

Hybrid Data Assimilation

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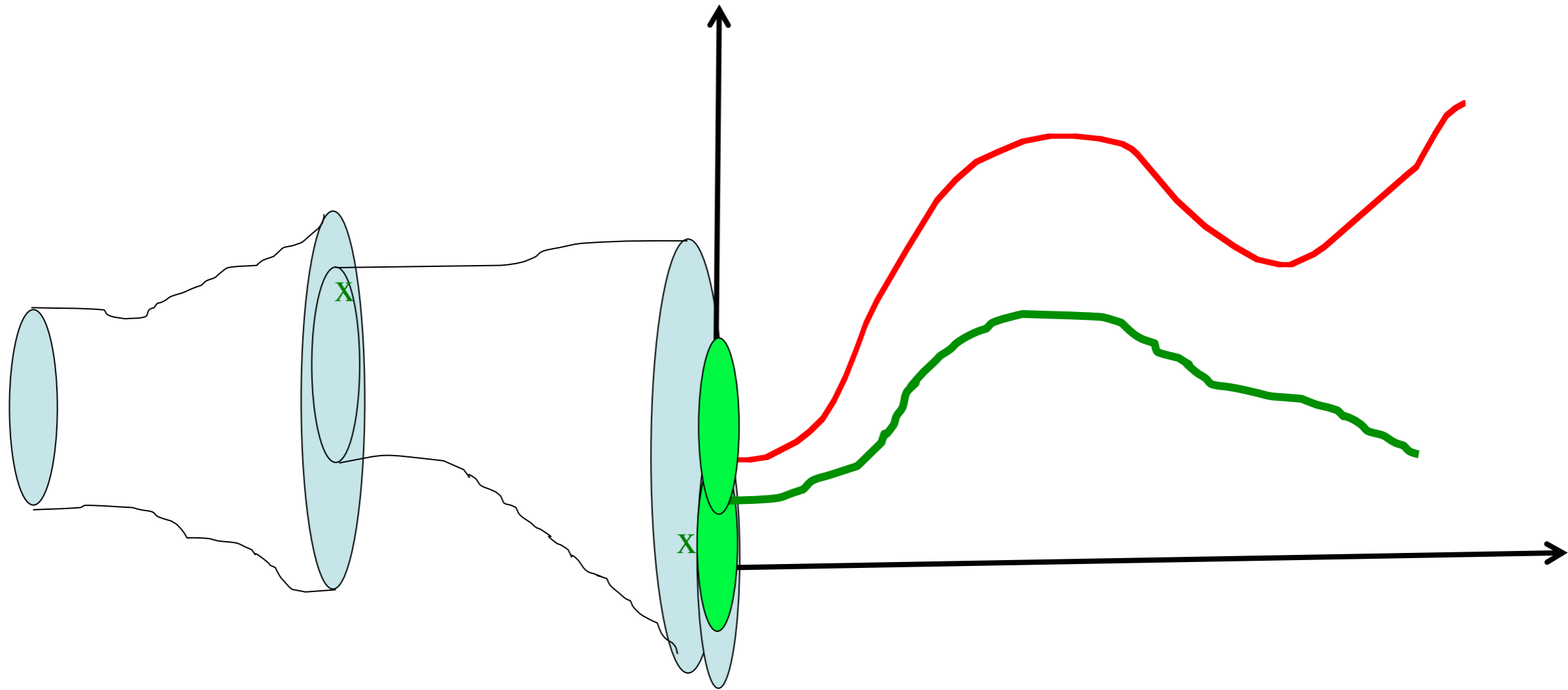
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Operational Hybrid methods

