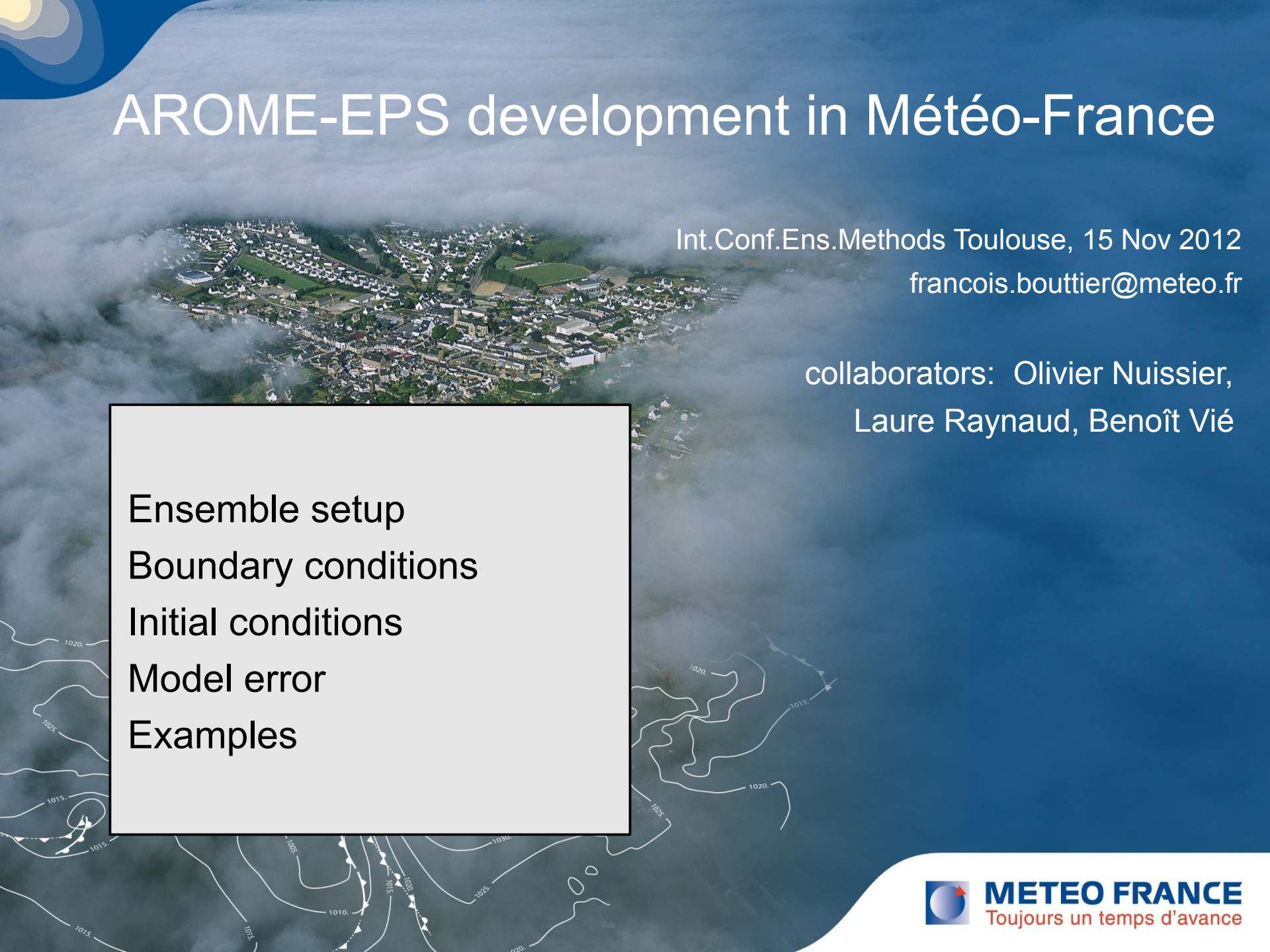


# AROME-EPS development in Météo-France

Int.Conf.Ens.Methods Toulouse, 15 Nov 2012

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collaborators: Olivier Nuissier,  
Laure Raynaud, Benoît Vié

An aerial photograph of a town, likely in the Pyrenees region, is partially obscured by a white rectangular box. The town features a mix of residential and commercial buildings, green spaces, and a church spire. The background is a blue gradient with faint white contour lines and arrows, suggesting a meteorological map. The white box contains a list of topics related to the AROME-EPS development.

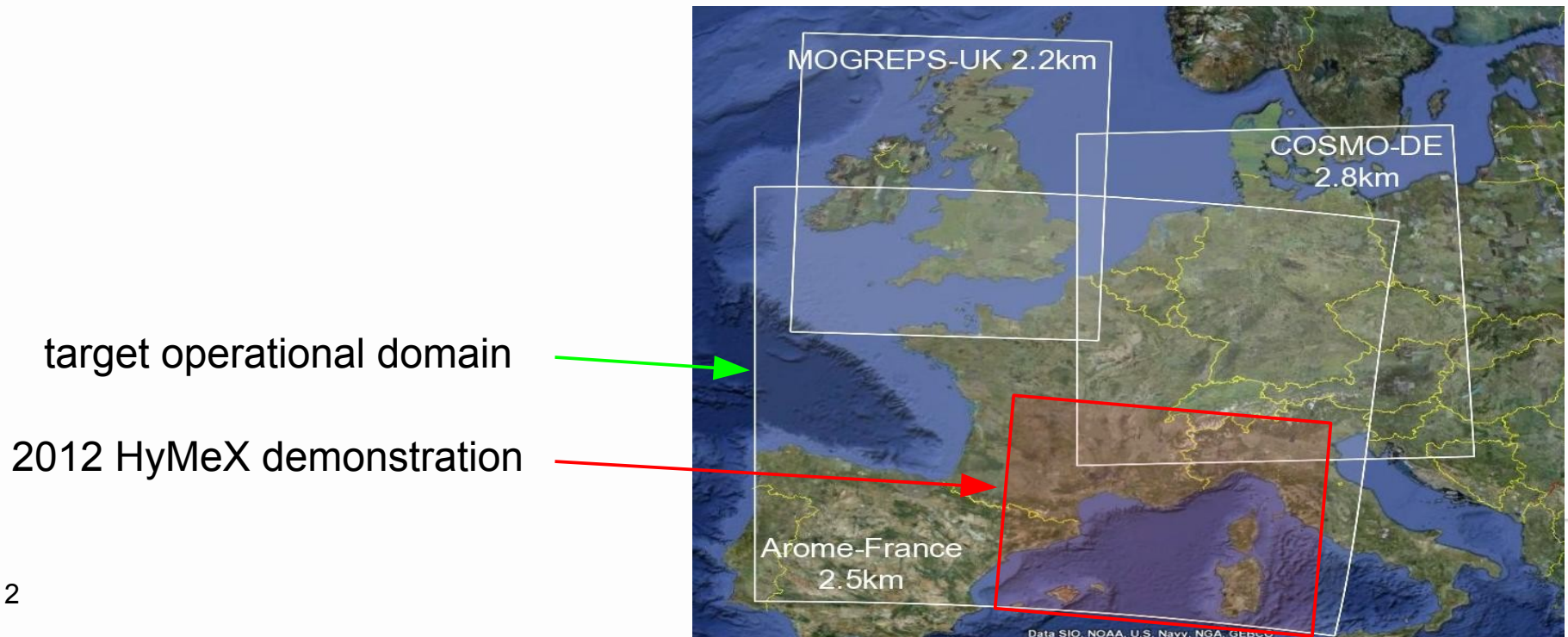
Ensemble setup  
Boundary conditions  
Initial conditions  
Model error  
Examples



