

An aerial photograph of a town, likely in the Pyrenees region, is shown from a high angle. The town is surrounded by green hills and is partially obscured by a thick layer of white clouds or fog. Overlaid on the bottom left of the image is a weather map showing isobars (lines of equal atmospheric pressure) and wind vectors (arrows). The isobars are labeled with values such as 1010, 1015, 1020, 1025, 1030, 1035, and 1040. The wind vectors are represented by white arrows with black outlines, indicating the direction and strength of the wind. The background of the slide is a dark blue gradient with a stylized sun and cloud icon in the top left corner.

A possible implementation of the 4D-Var based on a 4D-ensemble

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dépasser les frontières



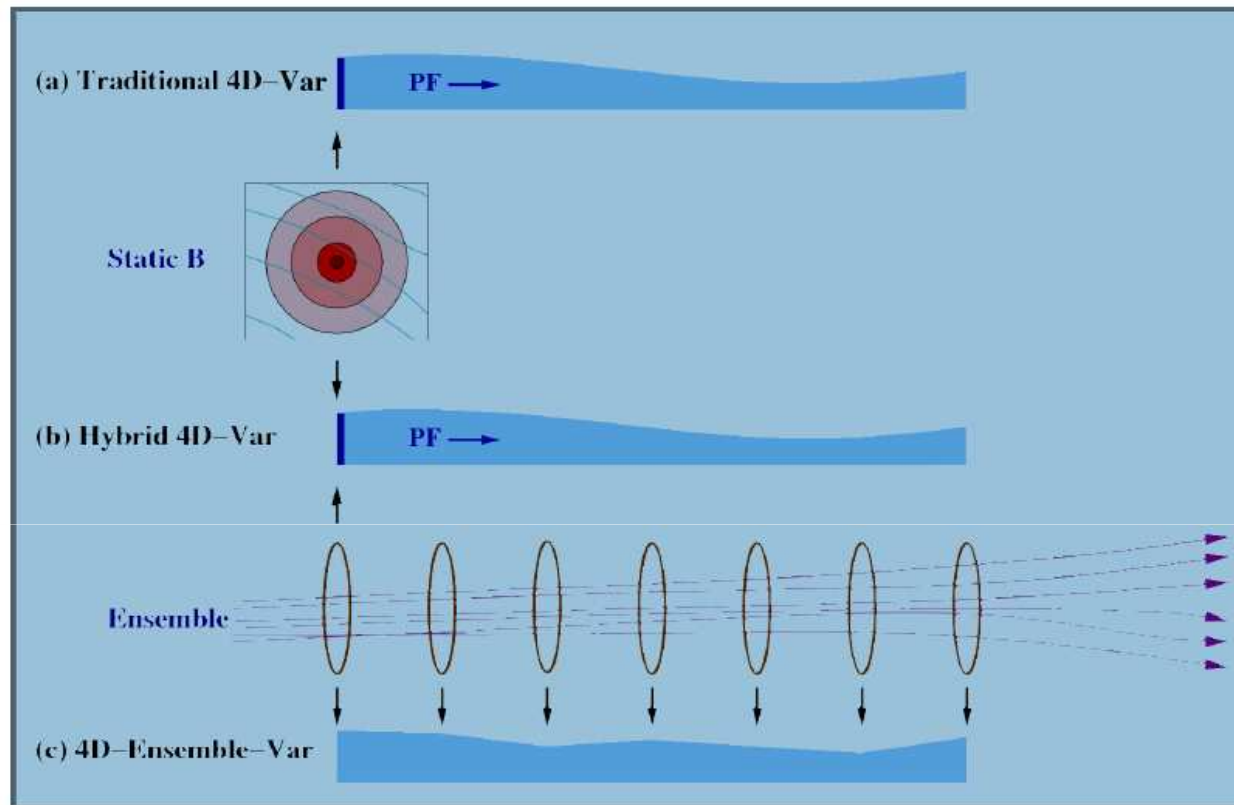
METEO FRANCE
Toujours un temps d'avance

Introduction

- 4D-Var
 - ✓ Simplified description of \mathbf{B} at initial time.
 - ✓ Possible improv. via an ens. of pert. 4D-Var (Météo-France, ECMWF) : spat./temp. variations of error variances and correlations (wavelets).
 - ✓ Difficult development and maintenance of TL/AD.
 - ✓ Poor scalability of TL/AD
 - ✓ Linear evolution of covariances.

