### **AROME-EPS** developments in Météo-France

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- Overview of AROME-EPS project
- Model error
- Initial conditions
- Boundary conditions
- Verification & postprocessing
- Demonstration projects: HyMeX & SESAR11.2.2

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### **Overview of AROME-EPS**

- model= AROME-2.5km
- forced by ARPEGE EPS & EDA
- schedule:
  - 2011: 5 weeks research experiments
  - 2012: 8 extra weeks + real-time production since Sept 2012 over smaller HyMeX domain
  - 2013: start real-time production over full domain (on new computer)
  - 2014: operational
- collaborations:
  - Hungary on wintertime weather (ECMWF special project)
  - DWD & UK on aviation & merged convection products (SESAR project)
  - (Spain on stochastic physics)



### **Technical architecture**

(details may change until operational stage)





### Model error

- in AROME atmosphere: SPPT stochastic physics tendencies
  - multiplicative noise, large-scale, slowly evolving
  - not active in PBL (under testing)
  - increased spread & rmse, most probabilistic scores are improved
  - also tried: microphysics & turbulence parameter perturbations
  - plan: better understand the physical impact of SPPT, improve SPPT tuning
- in AROME EDA: adaptive inflation of perturbations (see later)
- in AROME surface: perturb surface initial conditions



# **Surface perturbations**

works well: SST, Wg, Rsmin, LAI, C<sub>veg</sub>, Zo<sub>orog</sub>



Toujours un temps d'avance

# Impact of stochastic physics

#### Nice impact on reliability of precip

#### (parameter & obs not fully consistent with Arome plot)

#### **PE Arome**

#### **COSMO-DE EPS**



# **Initial perturbations**

- EDA (ensemble data analysis)
  - same AROME-2.5km 3-h 3DVar as deterministic assim
  - coupled to ARPEGE EDA
  - randomly perturbed obs including surface analysis
  - also used to derive Jb covariances for AROME 3DVar
- EDA much better than unperturbed ICs until ~6h range...but expensive (would cost 1 ensemble forecast member in operations)
- cheaper ARPEGE EDA-based IC perturbations under testing
- later: revisit breeding/ETKF



#### **Initial spread from AROME EDA**

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



Error std of the day (T 900 hPa) (Arome EnVar, 6 members) 4"E 8"E 4"W 12"E 0" 10 7 12°E 5 3 2 42° N 0.5 0

# model error in EDA, using adaptive ensemble inflation

- total forecast error covariances estimated by innovations:  $cov(y-Hx_b)=cov(Me_a + b_b)$
- $e_m$ ) using Desroziers' a posteriori variational diagnostics (Jbmin)
- compare with EDA-predicted variances cov(Me<sub>a</sub>)
- inflation of forecast perturbations:  $e'_i = e_i + \alpha (e_i e_i)$ ,  $\alpha > 1$
- $\alpha$ ~1.15 is applied every 3 hours
- yields 'more realistic' initial spread, and better ensemble scores later on



# **Boundary conditions**

- From ARPEGE EPS (same short-range quality as EPS)
- AROME model is coupled at lateral & upper boundaries
- How to select the 'best' ARPEGE-EPS runs for AROME-EPS:
  - reference: randomly pick ARPEGE-EPS members
  - with "one ARPEGE physics package per member": neutral
  - with K-means clustering: improves wind scores, precip spread
  - with hierarchical complete link (similar to COSMO-LEPS): even better
  - impact on objective scores is small.



### **Boundary conditions**

• Example of Kmeans clustering (plotted here in 2D PCA space)



# **Boundary conditions**

typical impact of selection method (Kmeans vs random choice) on precip scores





### Verification

- Lessons learned:
  - analyses is a poor truth. **Real observations** are needed.
  - significant observation errors & model **biases**. Strong **diurnal cycle** of bias.
  - forecast error growth over first 24h is much smaller than step-zero errors
  - lack of upper-level high-frequency observations for verification (satellite, radars have difficult observation operator issues)
- New MF ensemble verification package:
  - observation-based, accounts for blacklists & observation errors
  - uses surface obs & **aircraft** T,U,V (plan: use 3D radar reflectivities)
  - probabilistic scores with **significance testing**
- Plans:
  - need to improve precip verification (because complex score behaviour)
  - evaluate value for real-time decision-making

# **Post-processing aspects**

- Kernel dressing of output probabilities
  - unavoidable because ensemble is so tiny
  - plan: account for location uncertainties in output
- Lagging:
  - tests show BMA not useful (ie quality of old runs is not much worse than latest ones)
  - benefits of lagging:
    - lagging gives better spread & resolution (independently of ensemble size)
    - with <20 members, ensemble size is the top improvement opportunity
- **Calibration:** need to span several years (to sample 'extreme' weather)



### SESAR11.2.2 'superensemble' demonstration

- demonstrate convection probability products for aviation
- 'seamless' products spanning several model domains
- over two 40-day periods





### HyMeX 2012 SOP demonstration

- real-time runs of AROME-EPS over small domain
- focus on heavy rain & strong wind events
- coupled to hydrological EPS system
- intercomparison with other ensembles & models
- reruns & physical studies planned for 2013





#### pre-HyMeX test on high precip cases



#### other pre-HyMeX test cases, with web plots







#### AROME ensemble performance so far

- is much better than lower-resolution systems for low levels & precip
- Iooks comparable to available DWD & UK ensemble results
- lack of dispersion on some parameters
- precip is overdispersive (model bias issues)
- need more weather-type-specific physical understanding

### **Final thoughts**

#### Open issues

- Some forecasters are mainly interested in choosing between several deterministic models.
- Strategic link with ensemble assimilation to mutualize CPU resources.
- We need reforecasts for calibration (which means heavy work !)
- Current EPS R&D is pragmatic, with some scientific flaws e.g.
  - o lagging is handy, but theoretically incorrect.
  - inconsistent notions of 'statistical consistency' between data assimilation and EPS worlds.

#### Thank you for your attention

#### **Questions ?**

#### **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 convectionpermitting 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: Hydrometeorological 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.

#### **Additional slides**

#### In EPS, it is said that a "statistically consistent ensemble"...

- has spread:  $cov(x_i \overline{x}_i) = cov(x_t \overline{x}_i) = the 'skill'$
- ie at 6-h range, the spread is B, the background error cov matrix

In data assimilation, a "well-built ensemble" at 6-h range...

- has  $cov(x_i \overline{x_i}) \sim cov(x_i x_b) = B = spread around the control$
- since by definition B = cov(x<sub>b</sub>-x<sub>t</sub>), it means that (under usual DA hypotheses):
- cov(x<sub>i</sub>-x<sub>t</sub>)=cov(x<sub>i</sub>-x<sub>b</sub>+x<sub>b</sub>-x<sub>t</sub>) ~ cov(x<sub>i</sub>-x<sub>b</sub>)+cov(x<sub>b</sub>-x<sub>t</sub>) = 2B because the control is not perfect
- I.e. DA ensembles tuned to have perfect covariances will be underdispersive whenever verified against observations (only nonlinear error growth can save us !)