



Overview of the COSMO Priority Project **KENDA** for Km-Scale Ensemble-Based Data Assimilation

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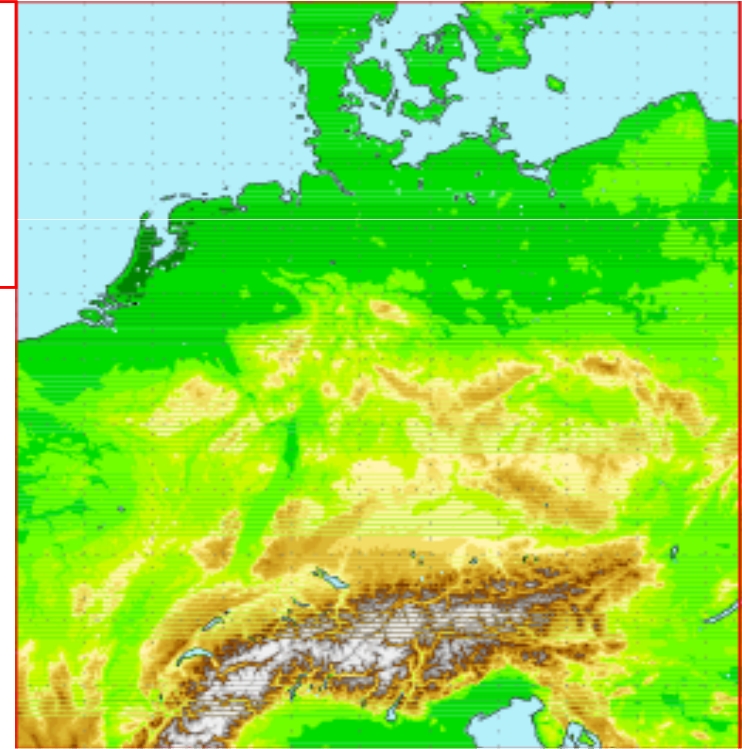
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KENDA : Why develop Ensemble-Based Data Assimilation ?

COSMO-DE: $\Delta x = 2.8 \text{ km}$
(deep convection explicit,
shallow convection param.)
domain size : $\sim 1250 \times 1150 \text{ km}$



convection-permitting NWP:

after 'few' hours,

a forecast of convection is a long-term forecast

→ deliver probabilistic (pdf) rather than deterministic forecast

→ need ensemble forecast and data assimilation system

→ forecast component: COSMO-DE EPS

KENDA : Why develop Ensemble-Based Data Assimilation ?



→ data assimilation: priority project within COSMO consortium

Km-scale ENsemble-based Data Assimilation (KENDA):

→ **Local Ensemble Transform Kalman Filter (LETKF)** ,
(because of its relatively low computational costs)