- 4D-Var based on a 4D ensemble : 4D-En-Var
 - ✓ Similar to EnKF.
 - ✓ Keeps benefits of 4D-Var (global analysis, add. terms, outer-loop, ...)
 - ✓ Localization of the raw covariances made in model space.
 - ✓ Minimization cost similar to 3D-Var.
 - ✓ Natural parallelization.
 - ✓ Covariances given by NL forecasts.

4D-Var / Hybrid 4D-Var / 4D-En-Var



Barker and Clayton, 2011

4D-En-Var formulation

- Minimization of

$$J(\underline{\delta x}) = \underline{\delta x}^T \underline{B}^{-1} \underline{\delta x} + (\underline{d} - \underline{H} \underline{\delta x})^T \underline{R}^{-1} (\underline{d} - \underline{H} \underline{\delta x}), \text{ with}$$

- d 4D vector of the innovations distributed in time,
- H 4D linearized observation operator,
- R 4D (but diagonal!) covariance matrix of obs. errors,
- δx 4D vector of the increments to be added to the 4D bg x^b ,
- B 4D covariance matrix of bg errors, given by an ensemble.

(Lorenc, 2012)

4D-En-Var formulation

$$\underline{\mathbf{X}}^f = (\underline{\mathbf{x}}^{f_1}, \dots, \underline{\mathbf{x}}^{f_L}),$$

where L is the ensemble size and

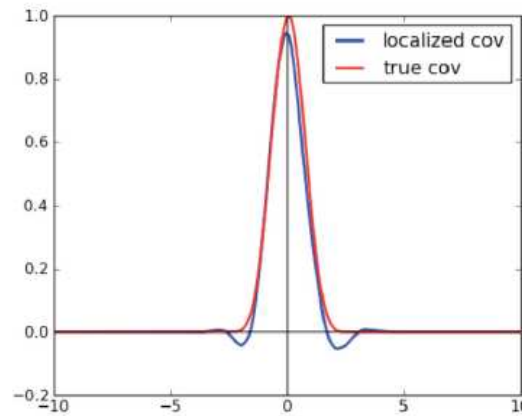
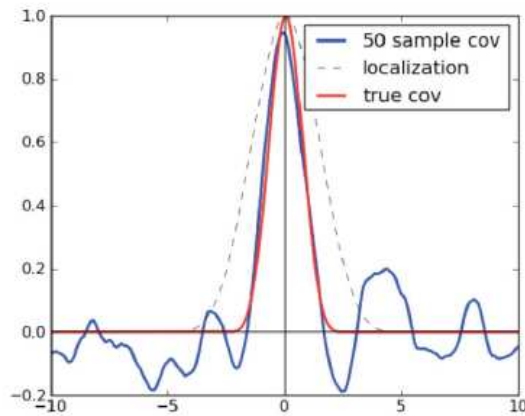
$$\underline{\mathbf{x}}^{f_\ell} = \underline{\mathbf{x}}_{\ell}^f - \langle \underline{\mathbf{x}}^f \rangle / (L-1)^{1/2}, \ell=1,L,$$

are the deviations of the 4D pert. forecasts from the mean ens. traject.

$$\begin{aligned} \underline{\mathbf{P}} &= \underline{\mathbf{X}}^f (\underline{\mathbf{X}}^f)^T \\ &= \begin{pmatrix} \mathbf{P}_{0,0} & \mathbf{P}_{0,1} & \dots & \mathbf{P}_{0,K-1} \\ & \mathbf{P}_{1,1} & & \\ & & & \\ & \mathbf{P}_{K-1,0} & \dots & \mathbf{P}_{K-1,K-1} \end{pmatrix} \end{aligned}$$

where K is the number of times in the assimilation period.