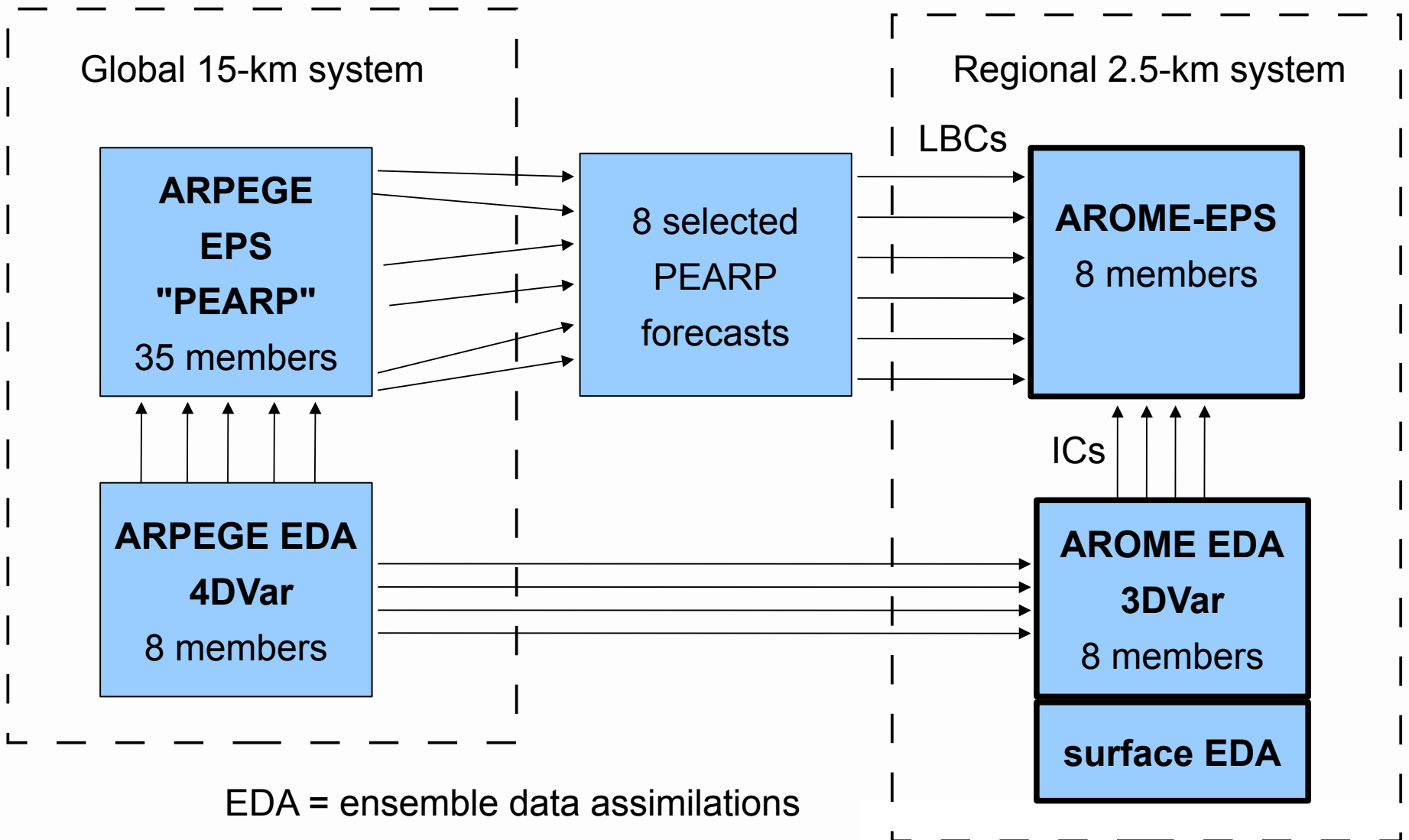
**METEO FRANCE**  
Toujours un temps d'avance

# Overview of AROME-EPS

- AROME-France model: 2.5km grid, non-hydrostatic, 36-h range
- 2011/2012: offline research experiments over several 20-day periods
- 2012: real-time demonstration over small domain (HyMeX field experiment)
- 2013: real-time preoperational over full domain
- 2014: operational production, at least 8 members every 6 hours



# AROME-EPS technical architecture



# Large-scale boundary conditions

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AROME model is hourly coupled at lateral & upper boundaries.

Several options for selecting LBCs in a bigger global ensemble:

