



Towards a NWC Application Based on HARMONIE-AROME

Carlos Geijo Guerrero, AEMET



Some considerations about NWP-NWC

- **LAM .vs. GLOBAL-NWP.** GLOBAL-NWP will always have it difficult to provide this kind of service. Strategic interest for LAM-NWP operators.
- **DA is at the core of a NWP-NWC system.** DA improved as the most practical way to tackle and give first benefits in forecasting atmos. phenomena with very short predictability time scales.
- DA developments in LAM-NWP and **DA legacy and directives from GLOBAL-NWP.** Time to break up ?
- Near future foreseeable scenario: **important increase in atmos. data flux.** (e.g. crowd sourced data). Given that management and technical issues are sorted out: what to do with all these data ?
- **Ensemble generation for NWC.** DA a key element
- A big challenge in many respects. **Comprehensive re-organization of production** after requirements of adaptability and early delivery. (UWC opportunity situation ?)

Towards a NWC prototype on HARMONIE-AROME

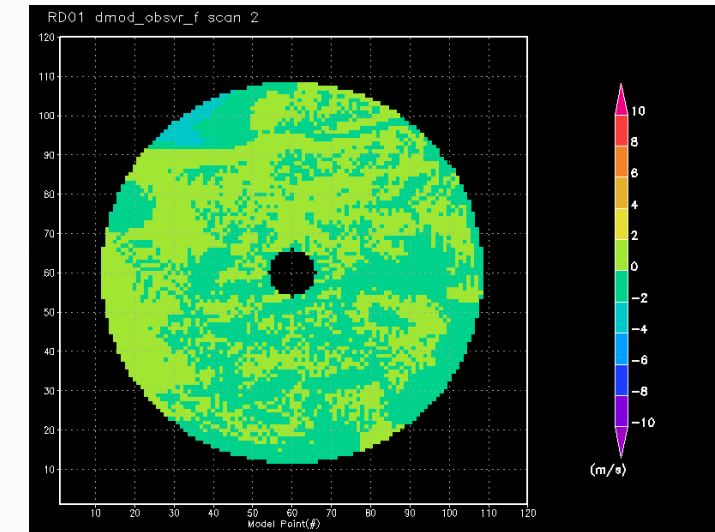
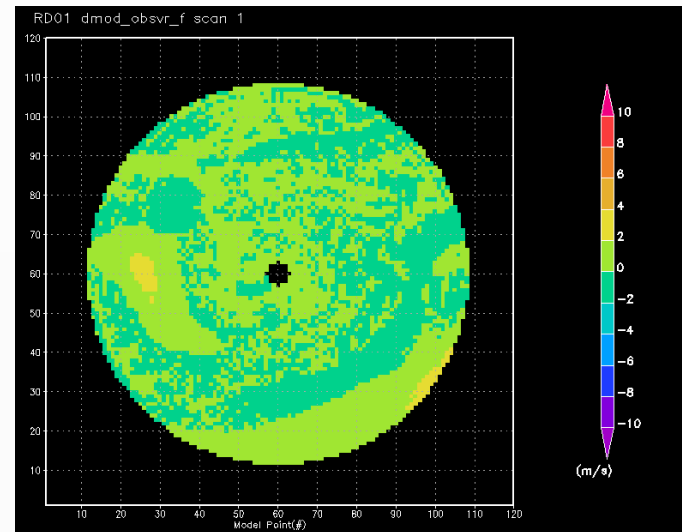
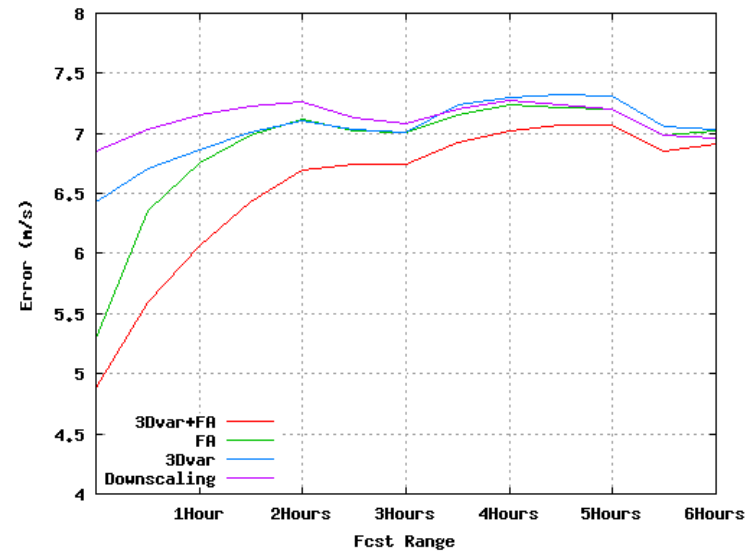
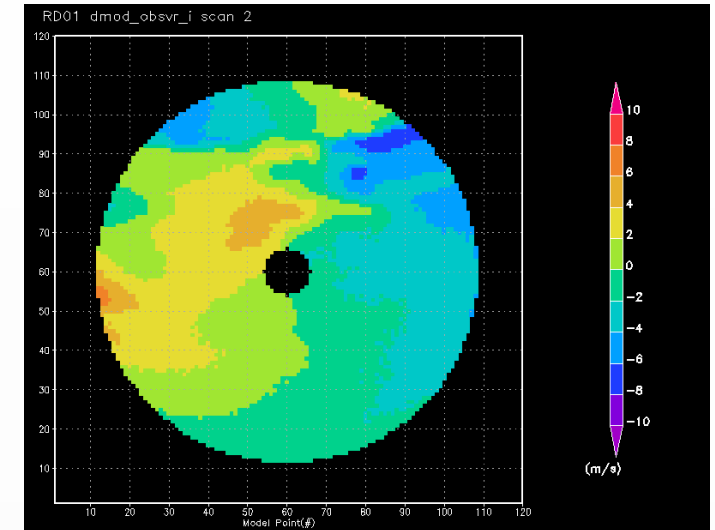
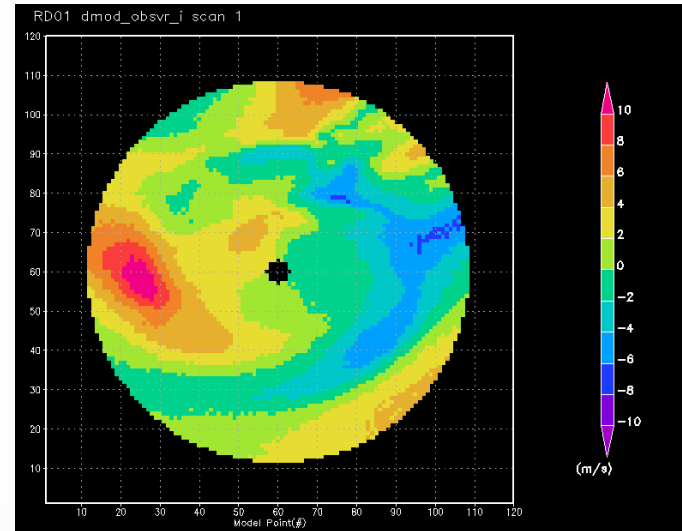
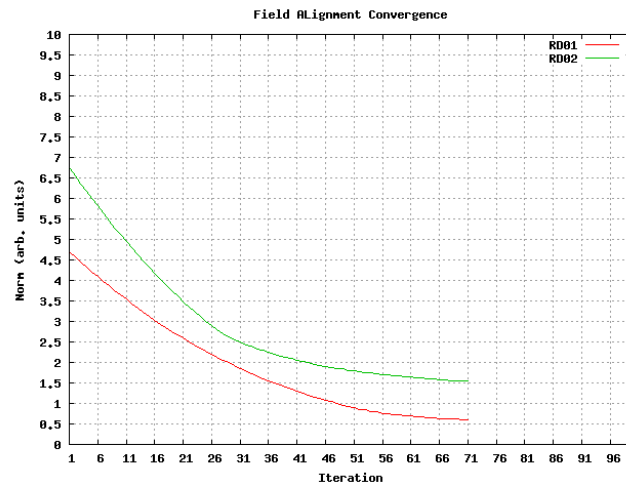
Work during last months to integrate in v40 some interesting **NWC oriented developments**

**(RWP 2019, WP DA-2 “Development of Flow-Dependent Algorithms”,
WP DA-5 “Development of DA suited for NowCasting”)**

1. The Field-Alignment software running in configuration 131 (variational job).
Also in configuration 2 (screening) ?
2. Variational Constraints to better balance, as dictated by SI dynamics, the initial conditions, also (to be) integrated in config. 131
3. Introduction of Flow-dependency in covariances by means of Gaussian Integrals ?

