

Towards the introduction of horizontal correlations of satellite observation errors into the DA system of the model AROME

Thomas Buey

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ITSC 2025 - Goa, India

May 12th, 2025

- **Data assimilation in AROME**

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- **Horizontal correlations of SEVIRI observation errors**

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- **Experimental framework**

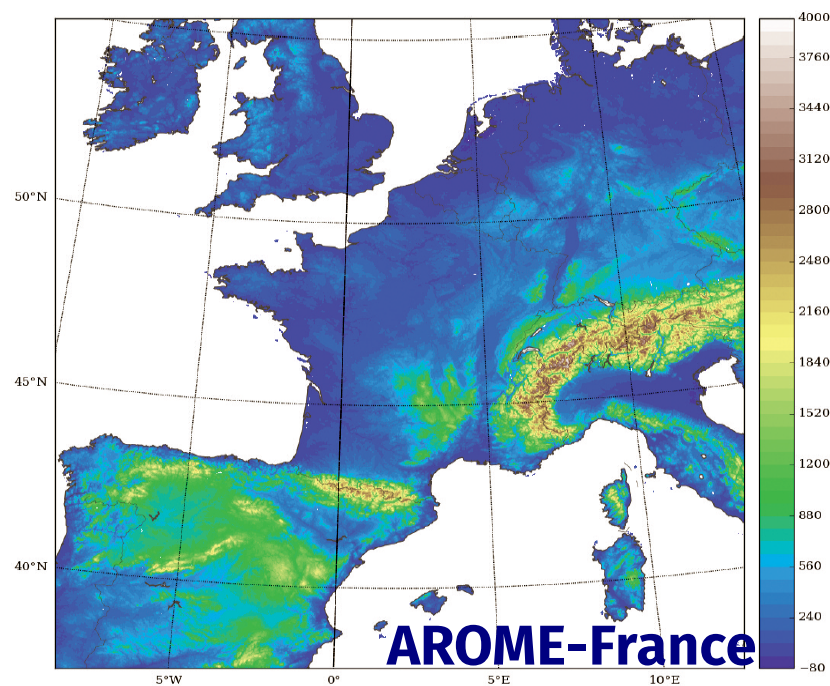
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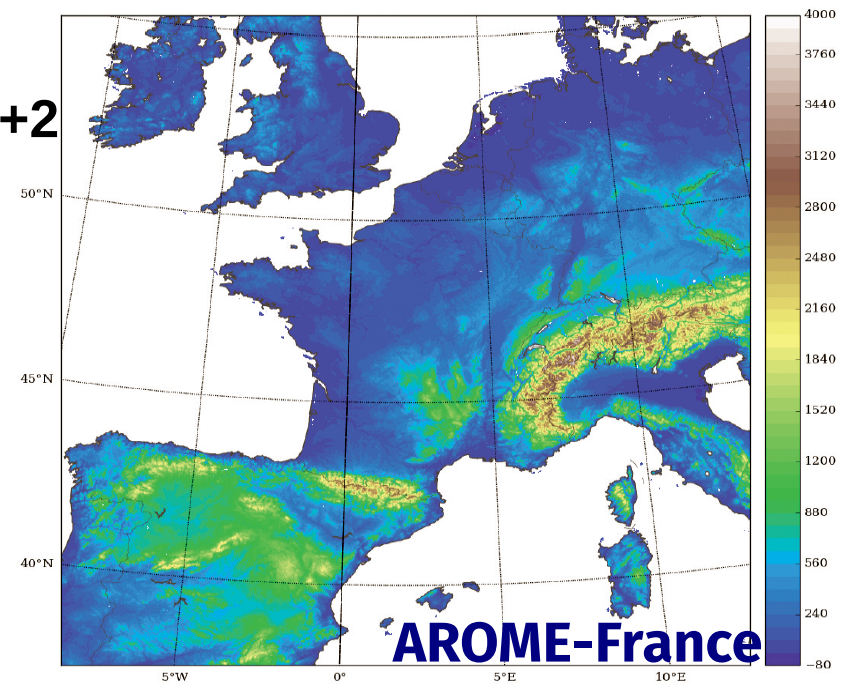
Limited-area model AROME