1. ETKF-3/4DVar with control variable transform
2. Ens4DVar 'ensemble of data assimilations' EDA
3. 4DEnsVar

1. ETKF-3DVar



$$B = B_c$$

$$B = \beta_c^2 B_c + \beta_e^2 P^b \circ C_{loc}$$

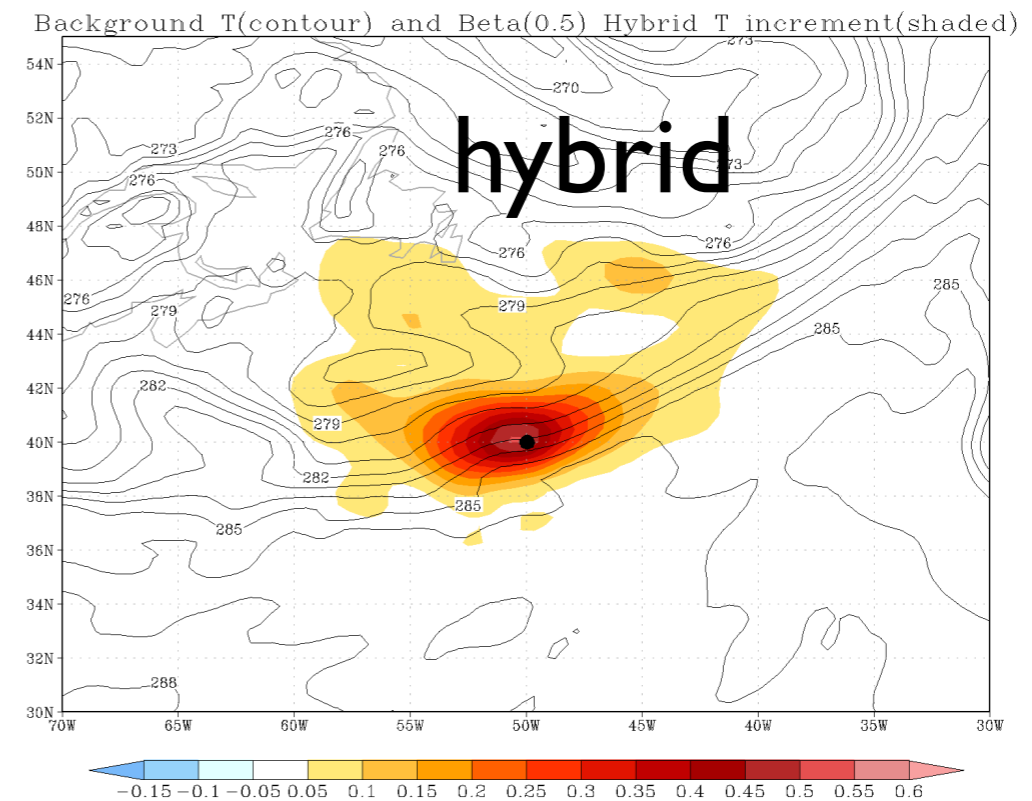
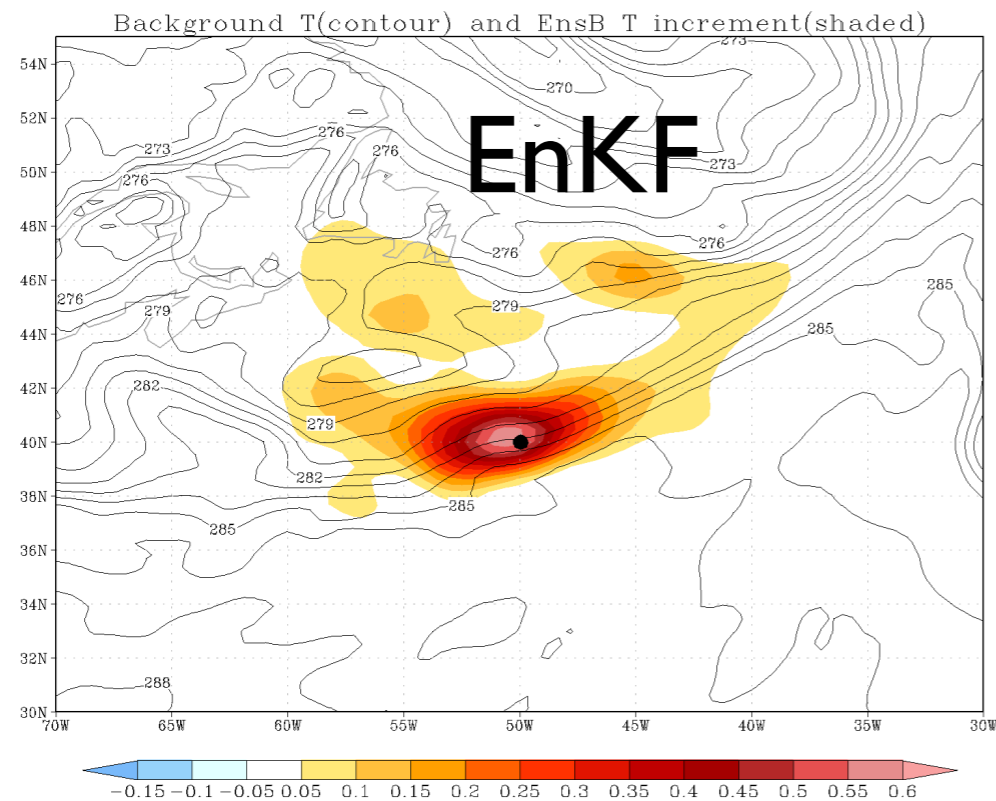
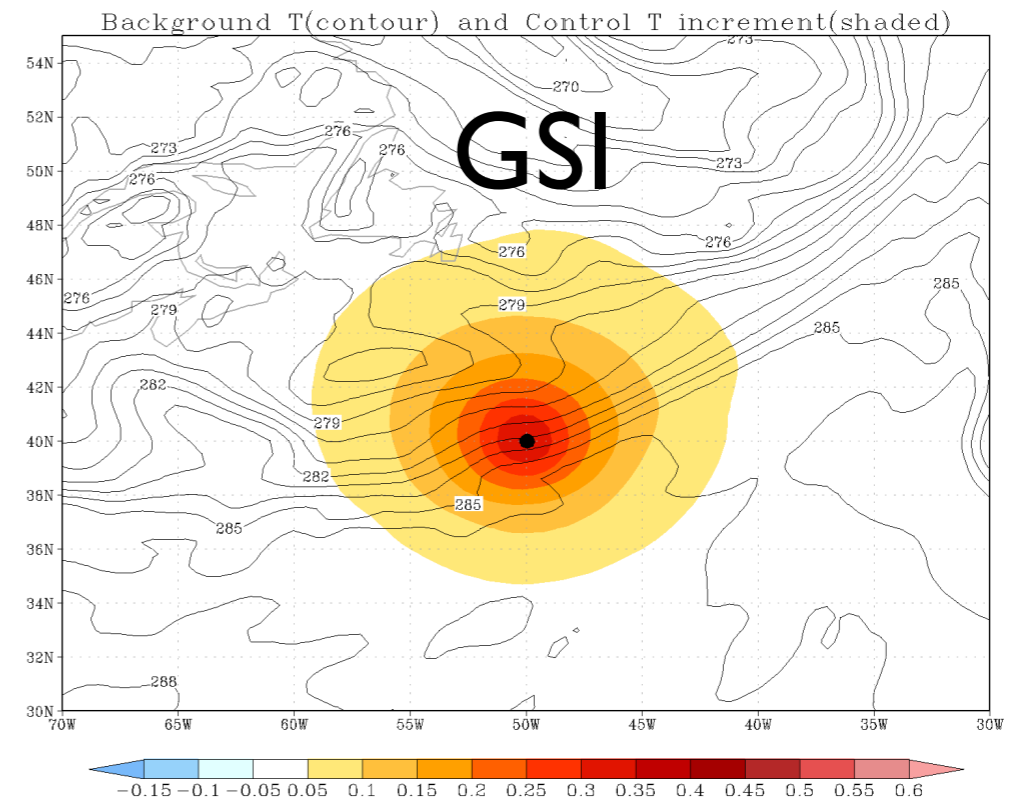
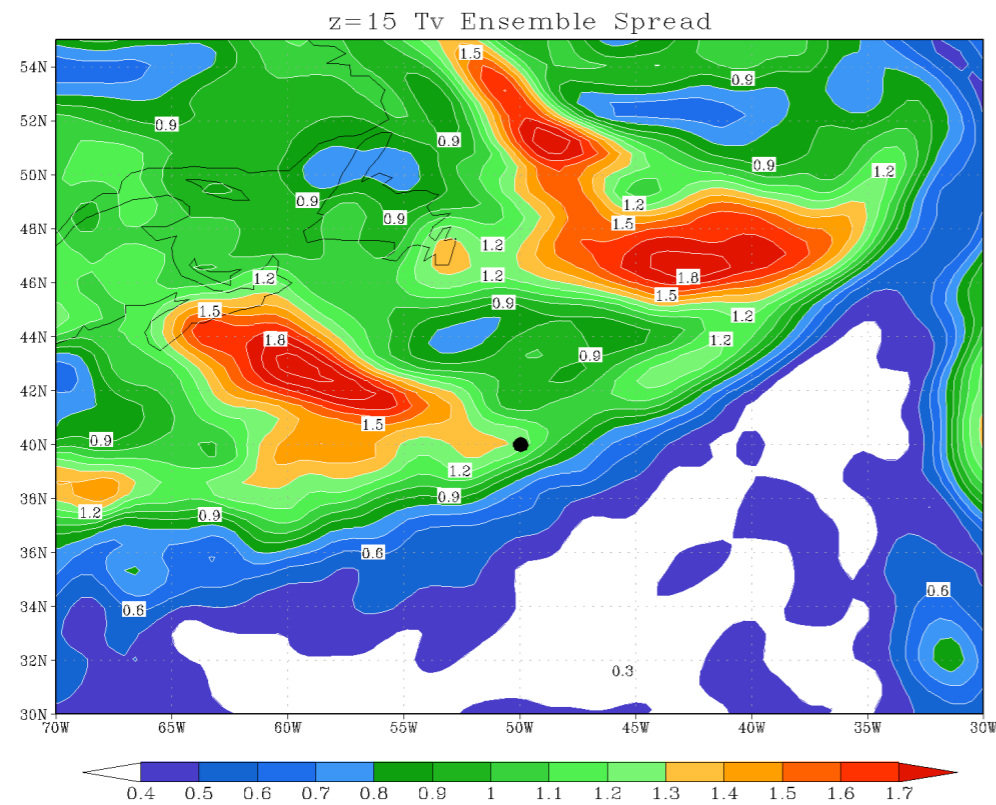
ETKF-3/4DVar algorithm