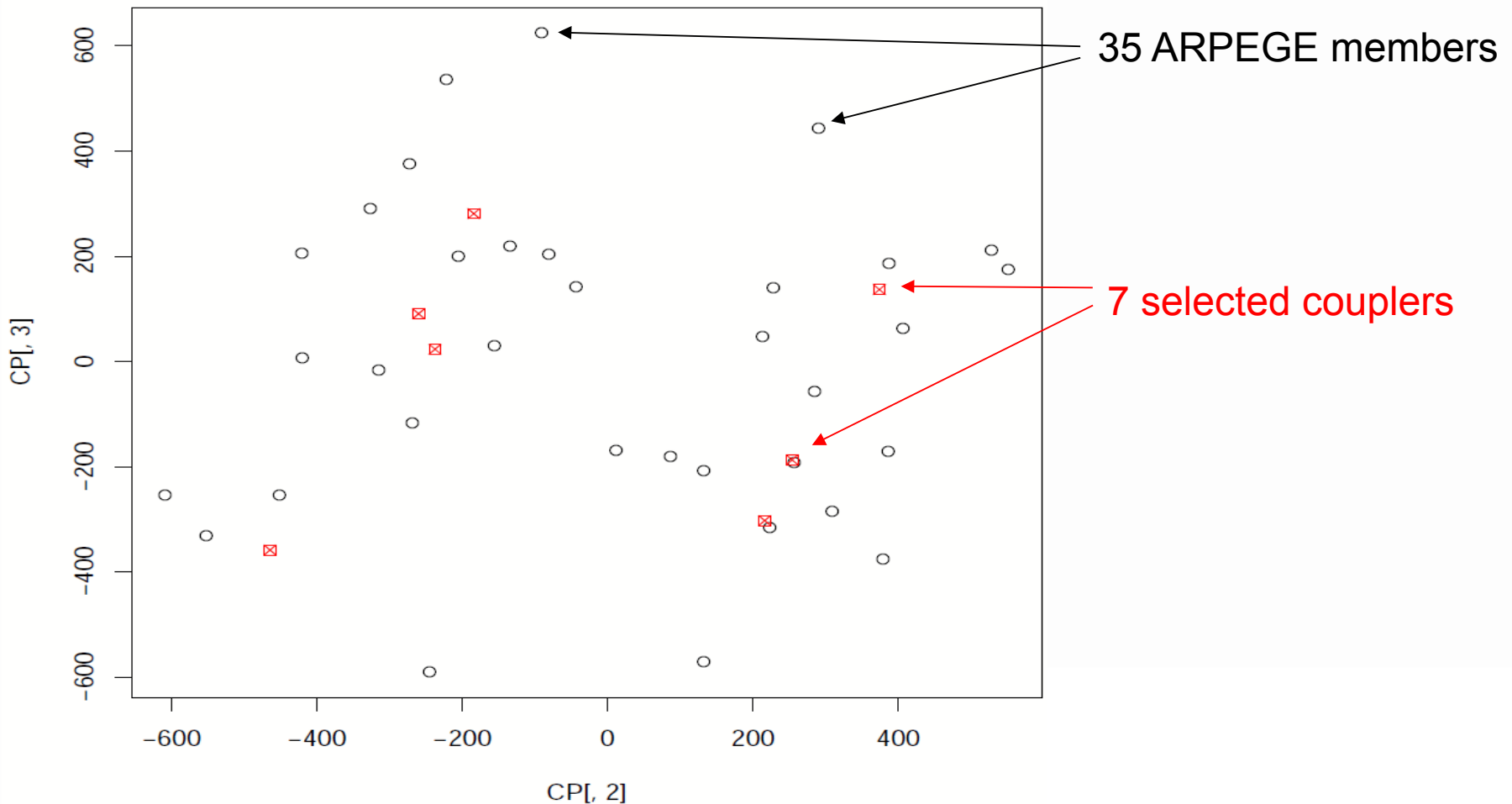
- random pick
- or, one PEARP multiphysics config per AROME member
- or, cluster PEARP forecasts using one of two methods:
  - K-means, more faithful to original PDF
  - hierarchical complete link (similar to COSMO-LEPS), maximizing interdistance between members

Results:

- impact on ensemble scores (ROC, Brier, CRPS) is tiny
- clustering improves a bit the high precip and wind scores, by promoting member diversity (ie sampling the PDF tails)

# Clustering example

- empirical idea: self-avoiding subsampling of forecast trajectories
- example of K-means clustering topology (plotted along 2 leading principal components)



# Initial perturbations: EDA in convective-scale model

Setup of AROME ensemble data analysis:

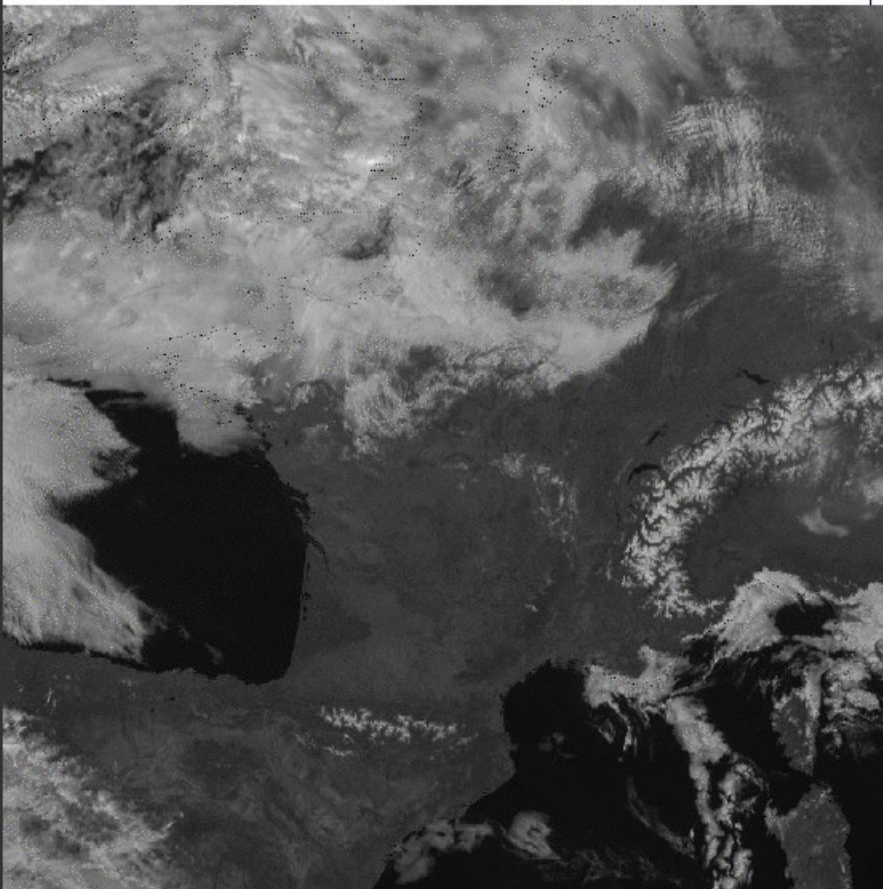
- 3DVar+surface DA system is like deterministic DA, except :
  - obs are randomly perturbed
  - LBCs taken from global ARPEGE EDA (4DVar)

Results:

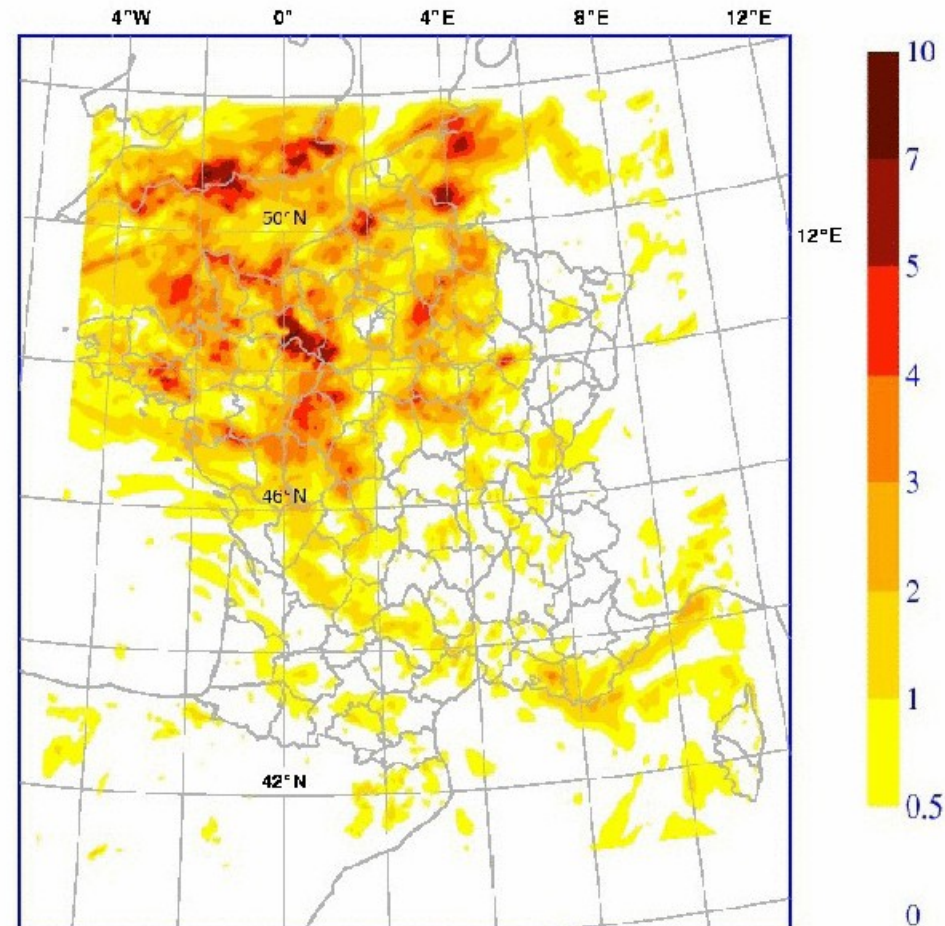
- provides acceptable forecast spread from step 0
- has physically believable low-level and surface spread
- no spin-up = ensemble suitable for probabilistic nowcasting ?
- (also used to produce background covariances for deterministic 3DVar, see posters)
- cost ~ 10% of AROME EPS forecast system