Localization of the covariances



Whitaker, 2011

$$\underline{\mathbf{B}} = \underline{\mathbf{P}} \circ \underline{\mathbf{C}}$$

$$= \begin{pmatrix} \mathbf{P}_{0,1} & \mathbf{P}_{0,1} & \dots & \mathbf{P}_{0,K-1} \\ & \mathbf{P}_{1,1} & & \\ & & & \\ \mathbf{P}_{K-1,0} & \dots & & \mathbf{P}_{K-1,K-1} \end{pmatrix} \circ \begin{pmatrix} \mathbf{C}_{0,1} & \mathbf{C}_{0,1} & \dots & \mathbf{C}_{0,K-1} \\ & \mathbf{C}_{1,1} & & \\ & & & \\ \mathbf{C}_{K-1,0} & \dots & & \mathbf{C}_{K-1,K-1} \end{pmatrix}$$

Implementations of 4D-En-Var

- $\delta \mathbf{x} = \mathbf{B}^{1/2} \boldsymbol{\chi}$
 $= (\mathbf{P} \circ \mathbf{C})^{1/2} \boldsymbol{\chi}$
 $= \sum_{\ell=1,L} \mathbf{x}_{\ell}^{f_{\ell}} \circ (\mathbf{C}^{1/2} \boldsymbol{\chi}_{\ell})$

$$\mathbf{J}^b(\boldsymbol{\chi}) = \sum_{\ell=1,L} \boldsymbol{\chi}_{\ell}^T \boldsymbol{\chi}_{\ell}, \dim \boldsymbol{\chi} = N(N_c) \times L \text{ (or } K \times N_c \times L \text{ in 4D with model error)}$$

Use of a Conjugate Gradient (CG) with $\mathbf{B}^{1/2}$ change of variables.

(Buehner 2005, 2010)

- $\delta \mathbf{x} = \sum_{\ell=1,L} \mathbf{x}_{\ell}^{f_{\ell}} \circ \boldsymbol{\alpha}_{\ell}$, with $\boldsymbol{\alpha}_{\ell} = \mathbf{C}^{1/2} \boldsymbol{\chi}_{\ell}$

$$\mathbf{J}^b(\boldsymbol{\alpha}) = \sum_{\ell=1,L} \boldsymbol{\alpha}_{\ell}^T \mathbf{C}^{-1} \boldsymbol{\alpha}_{\ell}, \dim \boldsymbol{\alpha} = N(N_c) \times L \text{ (or } K \times N_c \times L \text{ in 4D with mod. error)}$$

Use of a Double Preconditioned CG (DPCG) with \mathbf{C} preconditioning.

(Lorenc, 2003; Wang et al 2007; Wang 2010)

Possible alternative impl. of 4D-En-Var: 4D B preconditioning

Use of DPCG, with 4D covariance matrix B and Hessian $J'' = \underline{B}^{-1} + \underline{H}^T \underline{R}^{-1} \underline{H}$.
In this case, the dimension of the control variable $\delta \underline{x}$ is $K \times N$.

$$\delta \underline{x}_0 = \mathbf{0}$$

$$\underline{g}_0 = -\underline{H}^T \underline{R}^{-1} \underline{d}$$

$$\underline{h}_0 = \underline{B} \underline{g}_0$$

$$\underline{d}_{-1} = \underline{e}_{-1} = \mathbf{0}$$

loop over n

$$\underline{d}_n = -\underline{h}_n + \beta_{n-1} \underline{d}_{n-1}$$

$$\underline{e}_n = -\underline{g}_n + \beta_{n-1} \underline{e}_{n-1}$$

$$\underline{f}_n = \underline{e}_n + \underline{H}^T \underline{R}^{-1} \underline{H} \underline{d}_n$$

$$\alpha_n = \underline{g}_n^T \underline{h}_n / \underline{d}_n^T \underline{f}_n$$

$$\underline{g}_{n+1} = \underline{g}_n + \alpha_n \underline{f}_n$$

$$\underline{h}_{n+1} = \underline{B} \underline{g}_{n+1}$$

$$\delta \underline{x}_{n+1} = \delta \underline{x}_n + \alpha_n \underline{d}_n$$

$$\beta_n = \underline{g}_{n+1}^T \underline{h}_{n+1} / \underline{g}_n^T \underline{h}_n$$

