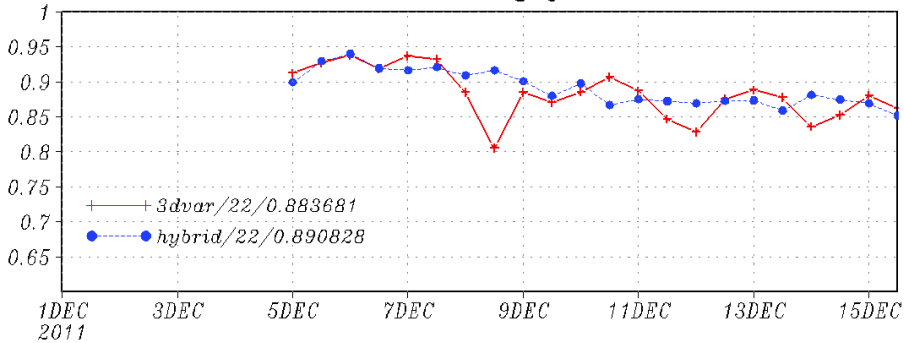
2012052612 single V=1/1ms-1 n-ops b=0.25 loc=800km



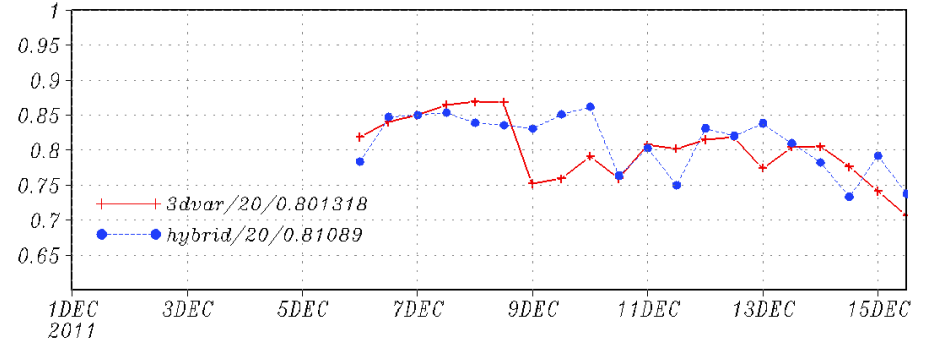
EXPERIMENT DESIGN: PARALLEL

- Data assimilation system: NCEP GSI operational version, may 2012
- Due to lack of EnKF system, the hybrid ensemble members are time lagged members: the forecasts of 6~72HR, per 6hr, total are 12 members
 - 6, 12, 18, 24, 30, 36, 42, 48, 54, 60, 66, 72HR forecasts
- Experiment period: 2011120100 ~ 2011121512
- Control (3dvar): Use NCEP operational GSI and do only 3dvar analysis
- Experiment (hybrid): Same as control exp., but turn on the hybrid capability. The members are time lagged members.
- $\beta_1=0.5$ and $\beta_2=0.5$, localization: 800 km

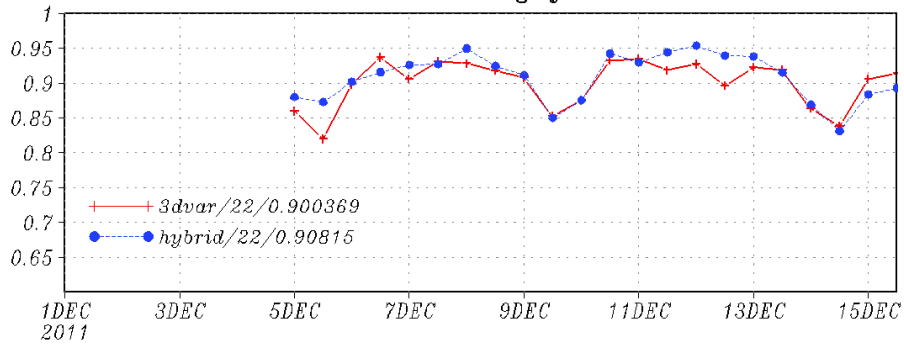
500 ach 4 day fest - NA



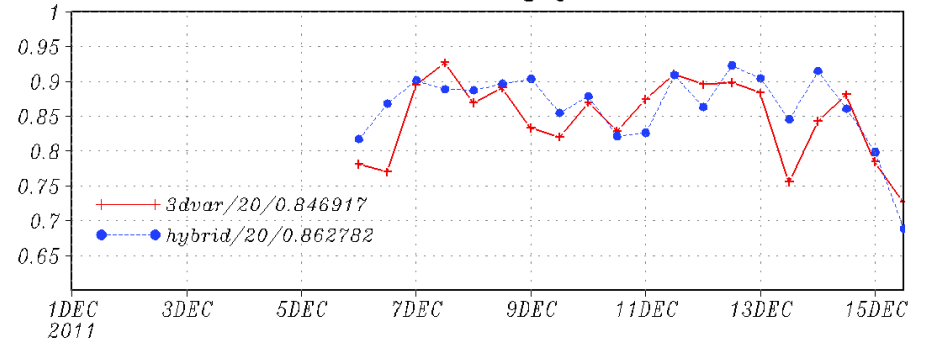
500 ach 5 day fest - NA



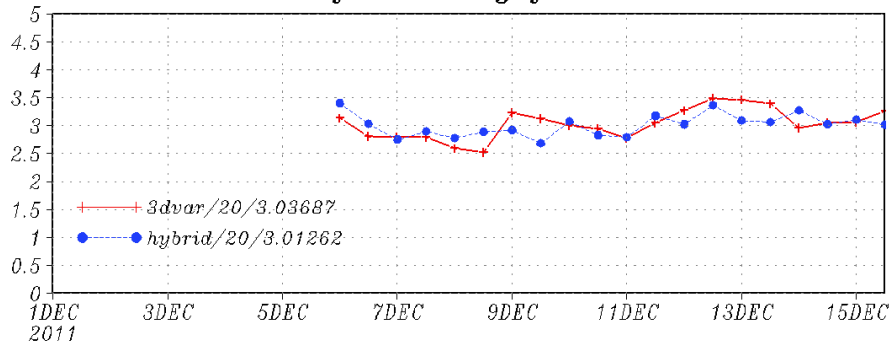
500 ach 4 day fest - SA



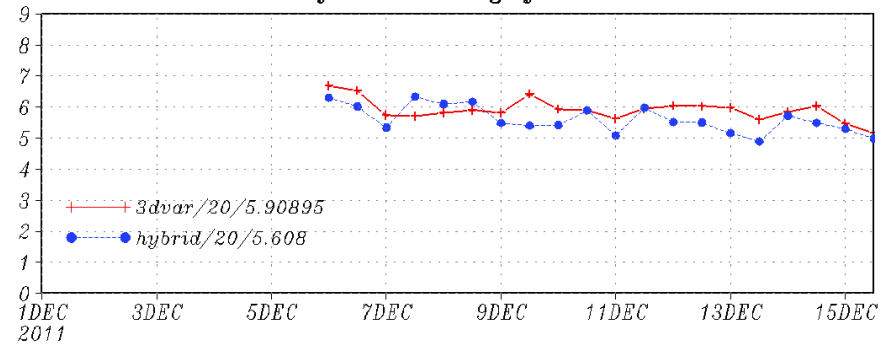
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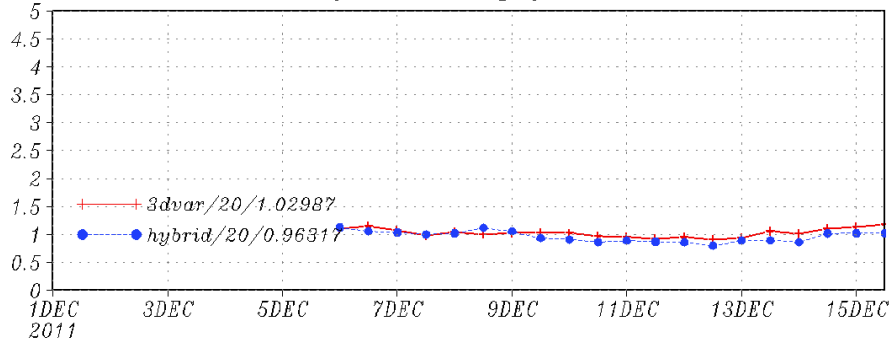
500 frt 5 day fest - NA