- Run ETKF to observation time (**low resolution**)
- Form new hybrid B matrix

$$B = \beta_c^2 B_c + \beta_e^2 P^b \circ C_{loc}$$

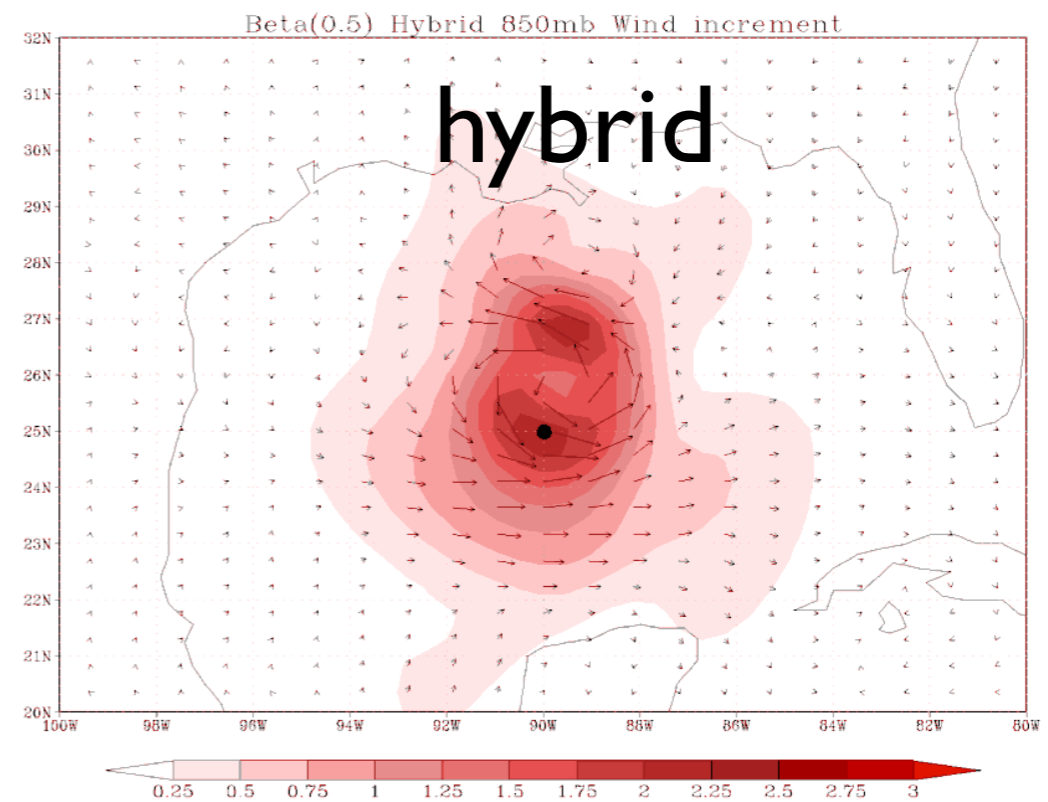
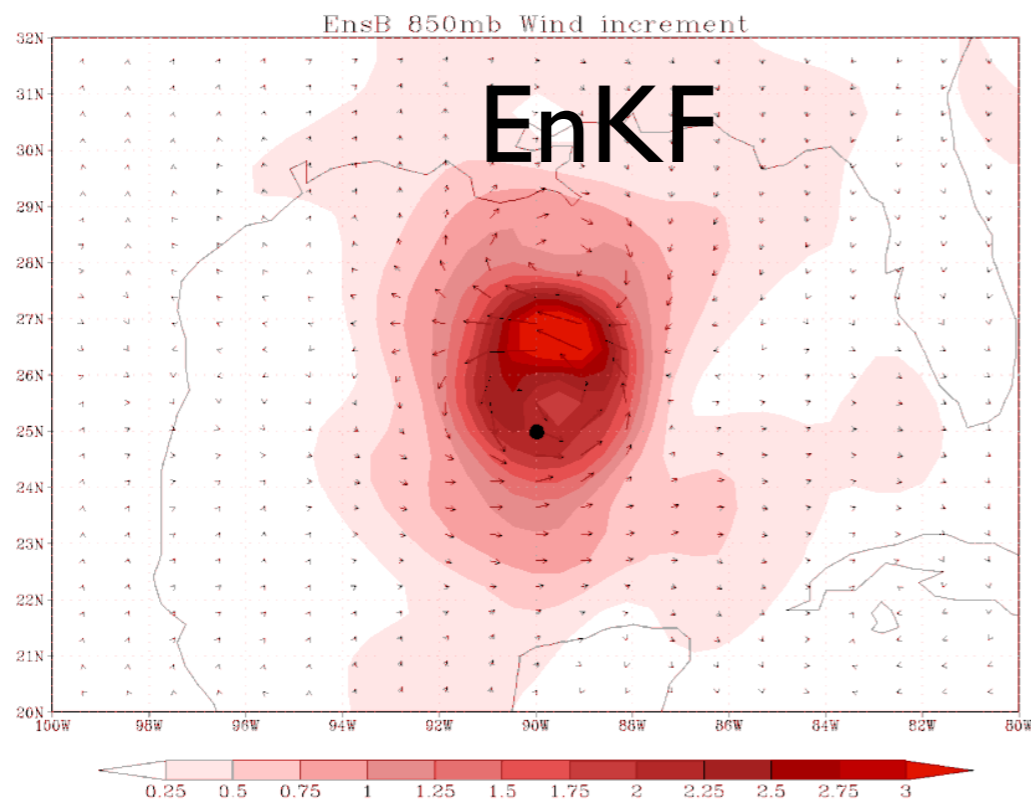