- Météo-France high-resolution model



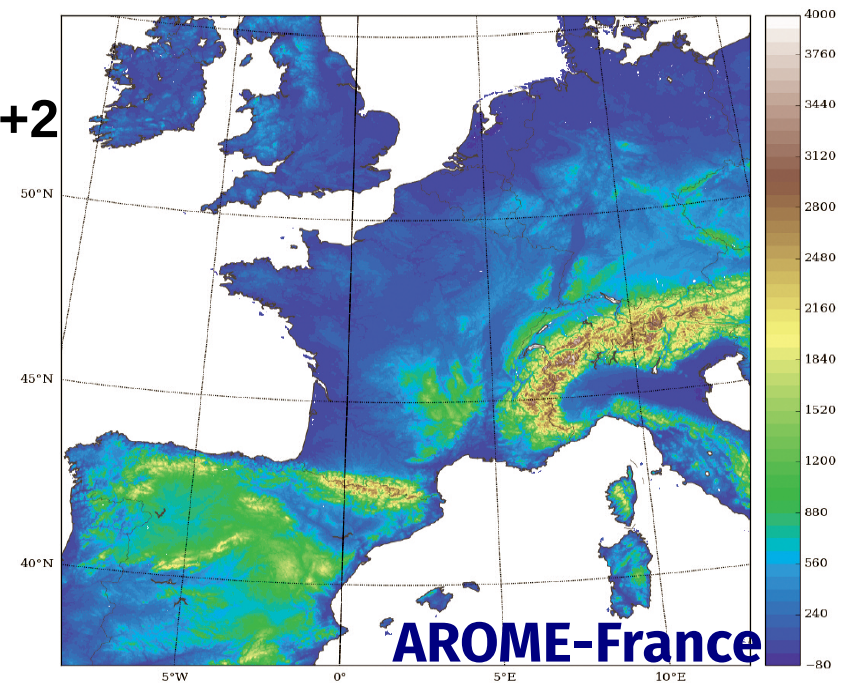
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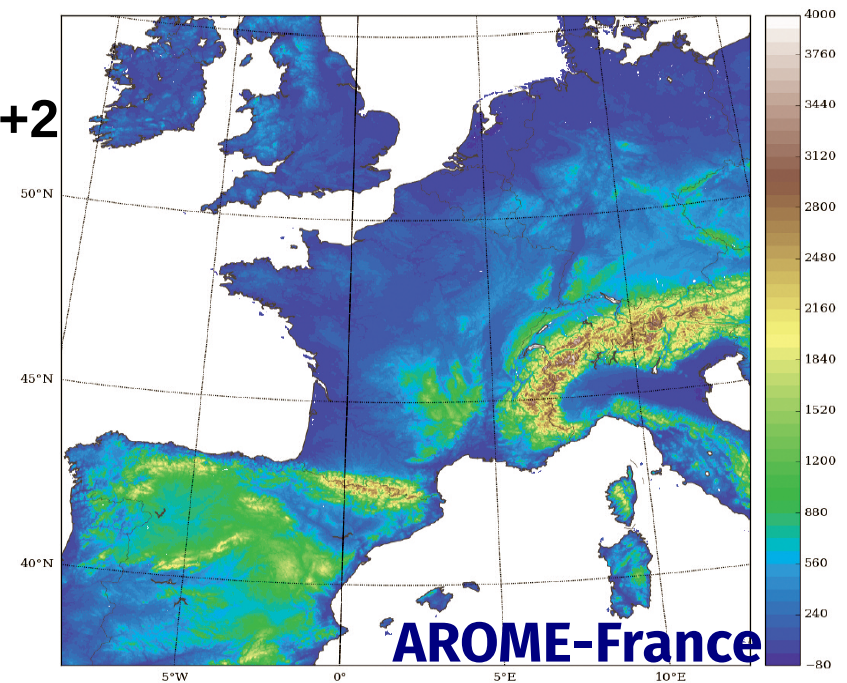
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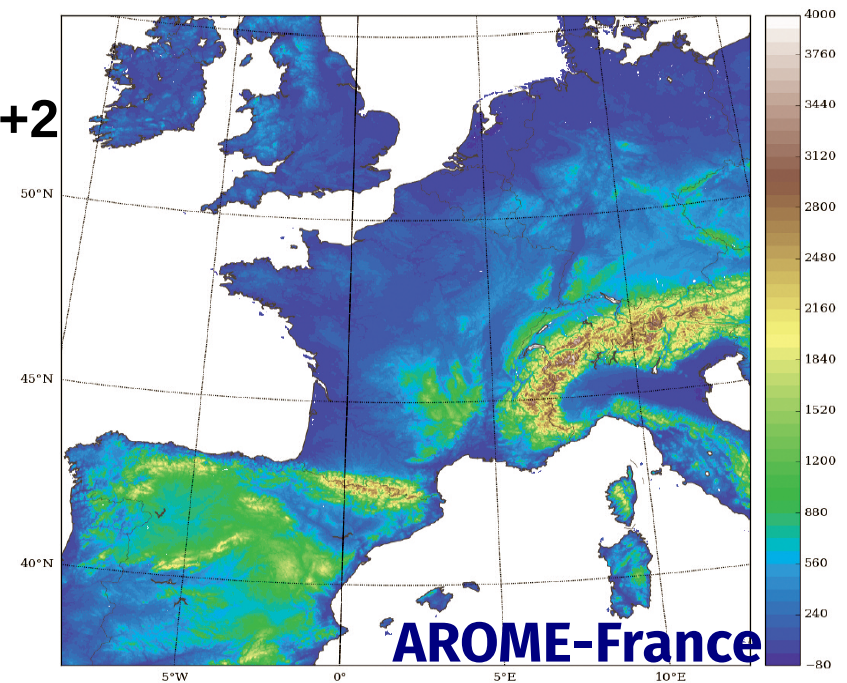
