



# Stochastic Parametrization of Deep Convection in a Regional Ensemble Prediction System

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## Rationale for stochastic parametrizations

- Stochastically Perturbed Parametrization Tendencies scheme (SPPT); Buizza et al. 1999, Charron et al. 2010) is very efficient method to represent model error but rather unsatisfactory from a more fundamental perspective.
- In the long-term, SPPT should be ideally replaced by more physically based approaches to error simulation – inherently stochastic schemes.
- At RPN-A/CMC, we are currently working on a stochastic deep convection scheme and investigating its application in the REPS.





## Stochastic deep convection scheme at RPN/CMC

- Approach is based on the Plant-Craig (PC) stochastic deep convection scheme (Plant and Craig, 2008).
- Plume model adopted from Bechtold scheme (Bechtold, 2001) :
  - A bulk mass-flux parametrization very similar to Kain-Fritsch (KF; used in the original PC scheme)
  - CAPE-type closure based on the assumption that 90% of CAPE is removed within a specified adjustment period ~ 60 min
  - The plume model used with this scheme, however, differ to some extent from KF (e.g., triggering mechanism, conservation of enthalpy and mixing ratio).
- Rationale for the use of Bechtold scheme :
  - Modular structure
  - Consistent deep and shallow convection representation possible extension to shallow convection

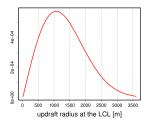






## Stochastic deep convection scheme at RPN/CMC

- Deterministic version of the Bechtold scheme: A single plume represents the mean properties of the entire subgrid-scale population of clouds.
- Stochastic version: in a given grid cell a cluster of convective activity with different intensities and sizes occurs.
  - Multiple plumes are randomly drawn from the radius distribution
  - Population size scaled by the closure assumptions.



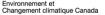




Realization 2







Canada

## Plume sampling function

 The clouds are generated based on the Plume sampling function – on average <N> plumes is generated during the specified cloud life time T (45 min in our case):

$$p(r)dr = \langle N \rangle \frac{\Delta t}{T} \frac{2r}{\langle r^2 \rangle} exp\left(-\frac{r^2}{\langle r^2 \rangle}\right)$$

where  $\langle N \rangle$  is the expected number of plumes generated over time T:

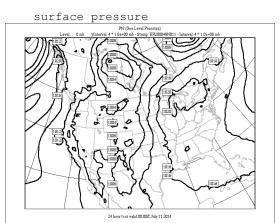
$$\langle N \rangle = \frac{\langle M \rangle}{\langle m \rangle}$$

- (M) is obtained from CAPE closure assumptions in the deep-convection scheme. In other words <N> number of clouds act to remove 90% of the CAPE in time T. This is equivalent to as having one single plume in the given area but with mass flux M.
- $\langle N \rangle$  can be changed by the tuneable  $\langle m \rangle$  but if  $\langle m \rangle$  is reduced the scheme becomes more costly. Therefore, larger grid areas could be problematic.





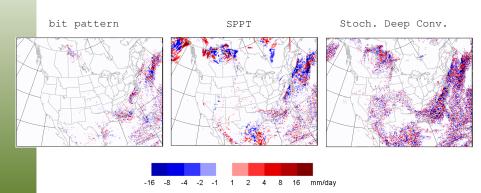
- Case of REPS forecast for 0000Z 10 July 2014
- REPS domain zoomed over the US (24-hr forecast valid 0000Z 11 July)







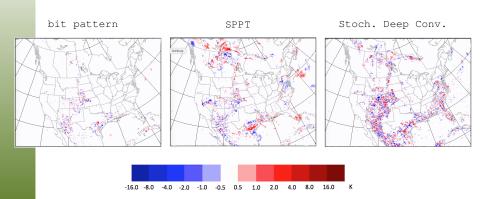
- Impact of various perturbations on 00-24h pcp accumulation (valid 0000Z 11 July 2014)
- One source of perturbations at a time.







- Impact of various perturbations on 24h screen-level temperature (valid 0000Z 11 July 2014)
- One source of perturbations at a time.

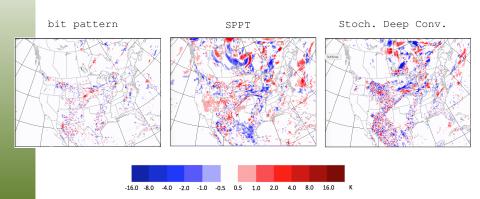








- Impact of various perturbations on 72h screen-level temperature (valid 0000Z 13 July 2014)
- One source of perturbations at a time.