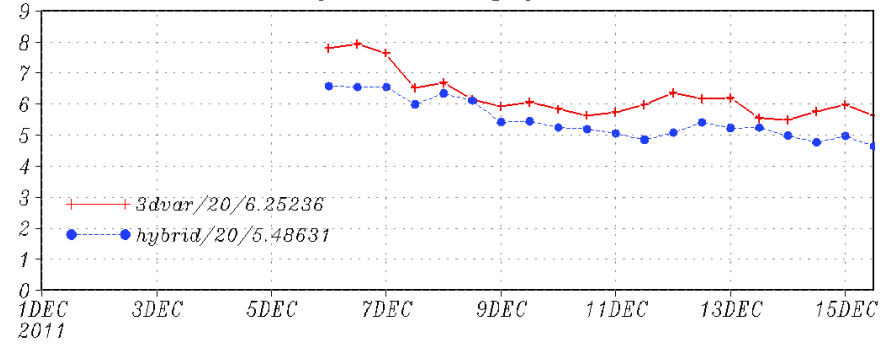
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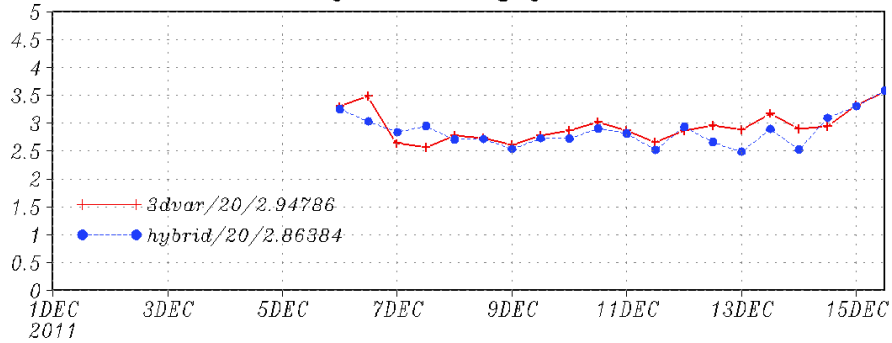
500 frt 5 day fest - TP