# AROME EDA: example of physically convincing initial spread

Low cloud cover on  
23/2/2008 (MSG)



Error std of the day (T 900 hPa)  
(Arome EnVar, 6 members)



# AROME EDA: inflation

Plain EDA is underdispersive: lacks representation of system errors (in model & in 3DVar algorithm).

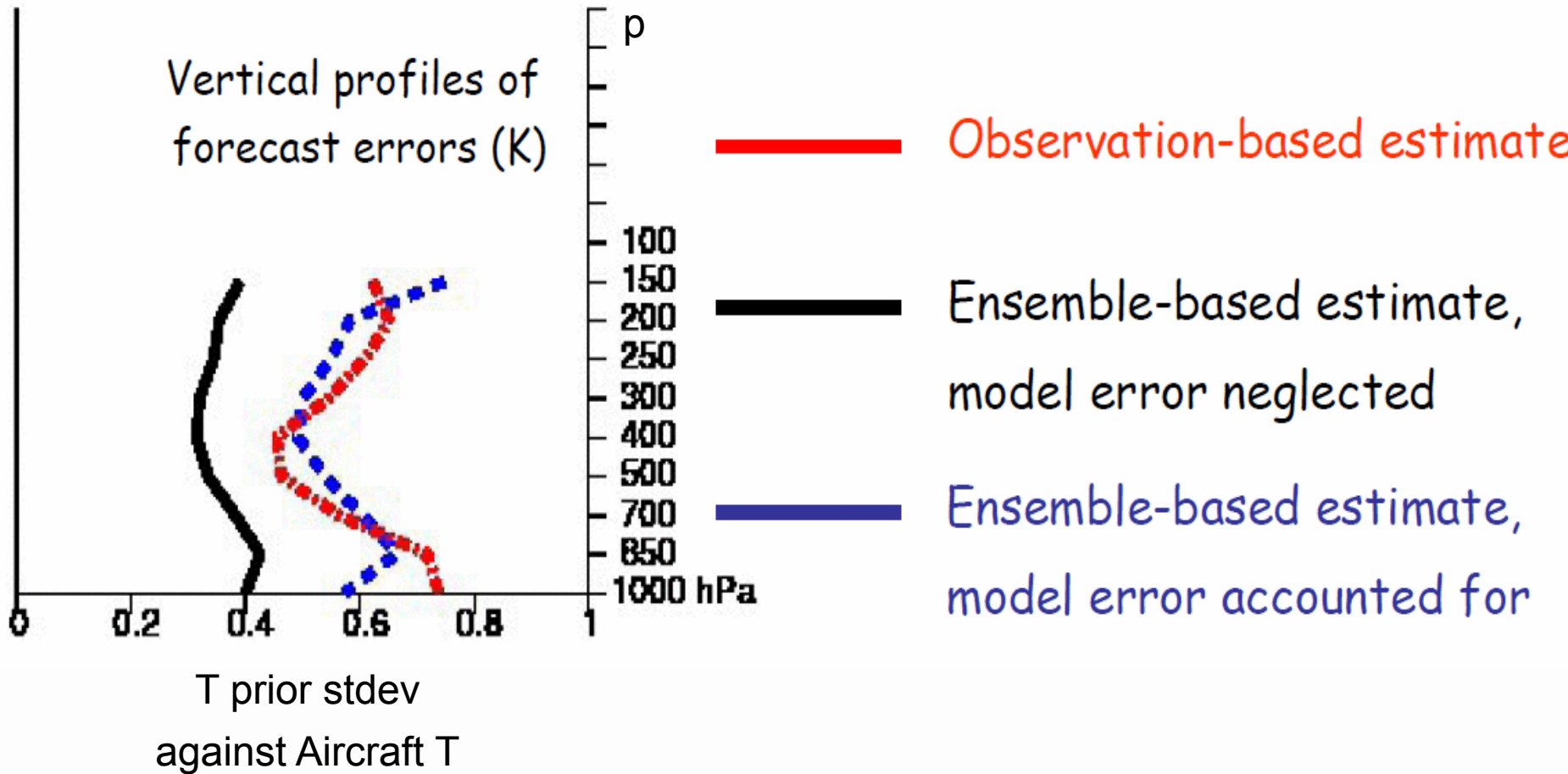
Adaptive inflation of background fields:

- compare variance of innovations  $cov(y-Hx_b)$  with EDA-predicted variances  $cov(M[x_{ai}])$
- inflate EDA background ensemble as a function of discrepancy:  
$$e_i' = e_i + a (e_i - \bar{e}_i), a > 1$$
- $a \sim 1.15$  is applied in each 3DVar cycle (every 3 hours)
- improves short-range spread/skill consistency, better ensemble scores (reliability, CRPS, Brier, ROC).
- Impact is significant up to 24h-range.



# Arome-EDA inflation

(This plot made using ARPEGE global EDA)



# Initial perturbations: cheaper alternatives to EDA

Consider IC perturbations of the form  $x_i = x_a + a (z_i - \bar{z}_i)$

- $x_i$  : initial condition of ensemble member  $i$
- $x_a$  : deterministic high-resolution analysis
- $z_i$  : externally supplied ensemble
- $a$  : scaling vertical profile (empirical tuning constant in mid troposphere, zero at top and bottom). Only applied to wind, T, ps.

## Results:

- with  $z_i =$  ARPEGE EDA : better than no perturb, but very underdispersive
- with  $z_i =$  PEARP ICs : better than ARPEGE EDA (because of SVs ?)
- with  $z_i =$  24-h breeding of AROME EPS: performance close to PEARP ICs
- AROME EDA easy winner until ~9h range or at low levels, not so superior otherwise.
- ranking is clear for spread/skill consistency and CRPS, not so clear for ROC.