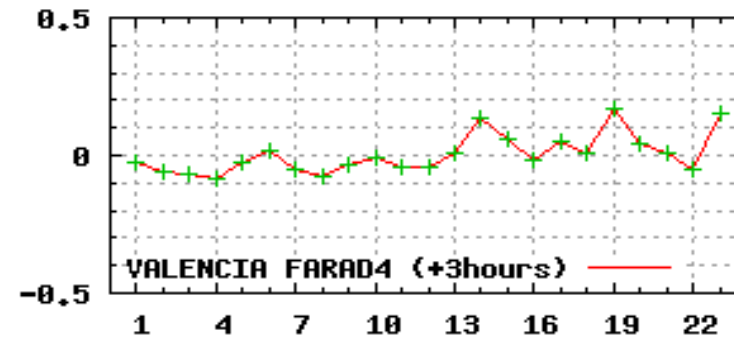
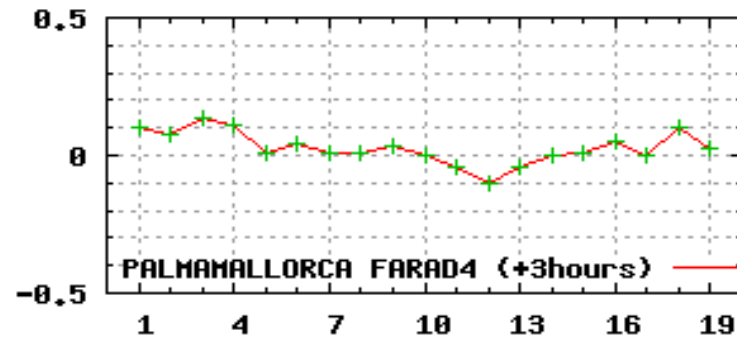
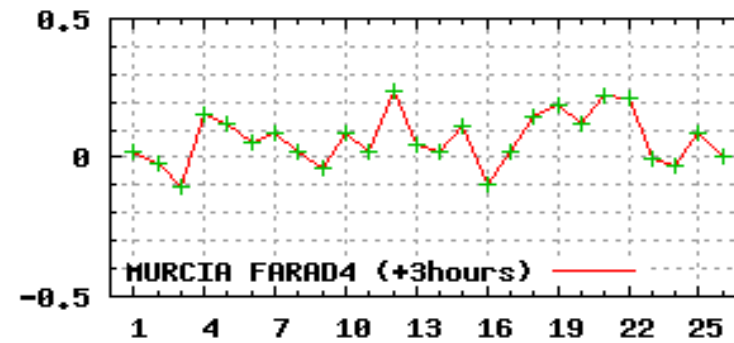
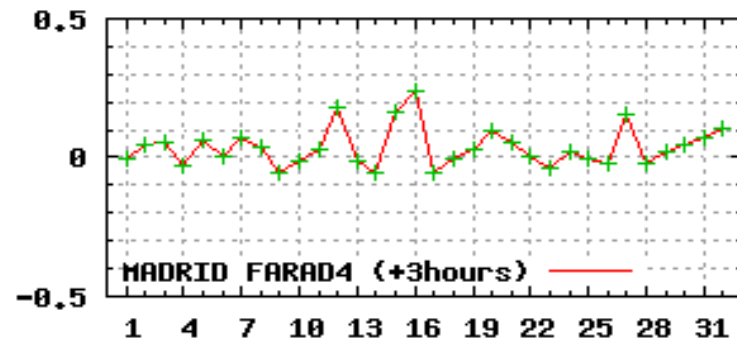
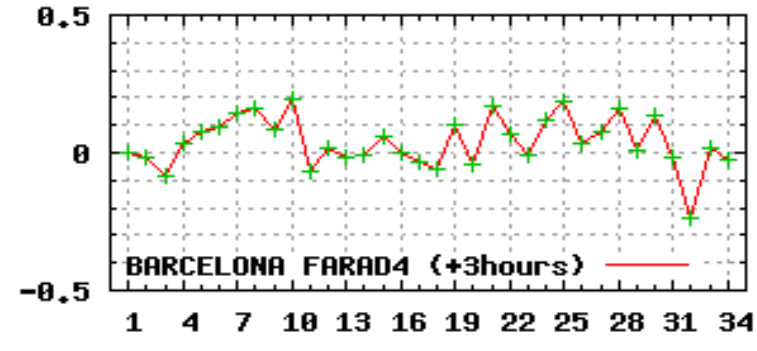
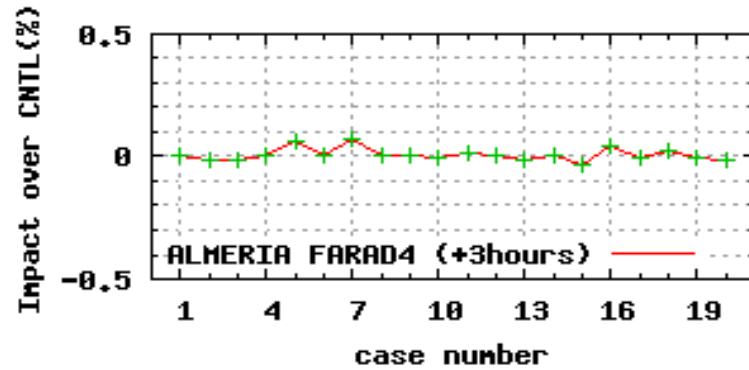
FIELD ALIGNMENT:

Ravela S., Emmanuel K. and McLaughlin D. (2007)



FIELD ALIGNMENT:

HYMEX –SOP1 (september-november 2012)



FIELD ALIGNMENT

Integration in v40

FA corrections are treated as full 2-D wind observations (*), which are after added to CMA ODB for Minimization, together with the rest of observations that may be available. This has required some re-ordering of the obs setup for NCONF=131