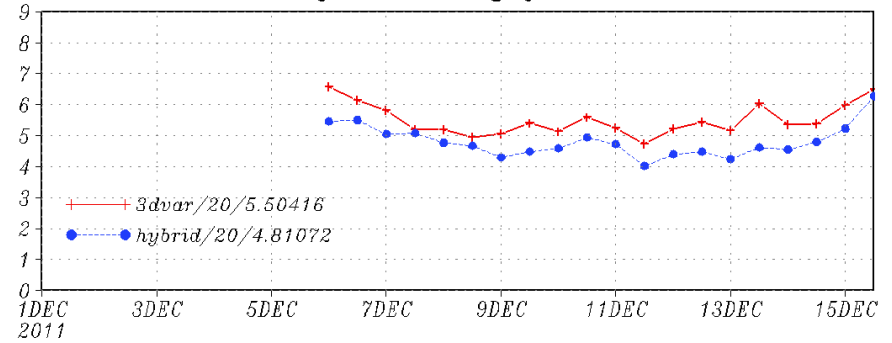
100 fru 5 day fest - TP



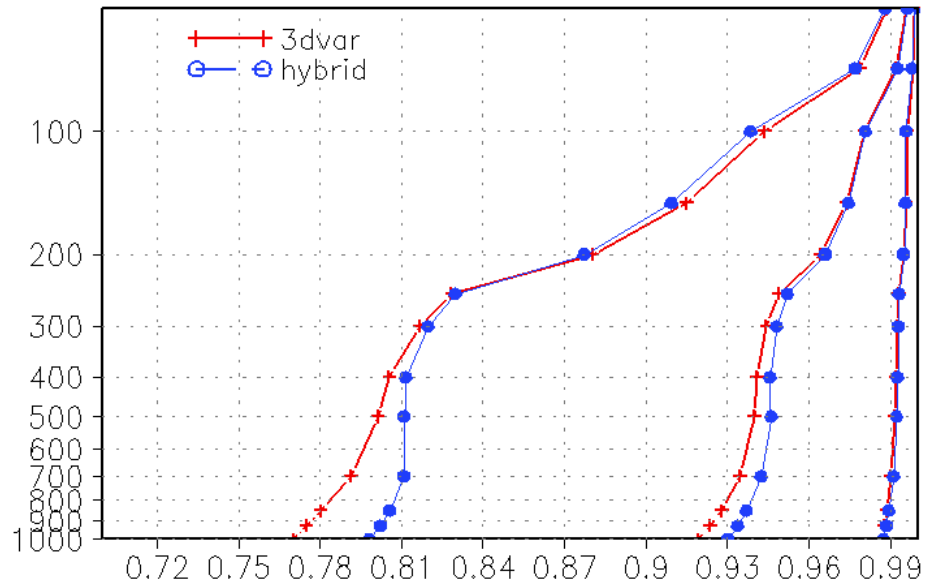
500 frt 5 day fest - SA



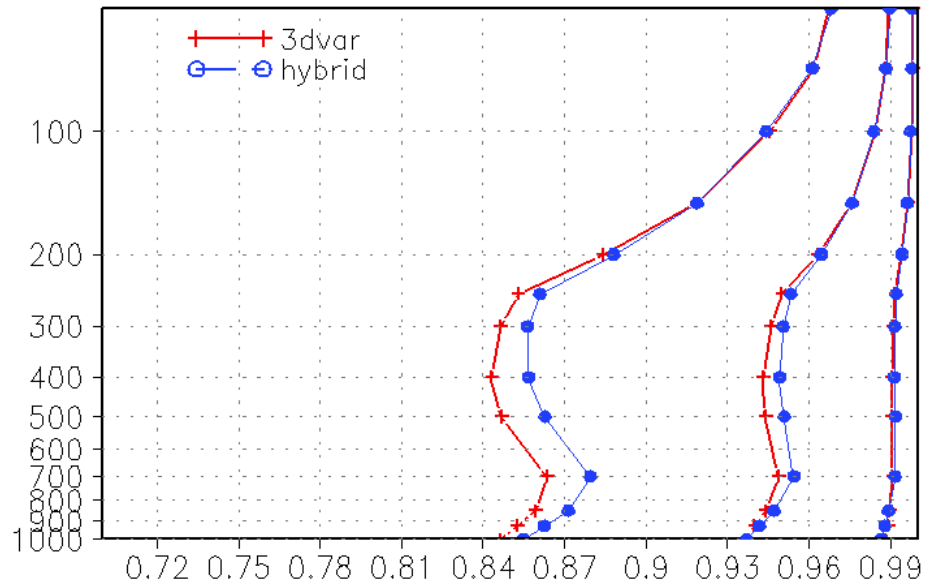
100 fru 5 day fest - SA



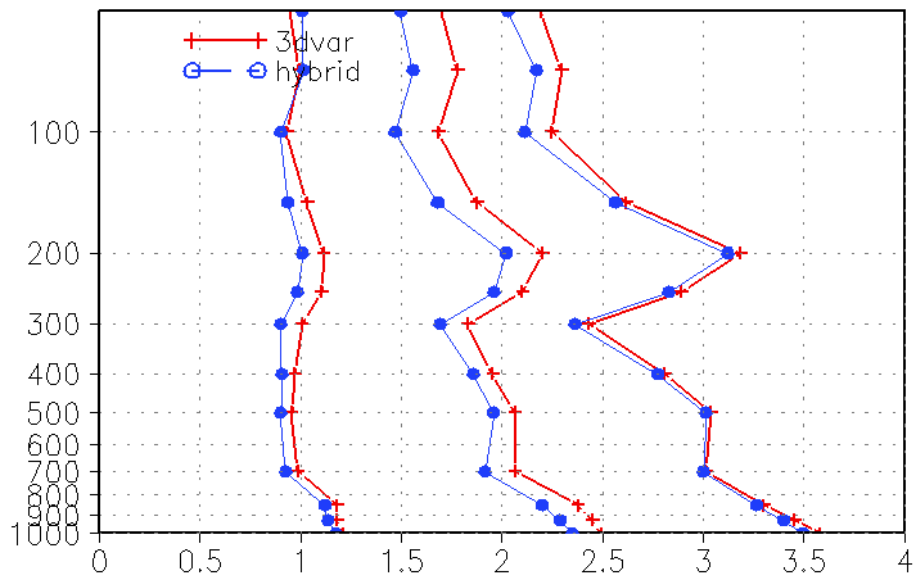
2011 Dec. AC score of Height, NH



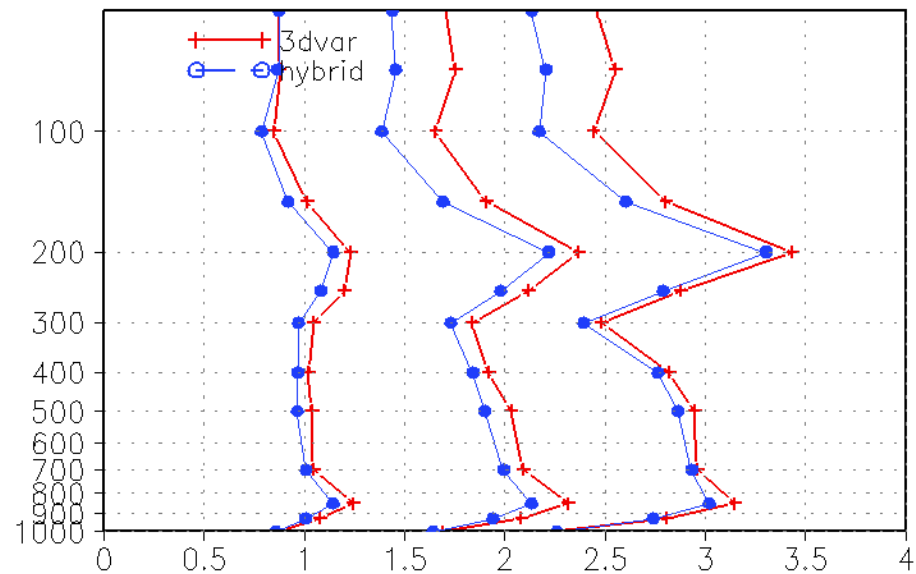
2011 Dec. AC score of Height, SH



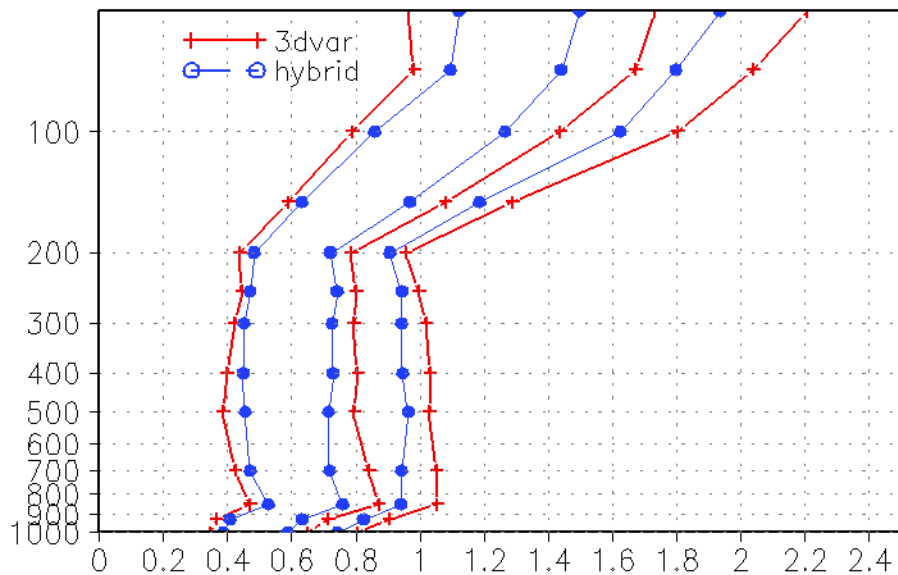
2011 Dec. RMSE of temperature, NH



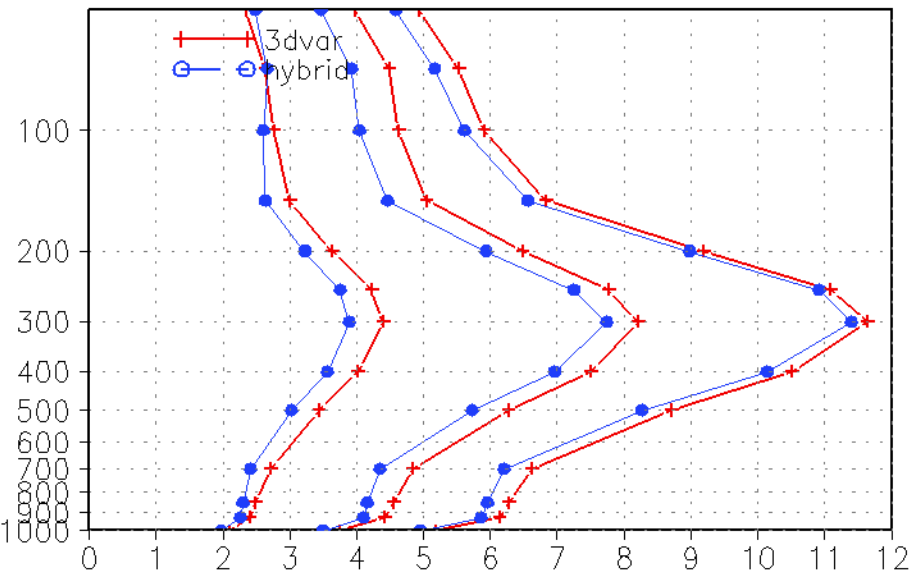