This talk: - method and implementation
- some scientific issues

talk by Hendrik Reich : preliminary experiments at DWD

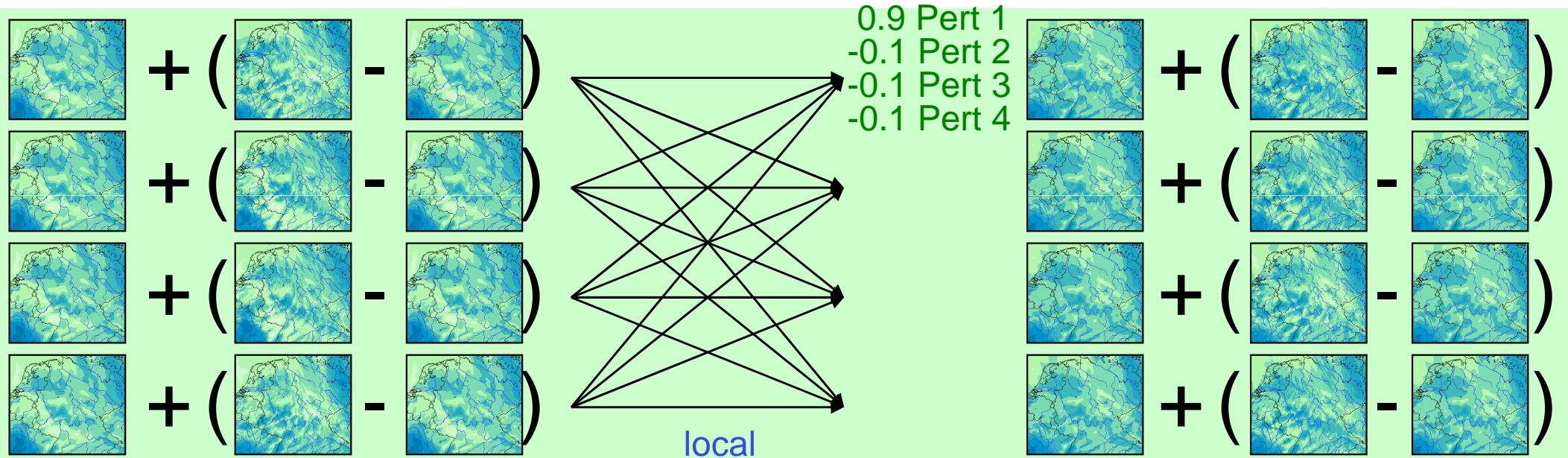


Assume: Gaussian errors

$$J(\mathbf{x}) = (\mathbf{x} - \bar{\mathbf{x}}^b)^T \mathbf{P}^{b-1} (\mathbf{x} - \bar{\mathbf{x}}^b) + [\mathbf{y}^o - H(\mathbf{x})]^T \mathbf{R}^{-1} [\mathbf{y}^o - H(\mathbf{x})]$$

for $\mathbf{P}^b = (k - 1) \mathbf{X}^b (\mathbf{X}^b)^T$, $J(\mathbf{x})$ is well-defined in sub-space S spanned by \mathbf{X}^b

Local Ensemble Transform Kalman Filter LETKF (Hunt et al., 2007)



ensemble mean forecast + k perturbed forecasts - ensemble mean fcst.

forecast perturbations \mathbf{X}^b

flow-dep background error cov.

$$\mathbf{P}^b = (k - 1) \mathbf{X}^b (\mathbf{X}^b)^T$$

in the $(k-1)$ -dimensional (!) sub-space S spanned by background perturbations :

$$\mathbf{x} = \bar{\mathbf{x}}^b + \mathbf{X}^b \mathbf{w}$$

set up cost function $J(\mathbf{w})$ in ensemble space,
explicit solution for minimisation (Hunt et al., 2007)

analysis mean $\bar{\mathbf{w}}^a$

analysis perturbations $\mathbf{W}^{a(i)}$

analysis error covariance
(computed only in ensemble space)