CNT0

SU0YOMA

SU0YOMB

```
IF ( LOBS. AND. (.NOT.LFA) ) THEN  
OBADAT  
ENDIF
```

(NCONF == 1,2) CNT1

(NCONF == 131) CVA1

CVA1

```
IF ( LOBS. AND. (.NOT.LFA) ) SUOBS (*)
```

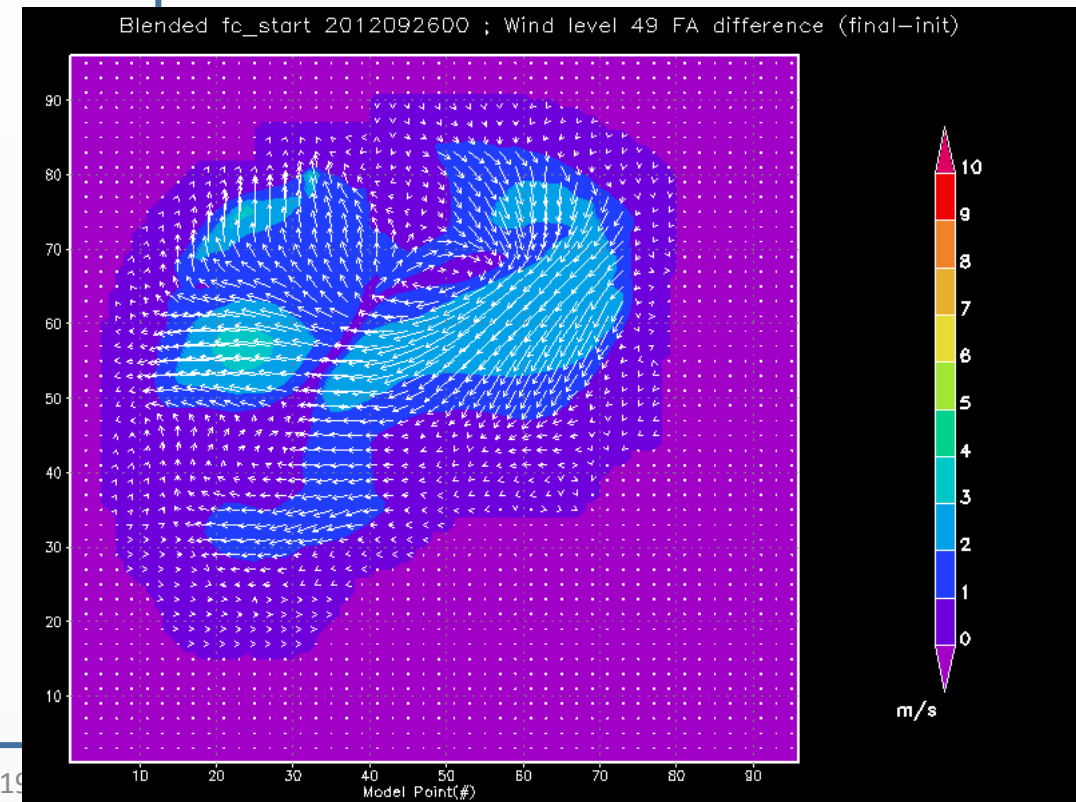
SU1YOM

```
IF ( LFA .AND. LOBS ) THEN  
FA  
OBADAT  
SUOBS  
ENDIF
```

CNT2

....

CVA2



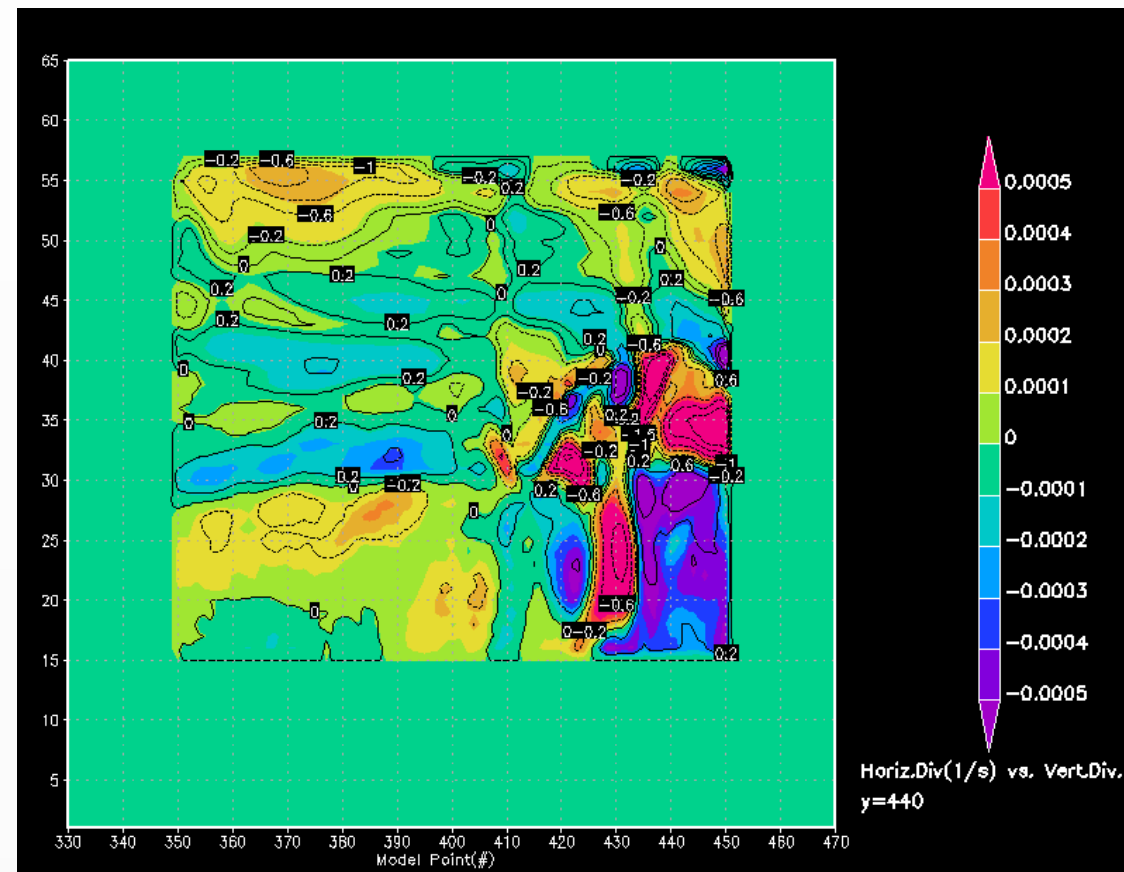
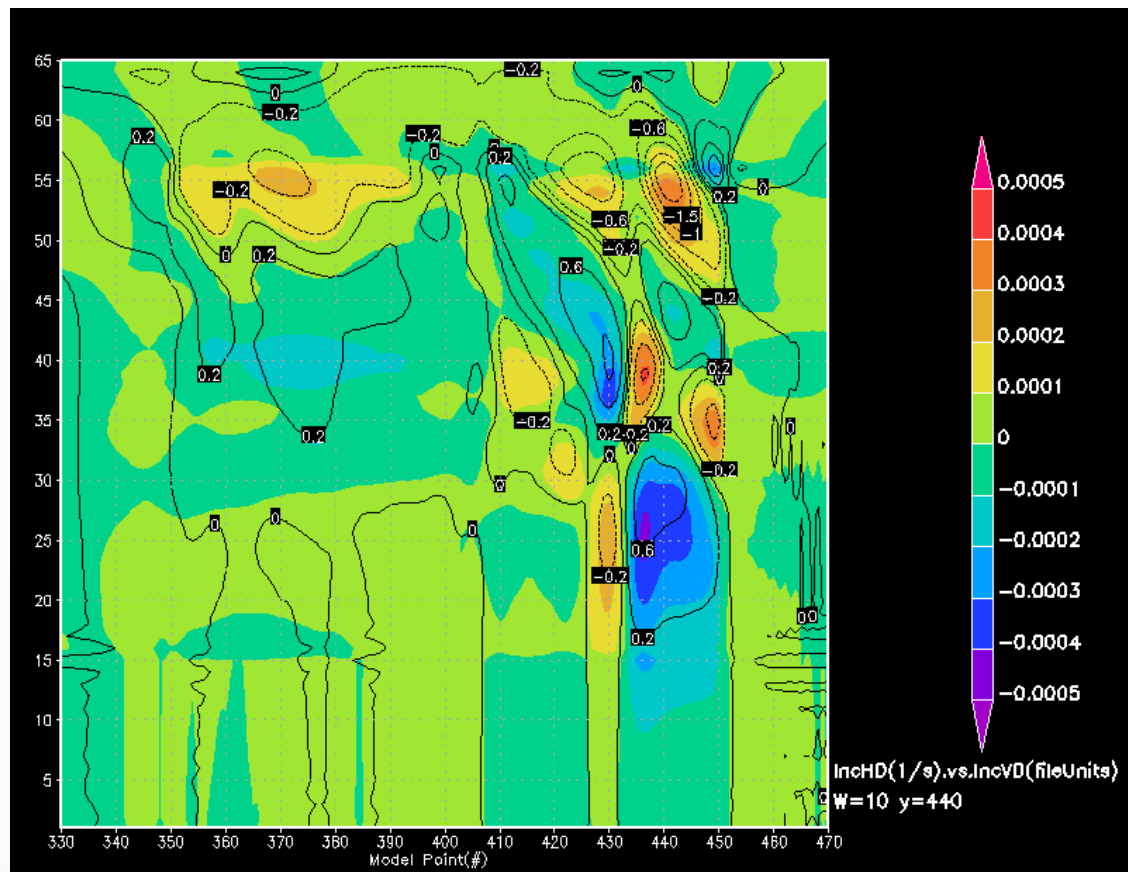
Variational Constraints (VC)

