Combining EDA and SVs to initialise Ensemble forecasts

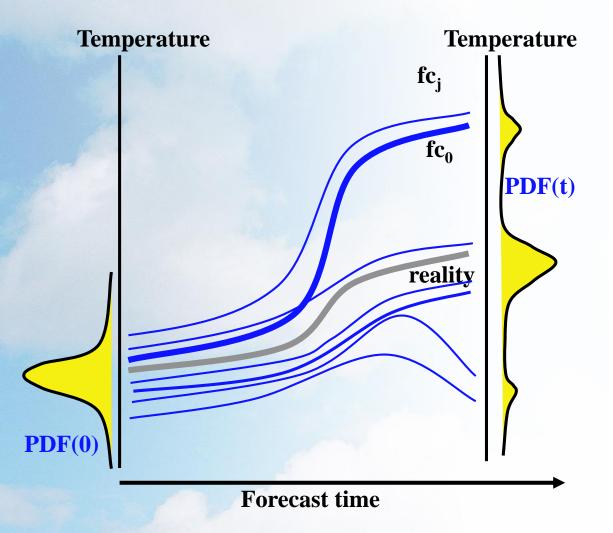
Simon Lang

Ack: Martin Leutbecher, Elias Holm, Massimo Bonavita, Roberto Buizza, Sarah-Jane Lock



Slide 1

3. Ensemble prediction systems



Sources of Uncertainty:

- Initial Conditions
- Model Formulation



The ECMWF Ensemble:

- 51 Members (50 perturbed + control member without perturbations), TCo639 (~ 18 km) to day 15
- 91 vertical levels
- Coupled to NEMO ocean model (1/4 degree) and LIM2
 ice model
- Initial perturbation via an ensemble of data assimilations and singular vectors, 5 member ocean data assimilation
- Model error representation (SPPT, SKEB)



Reliability of the ensemble spread

• Consider ensemble variance ("spread") for an *M*-member ensemble

$$rac{1}{M}\sum_{j=1}^M (x_j-\overline{x})^2$$

and the squared error of the ensemble mean

$$(\overline{x} - y)^2$$

- Average the two quantities for many locations and/or start times.
- The averaged quantities have to match for a reliable ensemble (within sampling uncertainty).



How to construct initial perturbations:

Methods that rely on the dynamics only, e.g.:

- bred vectors
- singular vectors (with total energy norm)

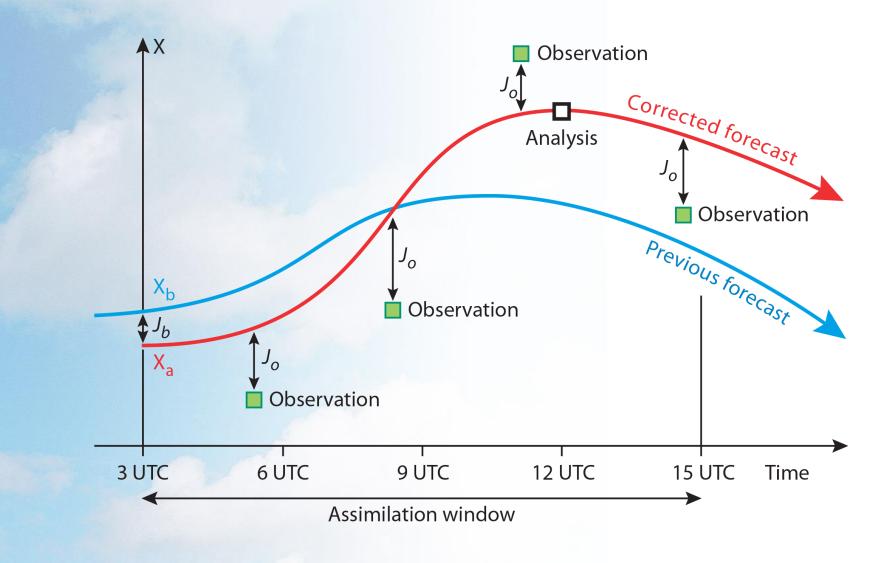
Ensemble data assimilation methods, e.g.:

- Ensemble of 4D-Var data assimilations (EDA)
- Ensemble Kalman Filter

ECMWF: combination of EDA and singular vectors



4D-Var assimilation





The Ensemble of Data Assimilations

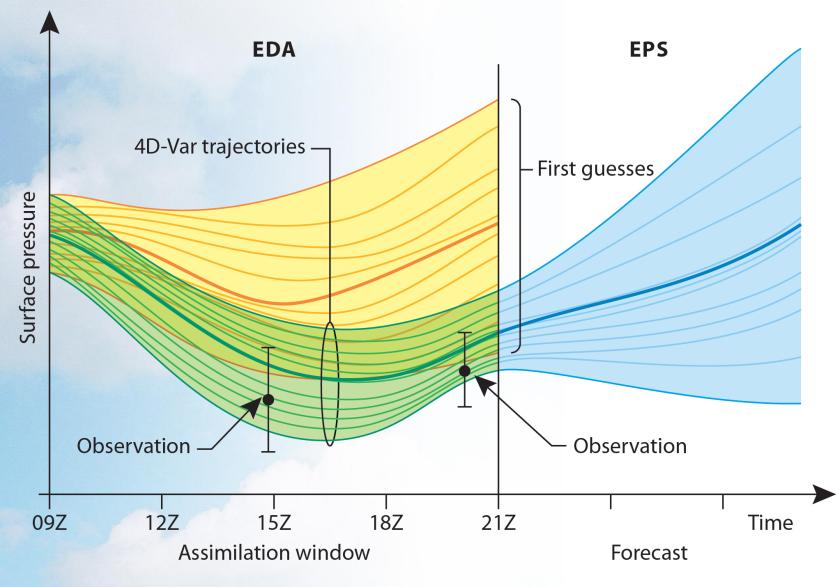
- 25 perturbed ensemble members + 1 control, TCo639 outer loops, 137 levels, TL191/TL191 inner loops. (HRES DA: TCo1279 outer loops, TL255/TL319/TL399 inner loops).
- Observations randomly perturbed according to their estimated error covariances (R)
- SST perturbed with climatological error structures
- Model error representation via Stochastically Perturbed Parametrization Tendencies (SPPT, see Sarah-Jane's Talk)

The EDA simulates the error evolution of the 4DVar analysis cycle:

- \rightarrow uncertainty estimates to initialize ensemble forecasts
- Flow dependent estimates of background error covariances for use in 4D-Var



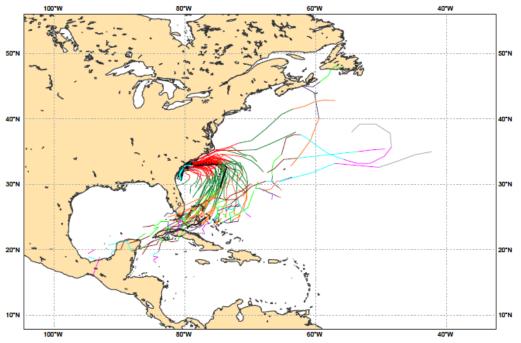
Ensemble assimilation and prediction





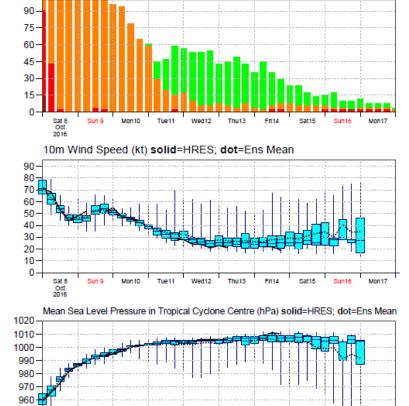
Date 20161008 00 UTC @ECMWF

Individual trajectories for **MATTHEW** during the next **240** hours tracks: **thick solid**=HRES; **thick dot**=CTRL; **thin solid**=EPS members [coloured] 0-24h 24-48h 48-72h 72-96h 96-120h 120-144h 144-168h 168-192h 192-216h 216-240h



List of ensemble members numbers forecast Tropical Cyclone Intensity category in colours: TD[up to 33] TS[34-63] HR1[64-82] HR2[83-95] HR3[> 95 kt]

+024	h : hr	ct 01	021	03 04	05	D6 C	17 0	8 09	10	11	12	13	14 1	15 1	16 1	17 1	81	9 20	21	22 :	23 2	24 23	52	6 27	28	29	30 3	31	32 3	33	4 35	36 3	373	8 39	40	41	42	43	44 4	45.4	6 47	48	49 5	0
+048	h : hr	ct D1	02	03 04	05	D6 C	17 0	8 09	10	11	12	13	14 1	15 1	16 1	17.1	81	9 20	21	22	- 2	24 25	52	6 27	28	29	30 3	31	32 3	33	4 35	36 3	373	8 39	40	41	42	43	44 4	45	47	48	49 5	٥
+072	h :	01	02			06 🕻	7		10	11				15 1	16 1	17.1	81	9	21		23		2	6 27	28	29	30 3	31	- 3	3	35	36		39	40	41		43	44	- 4	6 47		49 5	٥
+096	h : hr					06	0	8 09	10		12	13	14 1	15 1	16			20	21			2	52	6 27	28		30 3	31	- 3	3	35		37		40	41	42	43	44	- 4	6 47		49 5	0
+120	h : hr						0	8 09				13 '	14 1	15 1	16 1	7		20				2	52	6 27	28	29		31	- 3	3	35		37		40	41		43	44		47			
+144	h :		02				0	8 09				13		15	1	7		20							28		30 3	31	3	3	35	- 3	37			41	42		44		47		- 5	٥
+168	h :												14		16 1	7		20		22					28		- 3	31			35	- 3	37					43			47		- 5	0
+192	h:												14 1	15	1	7		20										31					37					43					- 5	0
+216	h:													15														31				- 3	37		40			43					- 5	0
+240	h :																											31				- 3	37											



Probability (%) of Tropical Cyclone Intensity falling in each category

TD[up to 33] TS [34-63] HR1[64-82] HR2 [83-95] HR3 [> 95 kt]