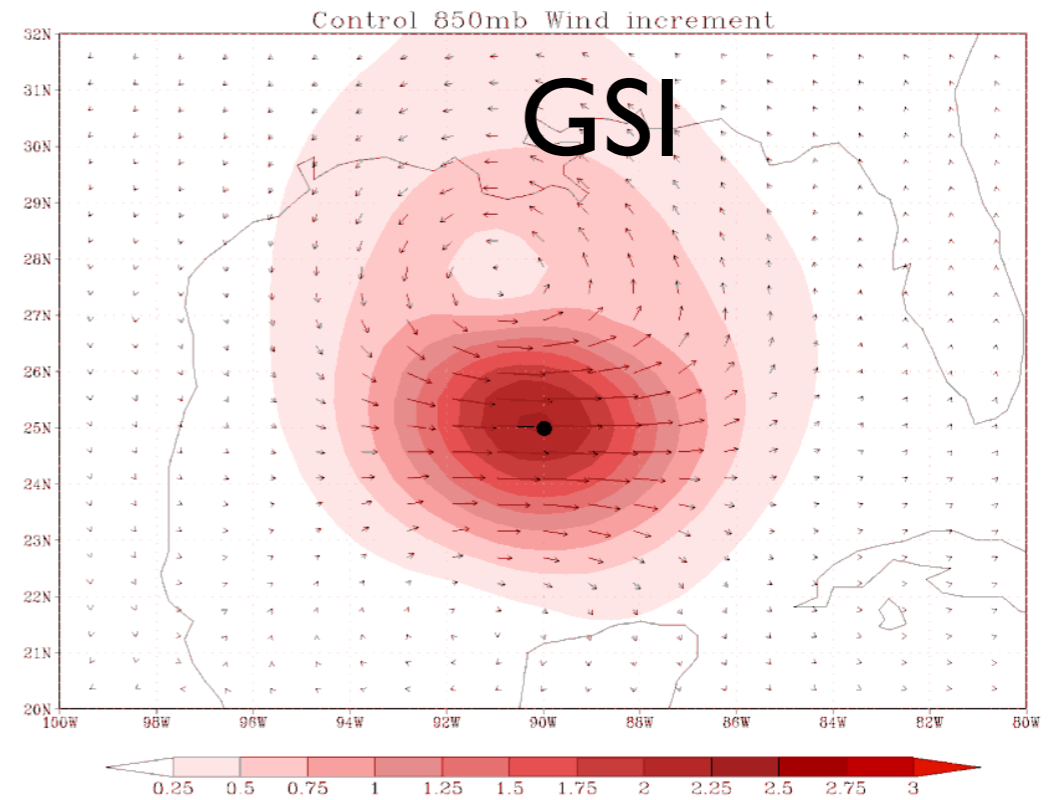
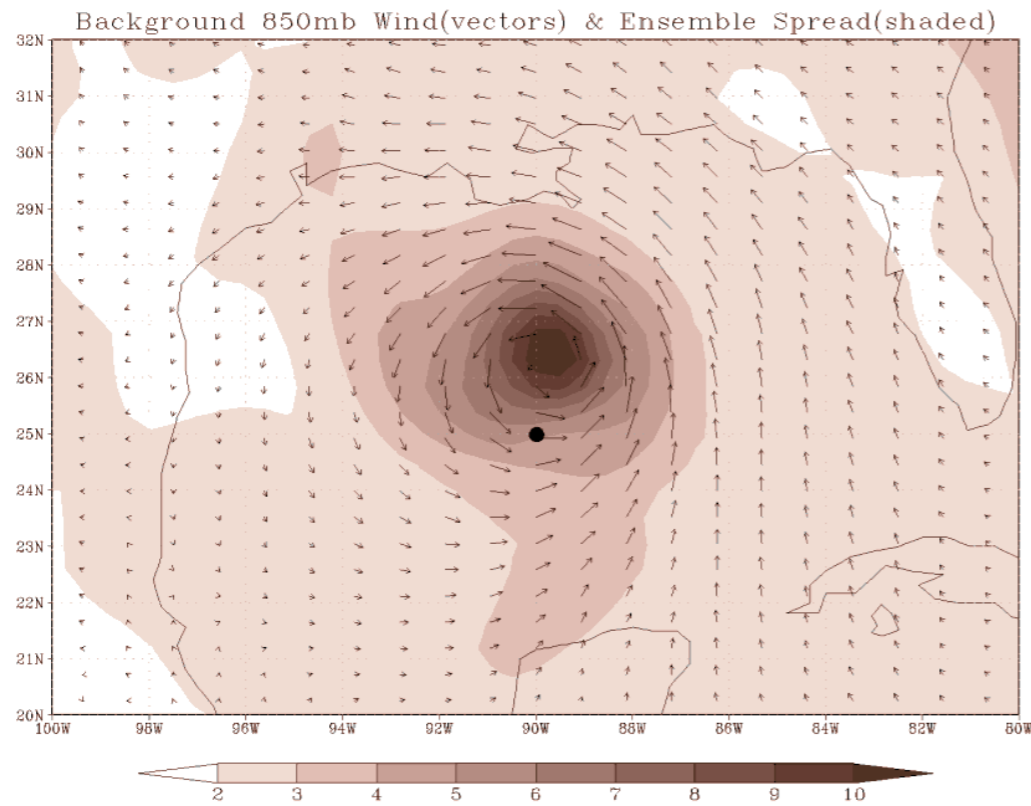
- Centre around 3/4Dvar forecast
- Precondition
- Calculate 3/4DVar solution(**high resolution**)
- Run ETKF to next observation time (**low resolution**)
- Etc.