$$\mathbf{w}^{a(i)} = \bar{\mathbf{w}}^a + \mathbf{W}^{a(i)}$$

perturbed analyses

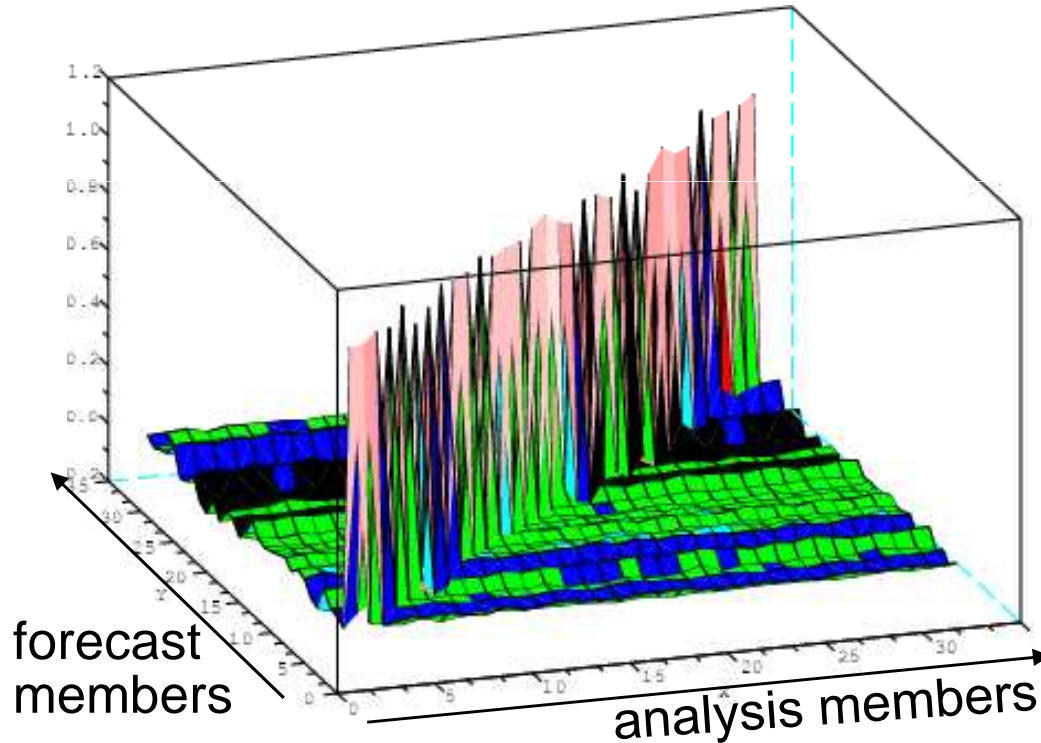




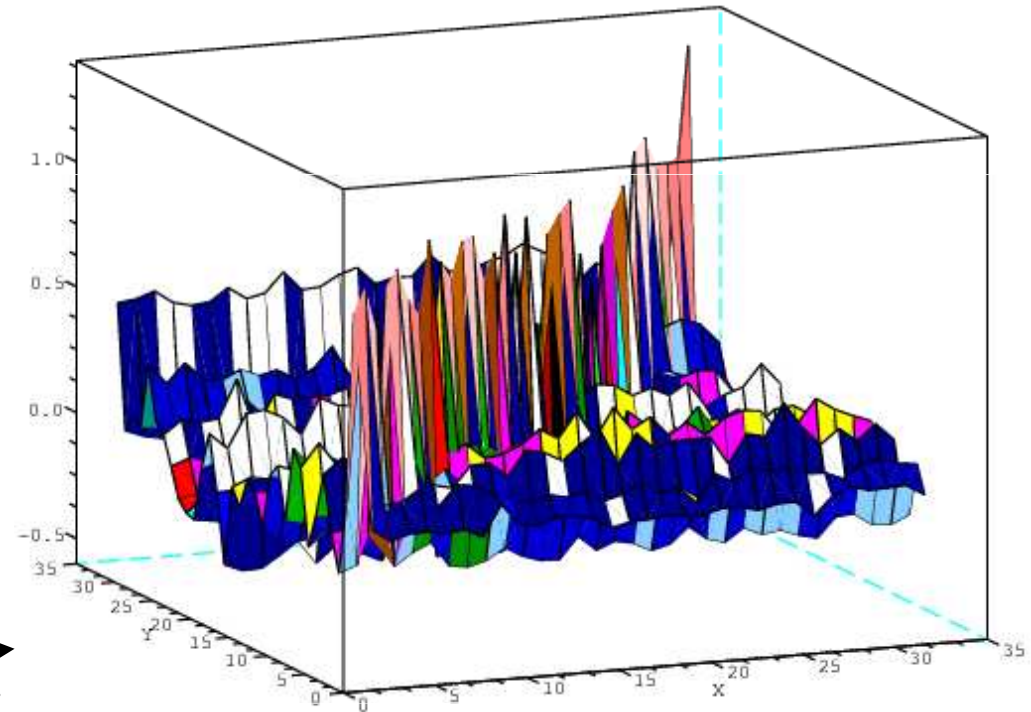
- implementation following Hunt et al., 2007
- basic idea: do analysis in the space of the ensemble perturbations
 - computationally efficient, but also restricts corrections to
subspace spanned by the ensemble
 - **explicit localization** (doing separate analysis at every grid point,
select only obs in vicinity)
 - analysis ensemble members are locally **linear combinations**
of first guess ensemble members