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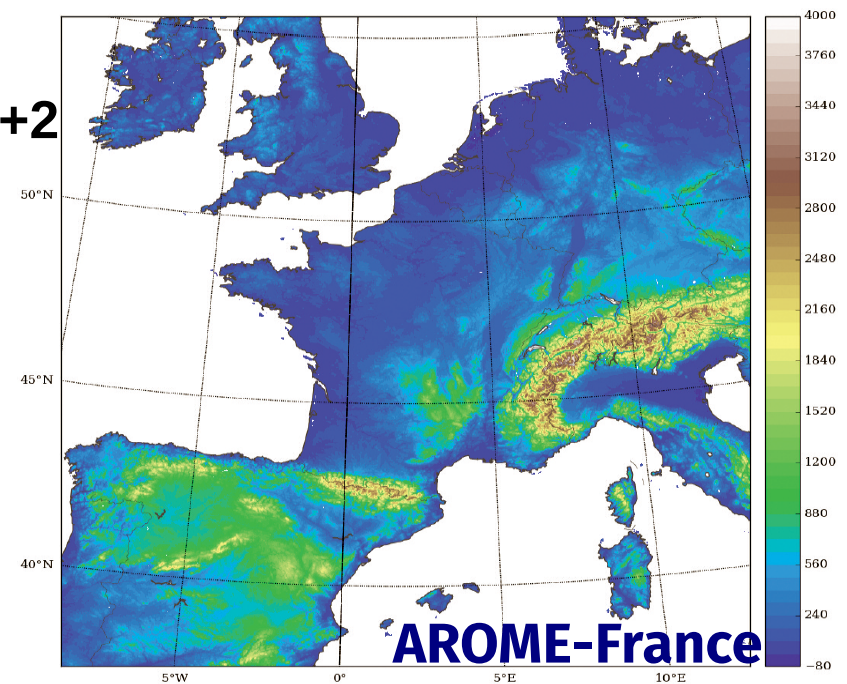
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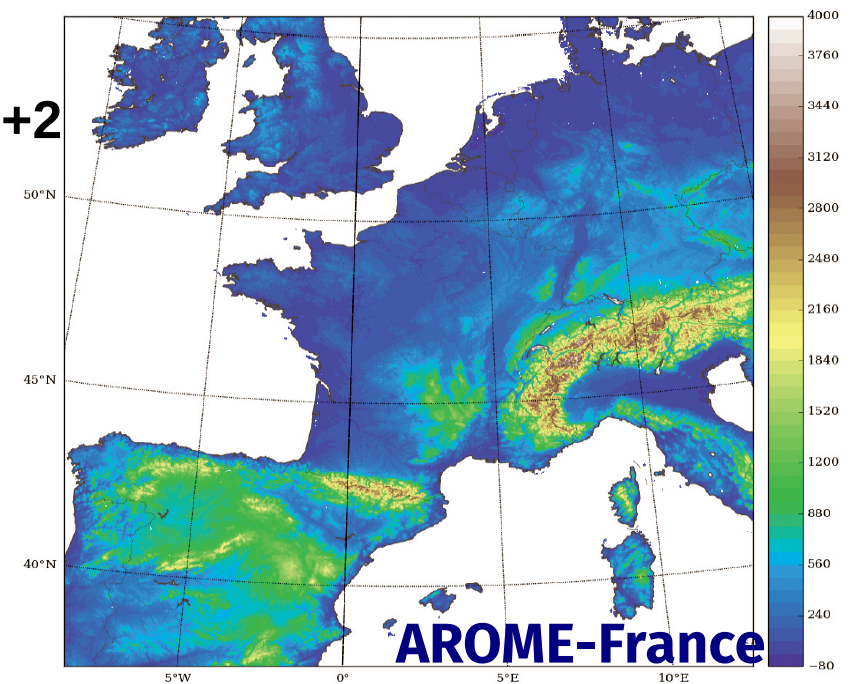
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 - × Satellite observations (MW, RO, AMV, IR)