2011 Dec. RMSE of temperature, SH



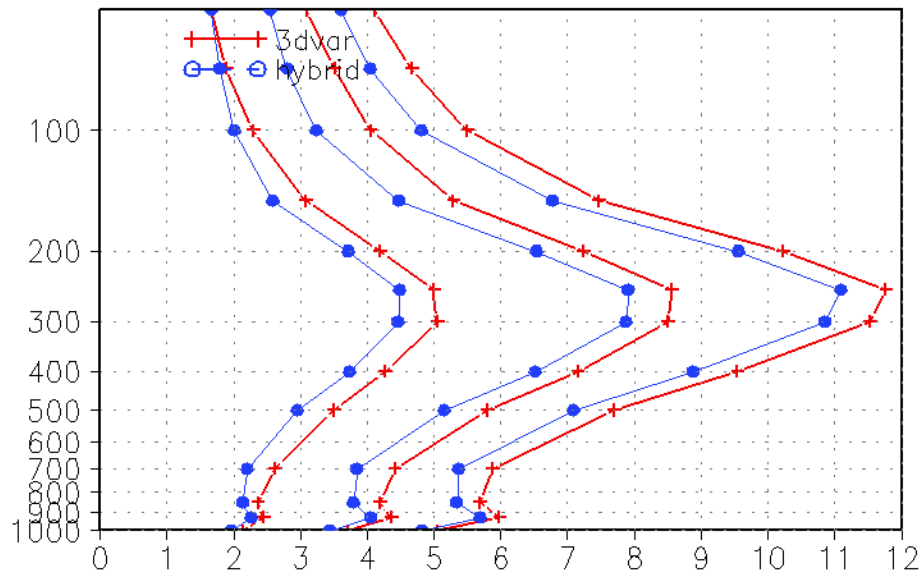
2011 Dec. RMSE of temperature, TR



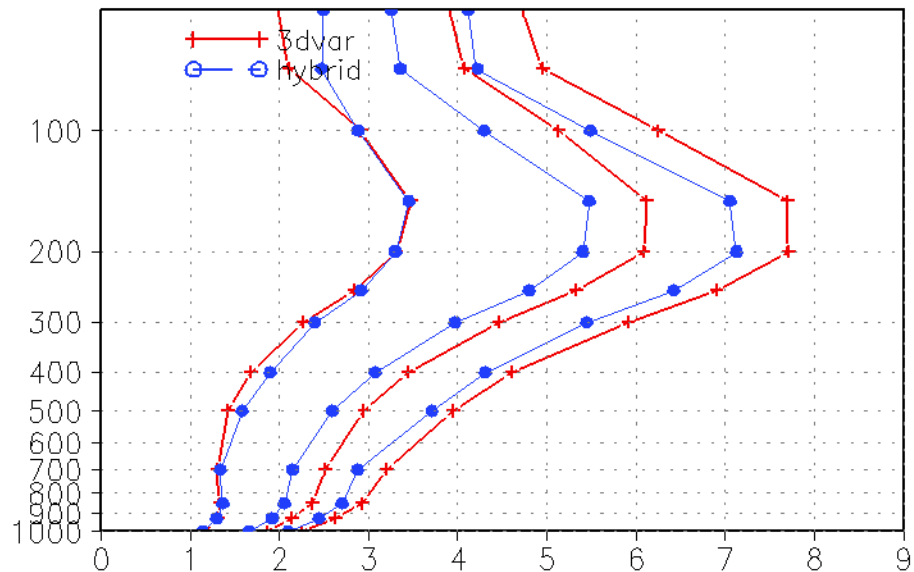
2011 Dec. RMSE of u-wind, NH



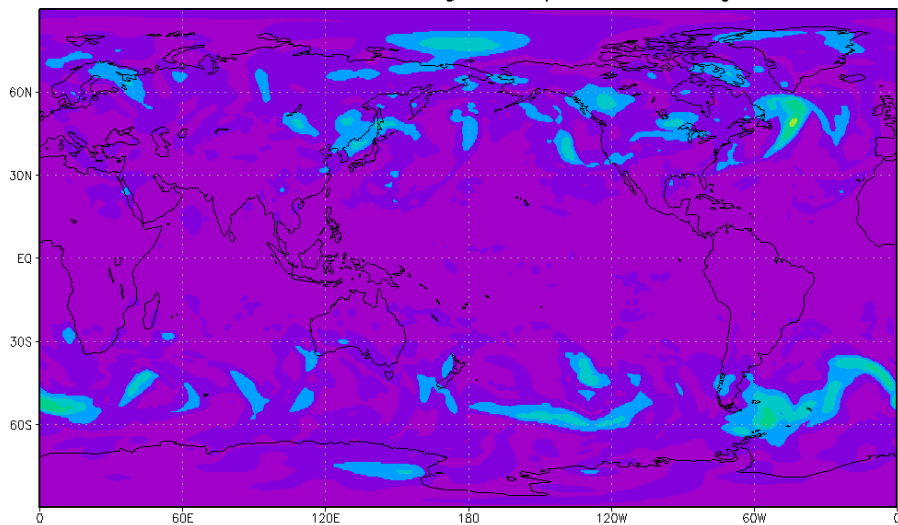
2011 Dec. RMSE of u-wind, SH



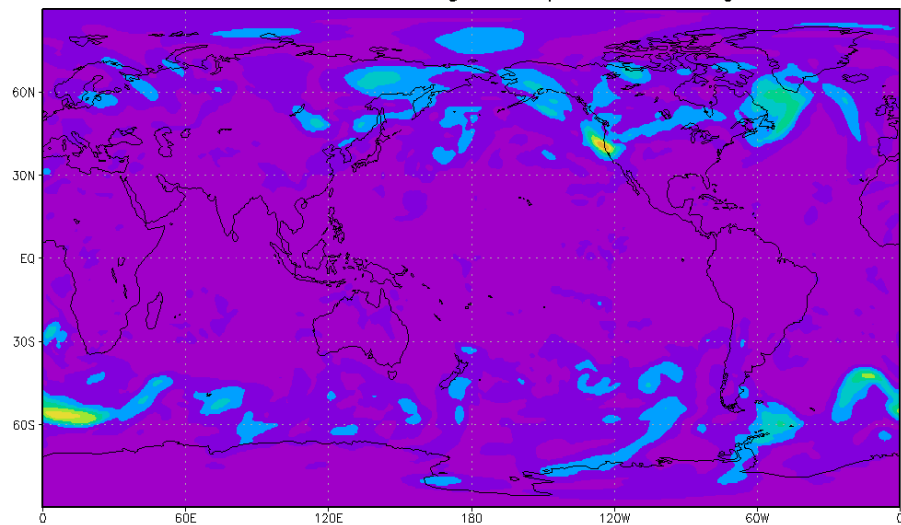
2011 Dec. RMSE of u-wind, TR



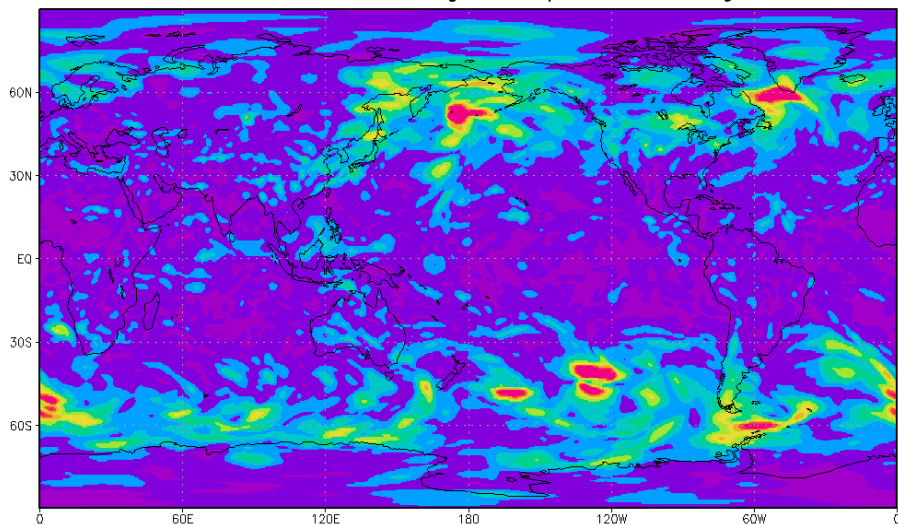
2011111900 T sig=11 spread timelag



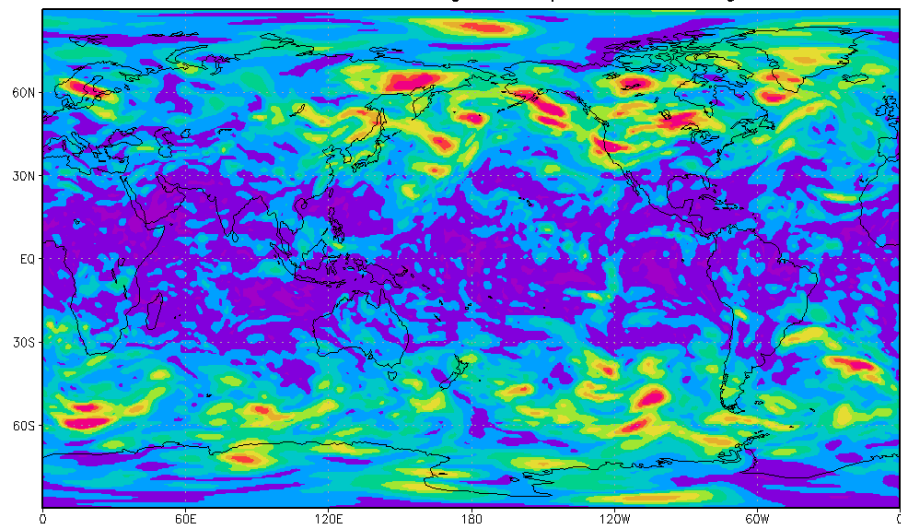
2011111900 T sig=21 spread timelag