(Derber and Rosati, 1989; El Akkraoui et al, 2012)

Multiplication by B

- Application of DPCG : $\underline{\mathbf{h}} = \underline{\mathbf{B}} \mathbf{g}$

- $\underline{\mathbf{B}} = \underline{\mathbf{P}} \circ \underline{\mathbf{C}}$

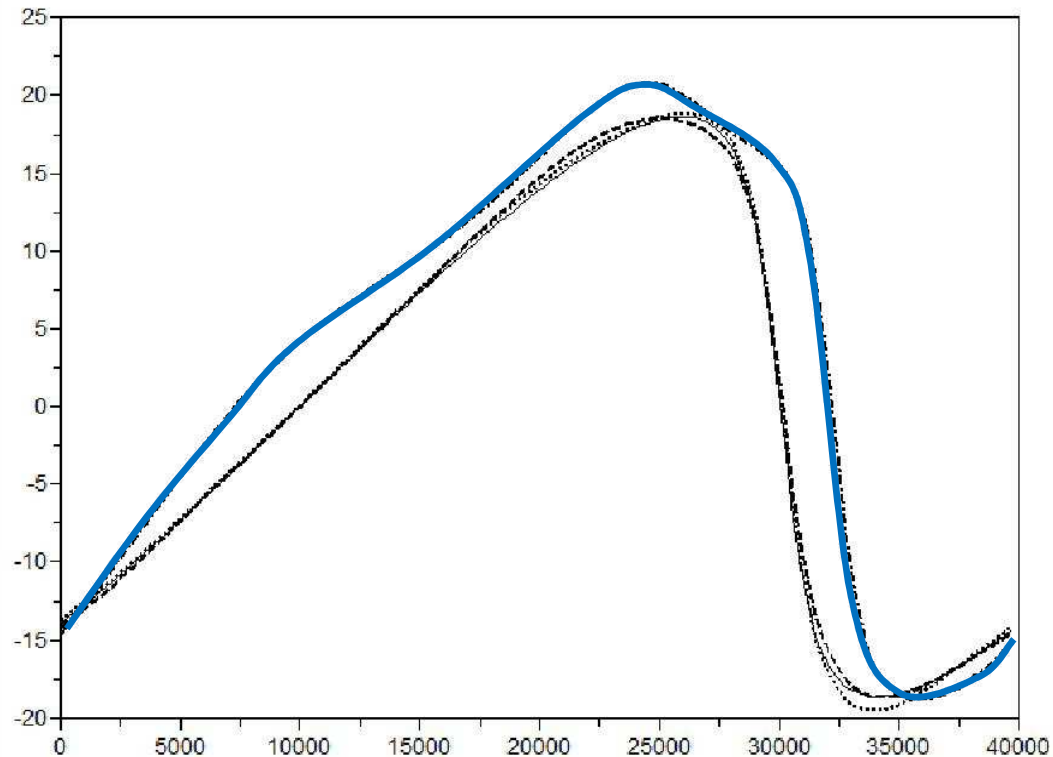
$$\mathbf{C} = \mathbf{S}^{-1} \mathbf{C}^s \mathbf{S}$$

- $\mathbf{B}_{k,k'} \mathbf{g} = (\mathbf{X}_{k'}^{f'} (\mathbf{X}_k^{f'})^T \circ \mathbf{C}) \mathbf{g}$
 $= \sum_{\ell=1,L} \mathbf{x}_{k,\ell}^{f'} \circ \mathbf{S}^{-1} (\mathbf{C}^s \mathbf{S} (\mathbf{x}_{k',\ell}^{f'} \circ \mathbf{g})),$

where k and k' are two time indexes, and omitting the vertical dimension.

- Compact and meaningful expression of blocks of B.
- No need of extra variables (α_ℓ of χ_ℓ , with $\alpha_\ell = \mathbf{C}^{1/2} \chi_\ell$).
- Can be parallelized over K (time), N (space), L (ensemble), ...