weight matrices (transform matrices) : determine linear combination



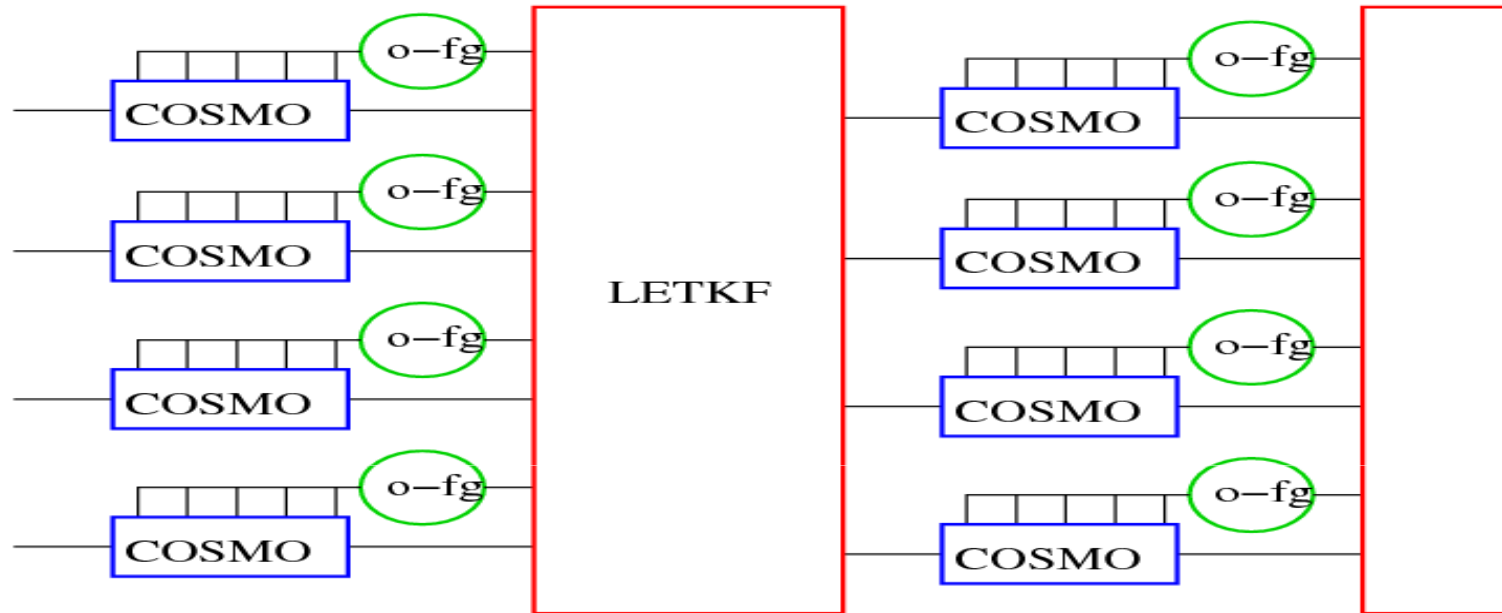
normal values for obs errors R
(for a grid point influenced by
> 200 conventional obs)



R divided by 100
(simulating many more obs)

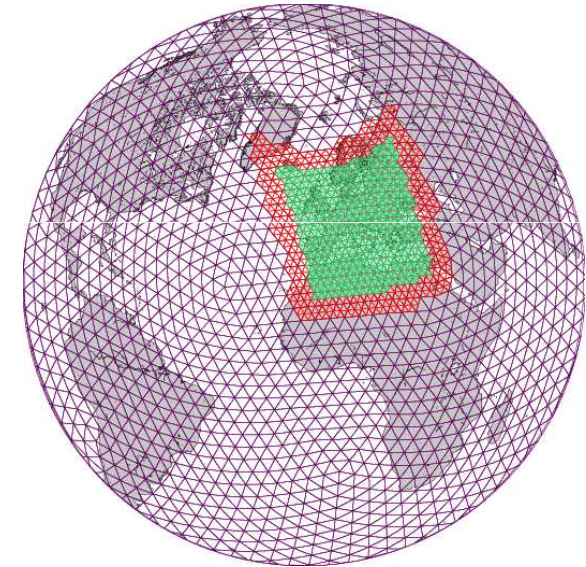
- diagonal elements \gg off-diagonal elements (→ analysis increments 'small')
- 'good' forecast members get larger weight in all analysis members