2011111900 U sig=11 spread timelag

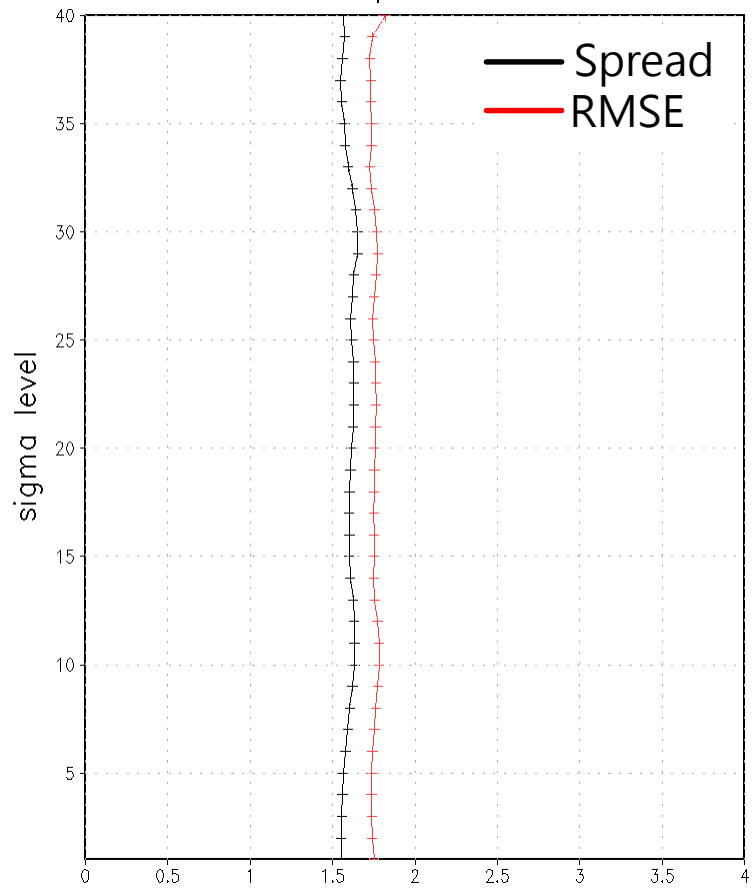


2011111900 U sig=21 spread timelag

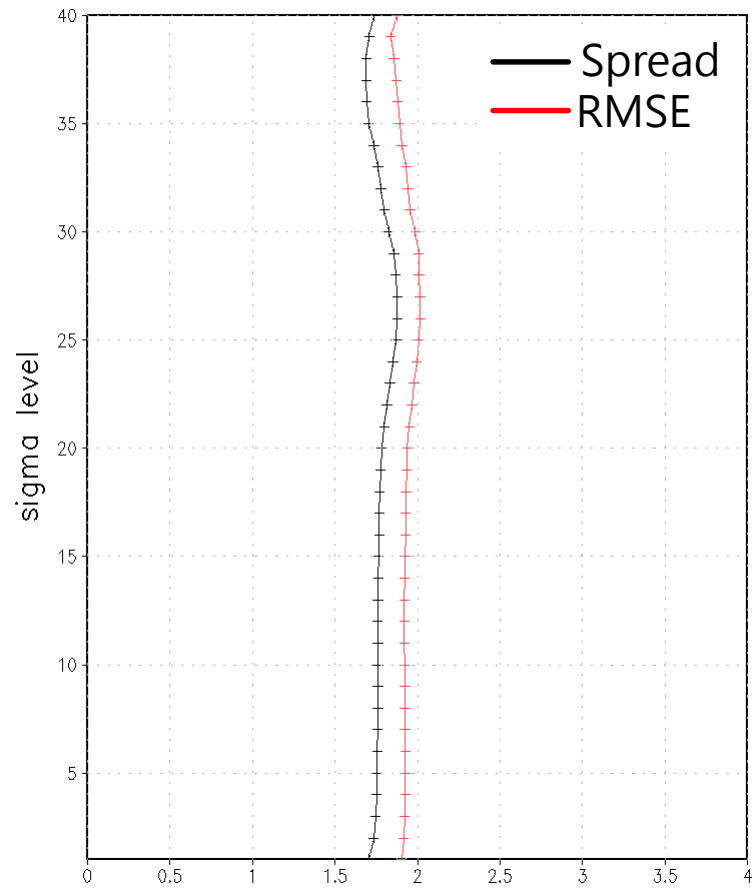




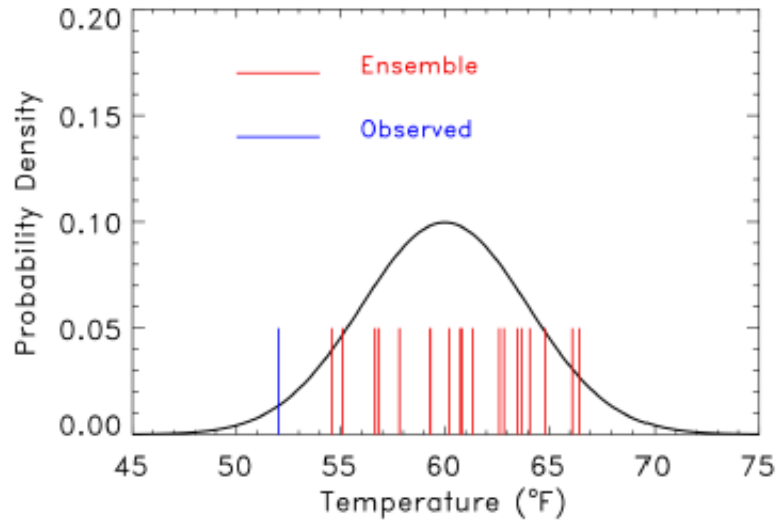
temperature



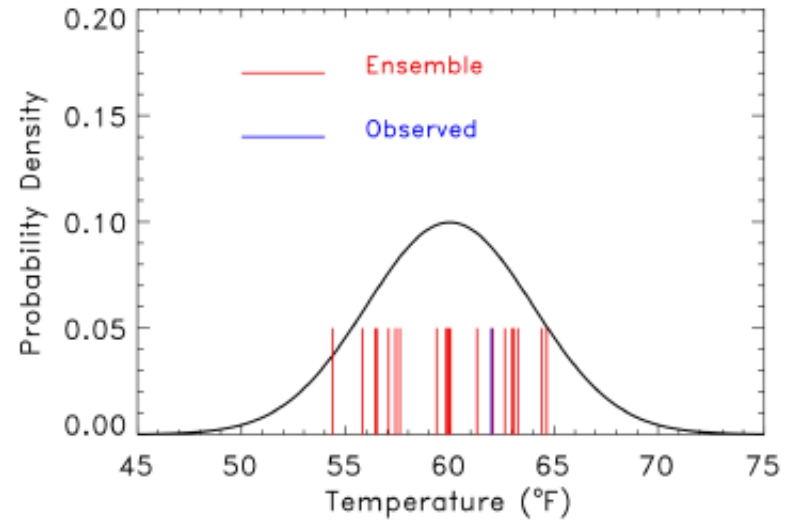
u wind



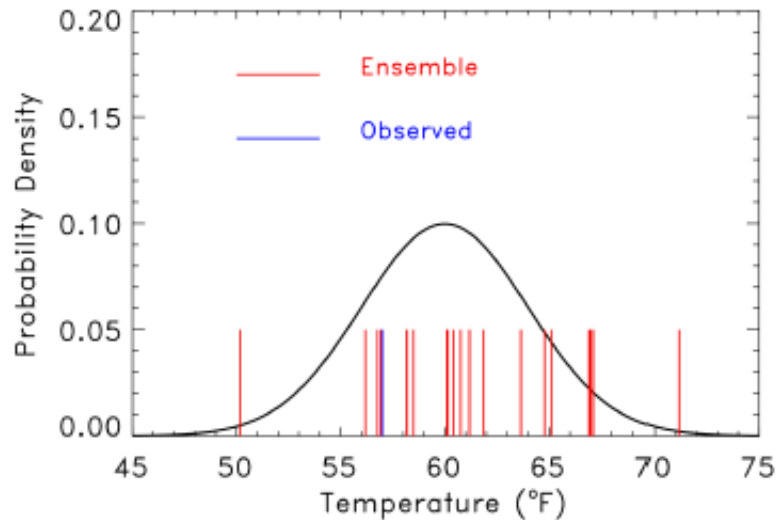
Rank 1 of 21



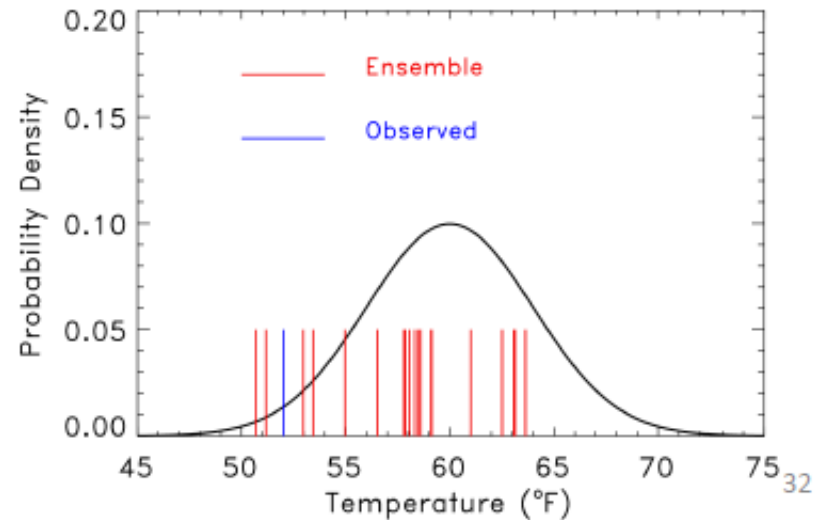
Rank 14 of 21



Rank 5 of 21

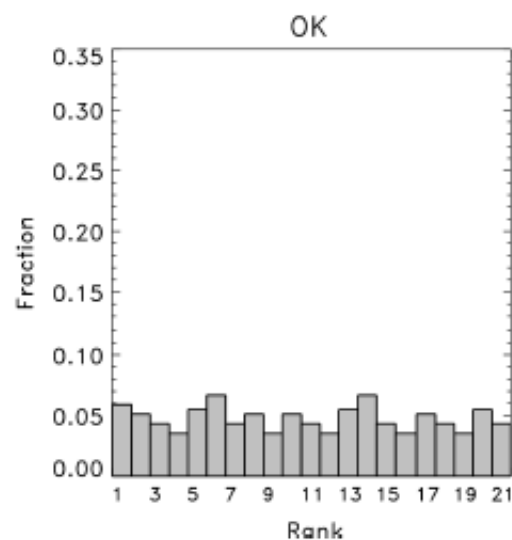


Rank 3 of 21

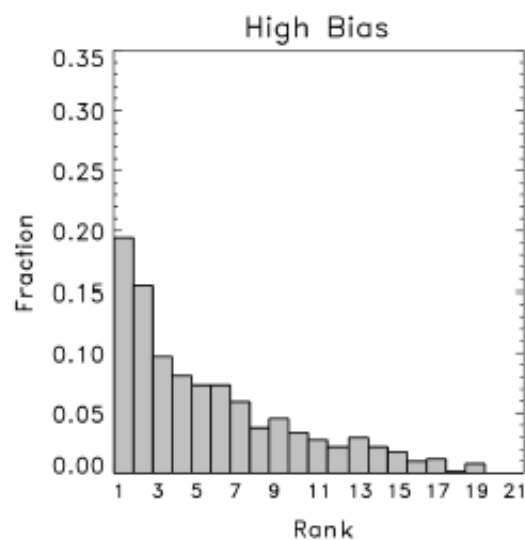