Single-observation increments



Single 850 hPa T_v observation (1K O-F, 1K error)

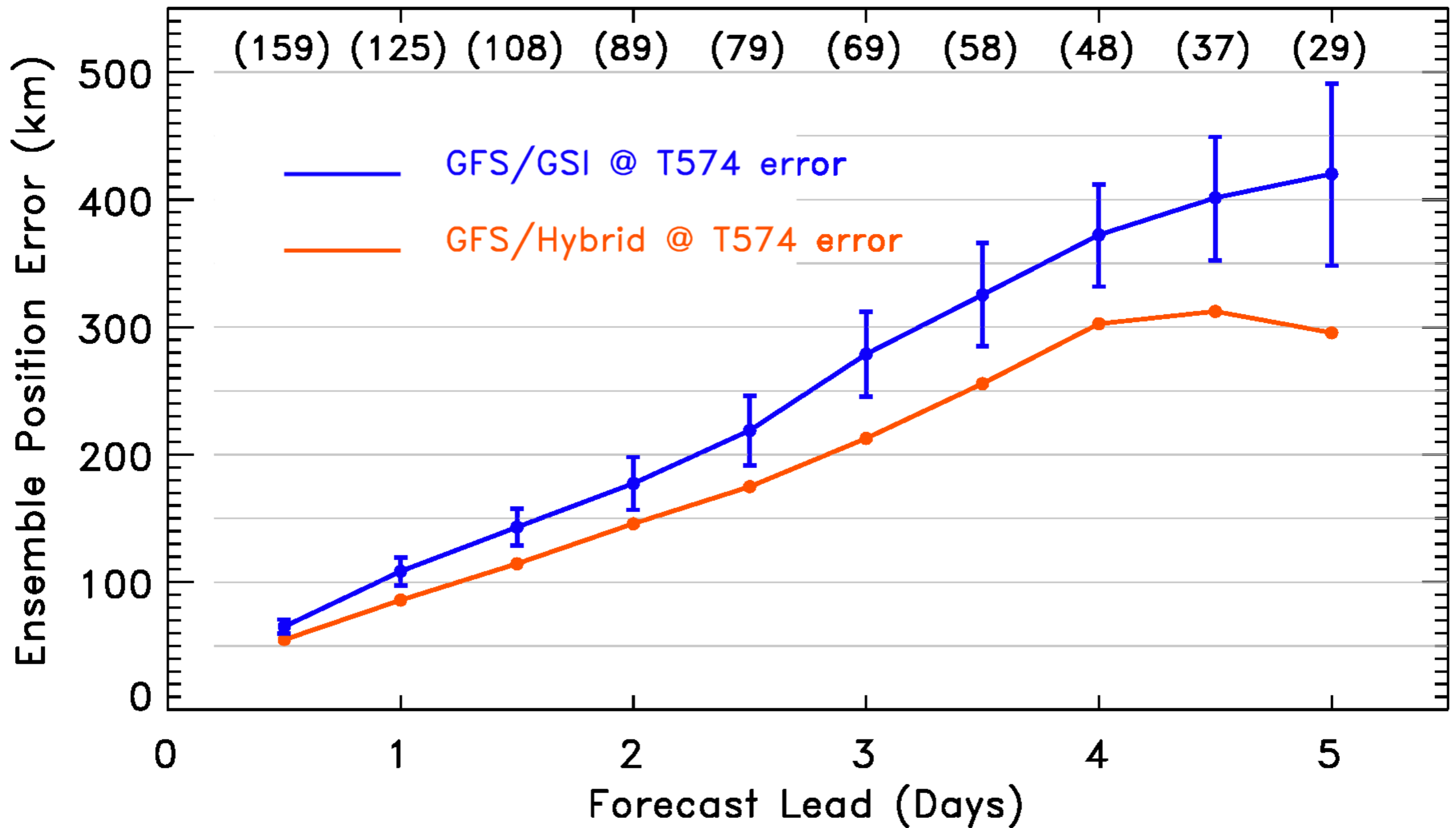
Single-observation increments



Single 850 hPa zonal wind observation (3 m/s O-F, 1m/s⁹error) Hurricane Ike

The excellent news

T574 GFS/GSI Deterministic vs.
T574 GFS/Hybrid Track Errors



Practical implementation

The 3DVar costfunction reads:

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

with now:
$$B = \beta_c^2 B_c + \beta_e^2 P^b \circ C_{loc}$$

Use alpha control variable transform:

$$\delta x = \beta_c B_c^{1/2} v + \beta_e X' \circ \alpha$$

Practical implementation

With this control variable transform

$$\delta x = \beta_c B_c^{1/2} v + \beta_e X' \circ \alpha$$

we find for the background term:

$$\begin{aligned} & \frac{1}{2} \delta x^T B^{-1} \delta x = \\ & \frac{1}{2} \left(\beta_c B_c^{1/2} v + \beta_e X' \circ \alpha \right)^T \left(\beta_c^2 B_c + \beta_e^2 P^b \circ C_{loc} \right)^{-1} \left(\beta_c B_c^{1/2} v + \beta_e X' \circ \alpha \right) = \\ & = \frac{1}{2} v^T v + \frac{1}{2} \alpha^T C_{loc}^{-1} \alpha \end{aligned}$$