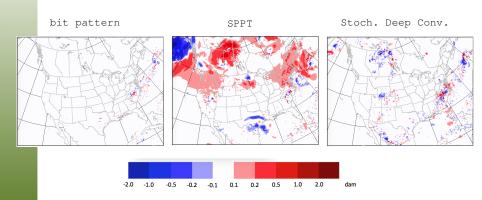








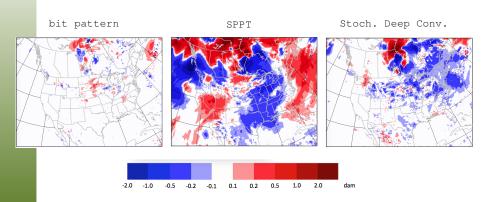
- Impact of various perturbations on 24h GZ 500 hPa (valid 0000Z 11 July 2014)
- One source of perturbations at a time.







- Impact of various perturbations on 72h GZ 500 hPa (valid 0000Z 13 July 2014)
- One source of perturbations at a time.







# Scale analysis of regional REPS error

- EPS field :  $x = x_m(i, j, k, t, \tau)$ i, j and k are horizontal and vertical grid indices, m ensemble member, t forecast issue time and  $\tau$  lead time.
- Analysis :  $v = v(i, j, k, t + \tau)$ .
- Model error  $\varepsilon_m \equiv x_m y$  is decomposed as

$$\langle \varepsilon \rangle = \langle \mathbf{x} \rangle - \mathbf{y}$$

$$\varepsilon_m^* = x_m - \langle x \rangle,$$

where

$$\langle x \rangle = \frac{1}{N_m} \sum_{m=1}^{N_m} x_m.$$

is the ensemble mean.





## Scale analysis of REPS error

 Using the Discrete Cosine Transform (Denis et al. 2002), the spatial variance of model error  $\varepsilon$  can be decomposed into contributions from different spatial scales  $\lambda$ :

$$\sigma_{\varepsilon\varepsilon}^2(k,m,t,\tau) \equiv \overline{(\varepsilon - \overline{\varepsilon})^2} = \sum_{q=1}^{N_q} \hat{\sigma}_{\varepsilon\varepsilon}^2(\lambda_q,k,m,t,\tau)$$

where

$$\overline{(\cdot)} = \frac{1}{N_i N_j} \sum_{i=1}^{N_i} \sum_{j=1}^{N_j} (\cdot)$$

is the spatial average.

It can be shown that the expectation of the error variance over the ensemble is additive in terms of error components:

$$\langle \hat{\sigma}^2_{\varepsilon\varepsilon} \rangle = \hat{\sigma}^2_{\langle \varepsilon \rangle \langle \varepsilon \rangle} + \langle \hat{\sigma}^2_{\varepsilon^* \varepsilon^*} \rangle$$





## **Error variance components**

 The two error variance components can be conveniently expressed in the form of the law of cosines:

$$\hat{\sigma}_{\langle \varepsilon \rangle \langle \varepsilon \rangle}^{2} = \hat{\sigma}_{yy}^{2} \left( 1 + (\hat{\gamma}\hat{\rho})^{2} - 2 (\hat{\gamma}\hat{\rho}) \hat{r} \right) \tag{1}$$

$$\langle \hat{\sigma}_{\varepsilon^* \varepsilon^*}^2 \rangle = \hat{\sigma}_{yy}^2 \left( \hat{\gamma}^2 (1 - \hat{\rho}^2) \right). \tag{2}$$

coherence and  $\hat{r}$  is the coherence between the ensemble mean and analyses:

$$\hat{\gamma}^2 \equiv \frac{\langle \hat{\sigma}_{xx}^2 \rangle}{\hat{\sigma}_{yy}^2} \,, \qquad \quad \hat{\rho}^2 \equiv \frac{\hat{\sigma}_{\langle x \rangle \langle x \rangle}^2}{\langle \hat{\sigma}_{xx}^2 \rangle} \,, \qquad \quad \hat{\varUpsilon} \equiv \frac{\hat{\sigma}_{\langle x \rangle y}^2}{\left(\hat{\sigma}_{\langle x \rangle \langle x \rangle}^2 \, \hat{\sigma}_{yy}^2\right)^{1/2}} \;. \label{eq:gamma_spectrum}$$

- The role of  $\hat{\rho}$  is twofold it represents :
  - Smoothing effect of averaging in the ensemble mean (Eq. 1)
  - Ensemble spread (Eq. 2)





# Modern-Era Retrospective analysis for Research and Applications (MERRA)

#### DA System

- Goddard Earth Observing System Model version 5 (GEOS-5) atmospheric DA system.
- Integrates a AGCM with grid-point statistical interpolation.
- $\bullet$  1/2  $^{\circ}$  x 1/3  $^{\circ}$  horizontal grid, 72 levels.

#### **Dataset**

- Fields: GZ, UU, VV, TT @ 500 and 250 hPa.
- Three-hourly time series (instantaneous values).
- Lat-lon grid 0.5° x 0.625° interpolated to the REPS grid (0.1375°).

