

Investigation of humidity variances and correlations towards the assimilation of WV radiances from MSG-SEVIRI and MTG-IRS into the DWD LETKF/VarEnKF



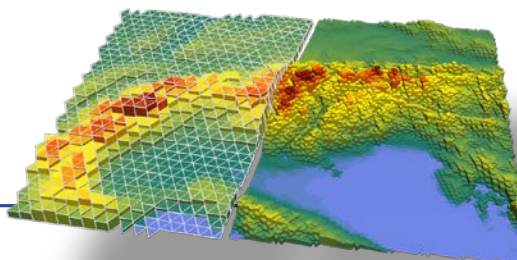
Dora Föhring, Ch. Köpken-Watts, A. Rhodin, R. Potthast

Overview

- Motivation
- The Local Ensemble Transform Kalman Filter (LETKF)
- The Ensemble Variational Data Assimilation (EnVar)
- The New Global ICON Model
- WV Assimilation
- WV Correlations & Comparison with 3dVar
- Outlook

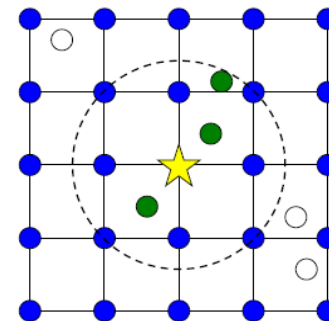
Motivation

- Infrared geostationary radiances provide information about humidity structures.
- Humidity fields have previously been shown to give information about the wind field in 4D-Var (Peubey and McNally 2009, Lupu and McNally, 2012).
- The ICON model at DWD has been made operational as of January 20 2015.
- The current 3dVar data assimilation system will be replaced by a EnVar/LETKF system by the end of the year. (13 km deterministic, 40 km ensemble with 40 members)
- The aim is to adapt the EnVar system to read and process water vapour CSR.
- Evaluation of cross-correlations of humidity with wind and temperature give an insight into how far the EnVar can derive meaningful wind increments from humidity information.



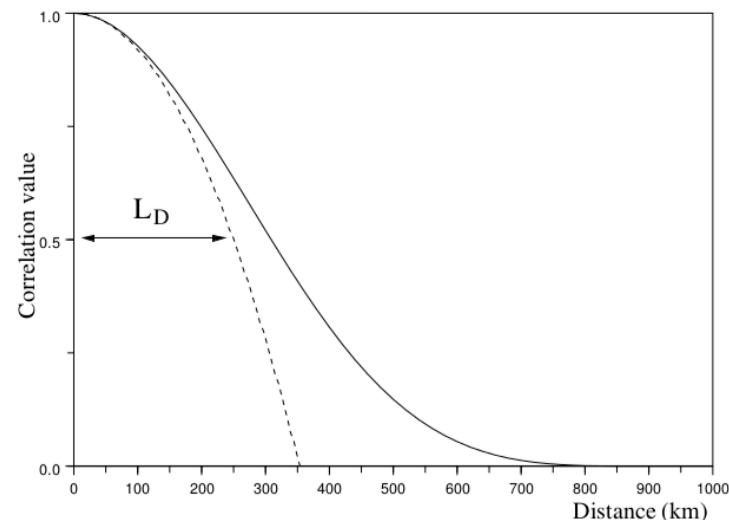
LETKF

- LETKF - Localized Ensemble Transform Kalman Filter
- Implementation following Hunt et al. 2007
- Independent analysis is performed at each grid point → essentially perfectly parallel
- Multiple observations are treated simultaneously.
- Analysis done in the space of the ensemble perturbations.
- Analysis is given by a linear combination of forecast ensemble.
- Only observations within certain vicinity (localization radius) are taken into account.



Localization

- Within the localization radius observations (i.e. their inverse observational errors) are smoothly weighted by the (Gaussian like) Gaspari & Cohn function.
- Localization is required to suppress noise of the sample covariance matrix due to the limited ensemble size.
- Pseudo-Gaussian that goes to zero at some $L_D = c\sqrt{0.3}$ off length
- Characterized by the Daley length-scale



The Kalman Filter and the Local Ensemble Transform

Kalman Filter

Kalman Filter

Model Forecast

$$\mathbf{x}_i^f = M(\mathbf{x}_{i-1}^a)$$

Forecast error:

$$\mathbf{P}_i^f = \mathbf{M}_{\mathbf{x}_{i-1}^a} \mathbf{P}_{i-1}^a \mathbf{M}_{\mathbf{x}_{i-1}^a}^T + \mathbf{Q}$$

Observations

$$\mathbf{y}_i^o = \underset{\substack{\uparrow \\ \text{Observation} \\ \text{operator}}}{H}(\mathbf{x}) + \underset{\substack{\uparrow \\ \text{Observational} \\ \text{error}}}{\varepsilon_i^o}$$



Kalman gain

$$\mathbf{K}_i = \mathbf{P}_i^f \mathbf{H}^T [\mathbf{H} \mathbf{P}_i^f \mathbf{H}^T + \mathbf{R}]^{-1}$$

Analysis:

$$\mathbf{x}_i^a = \mathbf{x}_i^f + \mathbf{K}_i (\mathbf{y}_i^o - H(\mathbf{x}_i^f))$$

$$\mathbf{P}_i^a = [\mathbf{I} - \mathbf{K}_i \mathbf{H}] \mathbf{P}_i^f$$

$$\mathbf{x} : [N] \quad \mathbf{P} : [N \times N]$$

Ensemble Kalman Filter

Ensemble Forecast

$$\mathbf{X}_i^f = [M(\mathbf{x}_{i-1}^{a(1)}) \dots M(\mathbf{x}_{i-1}^{a(m)})]$$

$$\equiv M(\mathbf{x}_{i-1}^a)$$

Number of ensemble members

Forecast error:

$$\delta \mathbf{X}_i^f = [\mathbf{x}_i^{a(1)} - \bar{\mathbf{x}}_i, \dots, \mathbf{x}_i^{a(m)} - \bar{\mathbf{x}}_i]$$

$$\mathbf{P}_i^f \approx \frac{\delta \mathbf{X}_i^f (\delta \mathbf{X}_i^f)^T}{m-1}$$

$$\mathbf{Y} = H(\mathbf{X}) = [\mathbf{y}_i^{a(1)} \dots \mathbf{y}_i^{a(m)}]$$

$$\delta \mathbf{Y} = [\mathbf{y}_i^{a(1)} - \bar{\mathbf{y}}_i^a, \dots, \mathbf{y}_i^{a(m)} - \bar{\mathbf{y}}_i^a]$$

Flow dependent!

$$\mathbf{x} : [N \times m] \quad \mathbf{P} : [N \times m]$$

LETKF vs 3dVar vs EnVar

- The EnVar is a hybrid variational analysis system which uses contributions from both the time invariant climatological and the flow dependent ensemble background error covariance matrix for a deterministic analysis.

Climatological P^f_{clim} used by 3dVar

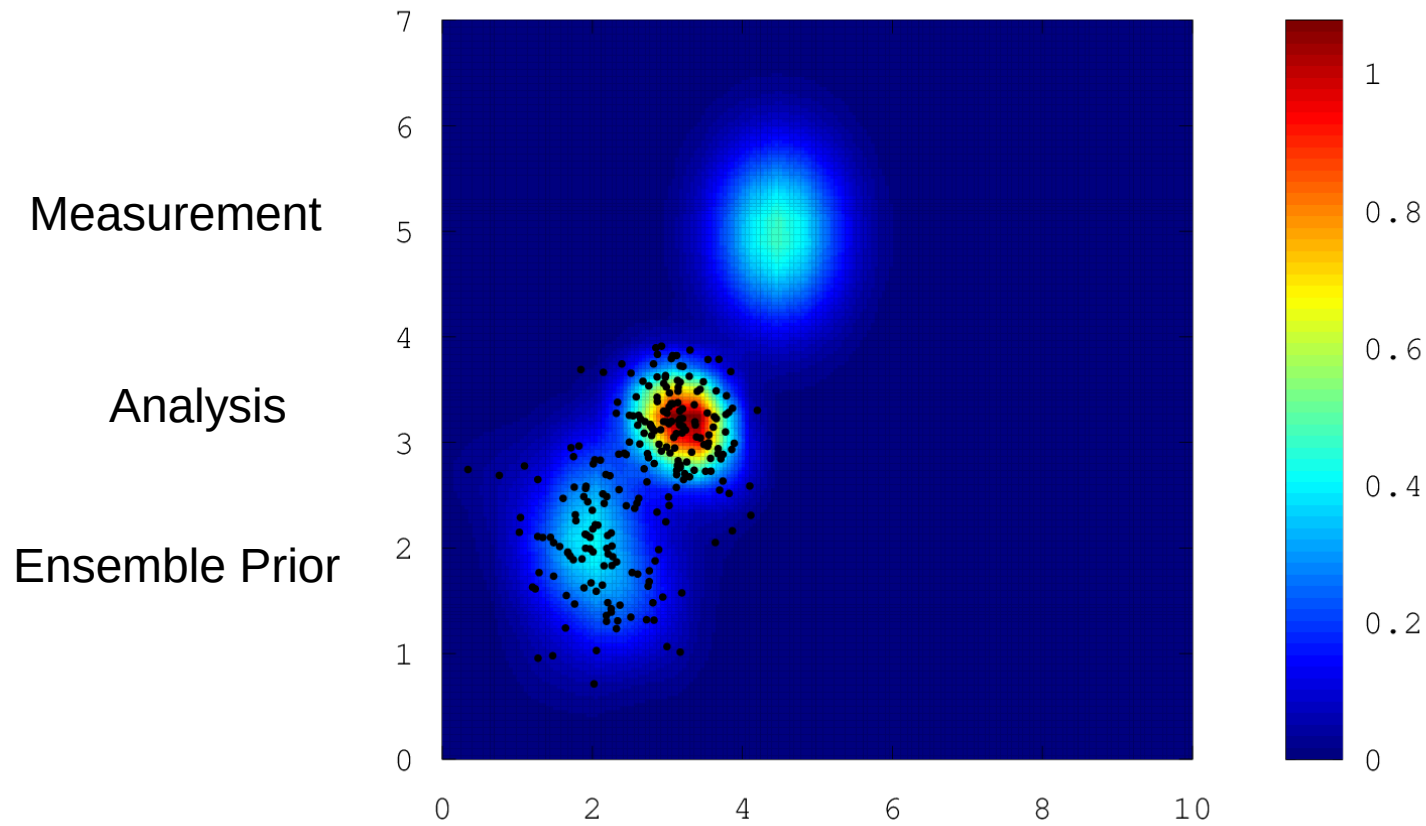
- Derived by the NMC method (differences of forecast with different lead time).
- No flow dependent correlations.
- Cross-correlations between temperature and humidity are assumed to be zero.