Fr114

Sat15

Sun16

Mon17

Thu13

Wed12

© ECMWF

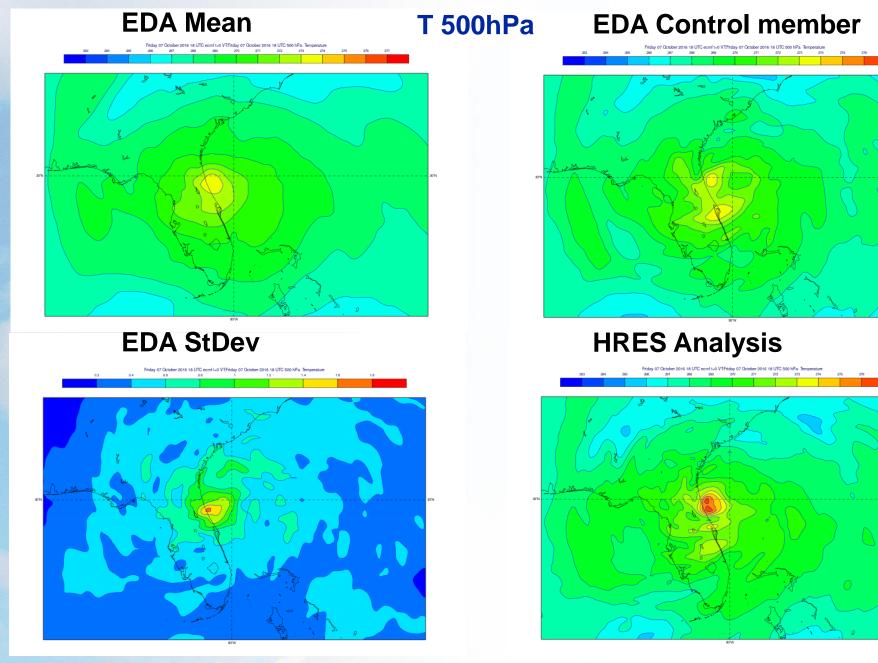
950 · 940 ·

> Sat 8 Oct 2016

Mon 10

Tue1

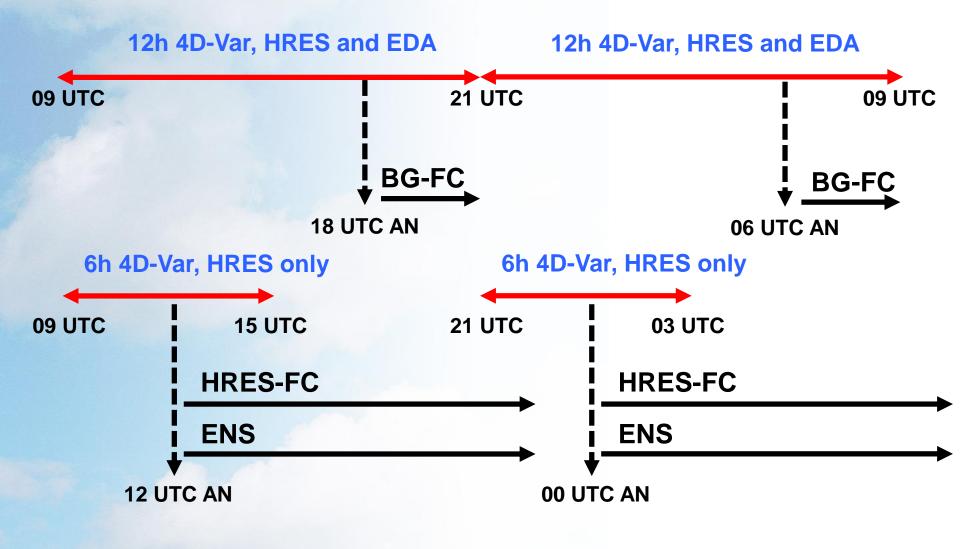
Sun 9





Slide 10

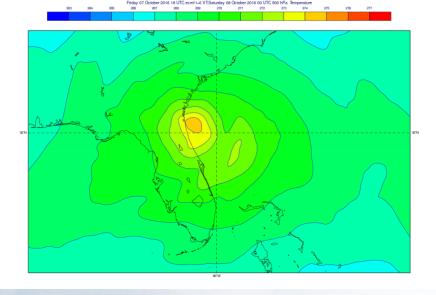
Early delivery Suite:





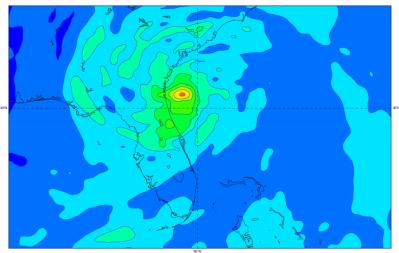
EDA Mean 18 UTC + 6h

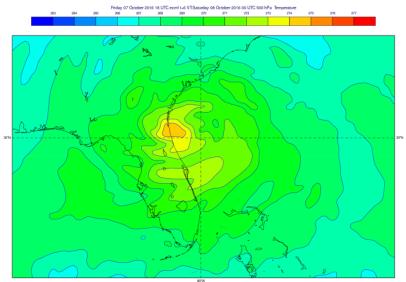
T 500hPa



EDA StDev 18 UTC + 6h

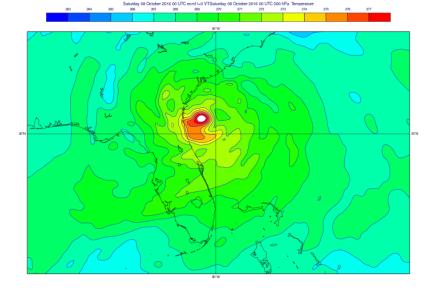






EDA Control 18 UTC + 6h

HRES Analysis 00 UTC





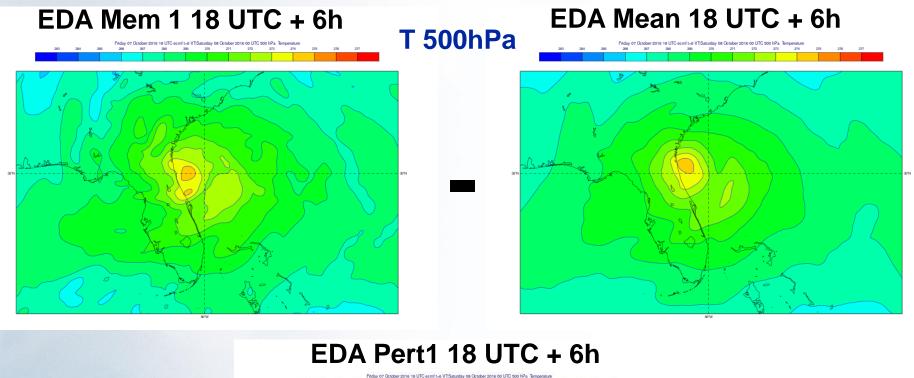
Slide 12

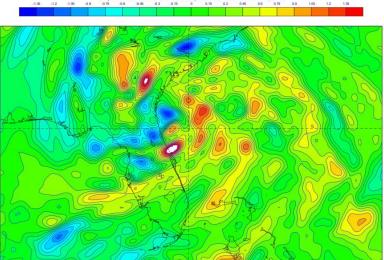
Generation of initial conditions for the ensemble:

$$AN_{pf} = AN_{Hres} \pm (EDA_i - \overline{EDA}) \pm SVPERT_j \qquad \begin{array}{l} i = 1..25\\ j = 1..25\end{array}$$

$$EDA \quad AN_{Hres} \quad \overline{EDA} \quad AN_{Hres} \quad EDA \quad SVPERT_j \quad EDA : 6h\\ Forecasts \quad EDA : 6h\\ Forec$$