- analysis step (LETKF) outside COSMO code
 - ensemble of independent COSMO runs up to next analysis time (collecting obs – f.g. → 4D -LETKF)
 - separate analysis step code, LETKF included in 3DVAR package of DWD



- basically for verification purposes, COSMO obs operators incl. quality control will be implemented in 3DVAR / LETKF environment
 - future: hybrid 3DVAR-EnKF approaches in principle applicable to COSMO

- lateral BC :
 - future: from global EnKF / EPS based on ICON (non-hydrostatic, with regional grid refinement)
 - currently: COSMO-SREPS (or deterministic)
- standard experimentation system not yet adapted to perform LETKF (but soon)
→ stand-alone scripts, only preliminary LETKF experiments up to now



→ talk by Hendrik Reich

(forecast / analysis) ensemble spread 'characterises' (forecast / analysis) error, but

- ensemble size is limited, ensemble can only sample but not fully represent errors
- model error is not accounted for by algorithm

→ lack of spread: (partly) due to model error and limited ensemble size
which is not accounted directly by the algorithm

to account for it: covariance inflation, what is needed ?

→ **multiplicative** $X_b \rightarrow \rho \cdot X_b$ (tuning, or adaptive ($y - \mathbf{H}(x) \sim \mathbf{R} + \mathbf{H}^T \mathbf{P}_b \mathbf{H}$))

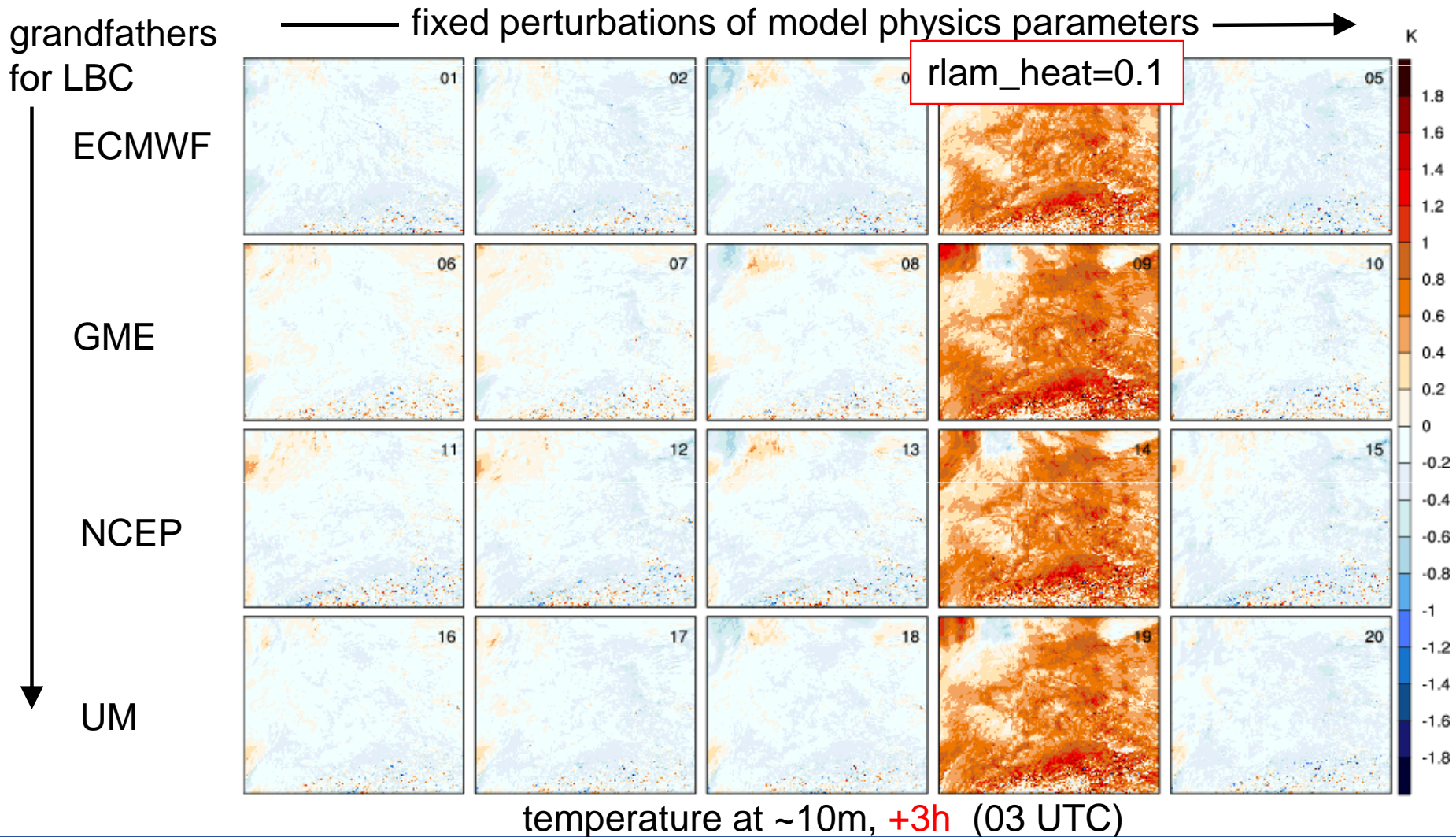
→ **additive** : perturbing the NWP model