- Variational technique to balance the IC, as dictated by SI dynamics
- Numerical solution found with Green Functions. Interesting analogy between GF and covariance matrix
- Tested in several contexts, (with synthetic obs, on Field Alignment increments, on LETKF increments, and as substitute or complementary to 3D-Var statistical balances)
- Potential utilization as nudging scheme (not tested yet)

Variational Constraints (VC)

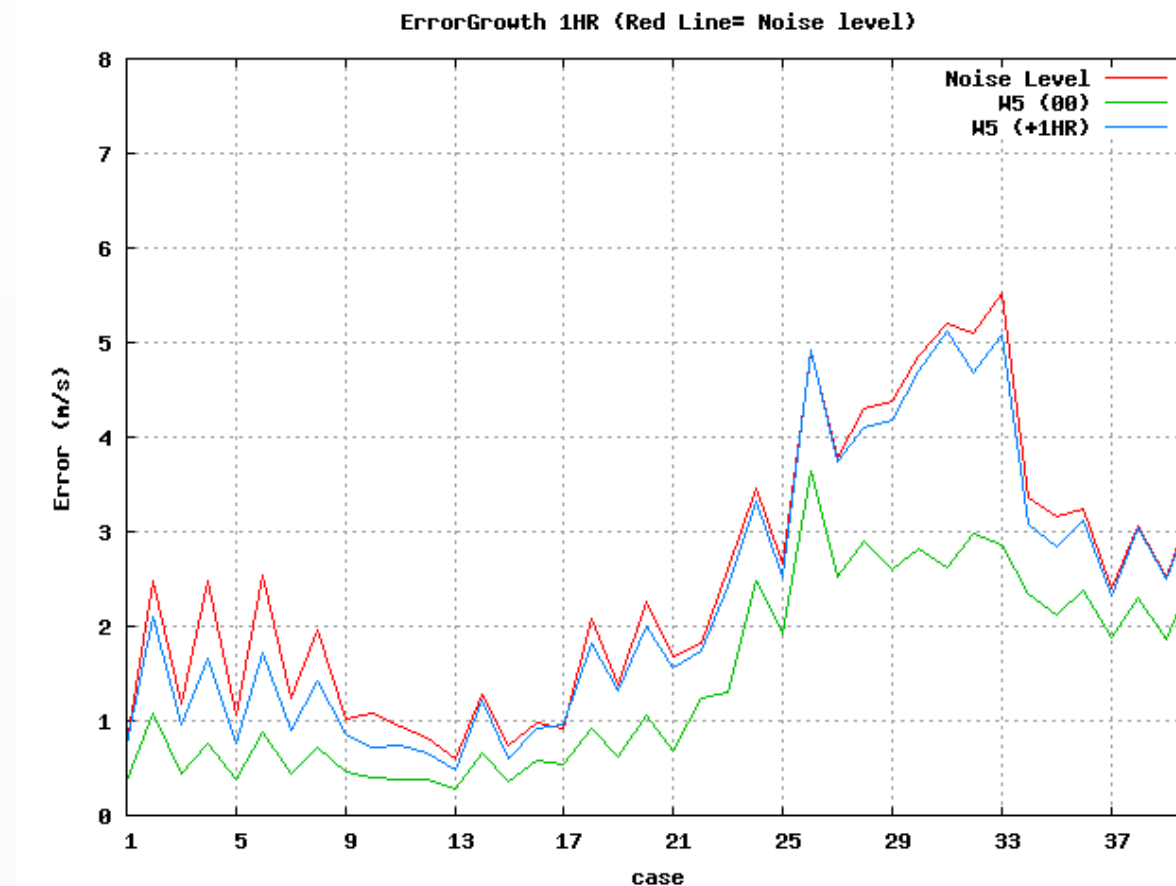
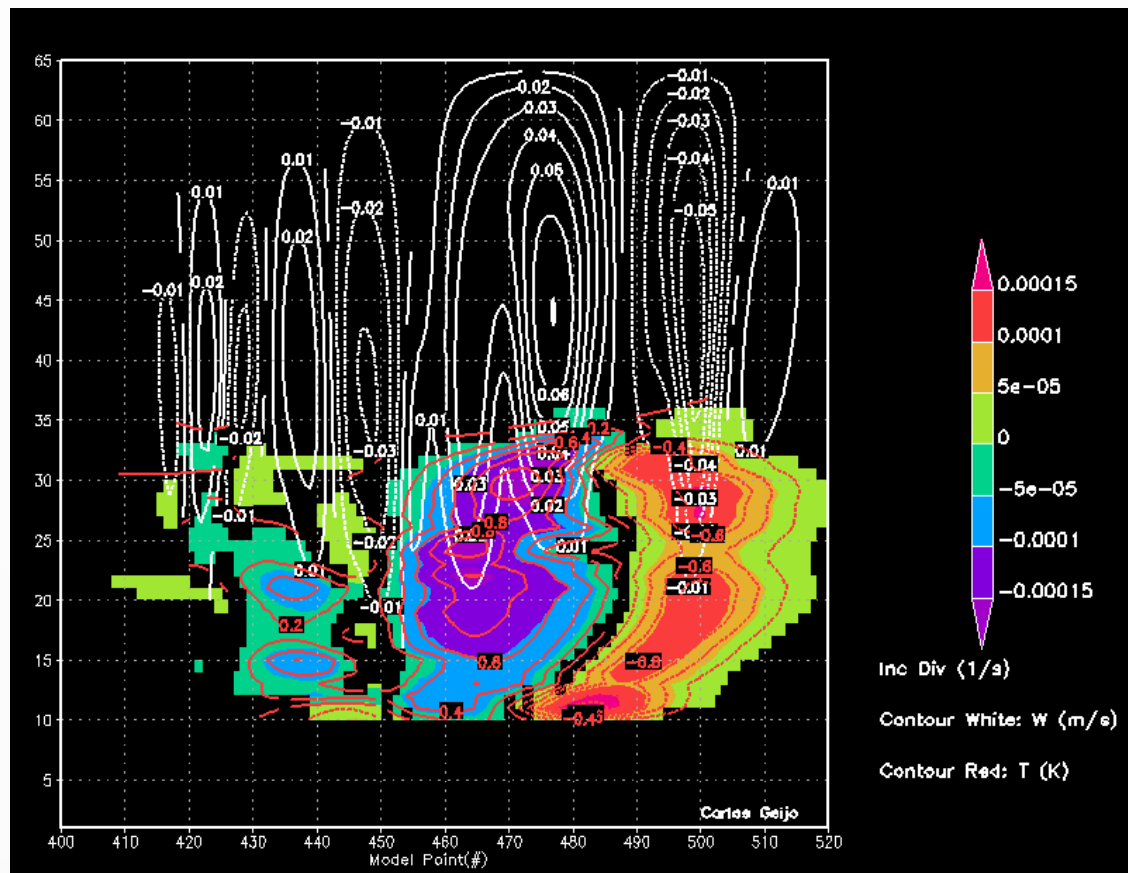


