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# Comparison of KENDA with nudging and impact of latent heat nudging

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COSMO General Meeting 2015, Wroclaw, Poland

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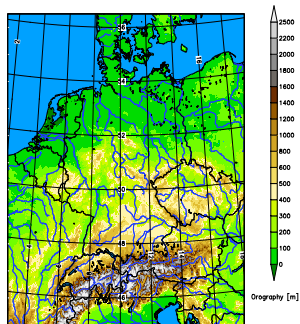
# Outline

- ▶ General overview on experiments
- ▶ Results
- ▶ Remaining problems, next steps

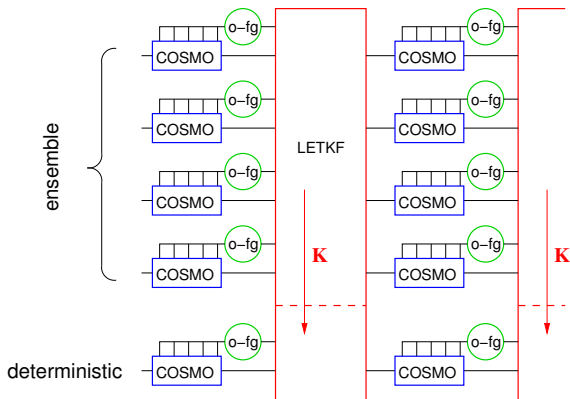
# KENDA experiments: general overview

- ▶ **KENDA**: Kilometer Scale **E**nsemble **D**ata **A**ssimilation
- ▶ implementation of **LETKF** (Local **E**nsemble **T**ransform **K**alman **F**ilter) following Hunt et. al.

*Orography of operational COSMO-DE domain used for KENDA-LETKF with 2.8 km horizontal resolution. The domain size is about 1170 km × 1280 km. Domain and resolution will be increased soon (COSMO-D2, 2.2 km resolution).*



# KENDA experiments: general overview



*KENDA system setup; 'o-fg' denotes observation minus first guess, 'K' the Kalman Gain for the analysis mean.*

# KENDA experiments: general overview

- ▶ **first goal:** replace (operational) nudging with deterministic LETKF analysis (second step: use as COSMO-DE EPS initial conditions)
- ▶ → focus on quality of deterministic analysis/forecast, compare with nudging (incl. LHN) as benchmark
- ▶ BACY (**b**asic **c**ycling, bash script environment) for ICON-LETKF and KENDA
- ▶ KENDA-BACY:
  - ▶ analysis cycle: LETKF incl. det analysis, nudg cycle with same obs set; verify against obs (surface/upper air)
  - ▶ forecast cycle: nudgecast (nudg analysis), det LETKF, verify against obs (surface/upper air/radar precipitation)
- ▶ speed  $\approx 2.0$  for KENDA, but needs large hard disk storage

# KENDA experiments: general overview

- ▶ BC for KENDA are taken from ICON-BACY, nudging and deterministic LETKF use same BC for analysis cycle and forecast
- ▶ ICON-BC (80 km resolution for ensemble members, 40 km for deterministic 3dVar-run):
  - ▶ 20120719-20120725,  
several experiments testing effect of soil moisture perturbations, latent heat nudging, RTPP
  - ▶ 20140517-20140615,  
compare LETKF/nudging within longer period
- ▶ preliminary tests with 20 km resolution BC from ICON-NEST;  
spread increased

## KENDA setup for 2014 periods

variable / feature	value
ensemble size $k$	40
deterministic run	1
horiz. resolution ens. + det. run	2.8 km
forecast frequency / length	6 h / 24 h
analysis update frequency	1 h
vert. localis. length scale (ln p)	0.075 - 0.5
horizontal localisation	adaptive
→ target weighted no. obs. $N_{loc}^{Oef}$	100
→ min. local. length scale $r_{loc}^{min}$	50 km
→ max. local. length scale $r_{loc}^{max}$	100 km
multiplicative covariance inflation	adaptive
→ lower / upper limit of $\rho$	0.5 / 3.0
RTPP relaxation weight $\alpha_p$	0.75

## model error: inflation/relaxation methods

- ▶ (1): compare “observed” with “expected” quantities:

$$\begin{aligned}\langle (y - H(x_b))(y - H(x_b))^T \rangle &= \mathbf{R} + \rho \mathbf{H} \mathbf{P}_b \mathbf{H}^T \\ \langle (H(x_a) - H(x_b))(y - H(x_b))^T \rangle &= \rho \mathbf{H} \mathbf{P}_b \mathbf{H}^T\end{aligned}$$