Things to try: 6-h breeding, surface breeding, adaptive tuning of  $a$ .

# Model error in AROME EPS

In atmosphere: SPPT stochastic physics tendencies (akin to ECMWF's)

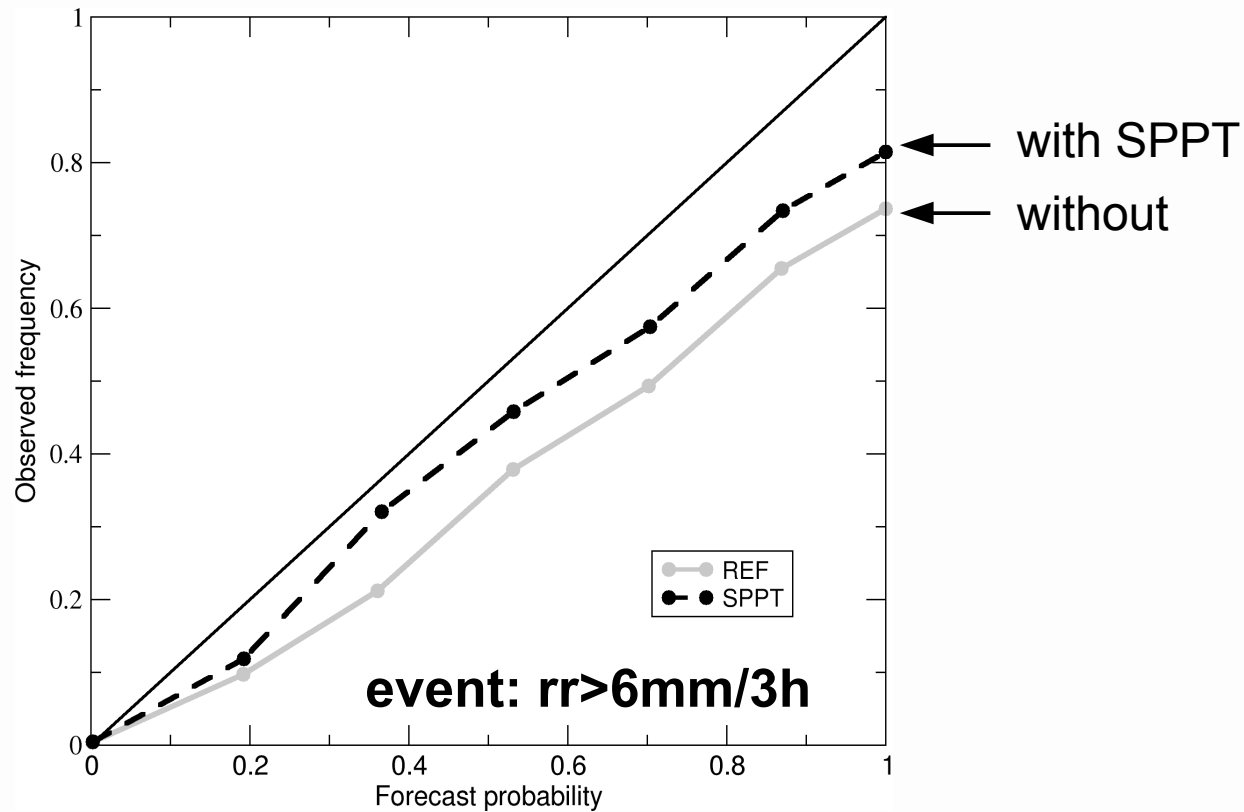
- multiplicative noise on model tendencies,
- noise autocorrelations : univariate, large-scale, slowly evolving
- recently extended into surface boundary layer
- increases spread & rms errors, most probabilistic scores are improved.
- drawback: drying in low troposphere
- not yet in EDA (model error represented by adaptive inflation)

Also tried physics parameter perturbations:

- in microphysics : negligible impact
- in turbulence : more promising

# Impact of SPPT model error scheme

## Reliability of precipitation over 5 weeks

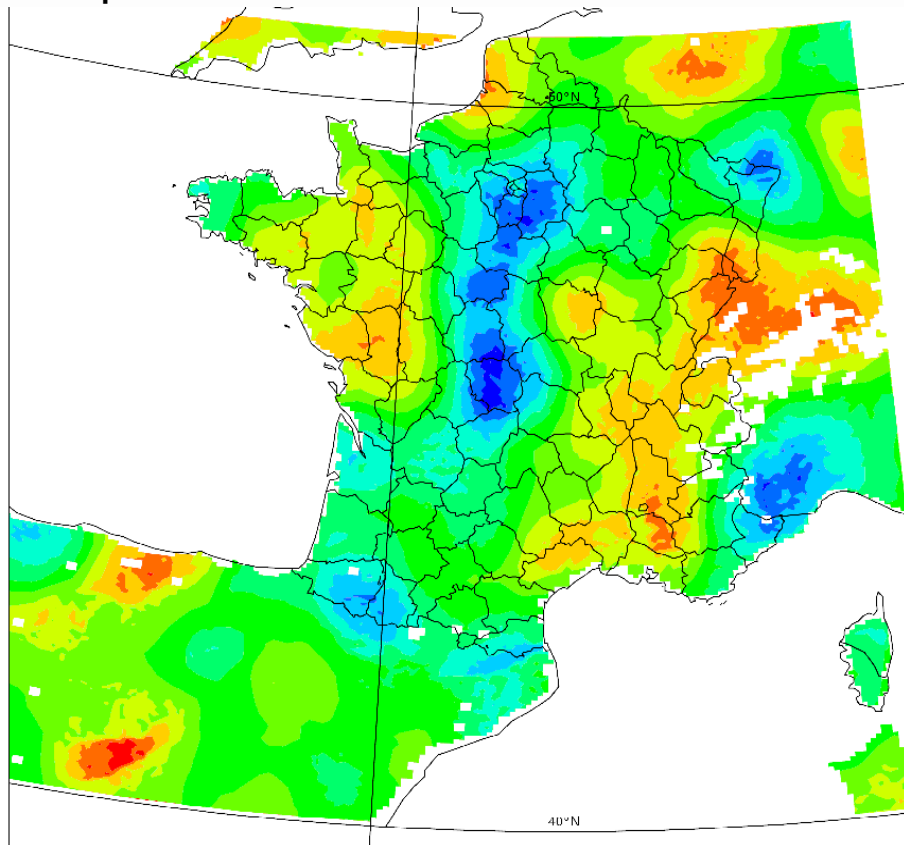


(without loss of sharpness)