Test with Synthetic Observations



Variational Constraints (VC)

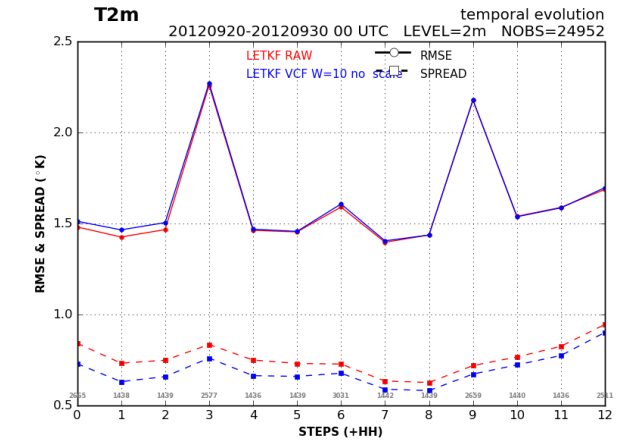
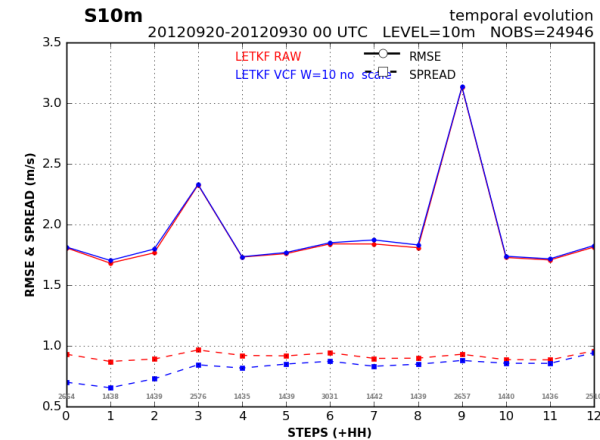
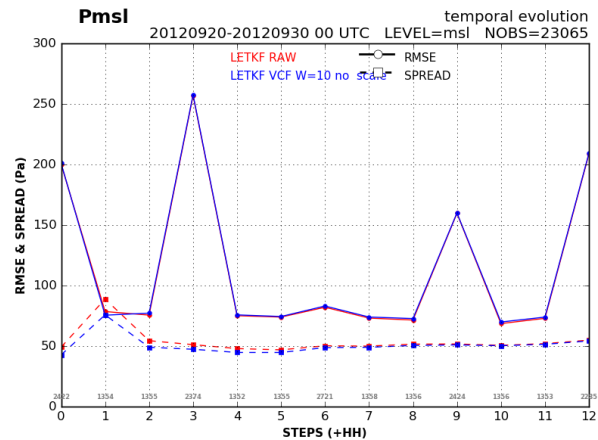
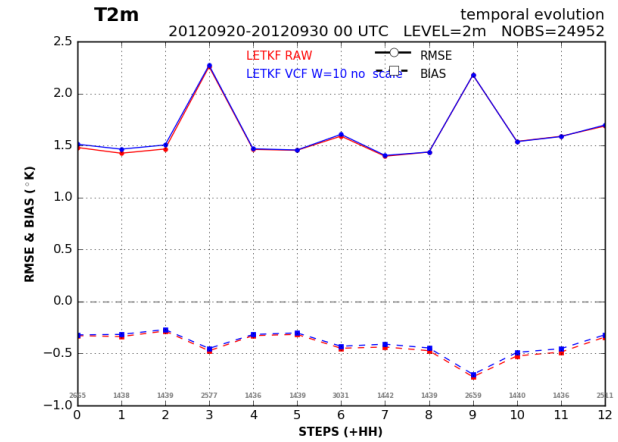
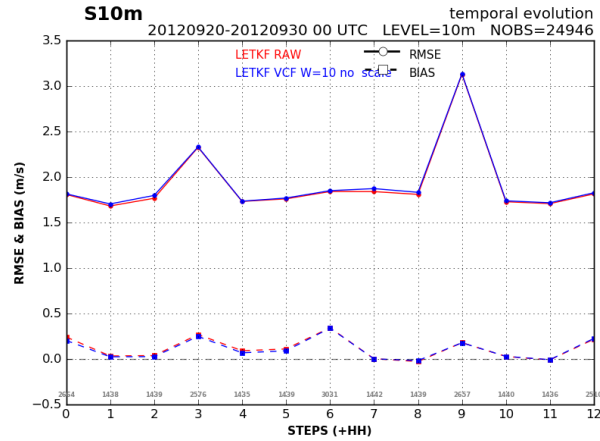
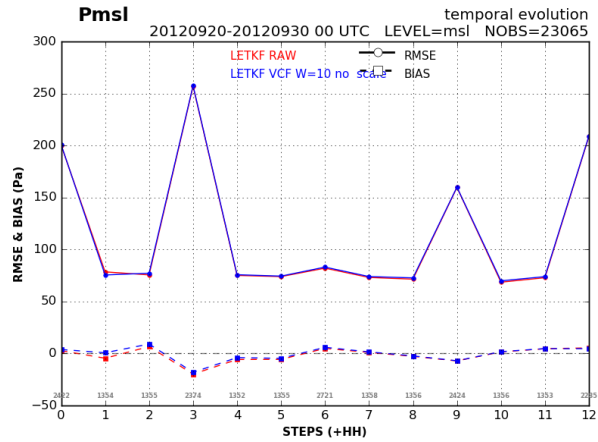
Test with Field-Alignment Increments



Variational Constraints (VC)



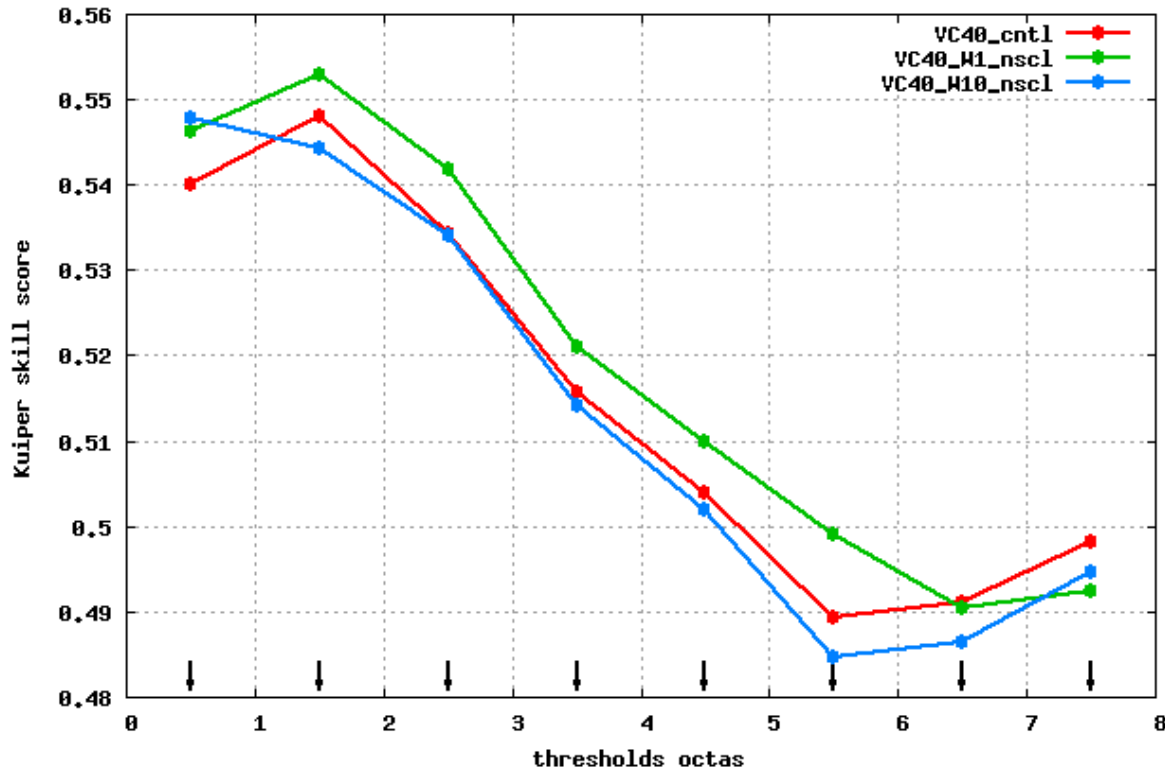
Test on LETKF increments FG + VCFilt [LETKF_analysis - FG]