- ▶ (2): “relaxation” methods: e.g. relaxation to prior spread (RTPS):

$$X_a^{i, infl} = \rho X_a^i, \quad \rho = \sqrt{\alpha \frac{\sigma_b - \sigma_a}{\sigma_a} + 1}$$

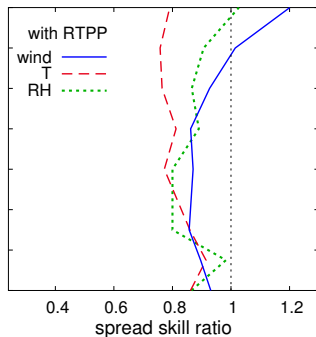
or relaxation to prior perturbation (RTPP):

$$X_a^{i, infl} = (1 - \alpha) X_a^i + \alpha X_b^i$$

- ▶ (1) works in observation space; tries to increase/decrease spread to fulfill statistical relations
- ▶ (2) works in model space; “corrects” reduction of spread due to assimilation of observations (RTPP: similar to additive pert., “directions” of fg pert partly remain; RTPS: inflates ana pert directions)



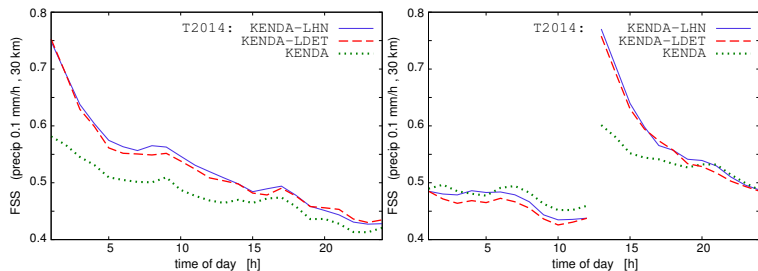
# Spread-Skill Ratio (RTPP + mult. inflation)



Spread-skill ratio  $r_s = \frac{\sigma_{f-\bar{f}}}{\sqrt{\sigma_{f_d-o}^2 - \sigma_{od}^2}}$ , where  $\sigma_{od}^2$  is the (diagnostic) observation error variance,  $\sigma_{f-\bar{f}}$  the spread and  $\sigma_{f_d-o}^2$  the RMSE.

→ With sufficient spread in the boundary conditions, RTPP (plus multiplicative inflation) gives reasonable spread-skill ratio

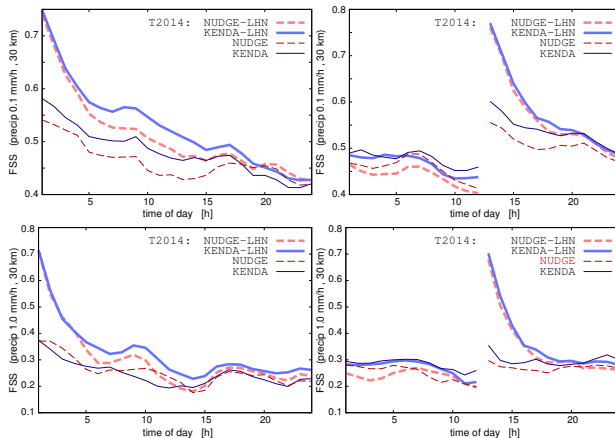
# Effect of LHN on radar-derived precipitation rates



- ▶ Fraction Skill Score (FSS) of 1-hourly precipitation (11 grid points,  $\approx 30$  km, 0.1 mm/h threshold), 00/12-UTC forecast (left/right)
- ▶ **KENDA** without LHN
- ▶ **KENDA-LDET** with LHN only in the det run
- ▶ **KENDA-LHN** with LHN also in the LETKF ensemble

Similar results for all scales and forecast start times.