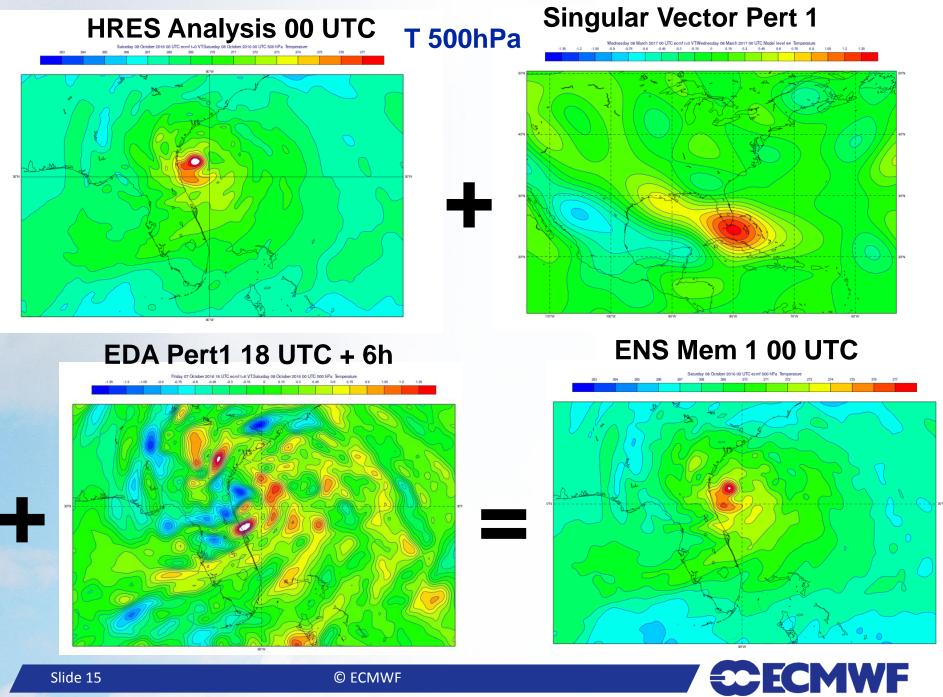




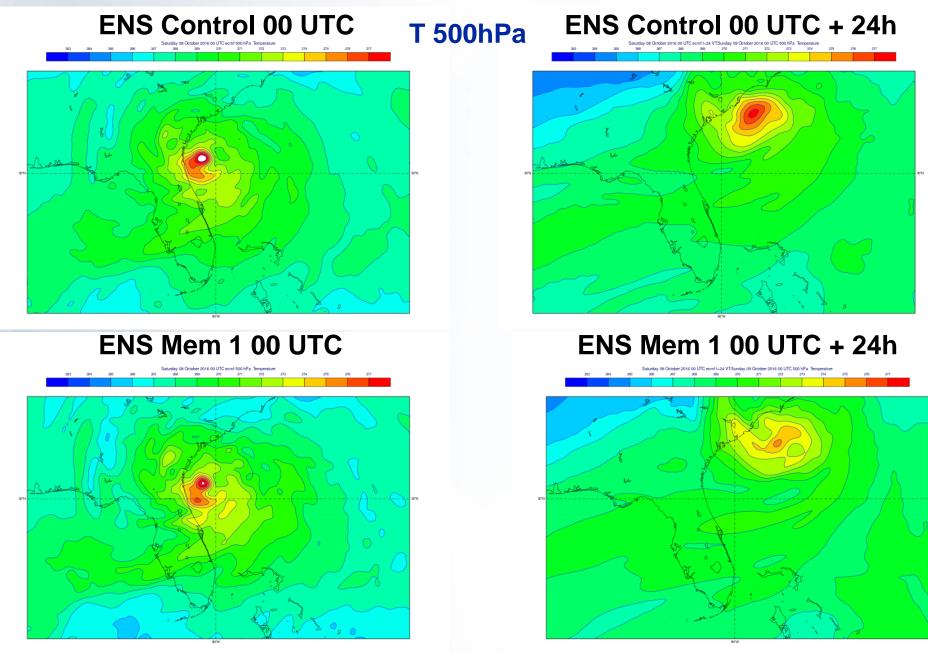








Slide 15

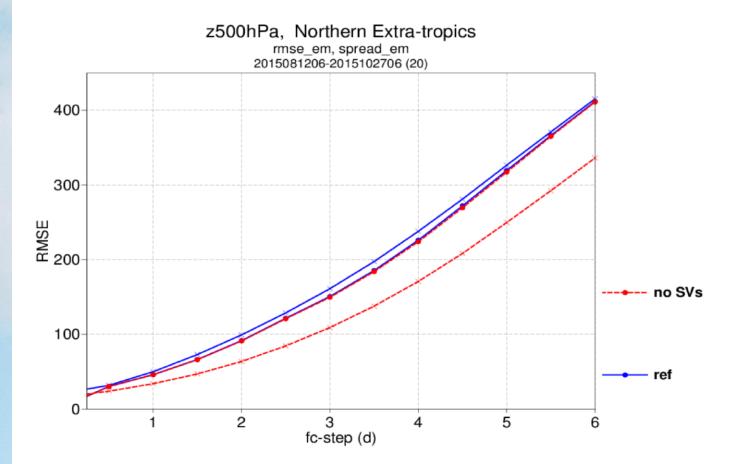




Slide 16

Why Singular Vectors?

Impact of SVs on ENS



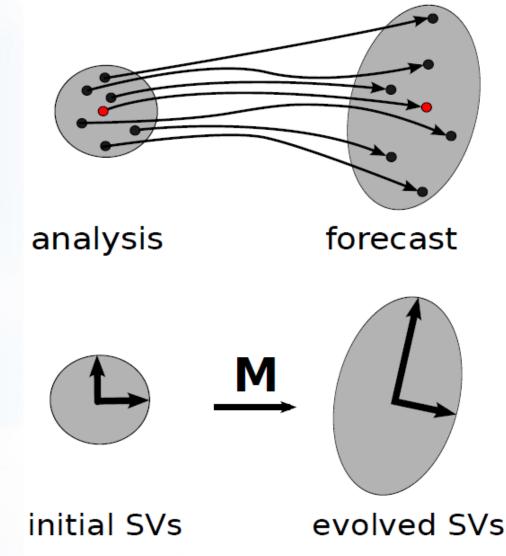
Oper like setup, TCo399, 20 Initial dates



Singular Vectors?

Directions of fastest growth over a finite time interval (optimisation interval)

EDA + Model Uncertainty representation produce substantial spread in the directions of the leading SVs but ensemble still under dispersive (Leutbecher and Lang (2013))





Singular Vectors?