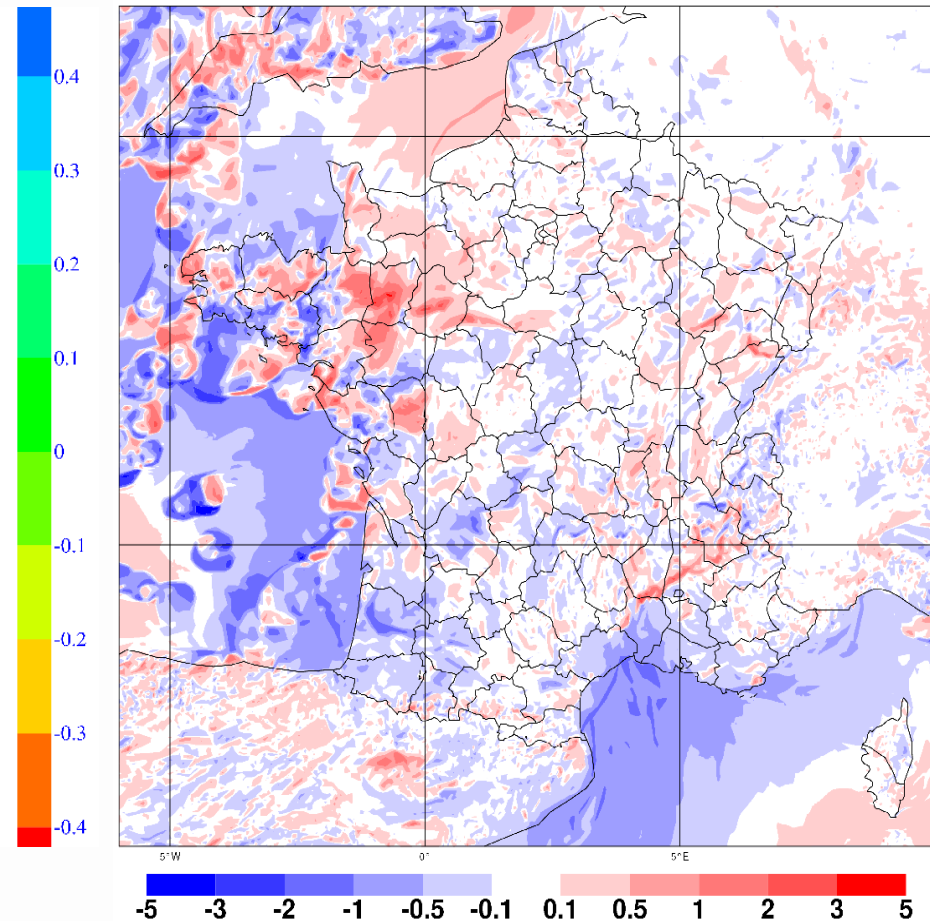
# Surface perturbations (work in progress)

found promising in sensitivity studies: SST, Wg, Rsmin, LAI,  $C_{veg}$ ,  $Zo_{orog}$

random perturbation generator:  
prescribed 2D Gaussian correlation



fc perturb T2m at 12h-range  
resulting from small SST changes



# Verification of short-range, high resolution ensembles

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## Lessons learned:

- Must use **real observations**, not analyses, as truth.
- **Observation errors** are significant
- **Model biases** are significant, with large diurnal cycle.
- **Error growth** over first 24h of forecasts is usually smaller than analysis errors.
- **Observation availability** limits significance of probabilistic scores (experiments are expensive, thus short)
  
- Usable: T2m, HU2m, 10m-wind, 10m-gusts, raingauge, aircraft wind & T
- Promising but more difficult: radar, satellite radiances (complex obs errors)

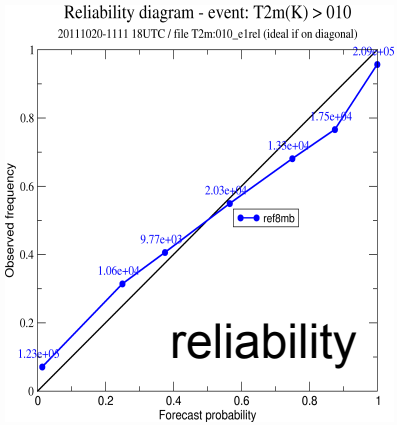
# Sample results from AROME EPS

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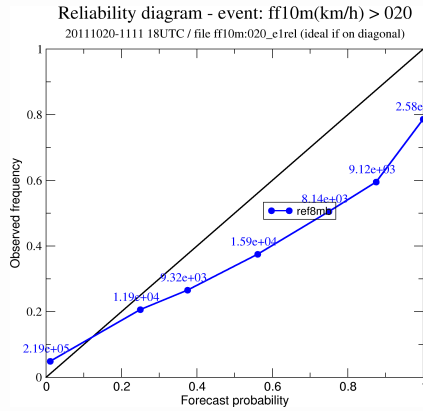
- Scores
- High-impact Mediterranean precipitation
- Scattered thunderstorms

# AROME-EPS sample scores (23 days in Oct 2011)

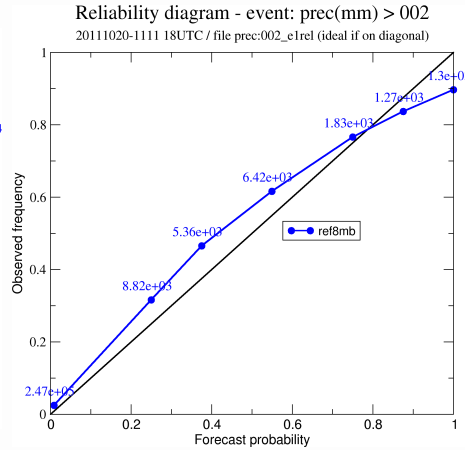
## T2m



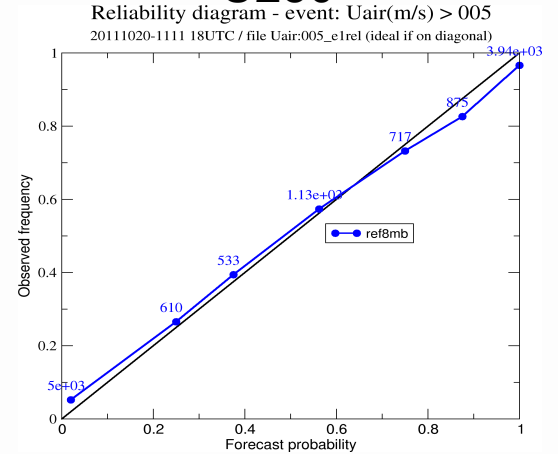
## ff10m



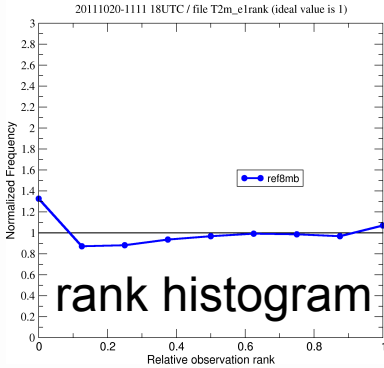
## prec3h



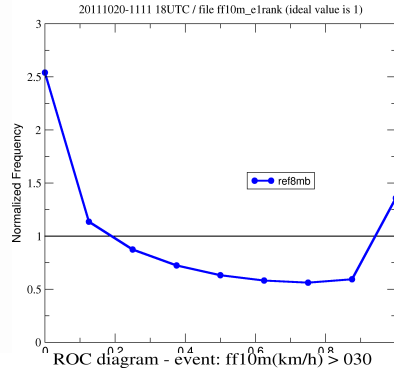
## U250



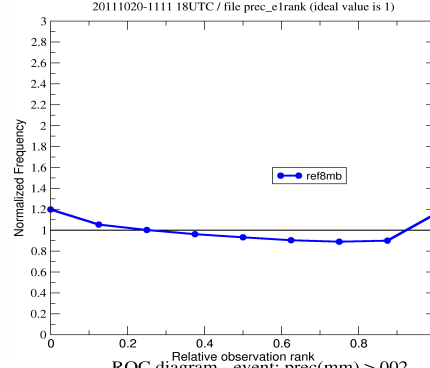
### Rank histogram - T2m(K)