Rank histograms

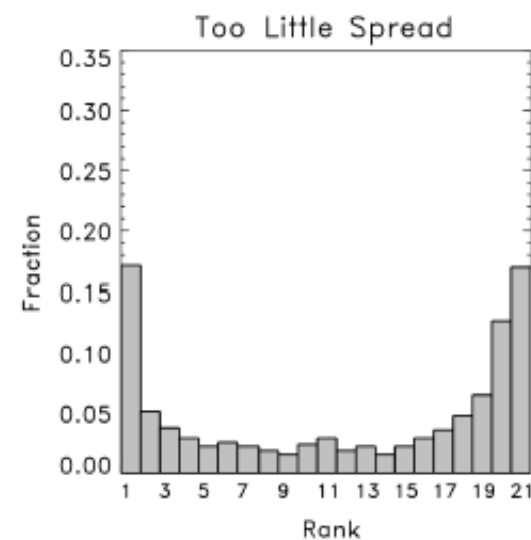
With **lots of samples** from many situations, can evaluate the characteristics of the ensemble.



Happens when observed is indistinguishable from any other member of the ensemble. Ensemble hopefully is reliable.



Happens when observed too commonly is lower than the ensemble members.



Happens when there are either some low and some high biases, or when the ensemble doesn't spread out enough.