Singular vectors are computed by solving an eigenvalue problem (e.g. Leutbecher and Palmer, 2008):

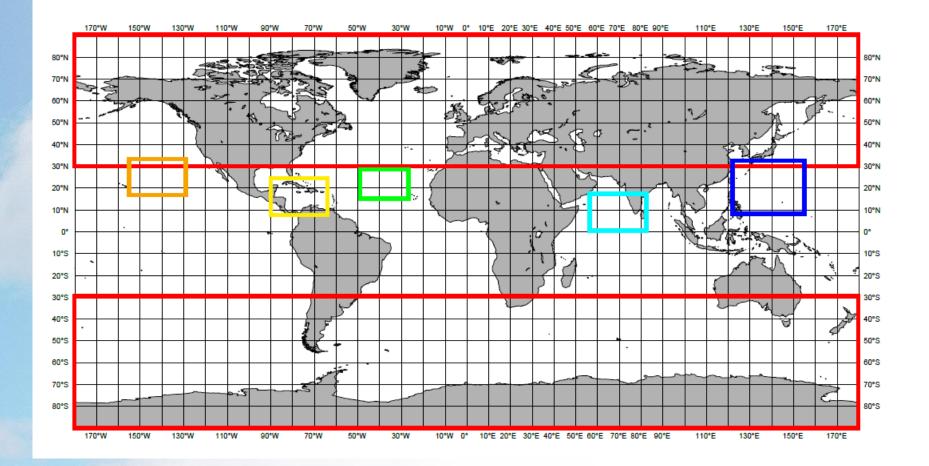
$$C_0^{-1/2} M^* P^* C_1 P M C_0^{-1/2} v = \sigma^2 v$$

- C_0 and C_1 initial and final time metrics
- M(0,t) linear propagator from time 0 to t and its adjoint M^*
- *P* and *P*^{*} local projection operator and its adjoint

$$\frac{1}{2} \int_{p_0}^{p_1} \int_{S} \left(u^2 + v^2 + \frac{c_p}{T_r} T^2 \right) dp \, ds + \frac{1}{2} R_d T_r p_r \int_{S} (\ln p_{\rm sfc})^2 ds$$