Ensemble P^f_{ens} used by LETKF (and EnVar)

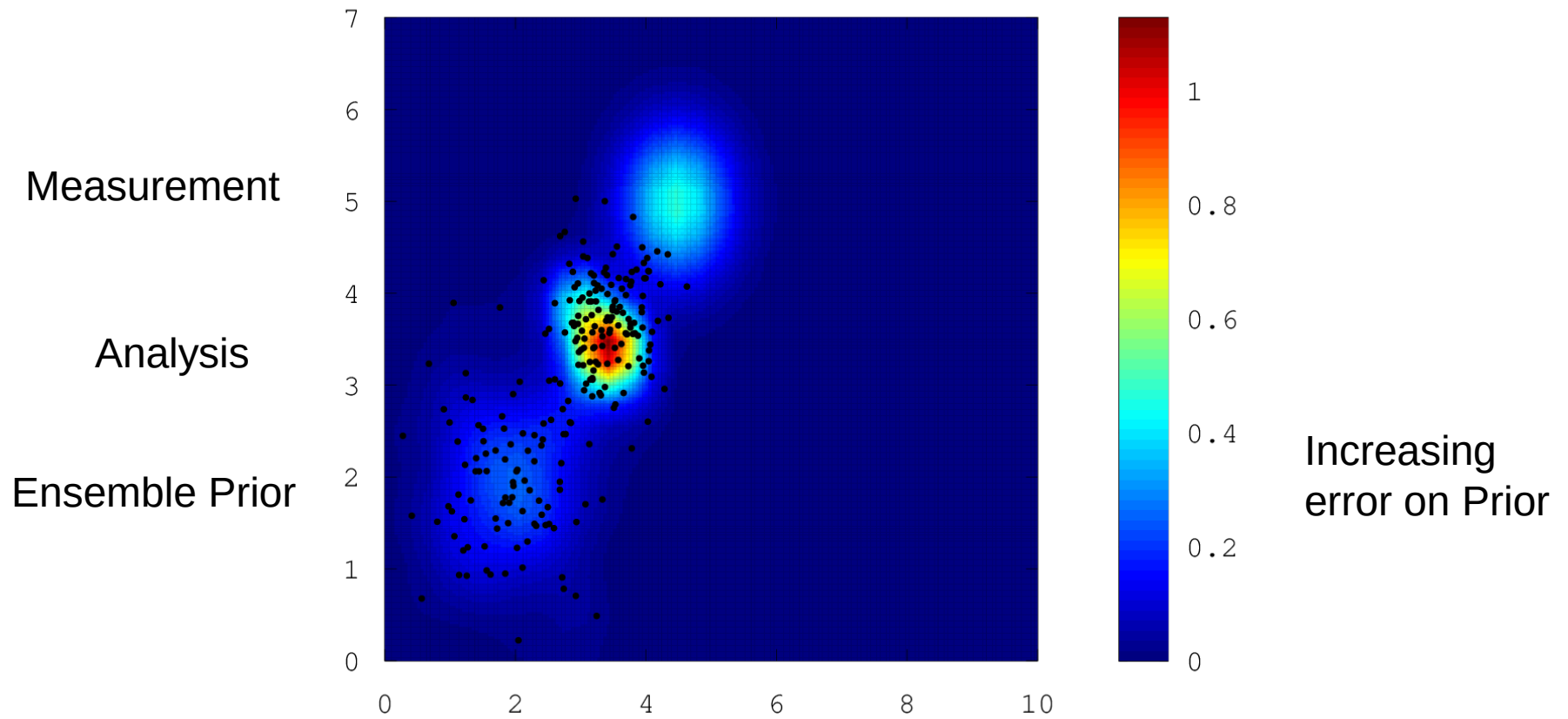
- Derived from differences of ensemble members
- Flow dependent correlations between all model variables

	3dVar	LETKF	EnVar
background error	climatological	flow dependent	combination of both
	P^f_{clim}	P^f_{ens}	$0.5 * (P^f_{\text{clim}} + P^f_{\text{ens}})$
analysis	deterministic	ensemble	deterministic

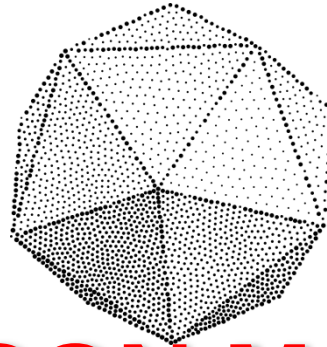
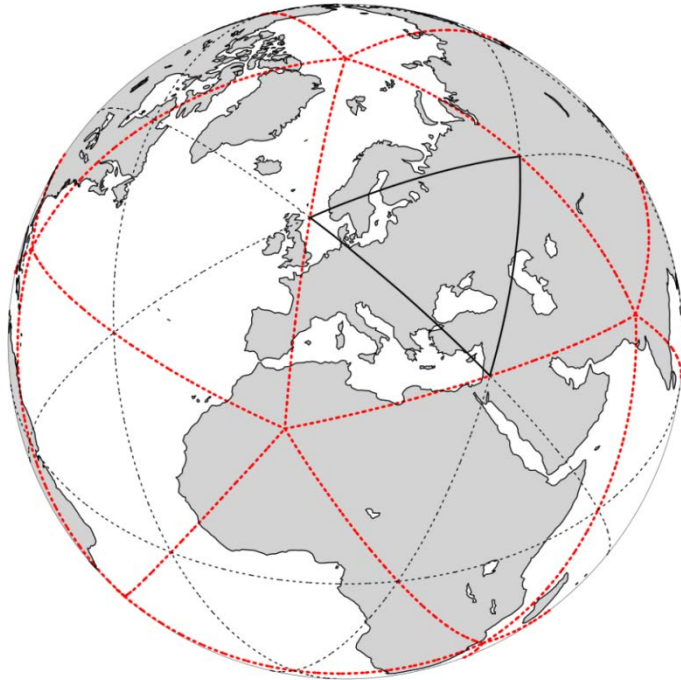
LETKF: demonstration



LETKF: demonstration



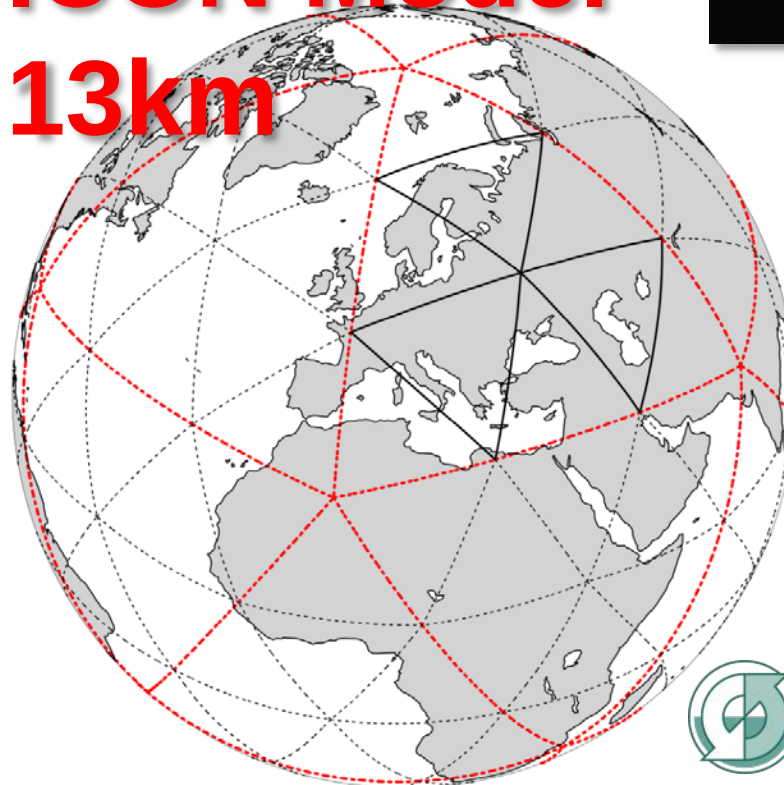
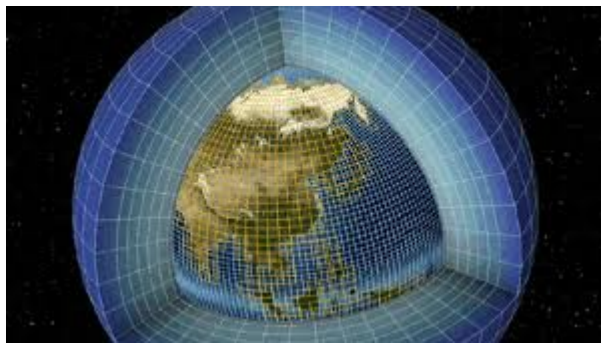
Global Modelling



ICON Model
13km



ICOsahedral
Nonhydrostatic



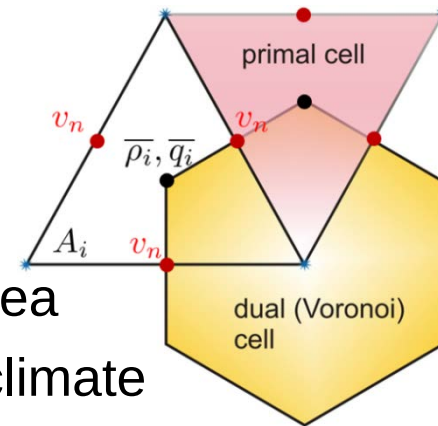
New Global Model **ICON**

- **Operational since Jan 20, 2015**
- Non-Hydrostatic
- 13km resolution globally
- 75km height
- 90 vertical levels
- Still a lot in the pipeline!!