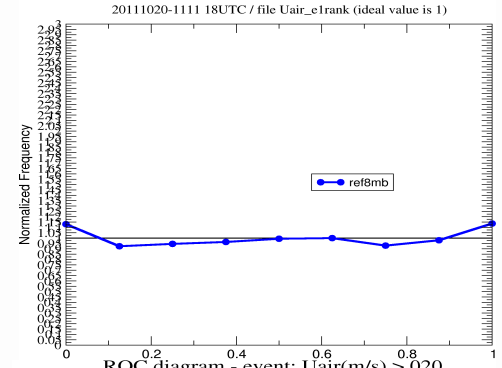
### Rank histogram - ff10m(km/h)



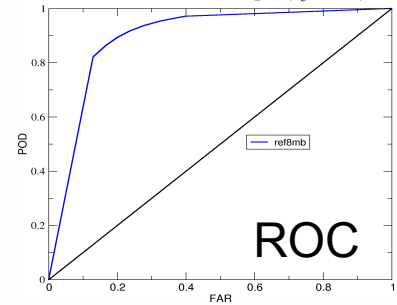
### Rank histogram - prec(mm)



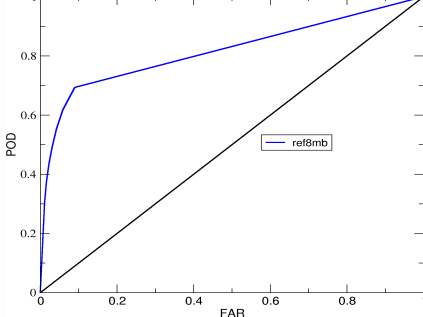
### Rank histogram - Uair(m/s)



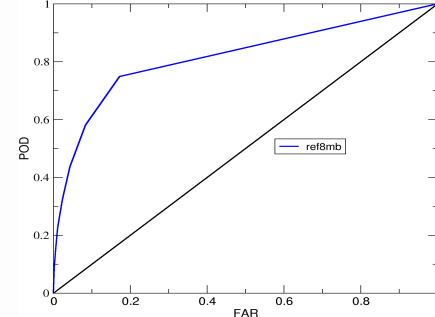
### ROC diagram - event: T2m(K) > 010



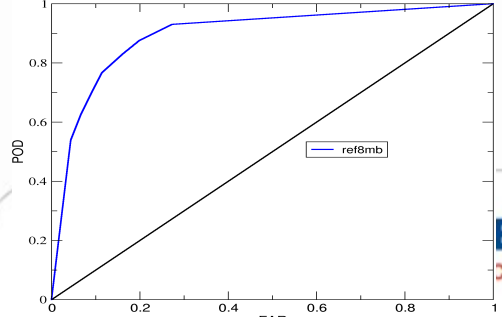
### ROC diagram - event: ff10m(km/h) > 030



### ROC diagram - event: prec(mm) > 002



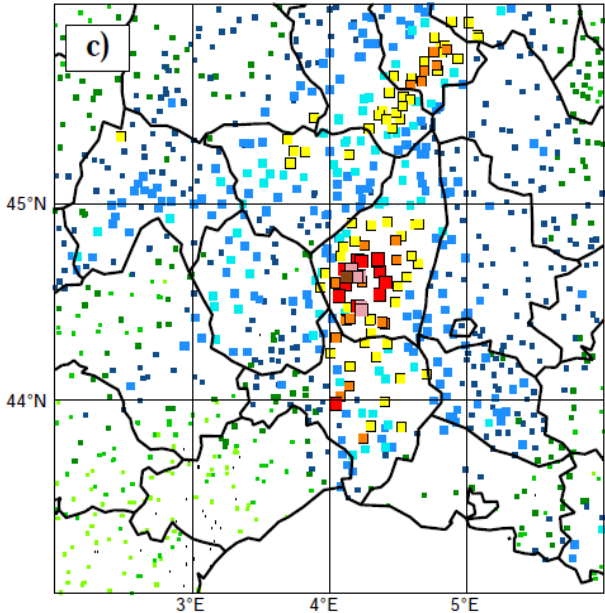
### ROC diagram - event: Uair(m/s) > 020



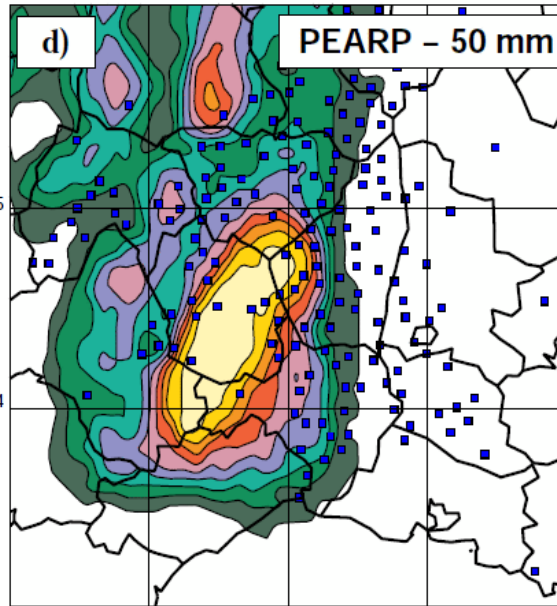


# Orographically-forced heavy convective precipitation (SE of France)

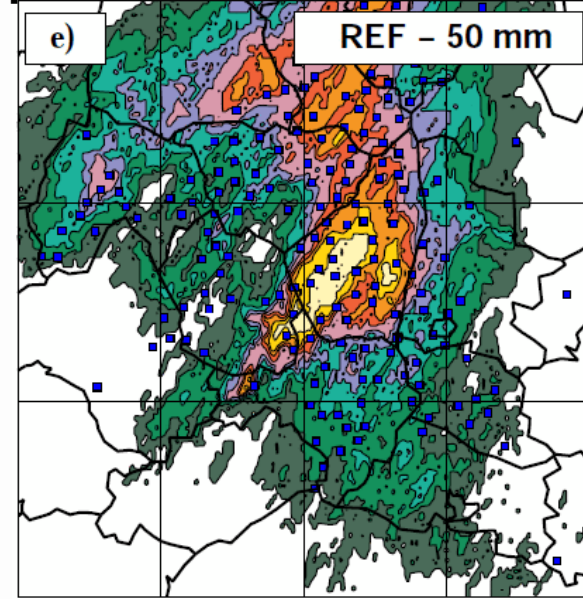
obs rr24



prob ARPEGE rr24>50mm



prob AROME rr24>50mm



1 5 15 30 50 75 100 150 200 250 300 350 400 450 500 (mm)