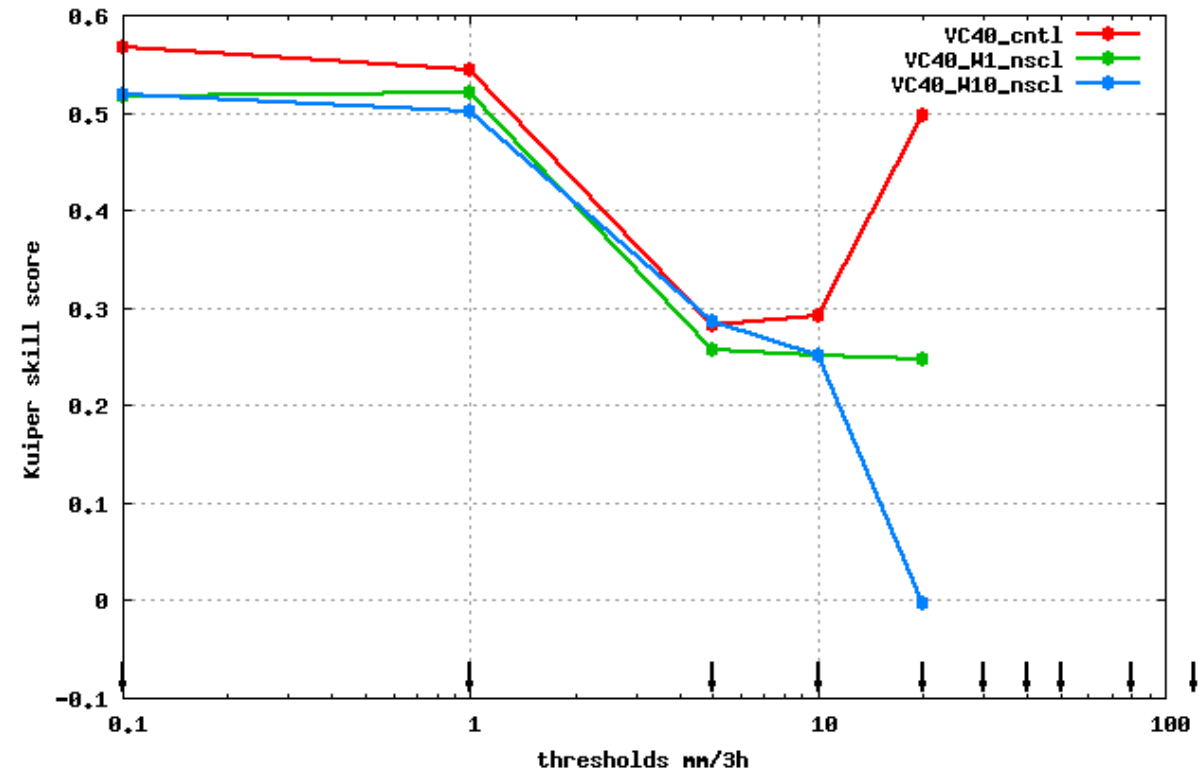
Variational Constraints (VC)

Test .vs. 3D-Var Stat. Bal FG + VCFilt [3D-VAR_analysis (uni.) - FG]

Kuiper skill score for Cloud cover (octas)
 Selection: ALL 149 stations
 Period: 20120920-20120930
 Used {00,03,...,21} + 06 12



Kuiper skill score for 3h Precipitation (mm/3h)
 Selection: ALL 63 stations
 Period: 20120920-20120930
 Used {00,03,...,21} + 06-03 12-09

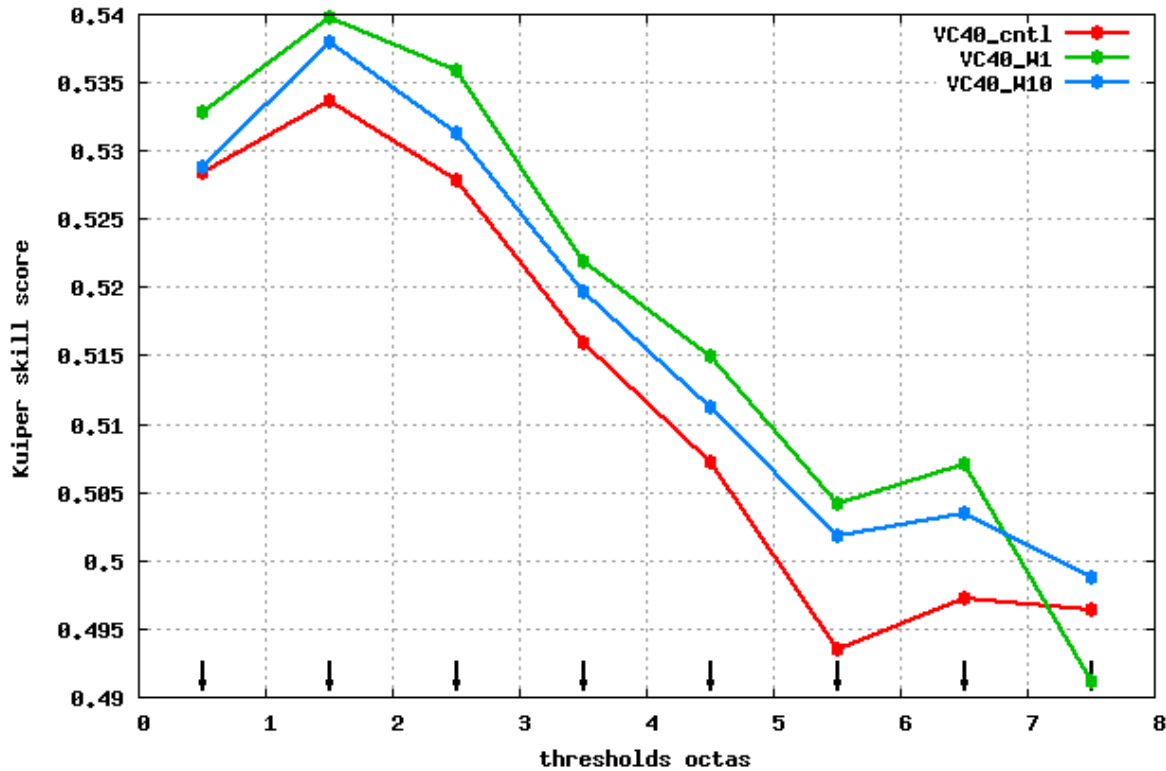


Variational Constraints (VC)