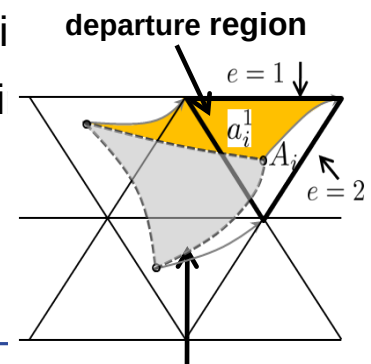
ICON: NWP and climate model

- Icosahedral grid on triangles (also hexagons)
- Non-hydrostatic (Gassmann, Herzog, 2008)
- Z-coordinate (terrain following)
- Two-way nesting (horiz. & vert.) and limited area
- Target resolution: 5-20km NWP, 100-200km climate
- LES option considered (100m-400m)

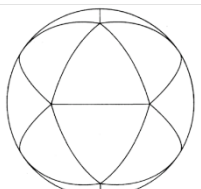
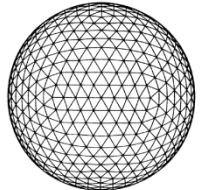
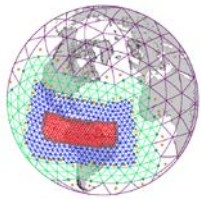


Numerics:

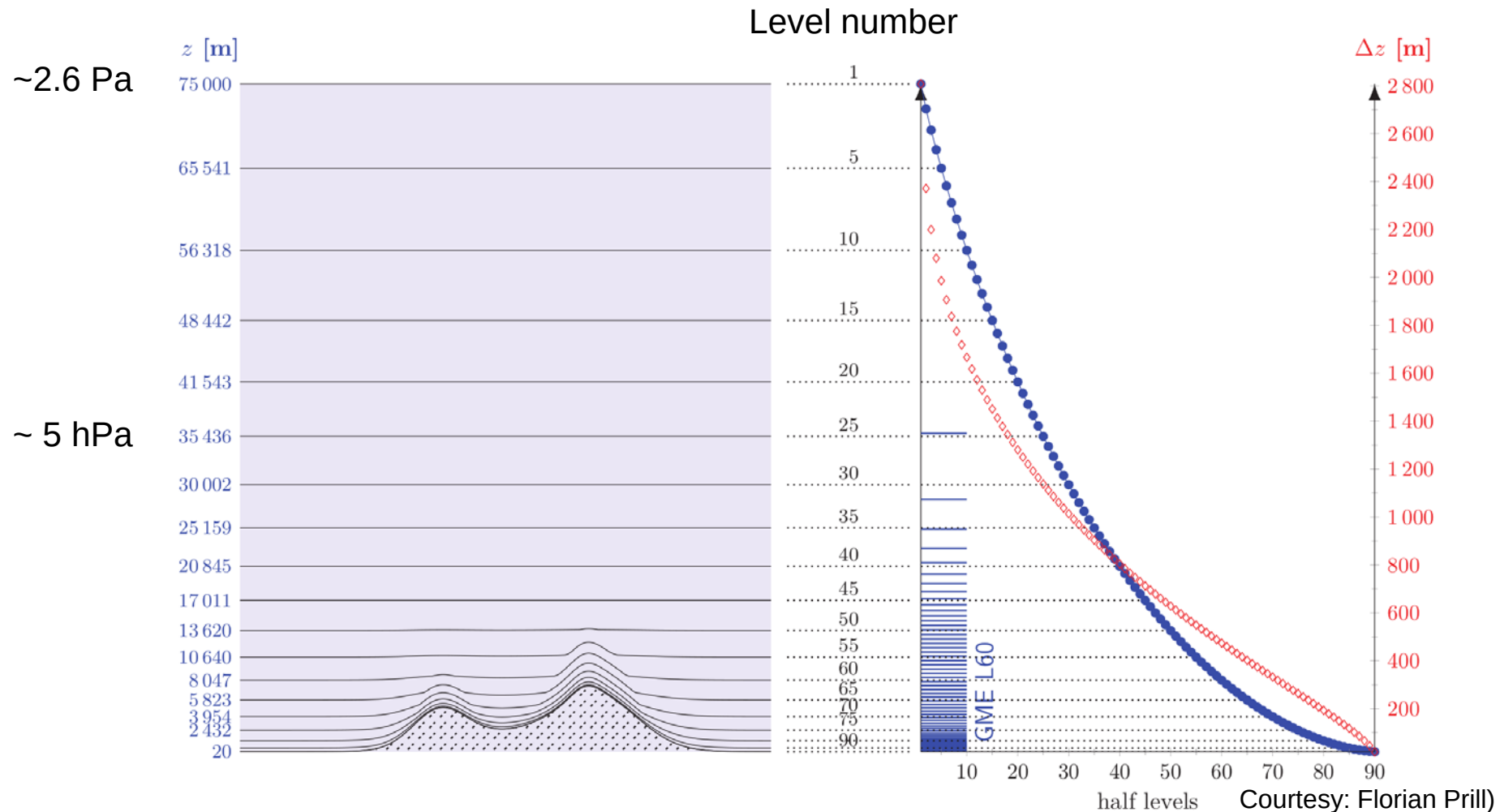
- C-type staggering
- Horizontally explicit, vertically implicit on sound wave (HEVI)
- Mahrer-type pressure discretization (Zängl 2012, MWR, i)
- Flux-form semi-Lagrangian (FFSL) advection based on fi (Lauritzen et al 2011, Lauritzen 2010, Miura 2007)
- Mass-conserving, consistent, optionally monotone



Lagrangian control volume



ICON : 90 Level version



Considerably higher model top beneficial for use of radiances !



Summary of ICON Atmosphere Analysis Fields

- VN – Velocity normal to edges
- U – Zonal wind
- V – Meridional wind
- W – Vertical wind
- DEN – Density of moist air
- THETA V
 - Virtual potential temperature
- T – Temperature
- QV – Specific humidity
- QC – Cloud mixing ratio
- QI – Cloud ice mixing ratio
- QR – Rain mixing ratio
- QS – Snow mixing ratio
- TKE – Turbulent kinetic energy
- P – Pressure

Assimilated data:

Operational & tested for implementation

(I) All standard conventional data:

- RS, PILOT, AIREP, SYNOP, BUOY

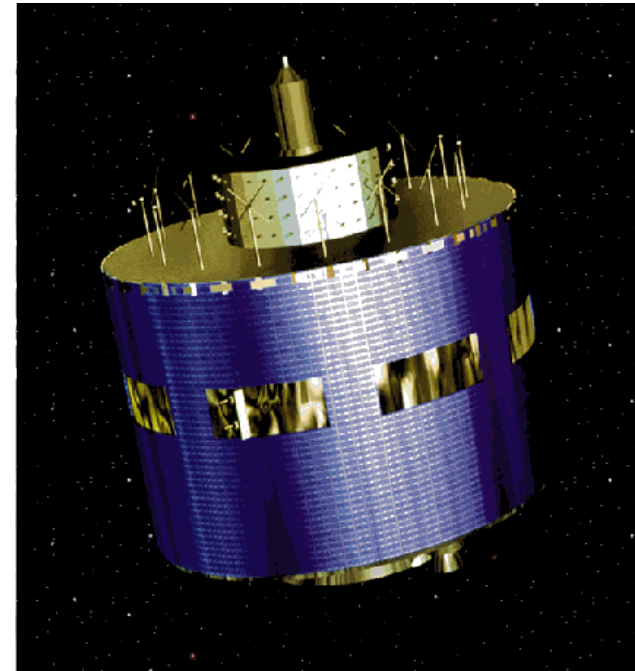
(II) Satellite data:

- Radiances/IR : HIRS, IASI (METOP-A/B)
CrIS
- Radiances/MW : AMSU-A (NOAA, METOP-A/B)
ATMS
MHS, ATMS, (FY-MWHS)
AMSR-2, SSMIS
- RO (bending angles): GRAS (METOP-A/B)
COSMIC, GRACE, TerraSar-X, C/NOFS
- AMVs - geo: GOES, METEOSAT, MTSAT (sea/land)
- polar: Aqua/Terra, NOAA, METOP-A/B
- Scat (wind vectors): ASCAT (METOP-A/B), Rapidscat