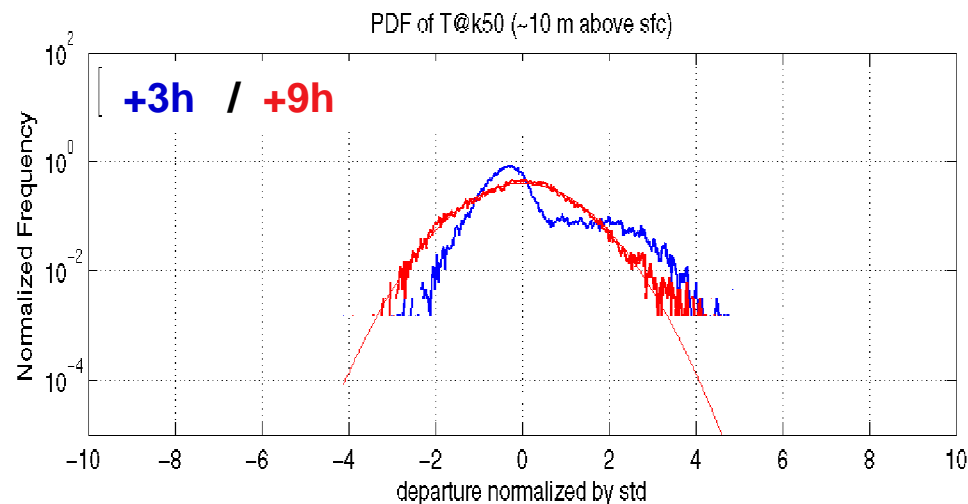
– fixed perturbations of model physics parameters

additive covariance inflation: model perturbations



ensemble forecast perturbations from COSMO-DE-EPS
here: a result from an old version without initial perturbations





ensemble forecast perturbations,
statistics over 9 days

$T_{10\text{ m}}$ over Northern Germany

→ $\leq +3\text{h}$: bi-modal pdf
but EnKF optimal
only for Gaussian pdf

→ physics perturbations that may be appropriate for the forecast component of an EPS need not be appropriate for DA (EnKF)