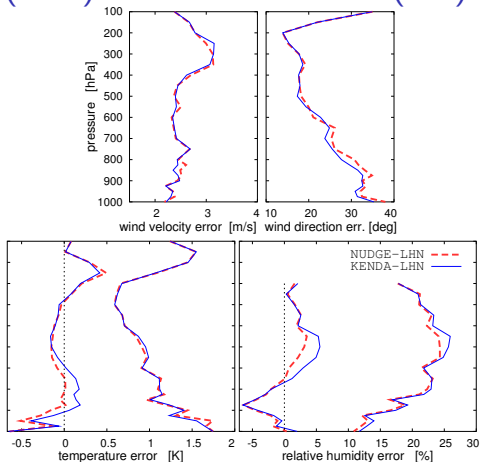
# KENDA-LHN vs. NUDGE-LHN: precipitation



- ▶ FSS as before, 0.1 mm/h and 1.0 mm/h threshold (upper/lower row), 00/12-UTC forecast (left/right)
- ▶ **KENDA-LHN** (LETKF + LHN)
- ▶ **NUDGE-LHN** (nudging + LHN)
- ▶ **KENDA** (LETKF without LHN)
- ▶ **NUDGE** (nudging without LHN).

Similar results for all scales and forecast start times.

# KENDA-LHN (blue) vs. NUDGE-LHN (red): 6h upper air



- ▶ Vertical profiles of bias and RMSE against radiosonde observations; 6-hour forecasts, wind speed and direction (upper row), temperature and relative humidity (lower row),) started from:
- ▶ **KENDA-LHN** analyses,
- ▶ **NUDGE-LHN** analyses

# Surface verification results

experiment	PS [hPa]	T2M [K]	TD2M [K]
KENDA-LHN	.53	2.03	3.33
NUDGE-LHN	.55	2.06	3.54
KENDA	.56	2.10	3.55
NUDGE	.56	2.15	3.89

- ▶ Root mean square errors (RMSE) of surface pressure ('PS'), 2-m temperature ('T2M'), and 2-m dewpoint depression ('TD2M') against observations from surface stations.
- ▶ Each of the RMSE values shown is an average over the 21 RMSE values valid for the forecast lead times from 1 to 21 hours.

KENDA gives clearly better results for TD2M and slightly better results for T2M and PS (with LHN).

# Soil moisture perturbations: Soil Moisture Index (SMI)

MeteoSwiss:

$$\text{SMI} = \frac{\text{WSO} - \text{PWP}}{\text{FC} - \text{PWP}} \quad (1)$$

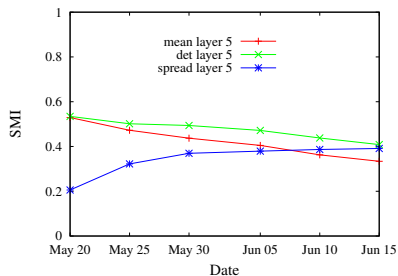
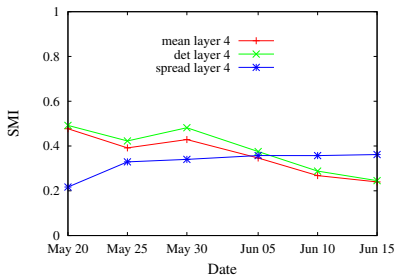
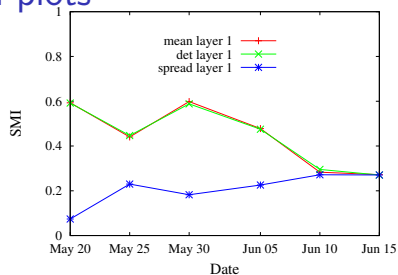
(PWP = Plant Wilting Point, FC = Field Capacity)

DWD:

$$\text{SMI} = \frac{\text{WSO} - \text{ADP}}{\text{PV} - \text{ADP}} \quad (2)$$

(ADP = Air Dryness Point, PV = Pore Volume)

# SMI plots



SMI (area mean of det run, ensemble mean and spread using Eq. (1) for layer 5 (54 cm), layer 4 (18 cm), layer 1 (0.5 cm)) → spread is too large, in layer 5 mean and det run diverge

# Conclusions

- ▶ ICON-BC: sufficient amount of spread at boundaries, but still only 80 km resolution! → preliminary tests with 20 km resolution BC from ICON-NEST; spread increased
- ▶ 24 h forecast of det run, nudging: deterministic LETKF forecast overall similar /slightly better quality than nudging forecast (except relative humidity), especially better results for precipitation
- ▶ plots shown are for 6h forecasts, but results also hold for 12h, 18h forecasts (differences get smaller)
- ▶ LHN: nearly no influence on upper air verification (wind slightly better); better results for Radar verification (precipitation, 00 and 12 UTC runs)
- ▶ soil moisture perturbations: positive impact on spread/rmse close to surface; but seems to introduce bias!! → tune parameters