LOCATION (μ error)

0.0

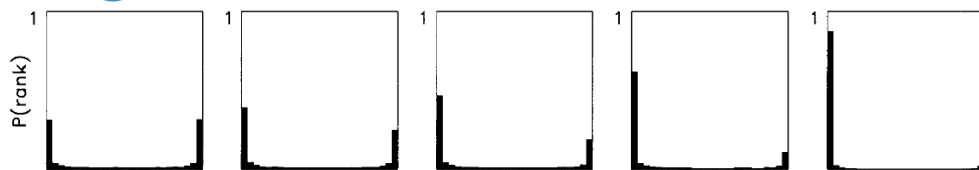
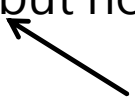
0.2

0.4

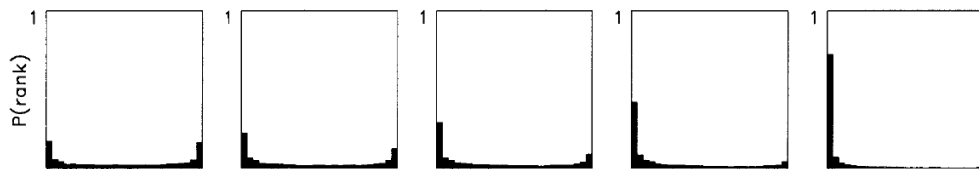
0.8

1.6

Indicate small spread but no bias

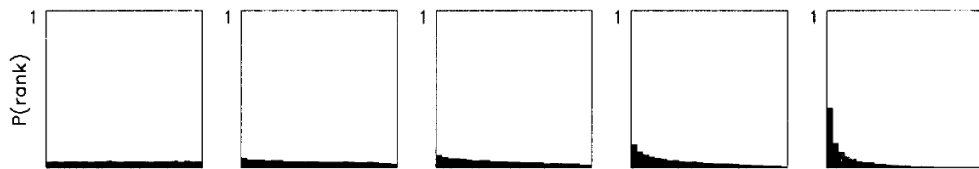


0.25

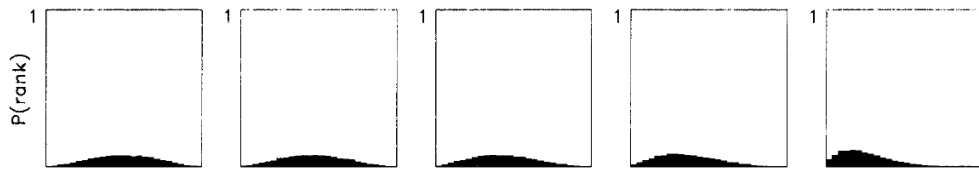


0.5

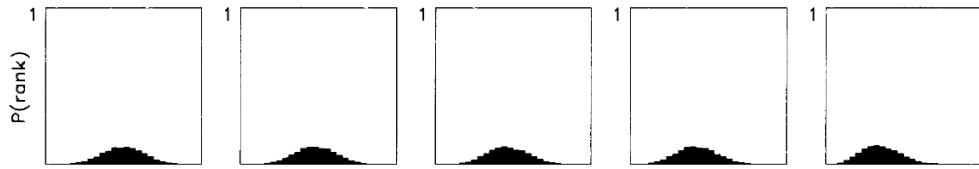
SCALE (σ error)



1.0



2.0

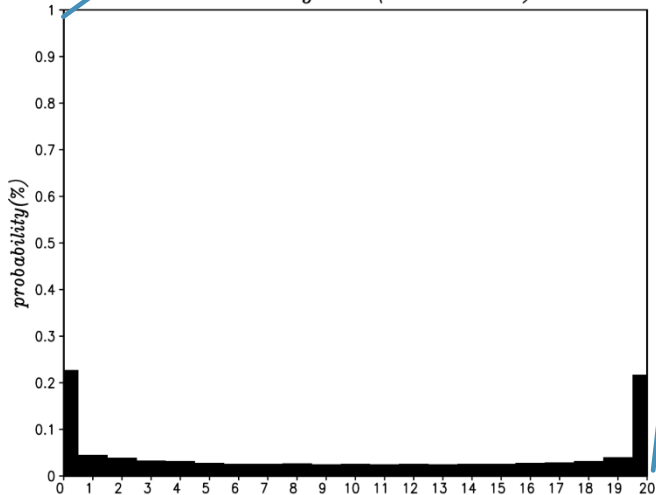


4.0

1 Rank 26 1 Rank 26 1 Rank 26 1 Rank 26 1 Rank 26

FIG. 1. Rank histograms where verification is sampled from a $N(0, 1)$ distribution and the ensemble ($n = 25$ members) is sampled from a $N(\mu, \sigma)$ distribution. The rank of the verification is tallied 10 000 times in each panel.

rank histogram (no satellite)



Time lag

Hamill (2001)

CONCLUSIONS

- When providing only time lagged members, the single observation tests works well and we can see the flow-dependent increment at front or typhoon cases.
- The performance of hybrid DA system with time lagged members are better than the 3DVAR experiment.
- The spread of time lagged members are large at mid-high latitude area and small at low latitude and tropics area.
- The time lagged members during experiment period have no bias but indicates small spread.
- NCEP operational GSI and EnKF systems are ported to CWB HPC system (Fujitsu FX10) and just works well last week. We will test the hybrid with EnKF system and evaluate the performance soon.