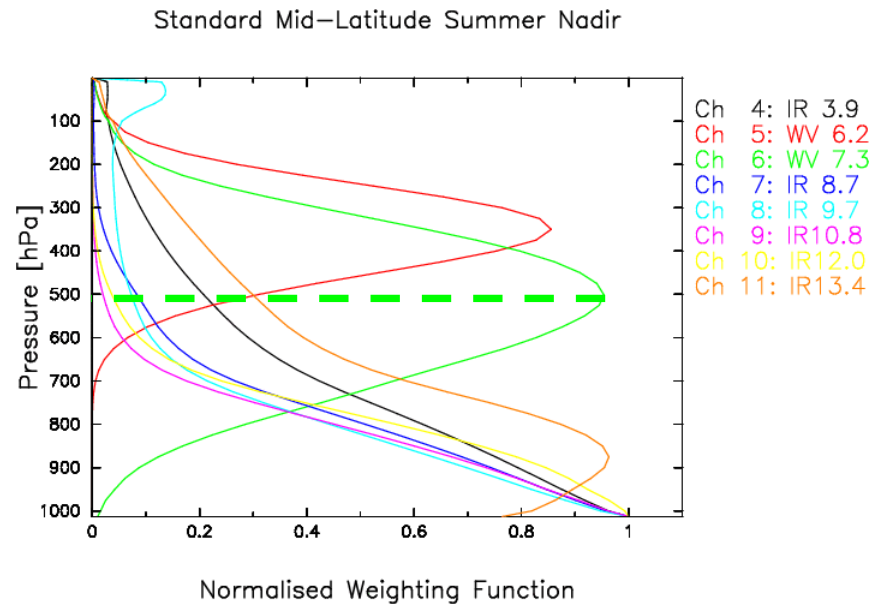
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Water Vapour Channels – MSG SEVIRI

- Channel 01: VIS 0.6 μ
- Channel 02: VIS 0.8 μ
- Channel 03: NIR 1.6 μ
- Channel 04: MIR 3.9 μ
- Channel 05: WV 6.2 μ
- Channel 06: WV 7.3 μ
- Channel 07: IR 8.7 μ
- Channel 08: IR 9.7 μ (Ozone)
- Channel 09: IR 10.8 μ
- Channel 10: IR 12.0 μ
- Channel 11: IR 13.4 μ (CO₂)
- Channel 12: HRV (High Resolution Visible)

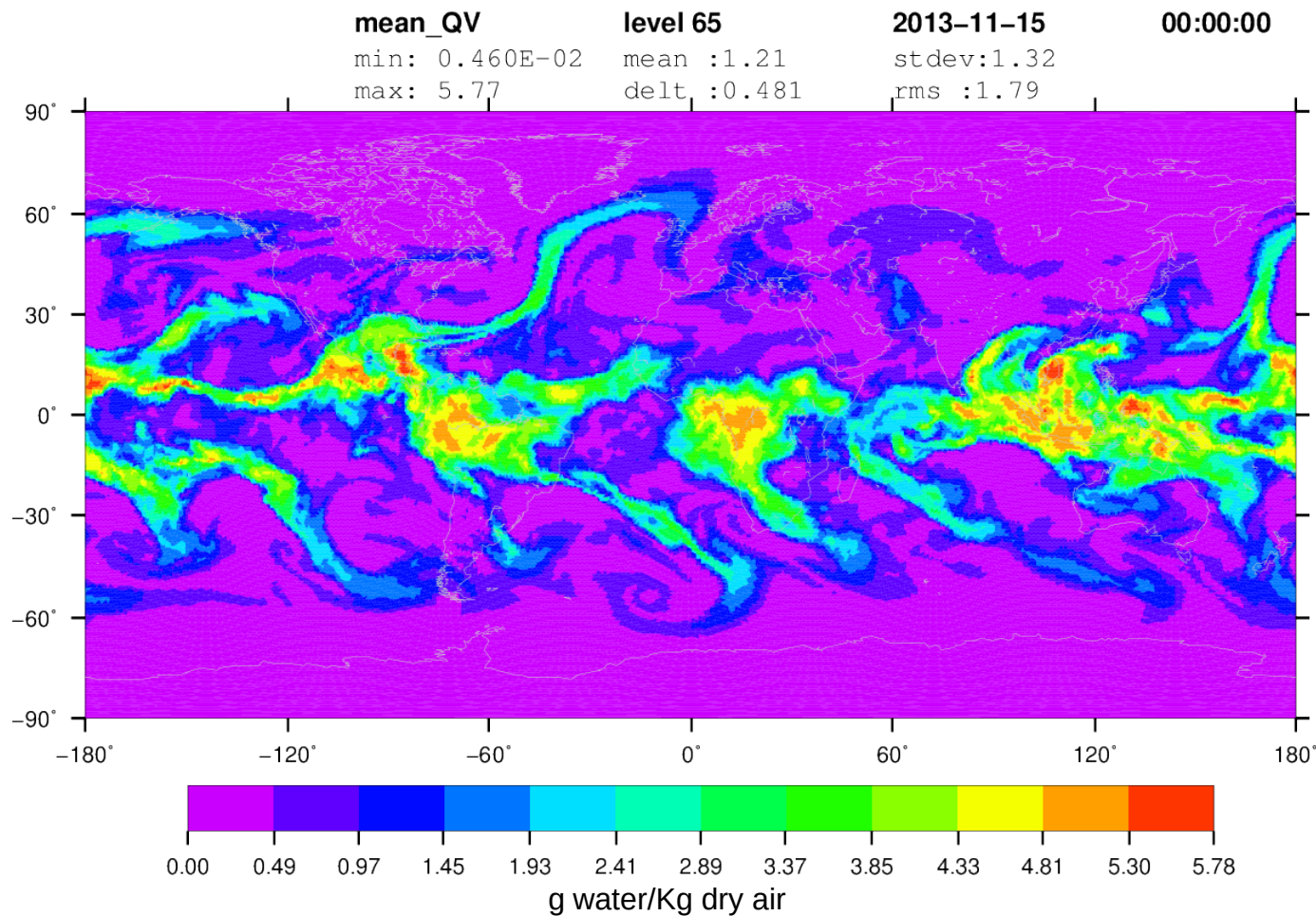


- Max signal in Ch 06 from approx. 500 hPa:
→ Corresponds to level 65 in the ICON model