SV-Opt Regions

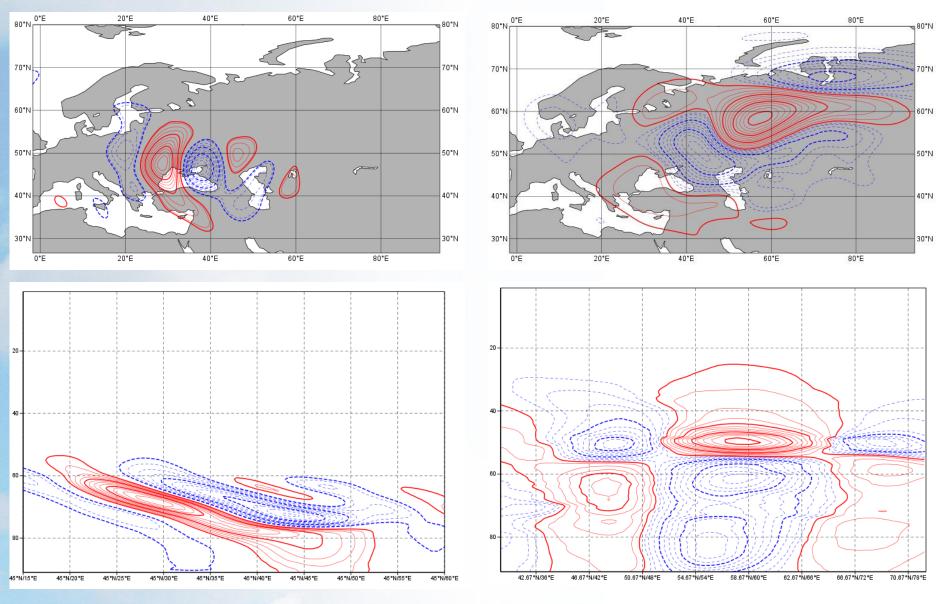




Slide 20

Initial SV, T mlevel 68

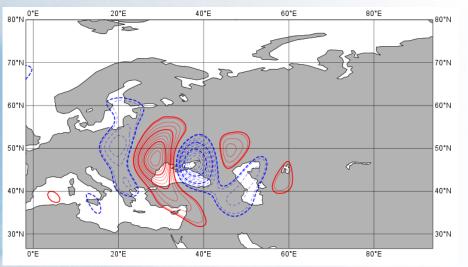
evolved SV, T mlevel 49



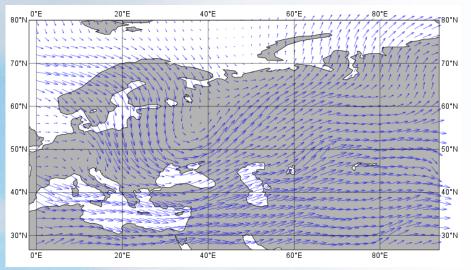


Slide 21

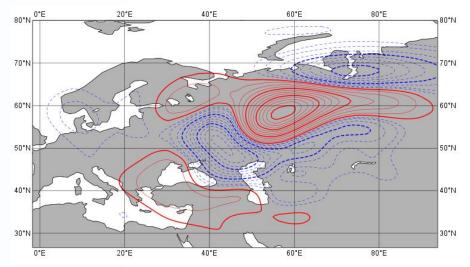
Initial SV, T mlevel 68



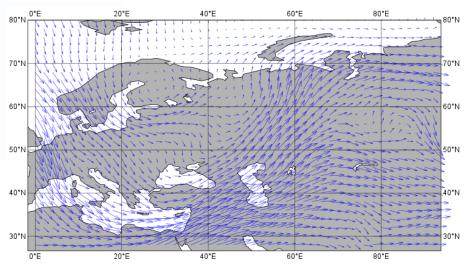
Analysis, Wind 200hPa



evolved SV, T mlevel 49



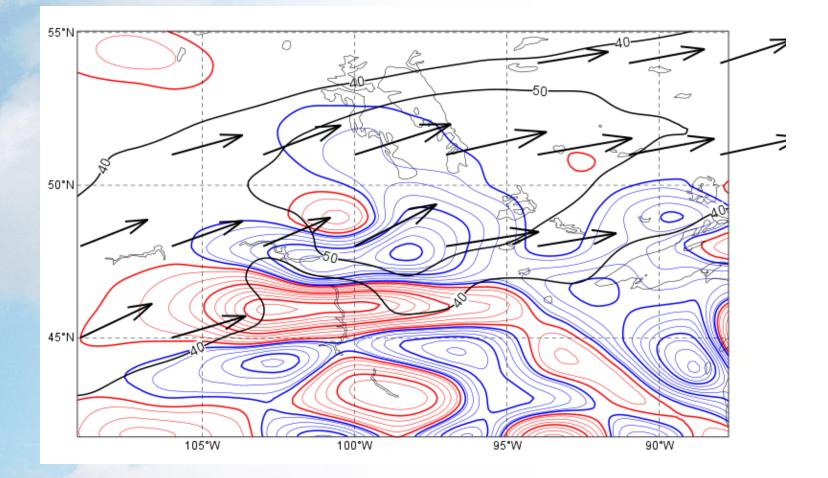
48h Fcst, Wind 200hPa





Slide 22

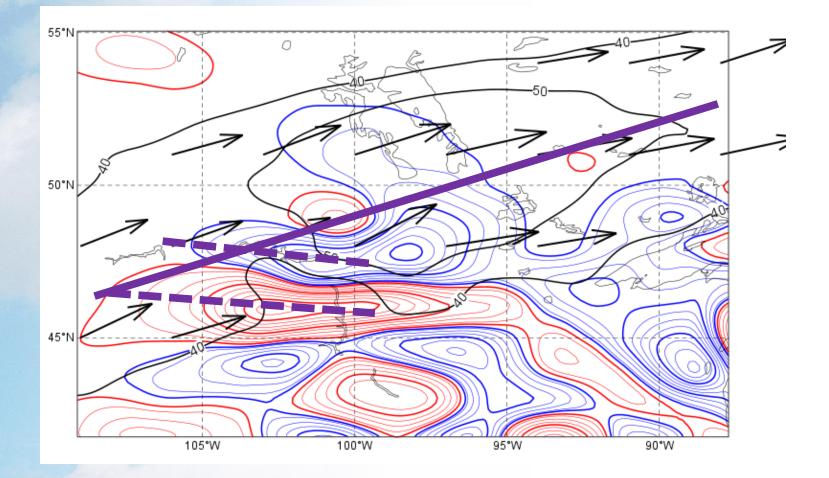
Analysis Increments (beginning of window), Vo mlevel 78







Analysis Increments (beginning of window), Vo mlevel 78





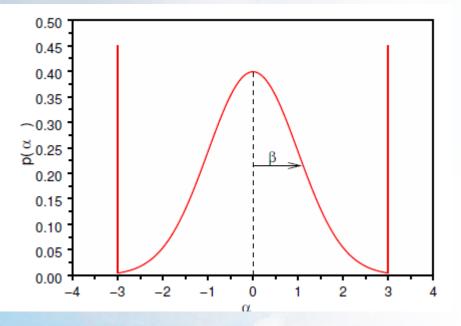


Singular vector ensemble perturbations

$$AN_{pf} = AN_{Hres} \pm (EDA_i - \overline{EDA}) \pm SVPERT_j$$

 $i = 1..25$
 $j = 1..25$

$$SVPERT_j = \sum_{l}^{NSET} \sum_{k}^{NSV_l} \alpha_{lk} SV_{lk}$$



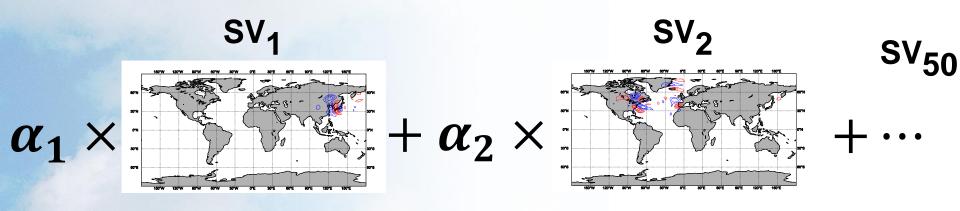
 α random number drawn from Truncated gaussian

NSET : nhem, shem, TCs1-6 NSV : 50 for nhem and shem, 5 for TCs

 β : standard deviation of Gaussian for a set of singular vectors depends on the EDA ensemble standard deviation of the day



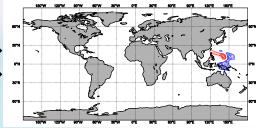
Combine SVs to construct Perturbations:



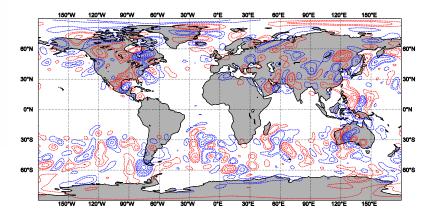
SV_{TC1}



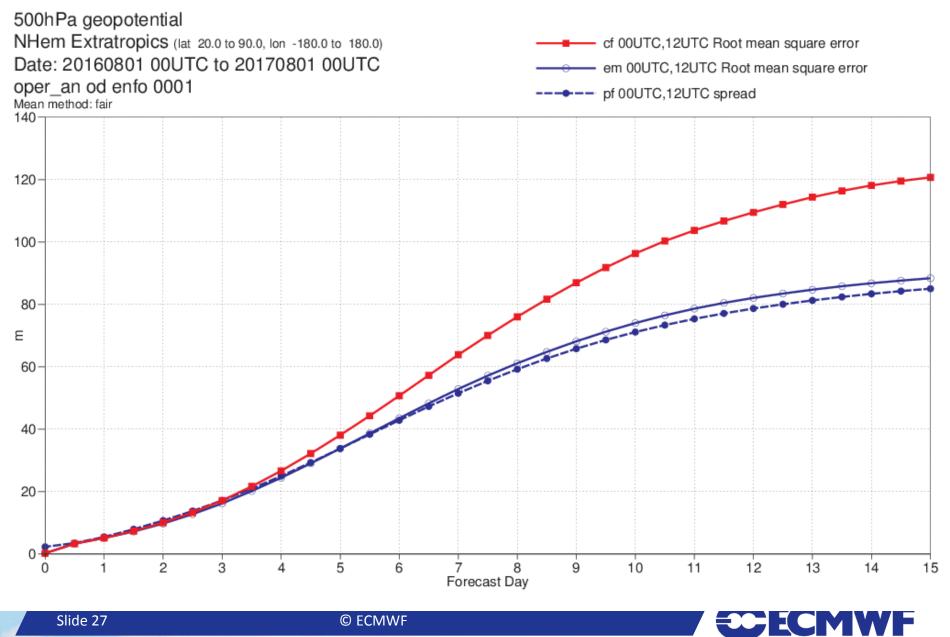




SVPERT 1



Ensemble Spread vs Error



Slide 27

How to improve?