Application to the Burgers model

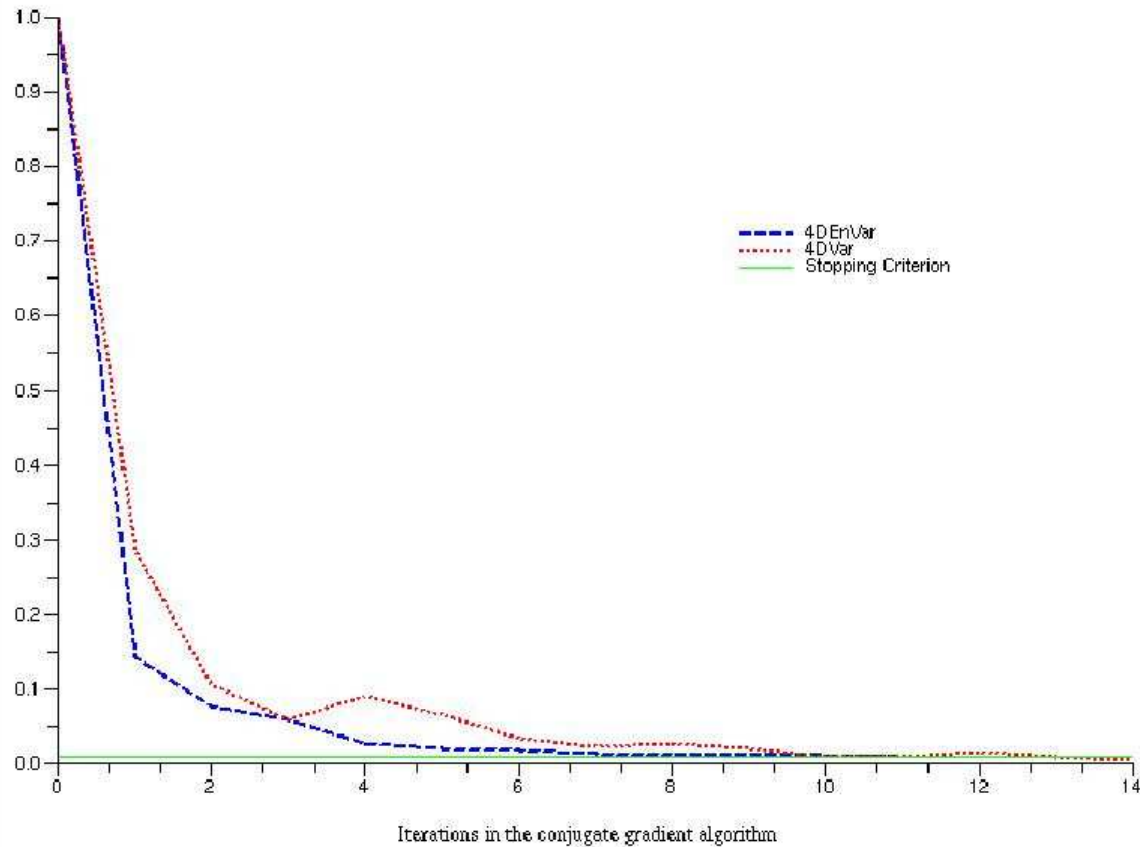


Forecasts from

- reference (black solid)
- x^b (blue dashed-dotted)
- x^a 4D-En-Var (black dashed)
- x^a 4D-Var (black dotted)