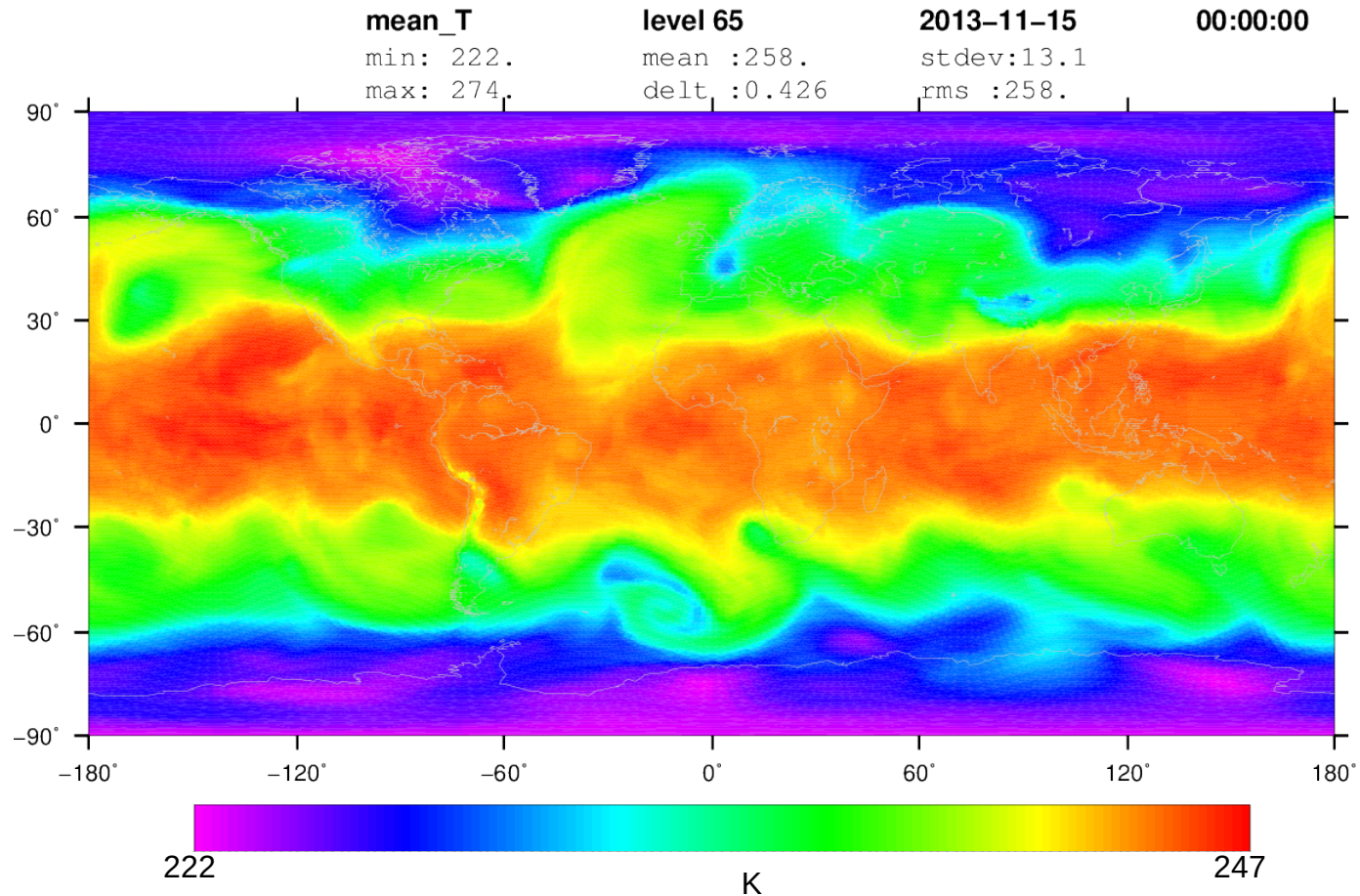


Weighting functions for the MSG SEVIRI thermal channels

Overview: Mean QV

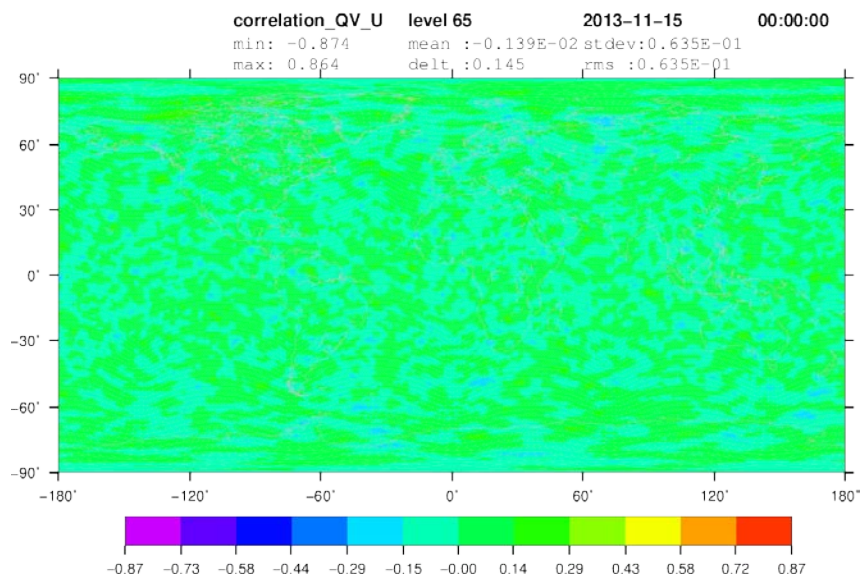


Overview: Mean T

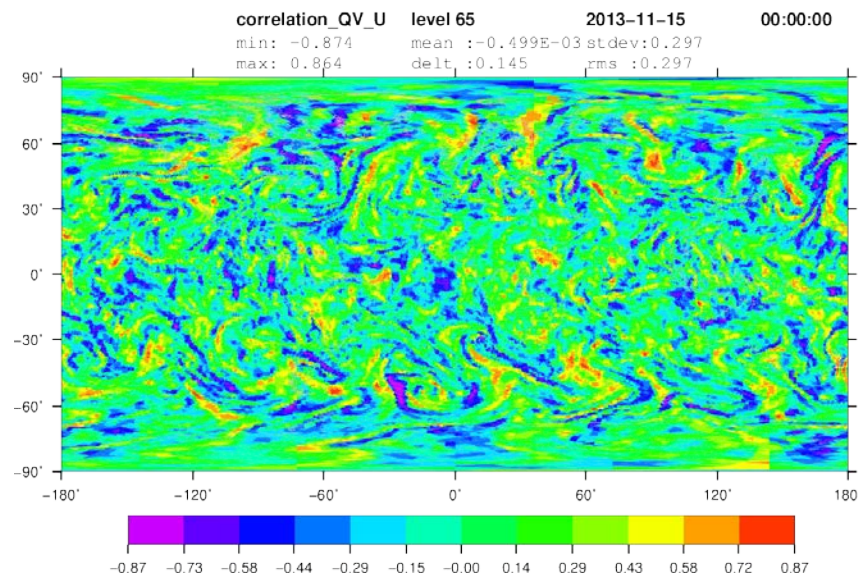


Correlations: Humidity - Wind

3dVar

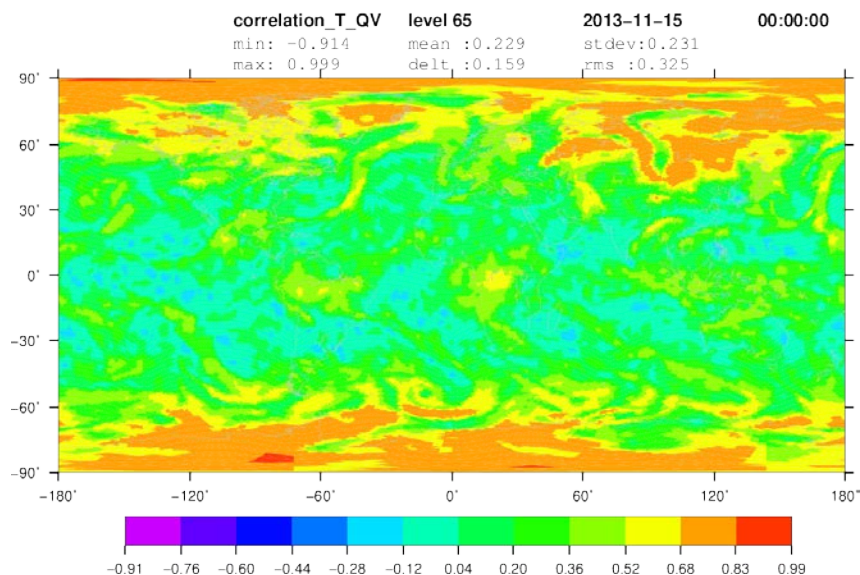


LETKF

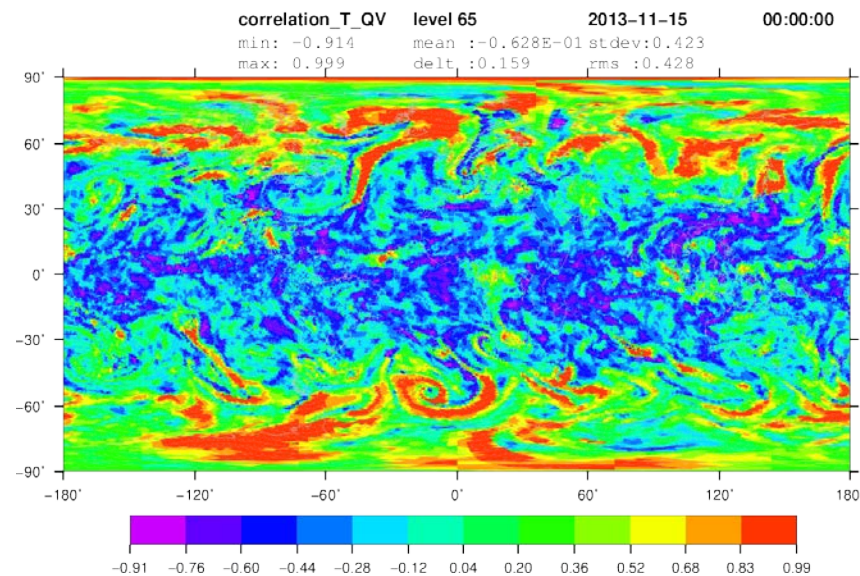