Improve SV perturbations:

- SV resolution
- Moist processes in SV computation
- Number of SVs

Improve EDA:

- Number of ensemble members
- Analysis quality and resolution
- Specification of observation errors
- Model uncertainty representation



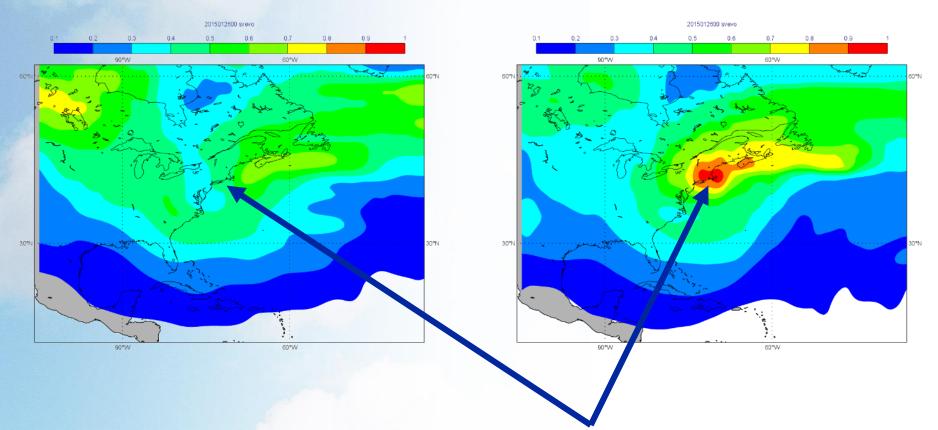
Singular Vector Configuration:

2015012600+48h

50 dry T42 evolved SVs

Slide 29

150 moist TL95 evolved SVs



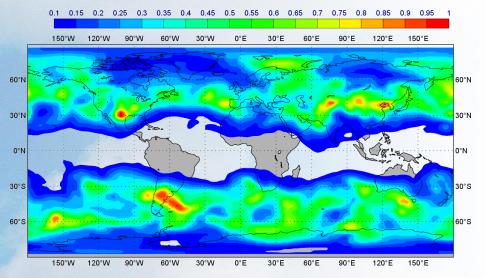
East coast blizzard, 27/28 Jan 2015

Normalised vertically integrated total energy: vte=vte/vte_max (global max)



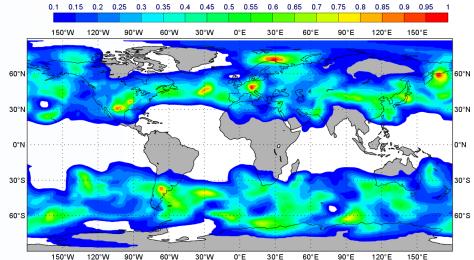
50 dry T42 initial SVs

Sunday 02 February 2014 00 UTC ecmft+0 VT:Sunday 02 February 2014 00 UTC Model level 1 U component of wind



50 moist TL95 initial SVs

Sunday 02 February 2014 00 UTC ecmft+0 VT:Sunday 02 February 2014 00 UTC Model level 1 U component of wind

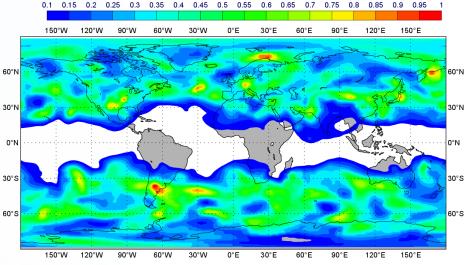


150 moist TL95 initial SVs

Sunday 02 February 2014 00 UTC ecmf t+0 VT:Sunday 02 February 2014 00 UTC Model level 1 U component of wind

Normalised vertically integrated total energy: vte=vte/vte_max

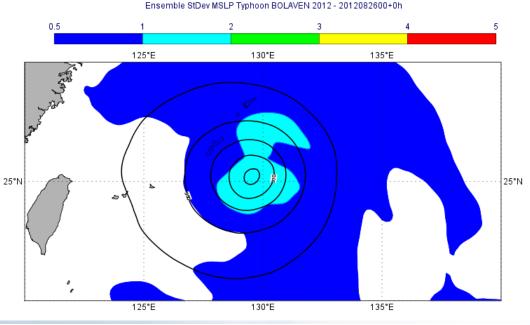
Date: 2014020200



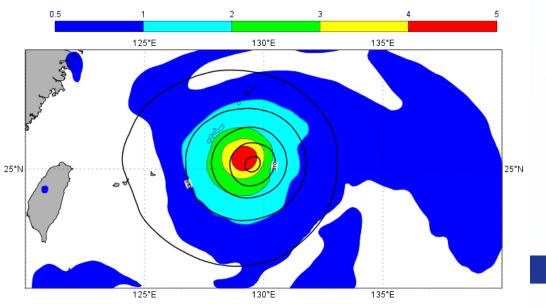




Example: EDA Resolution



Ensemble StDev MSLP Typhoon BOLAVEN 2012 - 2012082600+0h



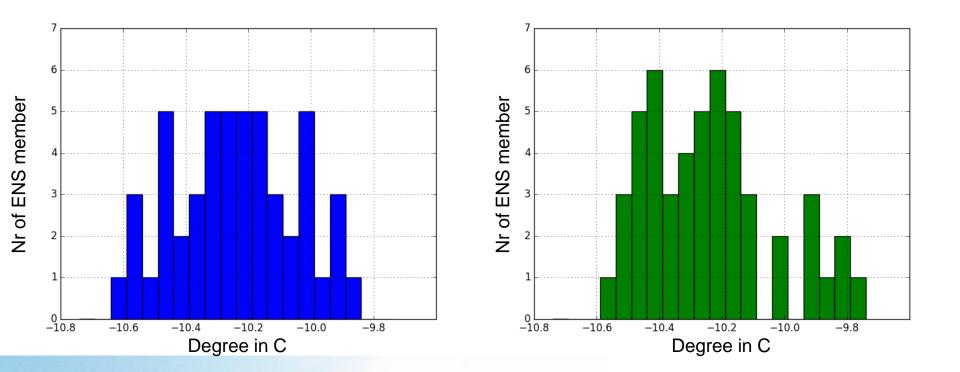
Typhoon BOLAVEN 2012 MSLP ENS StDev

ENS with perturbations from TL399 EDA

ENS with perturbations from TL639 EDA



ENS with 25 member EDA plus-minus vs. 50 member EDA



Grid Point Temperature values of Ensemble members at t=0

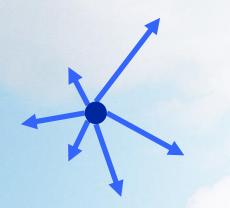


Re-centring with multiple centre analyses

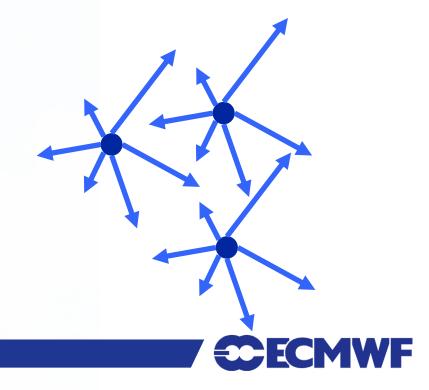
Wish: start from high quality real-time EDA (see Lang et. al 2015) -> very expensive

Here: centre on a small number of HRES like analyses with different satellite obs selection for thinning (ECMWF's Data Assimilation Strategy 2017). No model unc. representation and obs perturbations in the centre analyses to limit sampling uncertainty.