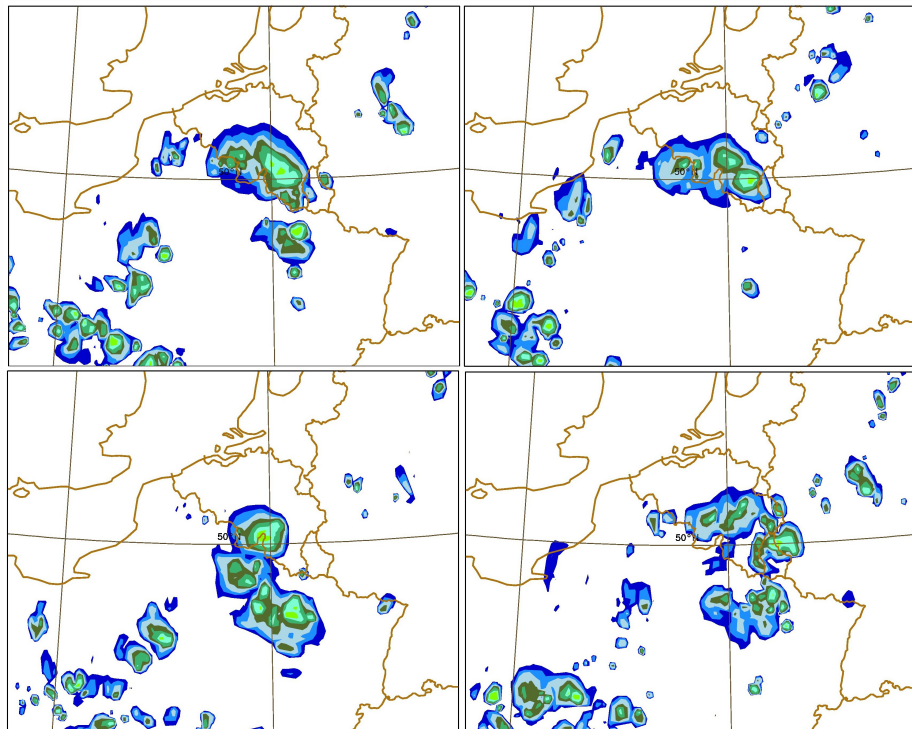
100km

AROME ens mean (blue)  
ratio spread/ensmean (grey)

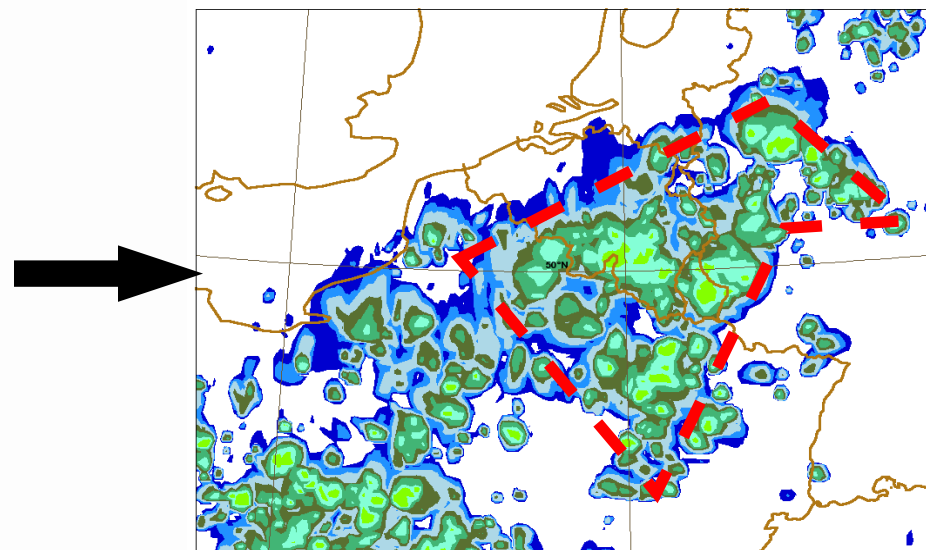
=good guidance (clear signal, limited spread,  
good detection, fewer false alarms than PEARP)

# Scattered thunderstorms over flat terrain: from members to probabilities

flight-level simulated reflectivity,  
4 random members



90% quantile, computed  
point-by-point using 12 members



All the structure inside the  
red area is sampling noise.

The *curse of convective dimensionality*: in a situation like this,  
~100 members would be needed for adequate sampling.  
PDF dressing is needed, with a mislocation error model.

# Summary

- nearly complete system setup (need surface perturbations)
- spread & reliability are mostly adequate,
  - we need to improve statistical resolution
  - low-level weaknesses on RH2m, ff10m
  - convection needs some kind of smoothing
- *ensemble size* is a major performance driver. Key events cannot be simulated using a lower-resolution model.
- model errors mostly seem *systematic* ones (diurnal biases, structure of precip field), they have a clear effect on probabilistic scores.
- is there a good perturbation tuning methodology ? currently, we risk compensation between various types of perturbations.



# Thank you for your attention

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## Recent papers

- Bouttier, F., B. Vié, O. Nuissier and L. Raynaud, 2012: Impact of stochastic physics in a convection-permitting ensemble. *Mon. Wea. Rev.*, early online release
- Brousseau, P., Berre, L., Bouttier, F. and Desroziers, G., 2012: Flow-dependent background-error covariances for a convective-scale data assimilation system. *Quart. Jour. Roy. Meteor. Soc.* 138, 310-322. doi: 10.1002/qj.920
- Nuissier, O., B. Joly, B. Vié and V. Ducrocq, 2012: Uncertainty on Lateral Boundary Conditions in a convection-permitting ensemble: A strategy of selection for Mediterranean heavy precipitation events. *Nat. Hazards Earth Syst. Sci.*, accepted.
- Raynaud L., L. Berre, G. Desroziers, 2012: Accounting for model error in the Météo-France ensemble data assimilation system. *Quart. Jour. Roy. Meteor. Soc.*, 138, 249-262. DOI: 10.1002/qj.906
- Vié, B., Molinié, G., Nuissier, O., Vincendon, B., Ducrocq, V., Bouttier, F., and Richard, E. 2012: Hydro-meteorological evaluation of a convection-permitting ensemble prediction system for Mediterranean heavy precipitating events, *Nat. Hazards Earth Syst. Sci.*, 12, 2631-2645, doi:10.5194/nhess-12-2631-2012.