Correlations: Humidity-Temperature

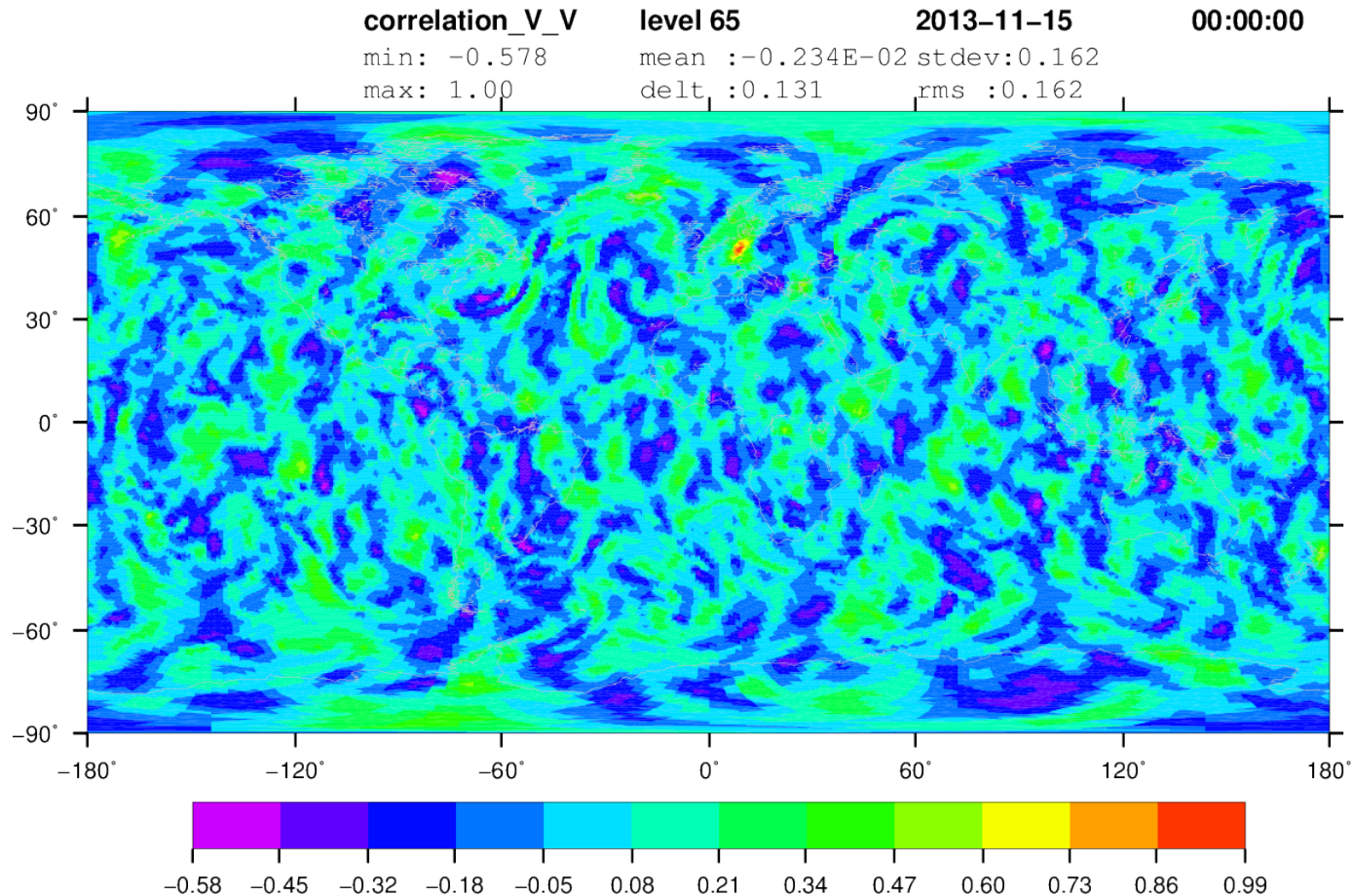
3dVar



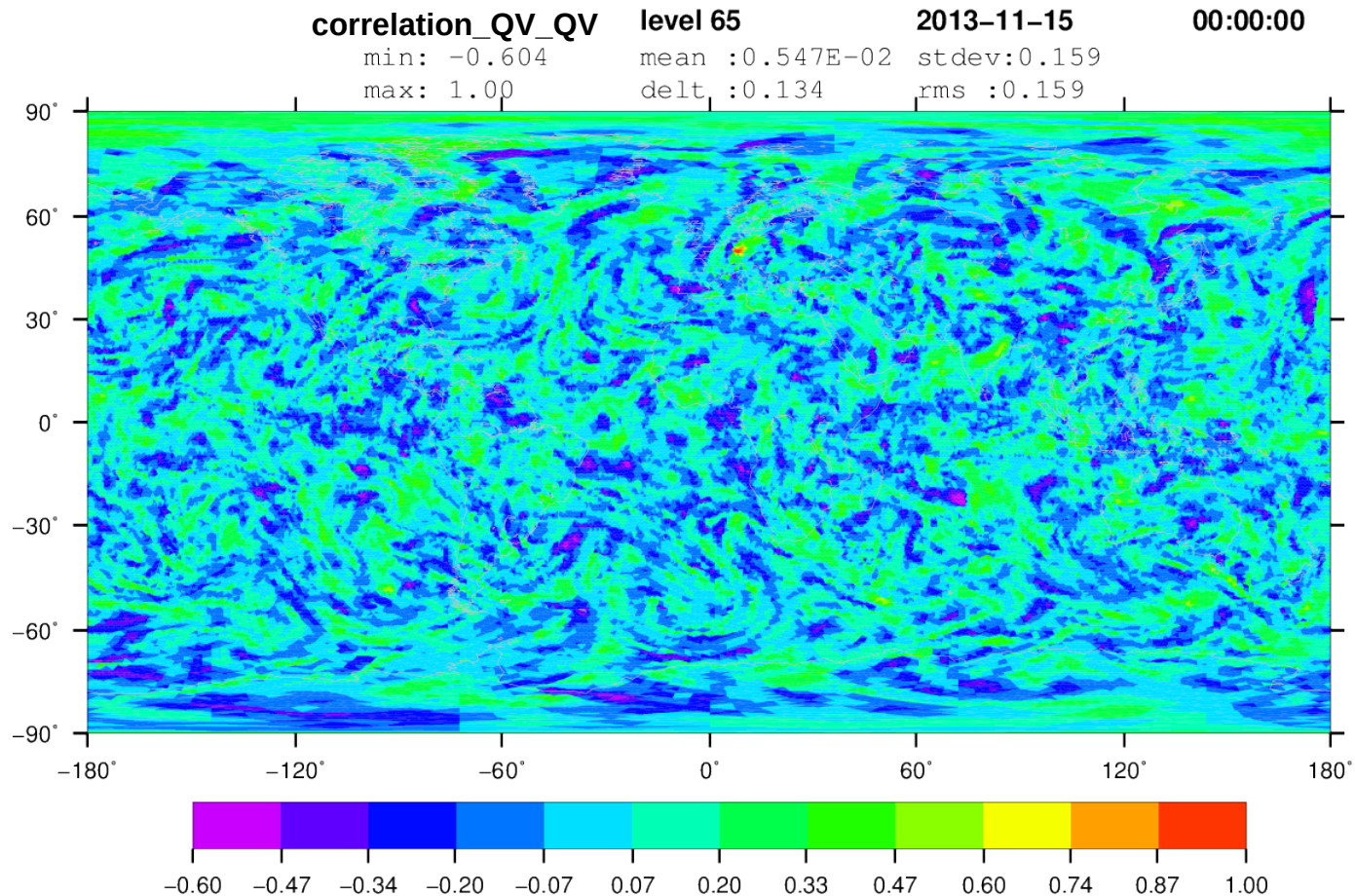
LETKF



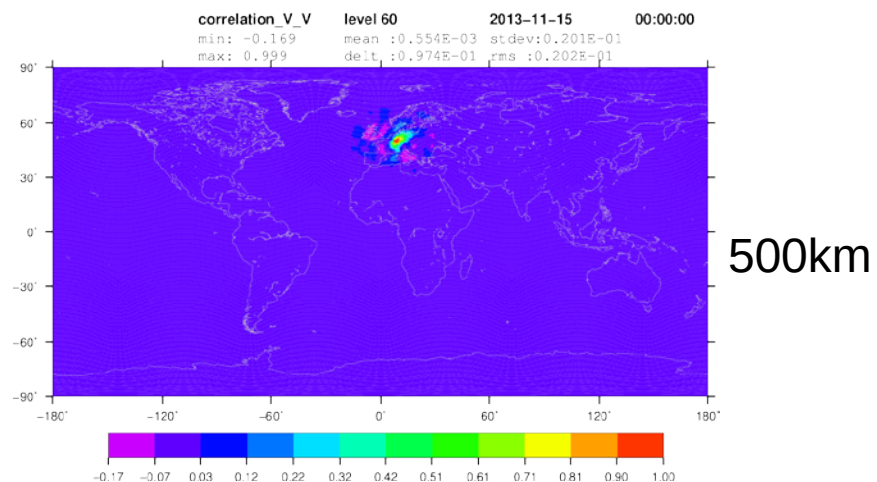
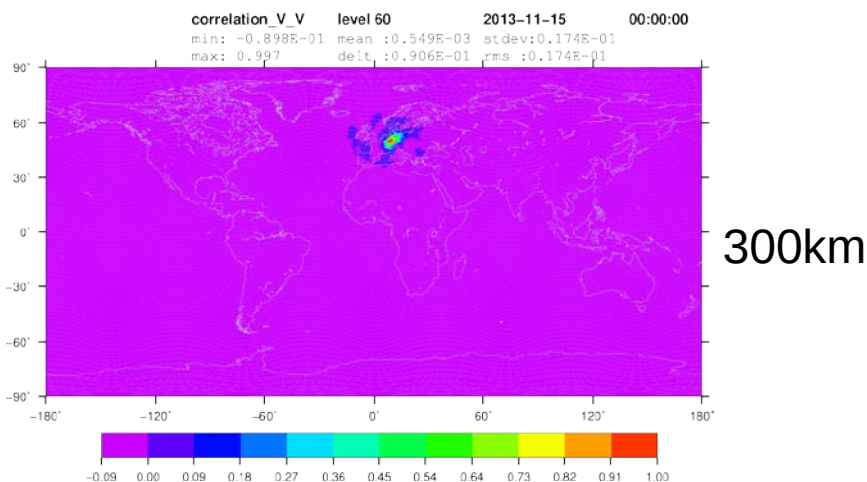
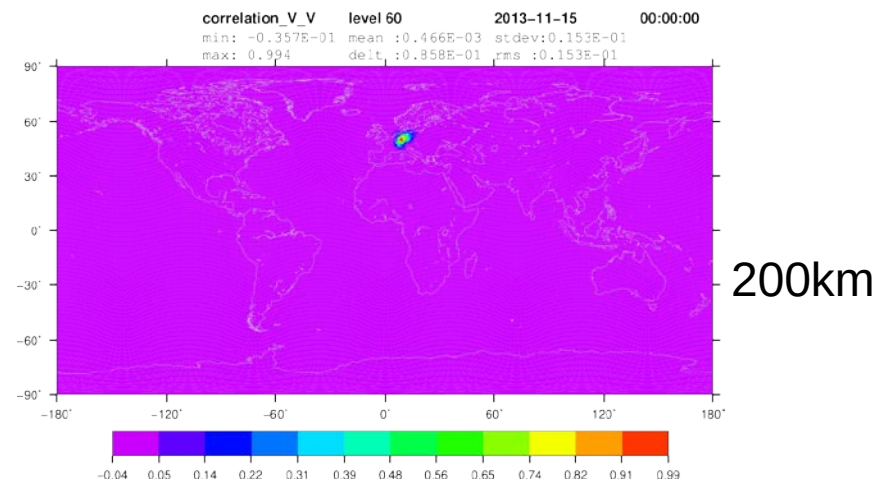
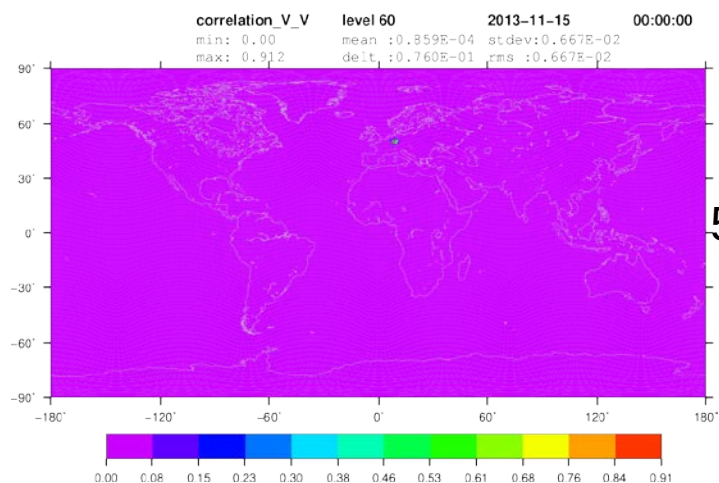
Autocorrelations: V