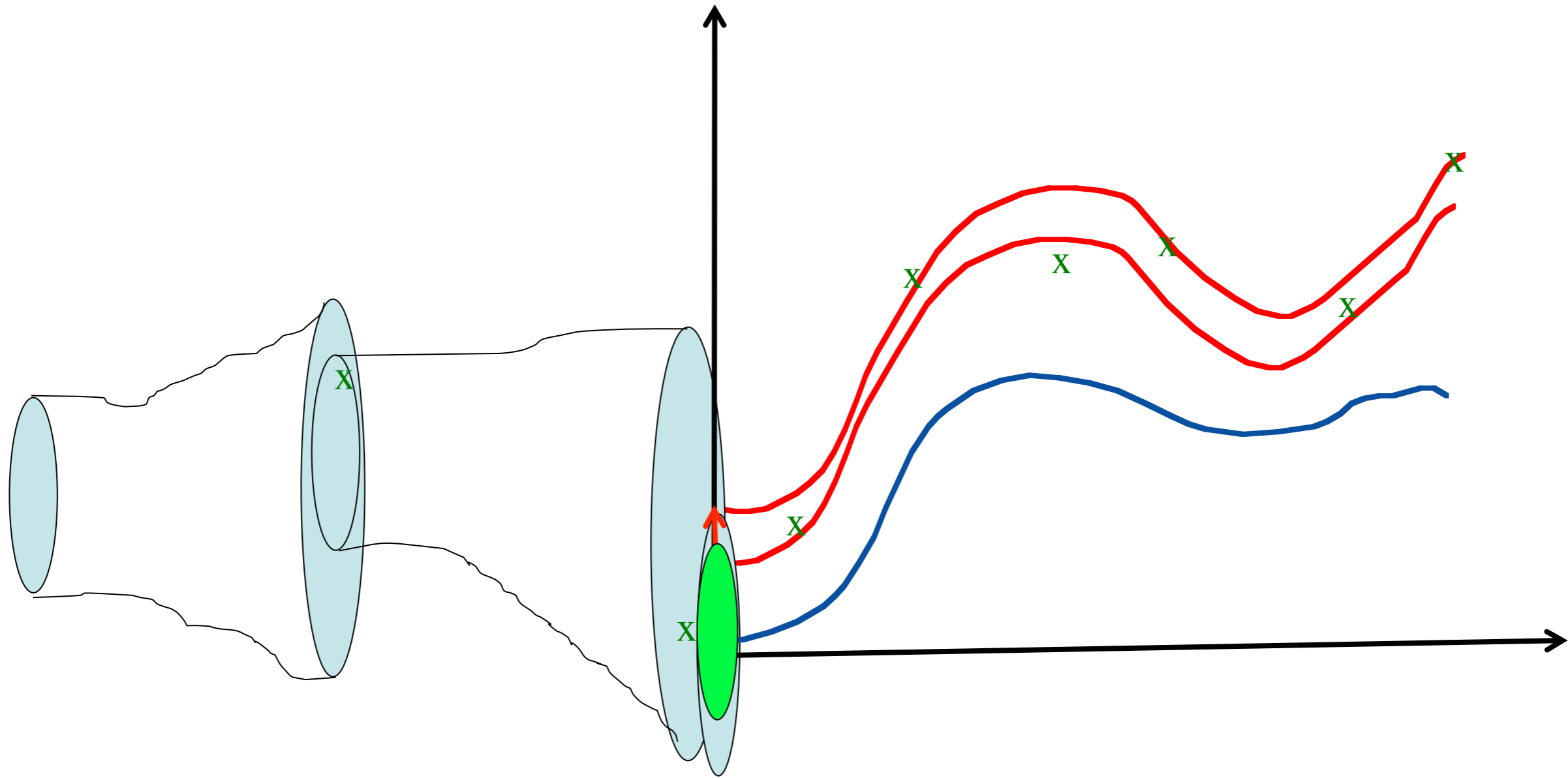
Meaning of control-variable transform

The control-variable transform can be written as:

$$\delta x = \beta_c B_c^{1/2} v + \beta_e X' \circ \alpha = \delta x_c + \delta x_e$$

- The increment is now a weighted sum of the static B_c component and the flow-dependent, ensemble based B_e
- The flow-dependent increment is a linear combination of ensemble perturbations X' , modulated by the α fields
- If the α fields were homogeneous δx_e only spans $N_{ens}-1$ degrees of freedom;
- We allow for varying α fields, which effectively increases the degrees of freedom
- C_{loc} is the localisation matrix for the flow-flow-dependent increments:
it controls the spatial variation of α