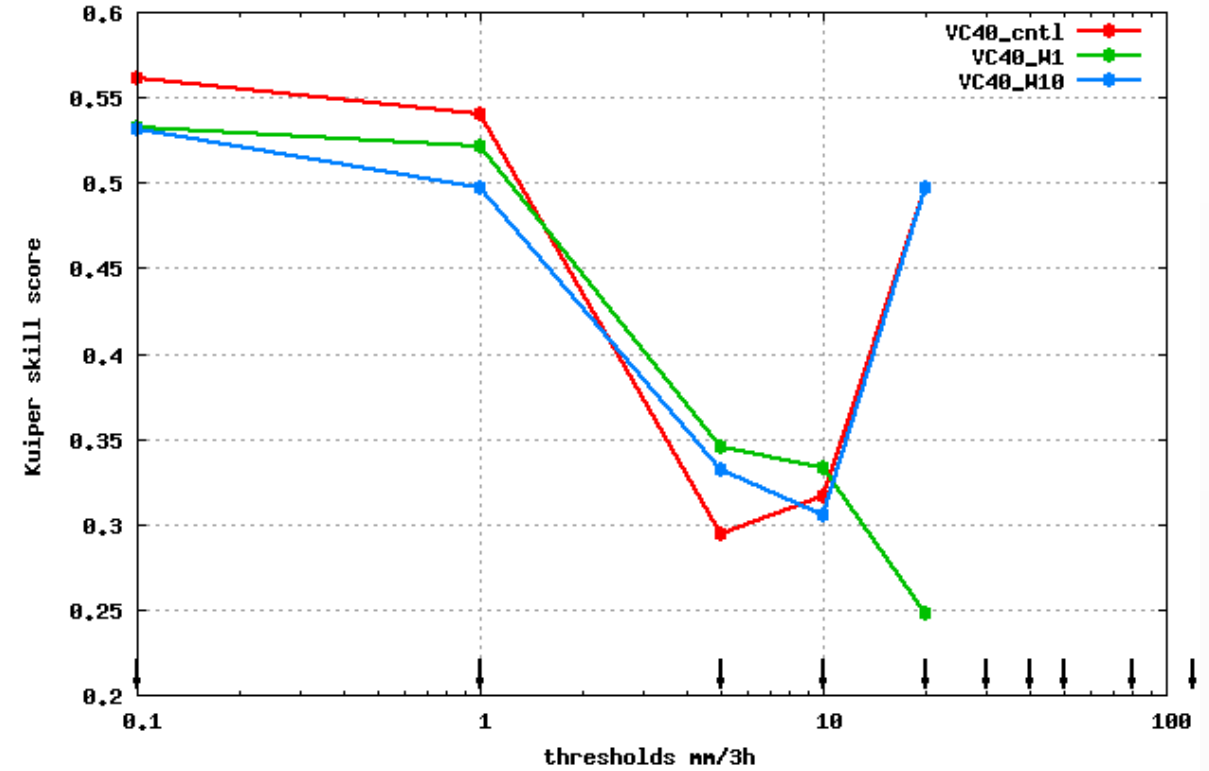


Test with 3D-VAR statbal FG + VCFilt [3D-VAR_analysis - FG]

Kuiper skill score for Cloud cover (octas)
 Selection: ALL 149 stations
 Period: 20120920-20120930
 Used {00,03,...,21} + 06 12



Kuiper skill score for 3h Precipitation (mm/3h)
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Modelling Flow-Dependent covariances with Gaussian Integrals

- Consider the analyses increments as a Gaussian random field on a grid with a source. The pdf looks much like (a discrete version of) the kernel in random (or quantum) field theory (QFT)

$$P(\Delta) \sim \exp(-J_S(\Delta)) \quad ; \quad J_S(\Delta) = \frac{1}{2} \Delta^T G^{-1} \Delta - \Delta^T S$$

- In this framework, the classical 3D-VAR results read :

$$\langle \Delta_i \rangle_S = \langle \Delta_i \Delta_j \rangle_{S=0} S_j = G_{ij} S_j$$

$$\langle \Delta_i^2 \rangle_S - \langle \Delta_i \rangle_S^2 = \langle \Delta_i^2 \rangle_{S=0} = G_{ii}$$

Modelling Flow-Dependent covariances with Gaussian Integrals

- GF as a covariance matrix \longrightarrow think the covariance matrix as a GF (“propagator”)
- As is common in QFT, introduce (non-random) external fields to calculate the effects of ambient or background fields on different probabilities

$$\langle \Delta_i \Delta_j \rangle_{S=0}^{\vec{V}} \sim \int d\Delta_1 \cdots d\Delta_N \Delta_i \Delta_j \exp\left(-\frac{1}{2} \Delta^T \underset{\text{a)}}{G^{-1}} \Delta - \frac{\mu}{2} \text{tr}\left(\left[\underset{\text{b)}}{\vec{V} \square \nabla \Delta}\right] \left[\vec{V} \square \nabla \Delta\right]^T\right)\right)$$

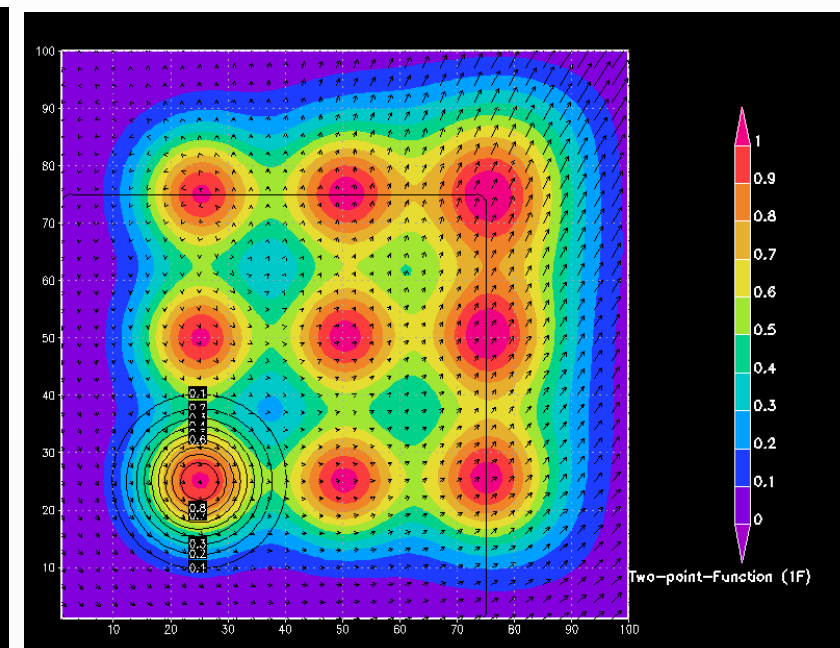
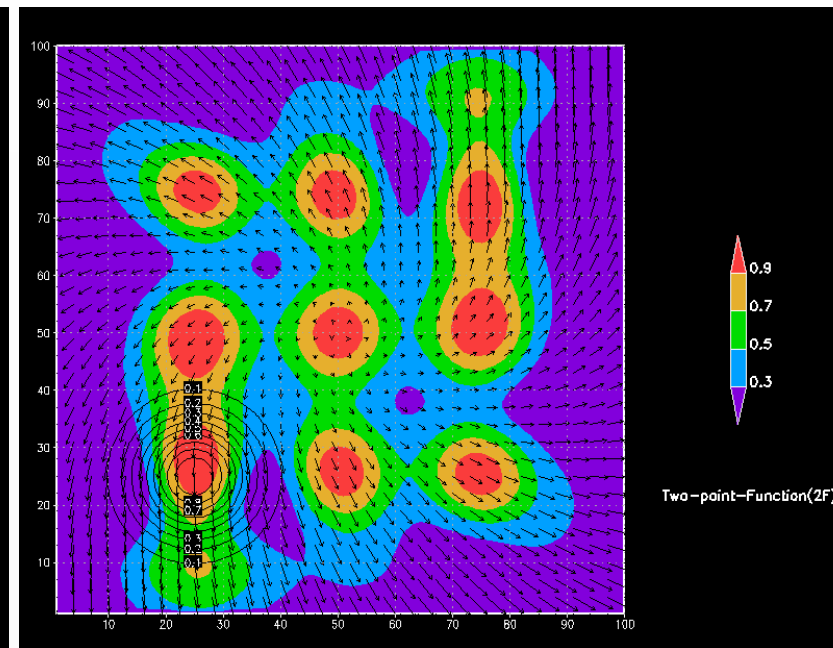
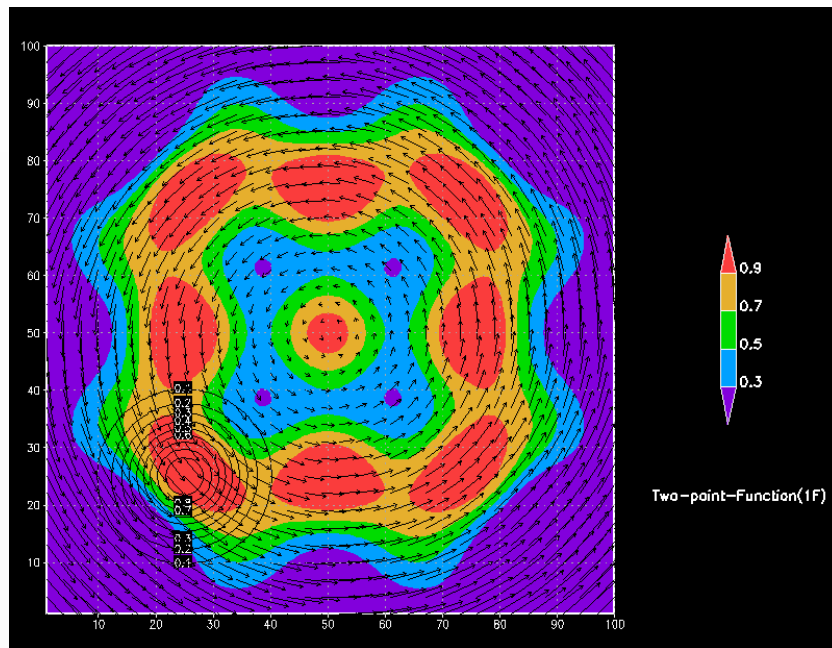
- Computations up to lower orders of μ can be done perturbatively