- additive covariance inflation: perturbing the NWP model
 - fixed perturbations of model physics parameters : no
 - stochastic physics (will be implemented)
 - statistical 3DVAR-B → hybrid schemes !
 - additive inflation which reflects model error as estimated by statistics (comparing forecast tendencies with observed tendencies, Gorin & Tsyrlnikov)





- localisation (multi-scale data assimilation,
successive LETKF steps with different obs / localisation ?)
 - update frequency $\Delta_a t$? $1 \text{ hr} \geq \Delta_a t \geq 15 \text{ min}$
non-linearity vs. noise / lack of spread / 4D property ?
 - perturbed lateral BC , how to deal with it ?
(\rightarrow source of noise)
(distort implicit error covariances in filter \rightarrow limit use of obs ?)
- note: shorter data cut-off & higher analysis frequency
for COSMO-DE than for driving global system ICON
- non-linear aspects, convection initiation (outer loop , (latent heat nudging) ?)



- **radar : radial velocity and (3-D) reflectivity**
- **ground-based GPS slant path delay** (direct use in LETKF , or tomography)
- **cloud information based on satellite and conventional data**
 - derive incomplete analysis of cloud top + cloud base, using conventional obs (synop, radiosonde, ceilometer) and NWC-SAF cloud products from SEVIRI, use obs increments of cloud or cloud top / base height or derived humidity
 - or use SEVIRI radiances directly

(Issues in LETKF: **non-Gaussian** distribution of obs increments, non-linear obs opr, non-local obs, obs error correlations / thinning ...)



thank you for your attention





$$J(\mathbf{x}) = (\mathbf{x} - \bar{\mathbf{x}}^b)^T \mathbf{P}^{b-1} (\mathbf{x} - \bar{\mathbf{x}}^b) + [\mathbf{y}^0 - H(\mathbf{x})]^T \mathbf{R}^{-1} [\mathbf{y}^0 - H(\mathbf{x})]$$

for $\mathbf{P}^b = (k-1) \mathbf{X}^b (\mathbf{X}^b)^T$, $J(\mathbf{x})$ is well-defined in sub-space S spanned by \mathbf{X}^b

- if \mathbf{w} is Gaussian random vector with mean 0 and covariance $(k-1) \mathbf{I}$, then $\mathbf{x} = \bar{\mathbf{x}}^b + \mathbf{X}^b \mathbf{w}$ is Gaussian with mean $\bar{\mathbf{x}}^b$ and cov. $(k-1) \mathbf{X}^b (\mathbf{X}^b)^T$

→ set up cost function in (low-dimensional!) ensemble space

$$J(\mathbf{w}) = (k-1) \mathbf{w}^T \mathbf{w} + [\mathbf{y}^0 - H(\bar{\mathbf{x}}^b + \mathbf{X}^b \mathbf{w})]^T \mathbf{R}^{-1} [\mathbf{y}^0 - H(\bar{\mathbf{x}}^b + \mathbf{X}^b \mathbf{w})]$$

→ apply nonlinear H to all forecast members and linearize around ensemble mean in observation space

$$\overline{H(\mathbf{x}^{b(i)})}$$

→ normal KF eq. in low-dim ensemble space, solve explicitly

→ analysis error cov. $\mathbf{P}_w^a = \left[(k-1) \mathbf{I} + (\mathbf{Y}^b)^T \mathbf{R}^{-1} \mathbf{Y}^b \right]^{-1}$
and analysis mean $\bar{\mathbf{w}}^a = \mathbf{P}_w^a (\mathbf{Y}^b)^T \mathbf{R}^{-1} [\mathbf{y}^0 - \bar{\mathbf{y}}^b]$

- need analysis ensemble members, with ensemble spread $(k-1) \mathbf{X}^a (\mathbf{X}^a)^T = \mathbf{P}^a$, choose: $\mathbf{X}^a = \mathbf{X}^b \mathbf{W}^a$, where $\mathbf{W}^a = \left[(k-1) \mathbf{P}_w^a \right]^{1/2}$