Single Centre Analysis



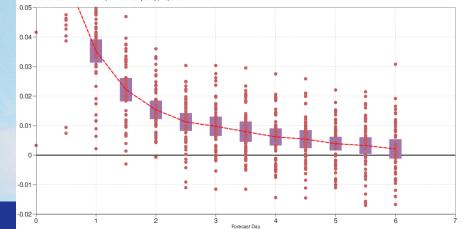


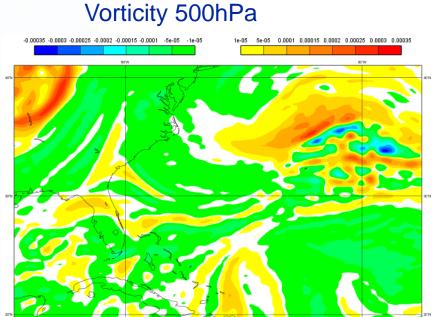


Re-centring with multiple centre analyses5-Ana ENS Mean Step 0,
Vorticity 500hPaRef ENS Mean Step 0,
Vorticity 500hPa

500hPa v component wind speed Continuous ranked probability score Tropics (at 20010 20.0, Im -180.0 to 180.0) Date: 20151201 00UTC to 20160120 00UTC 00UTC 1-0 1-12...T-144 (Confidence: [95.0] | Populator. 51

Slide 34







Summary

- Combining SVs with EDA results in a quite reliable system
- In order to obtain the best forecast skill it is currently necessary to re-centre on the HRES analysis
- Use of SVs justified as long the ENS lacks spread in the directions of the SVs

Options to improve the perturbed ENS initial condition in the future:

- singular vector resolution, number and moist processes
- Increase the number of EDA members and quality of the members, model uncertainty in EDA, obs. perturbations

-> Big challenge: how to use the available computer resources in the most effective way?

References:

- Bonavita, M, Trémolet, Y, Holm, E, Lang, STK, Chrust, M, Janiskova, M, Lopez, P, Laloyaux, P, De Rosnay, P, Fisher, M, Hamrud, M, English, S. (2017), A Strategy for Data Assimilation. ECMWF Tech. Memo 800
- Leutbecher, M. and Lang, S. T. K. (2014), On the reliability of ensemble variance in subspaces defined by singular vectors. Q.J.R. Meteorol. Soc., 140: 1453–1466. doi:10.1002/qj.2229
- Lang, S. T. K., Bonavita, M. and Leutbecher, M. (2015), On the impact of re-centring initial conditions for ensemble forecasts. Q.J.R. Meteorol. Soc., 141: 2571–2581. doi:10.1002/qj.2543



Additional Slides

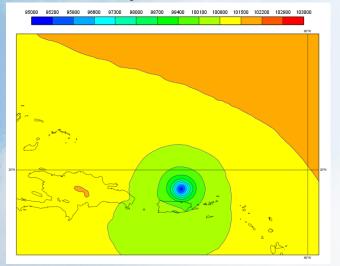


How can we adjust the ensemble spread? Pert* = a x Pert + b x Z multiplicative inflation additive inflation

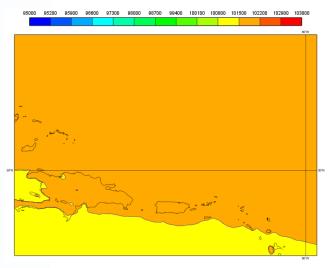


Flow dependency vs no flow dependency

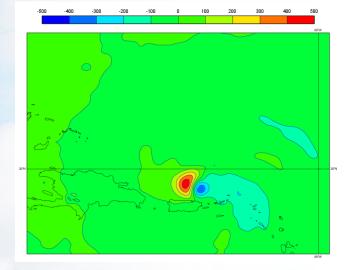
Analysis Irma



Analysis some other day



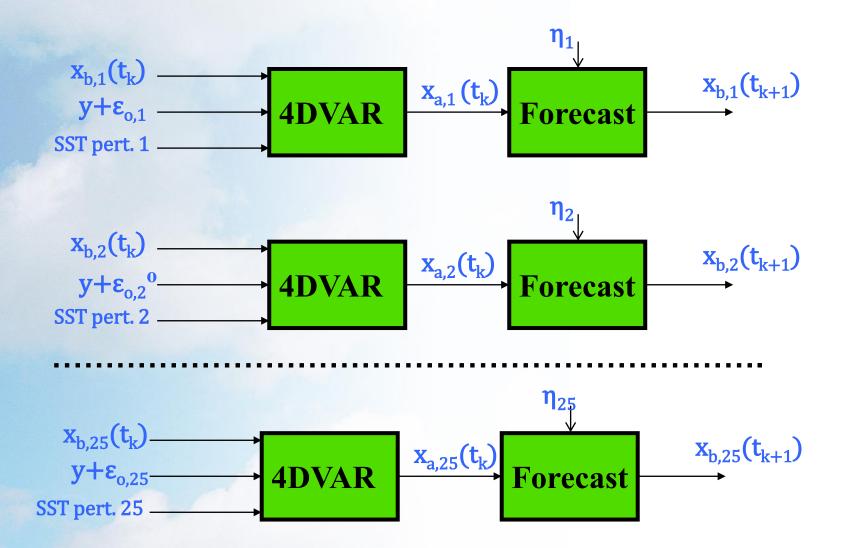
Perturbation EDA member 1





Slide 38

The Ensemble of Data Assimilations



From M. Bonavita's ECMWF DA training course lecture



Slide 39