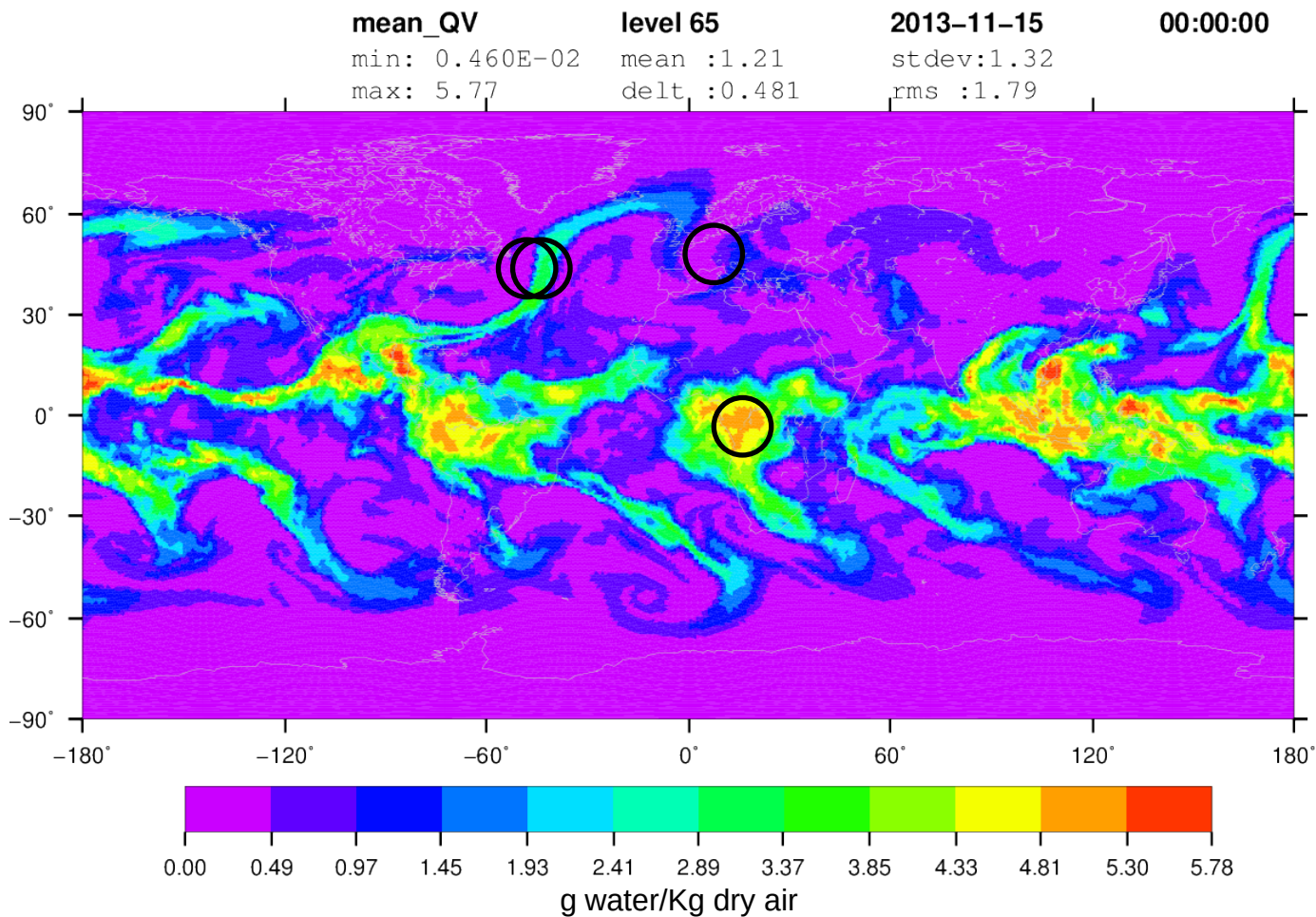
QV



Localization



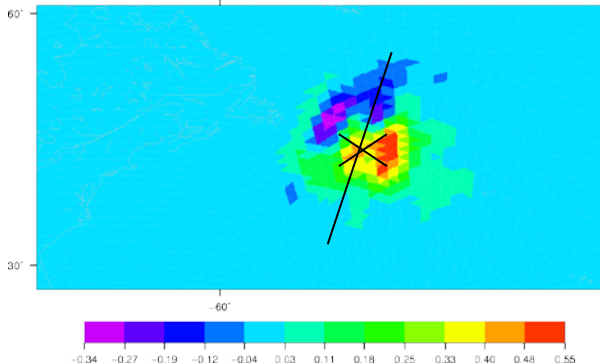
Overview: Mean QV



Correlations Across Fronts

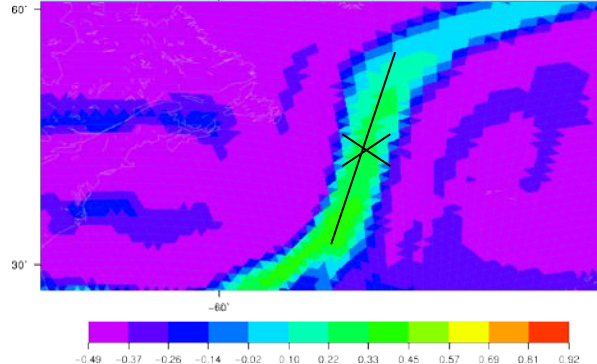
T-U

correlation_T_U level 65 2013-11-15 00:00:00
min: -0.342 mean: 0.417E-03 stdev: 0.135E-01
max: 0.554 delt: 0.756E-01 zms: 0.135E-01



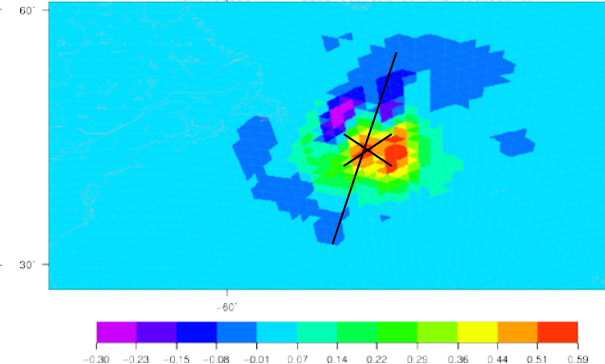
T-QV

correlation_T_QV level 65 2013-11-15 00:00:00
min: -0.492 mean: 0.556E-03 stdev: 0.210E-01
max: 0.924 delt: 0.518 zms: 0.210E-01

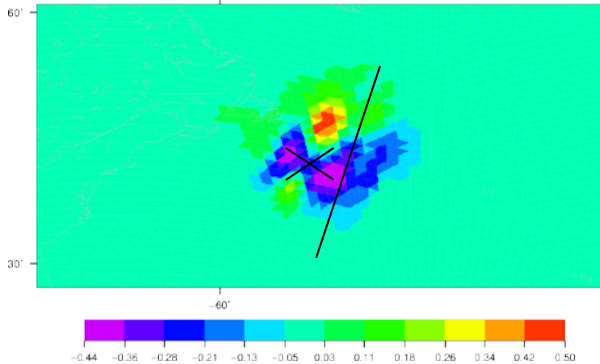


QV-U

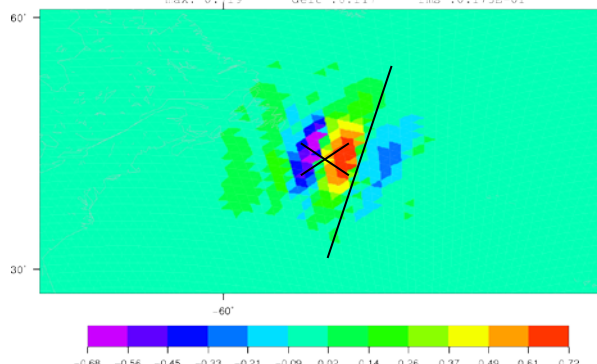
correlation_QV_U level 65 2013-11-15 00:00:00
min: -0.301 mean: 0.603E-03 stdev: 0.152E-01
max: 0.458 delt: 0.739E-01 zms: 0.152E-01



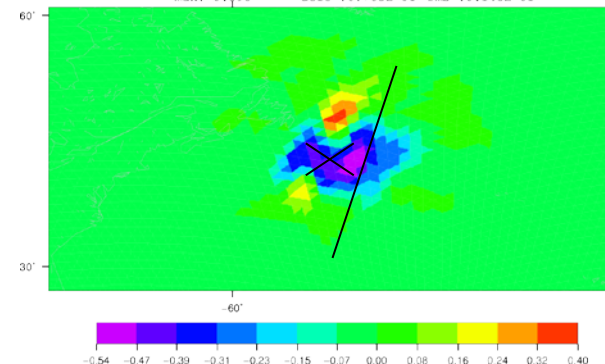
correlation_T_U level 65 2013-11-15 00:00:00
min: -0.440 mean: -0.309E-03 stdev: 0.128E-01
max: 0.196 delt: 0.780E-01 zms: 0.128E-01



correlation_T_QV level 65 2013-11-15 00:00:00
min: -0.679 mean: 0.824E-04 stdev: 0.113E-01
max: 0.719 delt: 0.117 zms: 0.113E-01