## Next steps

- ▶ include SST, SNOW analysis
- ▶ compute winter period
- ▶ COSMO-D2 experiment, using ICON-NEST with 20 km resolution as BC
- ▶ tests with pattern generator
- ▶ use of additional observations, e.g. radar radial winds, SEVIRI, radar reflectivity (Theresa Bick, paper will be submitted soon)
- ▶ compute/investigate ensemble forecasts

# LETKF basics

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

# LETKF Theory

- ▶ do analysis in the  $k$ -dimensional ensemble space

$$\bar{\mathbf{w}}^a = \tilde{\mathbf{P}}^a (\mathbf{Y}^b)^T \mathbf{R}^{-1} (\mathbf{y} - \bar{\mathbf{y}}^b)$$
$$\tilde{\mathbf{P}}^a = [(k-1)\mathbf{I} + (\mathbf{Y}^b)^T \mathbf{R}^{-1} \mathbf{Y}^b]^{-1}$$

- ▶ in model space we have

$$\bar{\mathbf{x}}^a = \bar{\mathbf{x}}^b + \mathbf{X}^b \bar{\mathbf{w}}^a$$
$$\mathbf{P}^a = \mathbf{X}^b \tilde{\mathbf{P}}^a (\mathbf{X}^b)^T$$

- ▶ Now the analysis ensemble perturbations - with  $\mathbf{P}^a$  given above - are obtained via

$$\mathbf{X}^a = \mathbf{X}^b \mathbf{W}^a,$$

where  $\mathbf{W}^a = [(k-1)\tilde{\mathbf{P}}^a]^{1/2}$

# LETKF Theory

- ▶ it's possible to obtain a *deterministic run* via

$$\mathbf{x}_a^{det} = \mathbf{x}_b^{det} + \mathbf{K} \left[ \mathbf{y} - H(\mathbf{x}_b^{det}) \right]$$

with the *Kalman gain*  $\mathbf{K}$ :

$$\mathbf{K} = \mathbf{X}_b \left[ (k-1)\mathbf{I} + \mathbf{Y}_b^T \mathbf{R}^{-1} \mathbf{Y}_b \right]^{-1} \mathbf{Y}_b^T \mathbf{R}^{-1}$$

- ▶ the deterministic analysis is obtained on the same grid as the ensemble is running on; the *analysis increments* can be interpolated to a higher resolution

# Assimilation of Radar-derived precipitation by LHN

Required relation:

precipitation rate  $\leftrightarrow$  model variables  
(observed)                      (info required by nudging)

precipitation  $\leftrightarrow$  condensation  $\leftrightarrow$  release of latent heat

→ **Assumption:** vertically integrated latent heat release  $\propto$  precipitation rate

**Approach:** modify latent heating rates such that the model responds by producing the observed precipitation rates → Latent Heat Nudging (LHN)

$$\frac{\partial T}{\partial t} = F(t) + \left. \frac{\partial T}{\partial t} \right|_{nudging} + \left. \frac{\partial T}{\partial t} \right|_{LHN}$$
$$\Delta T_{LHN} = (\alpha - 1) \cdot \Delta T_{LH} \quad \text{with} \quad \alpha = \frac{RR_{obs}}{RR_{ref}}$$

Use LHN in LETKF until assimilation of radar reflectivities is available

## Soil moisture perturbations

- ▶ perturb soil moisture (and SST) with defined spatial and temporal length scales and amplitude
- ▶ soil moisture: 2 length scales, 100 km (synoptic), 10 km (convection)
- ▶ cut-off if moisture is below zero or above capacity (→ bias)
- ▶ next step: for soil moisture, limit perturbation amplitude to “available capacity” (avoid bias)

```
name = 'W_SO'           ! disturb soil moisture
scales = 0.002 100 1 24 ! 0.004 of soil capacity, 100km,
                        ! 1m vertical, 24 hour
          0.002 10 1 24 ! 0.004 of soil capacity, 10km,
                        ! 1m vertical, 24 hour

name = 'SST'           ! disturb SST
scales = 1 100 0 24 ! 1K + 100 km length scale pattern,
                    ! 24 hour
```