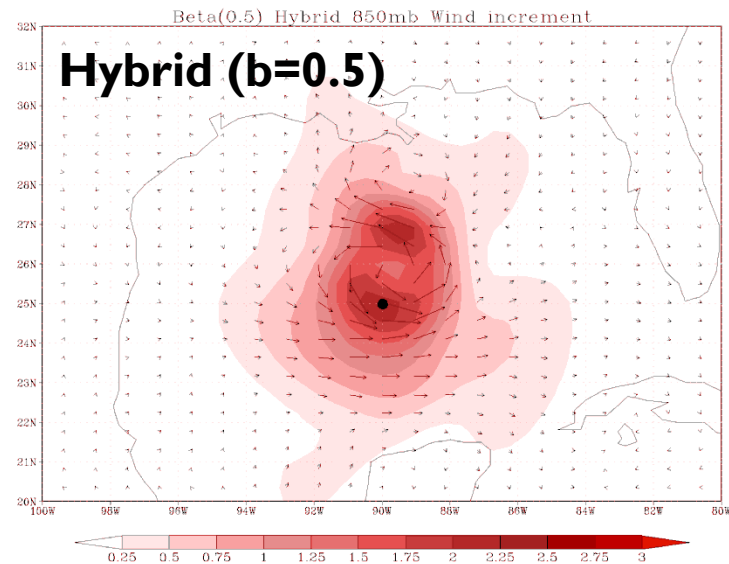
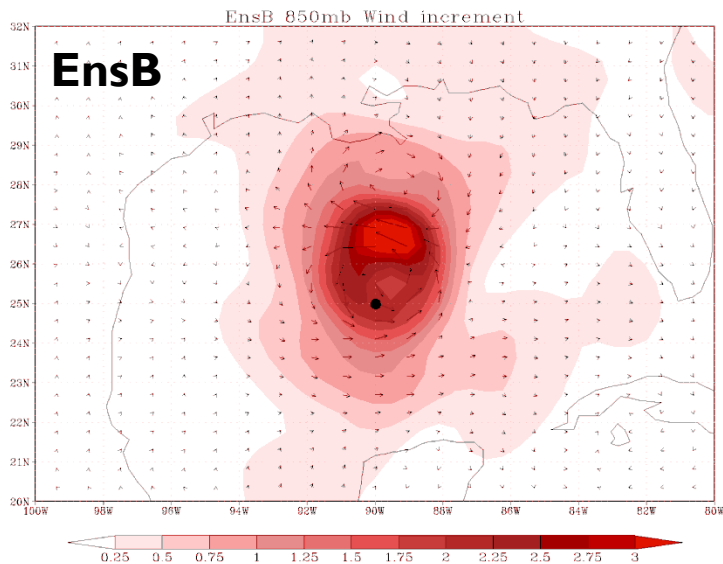
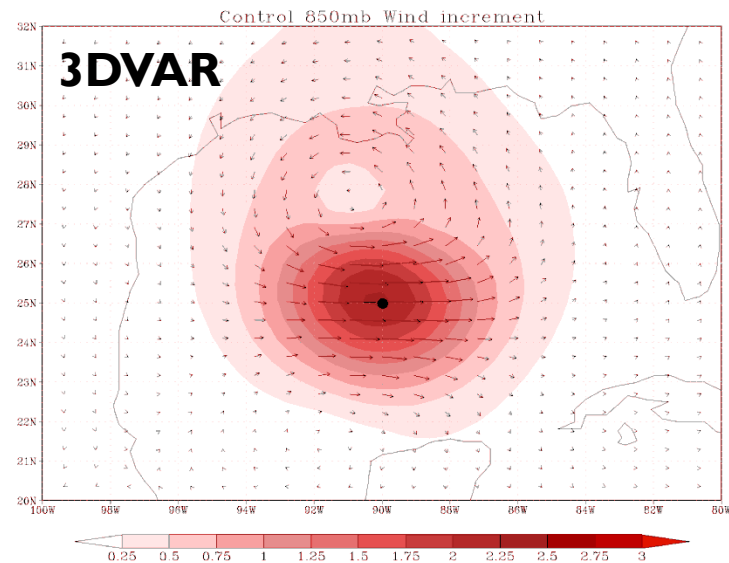
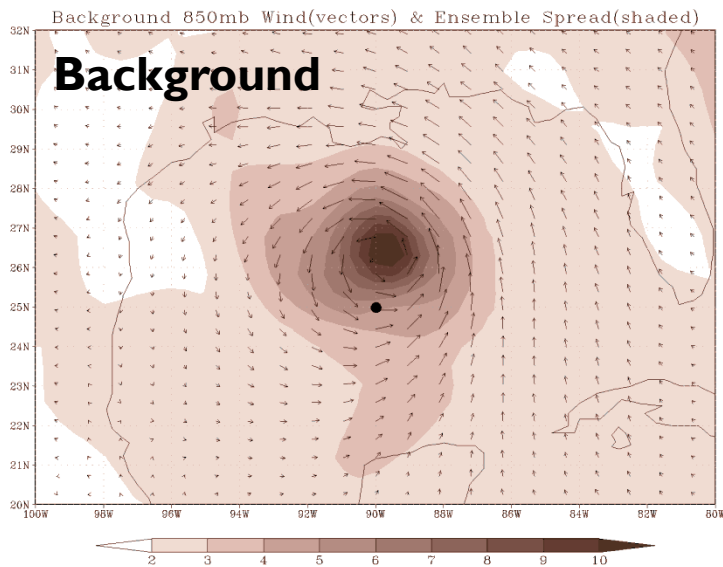
THE END

THANKS FOR YOUR ATTENTION.



SINGLE OBSERVATION

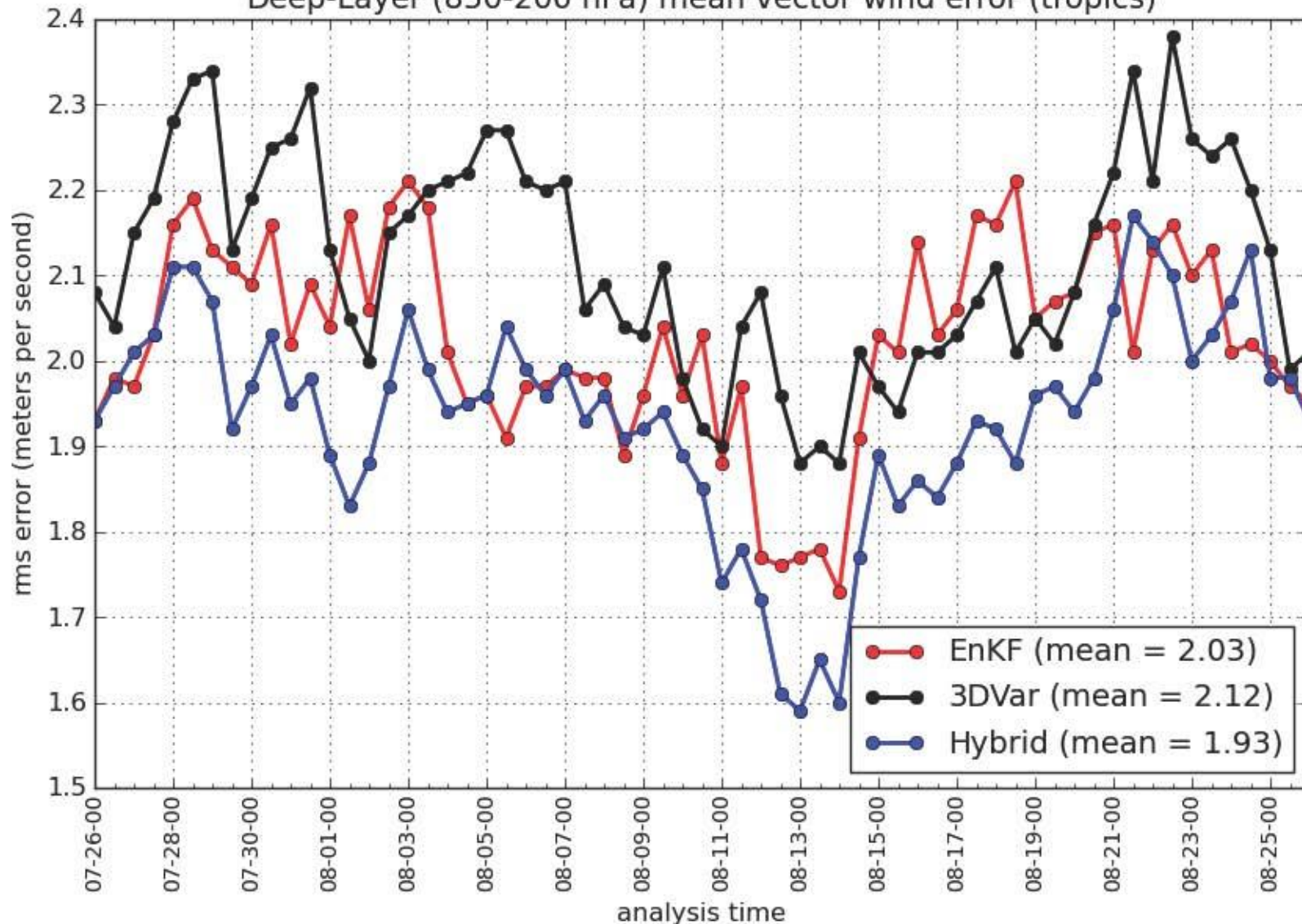
from D. Kleist



Single 850mb zonal wind observation (3 m/s O-F, 1m/s error) in Hurricane Ike circulation

Tropical Wind Errors (72-h) vs “consensus” analysis

Deep-Layer (850-200 hPa) mean vector wind error (tropics)



DLM == mean of vector wind at 850,700,500,400,300,200 hPa.

“Consensus analysis” = 1/3(ECMWF + UKMET + NCEP)

Kleist et al., 2011