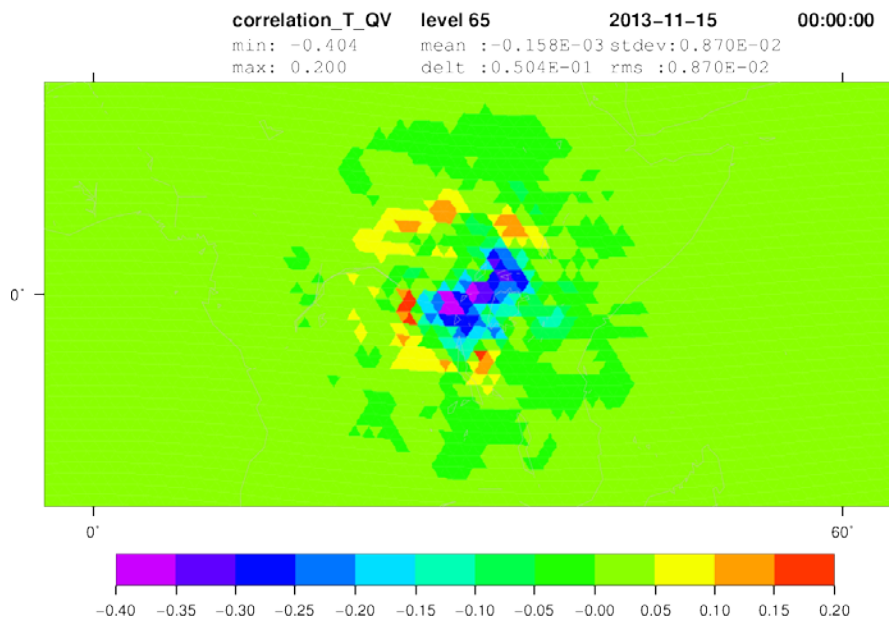


correlation_QV_U level 65 2013-11-15 00:00:00
min: -0.544 mean: 0.466E-03 stdev: 0.143E-01
max: 0.395 delt: 0.783E-01 zms: 0.143E-01

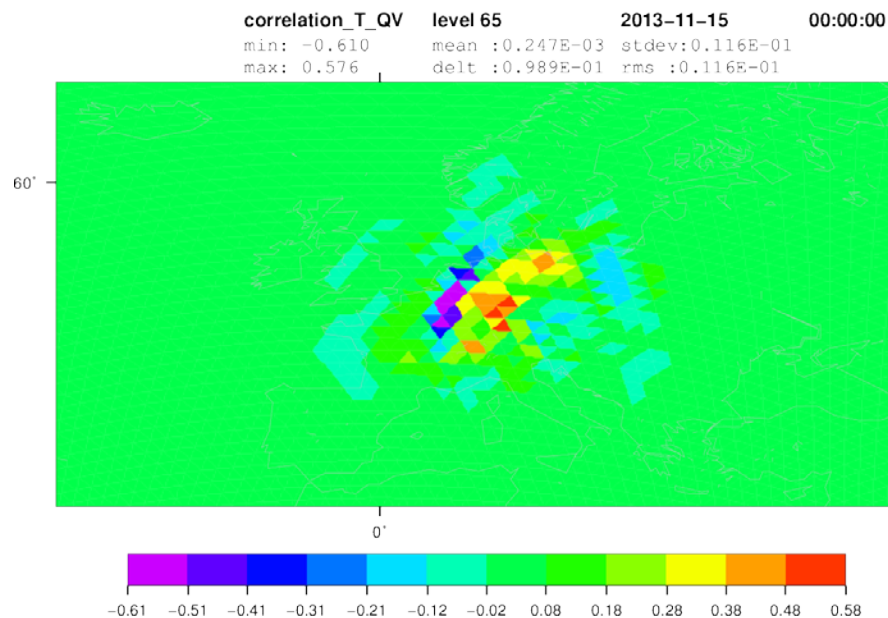


Correlations In Other Regions

Tropics



Extratropics

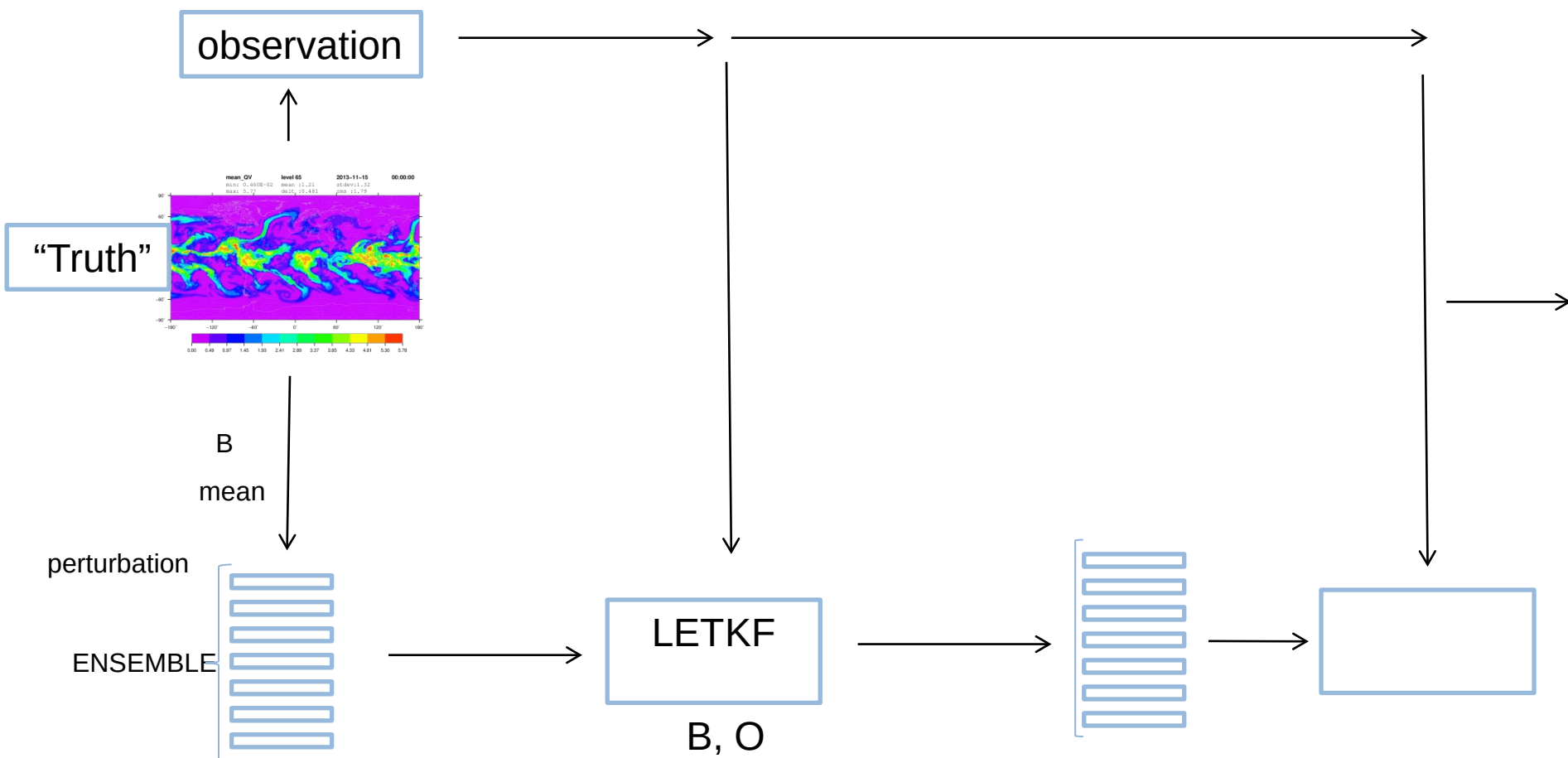


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Outlook

- Passive Monitoring of data:
 - Compare model equivalent of assimilation: $H(x)$
 - Evaluate the fit of the model to the CSR data
- Assimilate artificial data:
 1. Artificial humidity data generated by a ‘truth run’
 2. Synthetic single layer 1 channel CSR computed from a ‘truth run’ for the relevant SEVIRI water vapour channels
 3. Equivalent brightness temperatures
- Assimilate real radiances

Assimilating Artificial Humidity Data



Outlook

- Additional investigation of the effect of the weighted B matrix on the results may be required.
- A statistical investigation of the relationships between the different parameters within a frontal system is also of interest.

Summary

- Humidity fields have previously been shown to give information about the wind field in 4D-Var.
- The aim of the project is to adapt the EnKF system to read and process water vapour CSR.
- The ICON global model has been newly made operational since January 20 2015 for both for operational NWP and climate prediction applications at the DWD.
- In contrast to the climatological B matrix in 3Dvar, there is strong correlation between temperature and relative humidity, and between humidity and wind within ICON, due to the flow dependent B matrix in LETKF.
- Future work will include testing whether the LETKF/EnVar is able to infer meaningful wind increments from the sequence of CSR data.
- Testing the assimilation of artificial humidity data generated by a ‘truth run’.
- Adapting the EnKF system to read and process water vapour CSR to test with real data.

Investigation of humidity variances and correlations towards the assimilation of WV radiances from MSG-SEVIRI and MTG-IRS into the DWD LETKF/VarEnKF



DF would like to thank EUMETSAT for funding!