Convergence

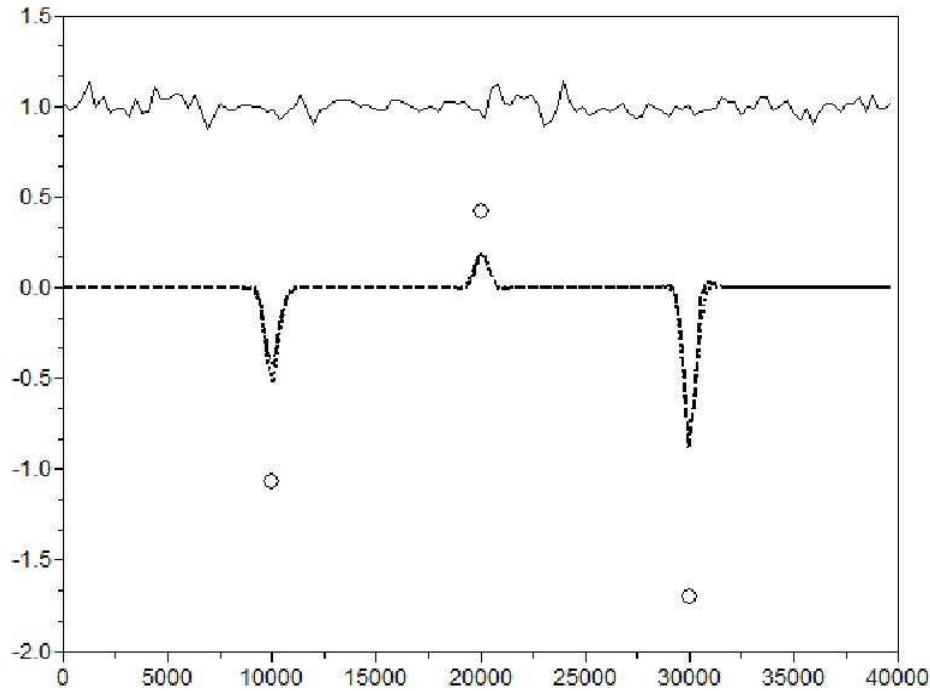
4D-En-Var / 4D-Var



Decrease of the gradient norm

- 4D-En-Var (blue dashed)
- 4D-Var (red dotted)

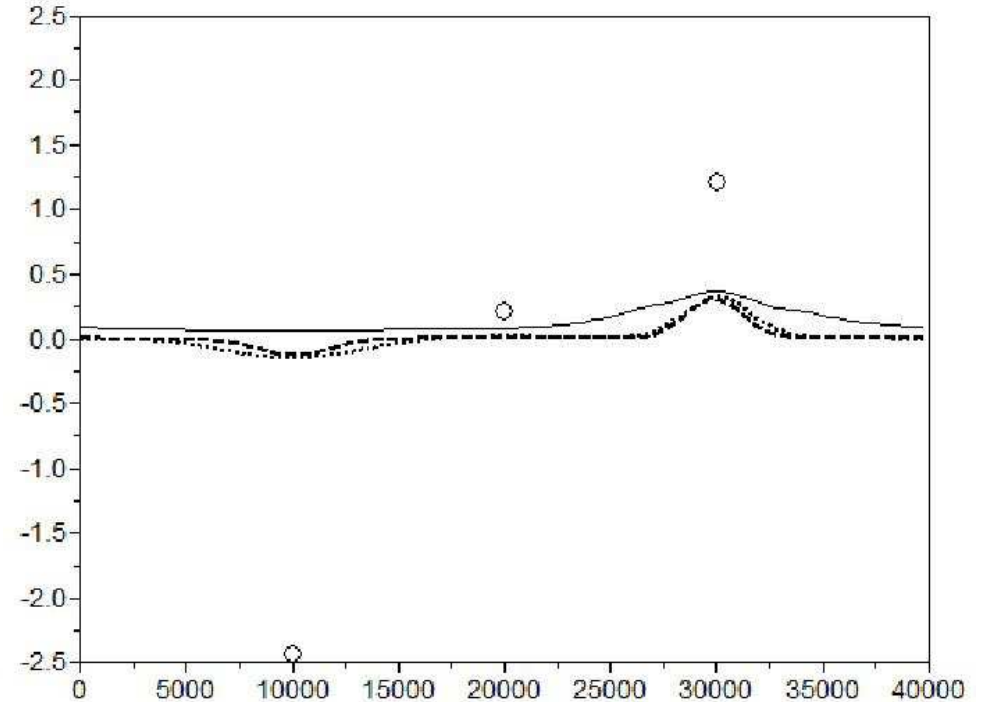
Increment comparison 4D-En-Var / 4D-Var



Observations at t_0

δx_0 at t_0 :