ETKF-4DVar, e.g. Met Office



$$B = \beta_c^2 B_c + \beta_e^2 P^b \circ C_{loc}$$

Practical implementation

ETKF-4DVar

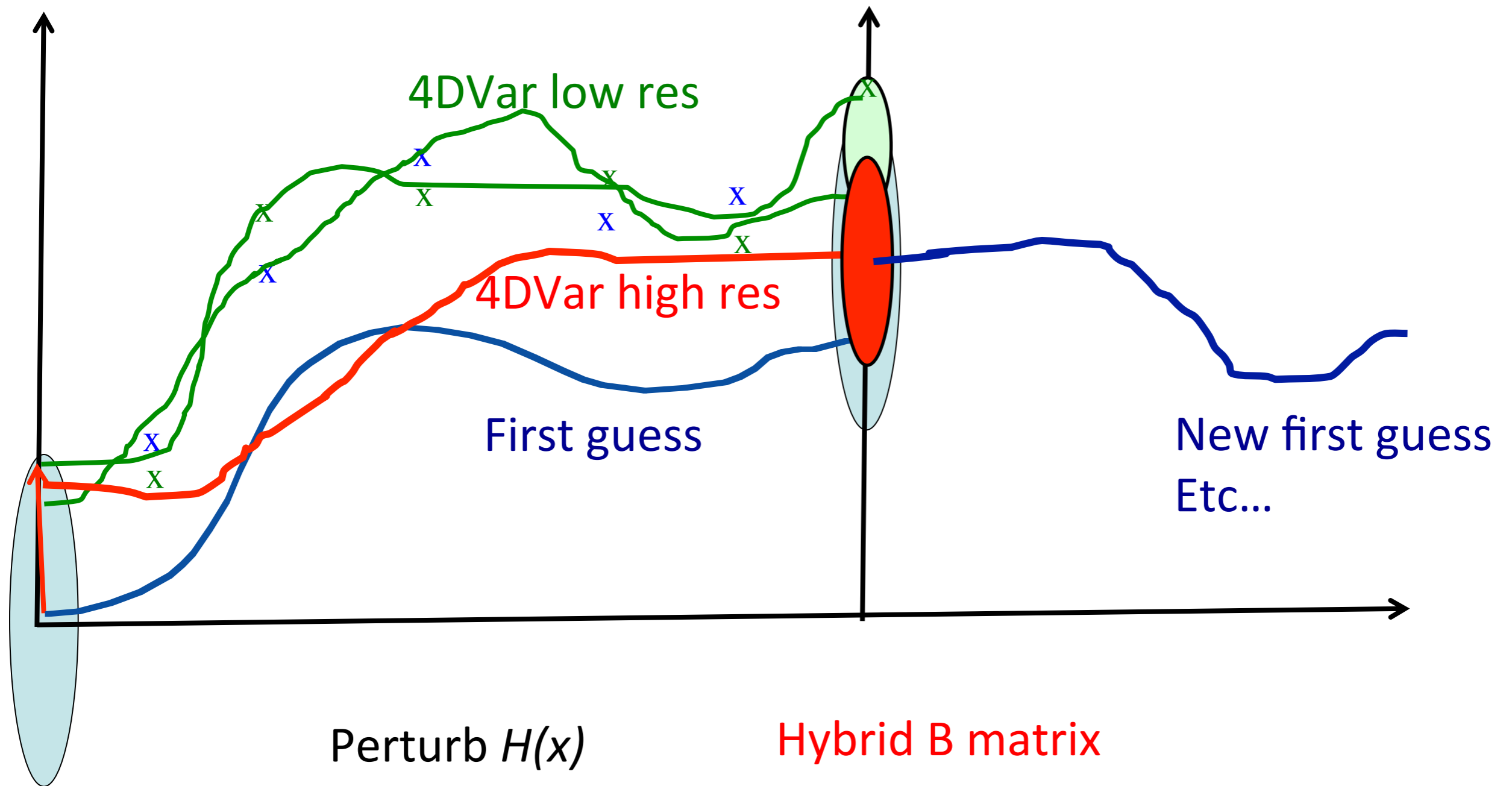
- Up to now this has only been implemented in strong-constraint 4Dvar
- Implementation similar to EnKF-3DVar.
- All usual tricks like preconditioning and incremental 4DVar are fully explored.

ETKF-4DVar

- 1) Prior is assumed Gaussian
- 2) B matrix is informed by previous observations, flow dependent B matrix
- 3) Error estimate from the EnKF ensemble
- 4) Natural ensemble prediction system
- 5) Possibility of getting stuck in local minima
- 6) Extra linearity by replacing ensemble mean by 4Dvar solution

2. EnsVars

'ensemble of data-assimilations' EDA



EnsVars algorithm

- Perturb $H(x)$ in 4Dvar costfunction N_e times ('perturb observations')
- Solve 4DVar for each of these costfunctions (**low resolution**)
- Solve high-res 4Dvar over same window
- Use ensemble of 4Dvar ensemble forecasts to form hybrid B matrix for 4DVars of next time window.
- All usual tricks like preconditioning and incremental 4DVar are fully explored.

Characteristics of EnsVars

- 1) Prior is assumed Gaussian
- 2) B matrix is informed by previous observations, flow dependent B matrix
- 3) Error estimate from the low-res 4DVar ensemble
- 4) Natural ensemble prediction system
- 5) Possibility of getting stuck in local minima
- 6) Extra linearity by replacing ensemble mean by high-res 4Dvar solution

Conclusions hybrid methods

Variational methods are the most popular data-assimilation scheme, but use static B matrix and need adjoint.

Ensemble-Kalman Filters are used to make B flow dependent: ETKF-4DVar, EnsVars (EDA), 4DEnsVar. The latter doesn't need an adjoint.

Strong-constraint 4DEnsVar, in which the space-time covariances are generated by an ensemble run has issues with localisation.

Weak-constraint 4DEnsVar partially solves localisation issue.

MORE WORK IS NEEDED !