$$\langle \Delta_i \Delta_j \rangle_{S=0}^{\vec{V}} \sim \int d\Delta_1 \cdots d\Delta_N \Delta_i \Delta_j \exp\left(-\frac{1}{2} \Delta^T G^{-1} \Delta - \frac{\mu}{2} \left(\sum_p (\vec{V} \square \nabla \Delta)_p\right)^2\right) = \int d\Delta_1 \cdots d\Delta_N \exp\left(-\frac{1}{2} \Delta^T G^{-1} \Delta\right) \Delta_i \Delta_j \left[1 - \frac{\mu}{2} \sum_p (\vec{V} \square \nabla \Delta)_p^2 + O(\mu^2)\right]$$

$$\langle \Delta_i \Delta_j \rangle_{S=0}^{\vec{V}} \sim \langle \Delta_i \Delta_j \rangle_{S=0} - \frac{\mu}{2} \sum_p u_p^2 \langle \Delta_i \Delta_j (\partial_x \Delta)_p^2 \rangle_{S=0} + \dots \quad \left\langle \Delta_i \Delta_j \left(\frac{\Delta_{p+dx} - \Delta_{p-dx}}{dx}\right)^2 \right\rangle_{S=0} \Rightarrow \langle \Delta_a \Delta_b \Delta_c \Delta_d \rangle_{S=0} = G_{ab} G_{cd} + G_{ac} G_{bd} + G_{ad} G_{bc}$$

Modelling Flow-Dependent Co-variances with Gaussian Integrals

$$G_{i,j} = G(r_{ij}) = \exp(-(r_{ij} / 10)^2)$$

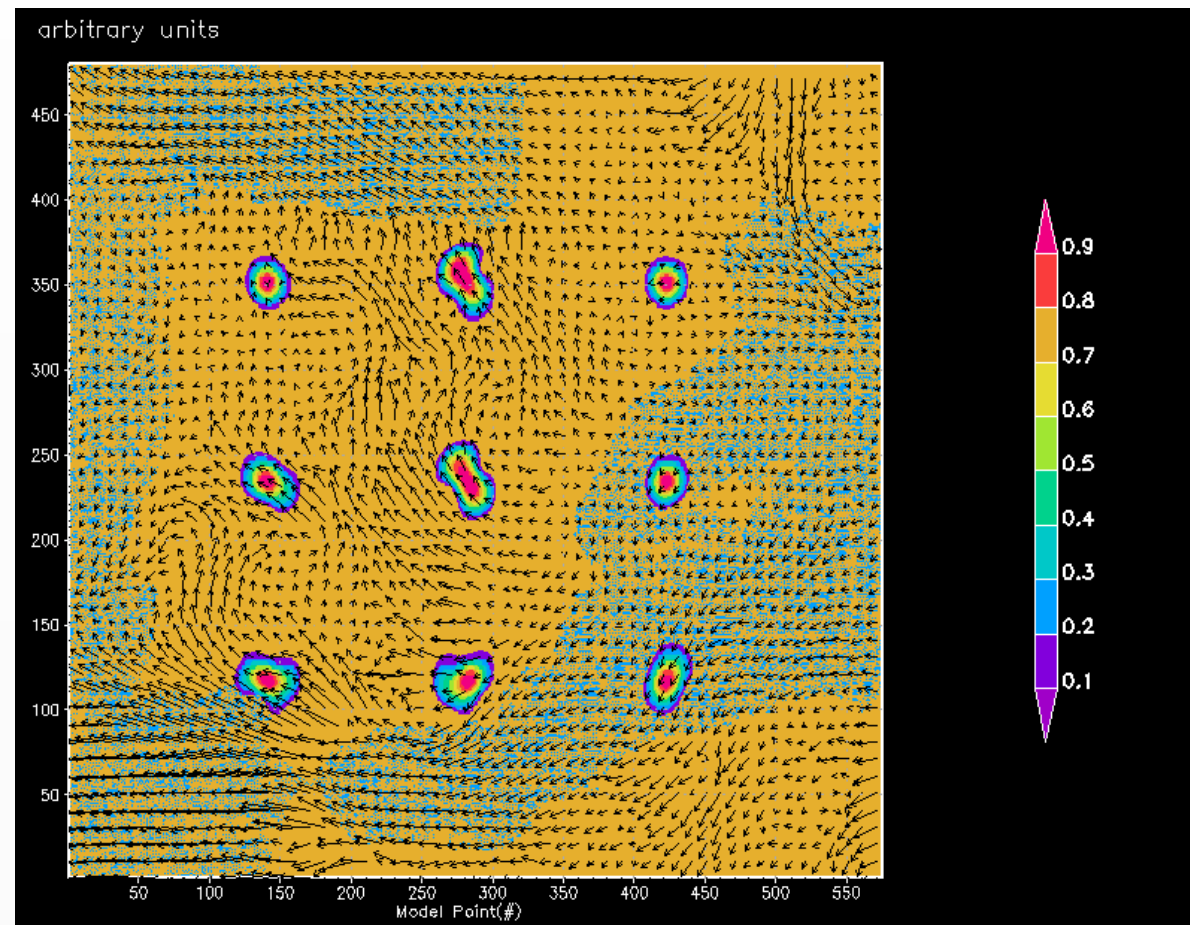


Modelling Flow-Dependent covariances with Gaussian Integrals

Correlations at model level 55
Modulated by the wind field

2D-test with 1 MPI-task and 32 OMP threads “-n1 -j1 -d32”

100 MPI-tasks -> 6000 obs/minute



Modelling Flow-Dependent covariances with Gaussian Integrals

Correlations at the surface
Modulated by orography

MUCHAS GRACIAS POR SU
ATENCIÓN !