- 4D-En-Var (dashed)
- 4D-Var (dotted)
- bg error square-root (solid)

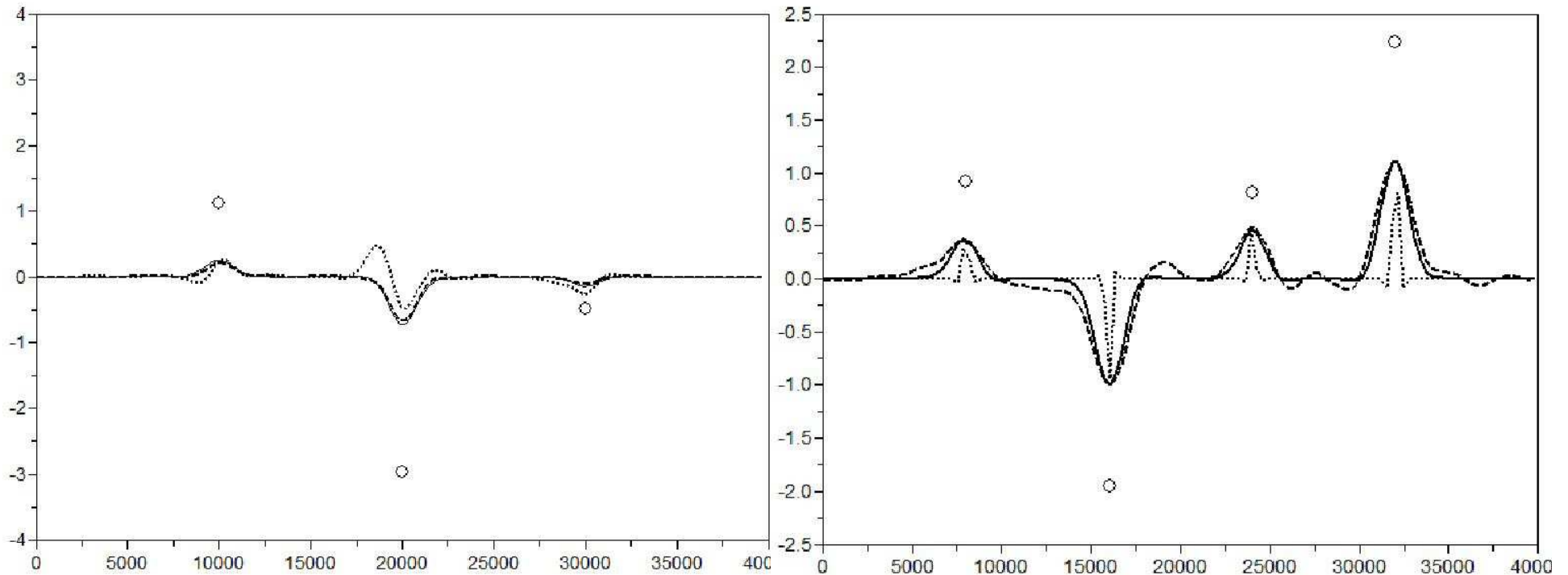


Observations at t_f ($t_0 + 6h$)

δx_f at t_f :

- 4D-En-Var (dashed)
- 4D-Var (dotted)
- bg error square-root (solid)

4D-En-Var parameters



Ensemble size:

- L = 5 (dotted)
- L = 100 (dashed)
- L = 1000 (solid)

Localization length scale:

- LC = 50 km (dotted)
- LC = 1200 km (solid)
- LC = 5000 km (dashed)

Conclusion

- 4D-En-Var : combination of 4D-Var and EnKF.
- Treats all observations simultaneously.
- Localization of covariances made in model space.
- Natural parallelization over time, space and ensemble.
- Possible representation of model error inside the assimilation window.
- Efficient preconditioning available (as in strong constraint 4D-Var).
- DPCG with 4D ens. $\underline{\mathbf{B}}$ matrix : possible implementation of 4D-En-Var.