#### Verification period

- Winter: January 01 31, 2014.
- Summer: July 01 31, 2014.



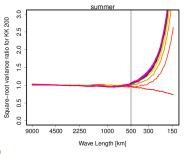


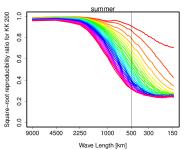


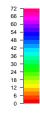
# Variance Ratio and Reproducibility, EKIN 200 hPa, summer)

$$\hat{\gamma} \equiv \left( \frac{\langle \hat{\sigma}_{xx}^2 \rangle}{\hat{\sigma}_{yy}^2} \right)^{1/2},$$

$$\hat{
ho} \equiv \left(rac{\widehat{\sigma}_{\langle x
angle \langle x
angle}^2}{\langle \widehat{\sigma}_{xx}^2 
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ight)^{1/2}$$







## Spread-to-error ratio

Members of a perfect EPS and perfect analyses would be statistically indistinguishable. Hence:

$$\langle \varepsilon \rangle = \langle x \rangle - y \sim \langle x \rangle - x = \varepsilon^*$$

which yields

$$\hat{\sigma}^2_{\langle \epsilon \rangle \langle \epsilon \rangle} = \langle \hat{\sigma}^2_{\epsilon^* \epsilon^*} \rangle$$

or

$$1 + (\hat{\gamma}\hat{\rho})^2 - 2(\hat{\gamma}\hat{\rho})\hat{r} = \hat{\gamma}^2(1 - \hat{\rho}^2).$$

In the spectral range in which we have confidence in the analyses we require:

$$\hat{\gamma} \approx 1$$

which yields the following necessary condition for an EPS to be balanced:

$$\hat{\rho} \approx \hat{r}$$
.



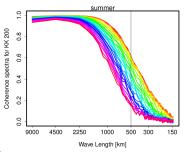


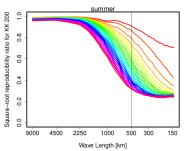


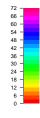
## Coherence vs. Reproducibility Ratio, EKIN 200 hPa, summer

$$\hat{\varUpsilon} \equiv \frac{\hat{\sigma}_{\langle x \rangle y}^2}{\left(\hat{\sigma}_{\langle x \rangle \langle x \rangle}^2 \, \hat{\sigma}_{yy}^2\right)^{1/2}}$$

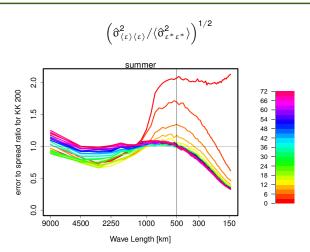
$$\hat{\rho} \equiv \left(\frac{\hat{\sigma}_{\langle x \rangle \langle x \rangle}^2}{\langle \hat{\sigma}_{xx}^2 \rangle}\right)^{1/2}$$





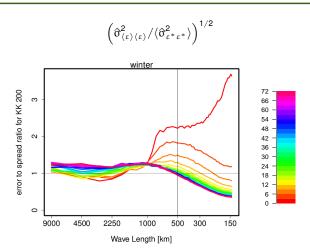


## Error-to-Spread Ratio, EKIN 200 hPa, summer



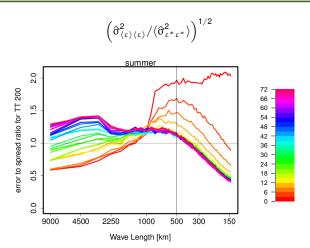


## Error-to-Spread Ratio, EKIN 200 hPa, winter





## Error-to-Spread Ratio, TT 200 hPa, summer



## Applications of the scale analysis method

## Study the contribution of different sources of perturbations

- Use to study the impact of perturbations in initial and lateral boundary conditions, SPPT
- Contrubution of stochastic parametrizations to the spread.

#### Uncertainty in the reanalyses

- Use other renalaysis (ERA-Interim, NARR higher spatial resolution (35 km), 3-hourly
- Use CMC analyses.

#### Apply to the Global EPS

- The method can be easily adapted to allow spectral transformations on the sphere
- More degrees of freedom for PTP and stochastic parametrizations than in the REPS.





#### Conclusions and Future Work

#### • Expectations from the stochastic deep convection scheme :

- Contribute to ensemble spread in weather situations with weak large-scale forcing – convection driven with the diurnal cycle.
- Increase the spread at scales below 1000 km in the early stages of the forecast.
- Hasten the upscale propagation of the inter-memember differences.
- It is not expected to have a large impact in situations with strong large-scale forcing.

#### Future work :

- Scheme adds fine-scale variability (grainy precipitation patterns) how this impacts the quality of the forecasts?
- A systematic evaluation of the scheme in REPS and proper scoring.
- Introducing stochasticity in other schemes (e.g., shallow convection, PBL vertical diffusion, gravity wave drag).