SEVIRI radiometer

- **Meteosat Second Generation (MSG)**

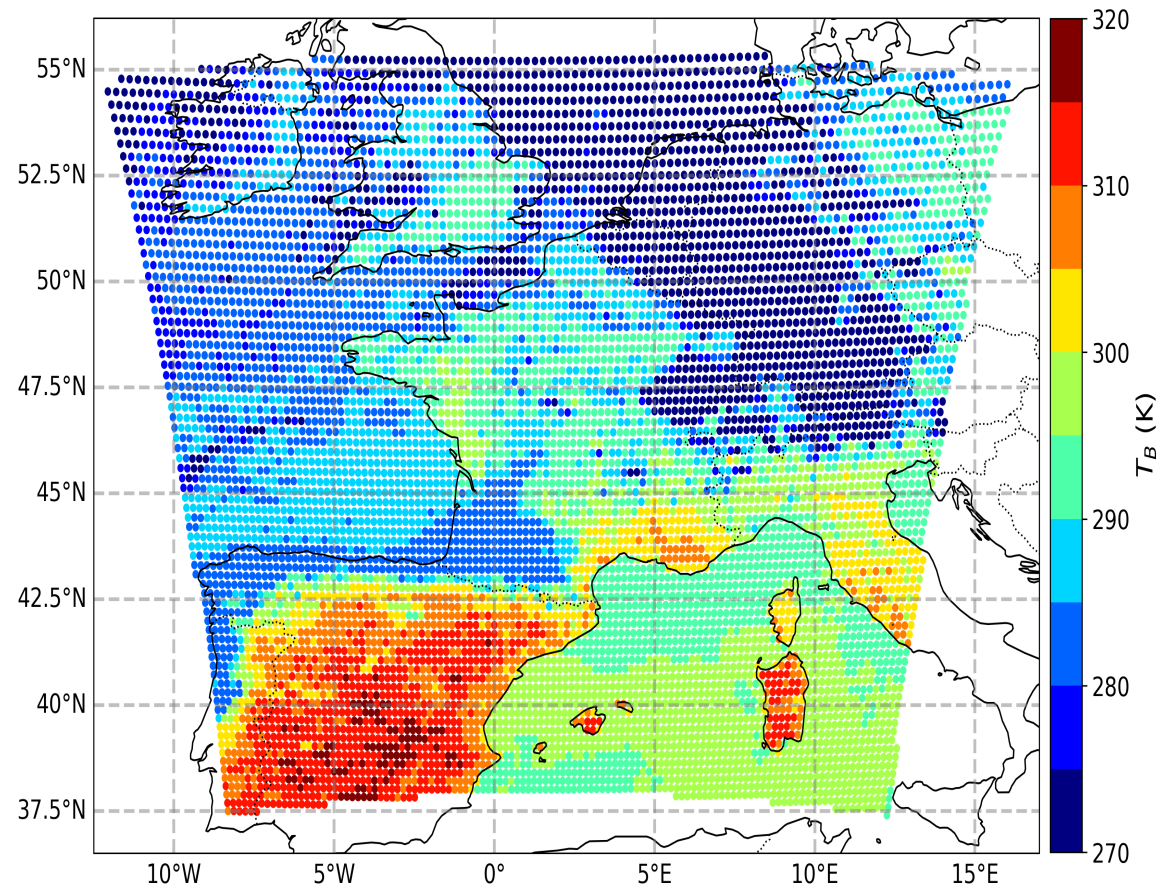
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- **Spatial resolution : 3 km at nadir**
- **Temporal resolution : 15 min**
- **Temperature and humidity information**

SEVIRI observations available



2024/08/01 12h UTC – SEVIRI radiances channel 8.7 μm

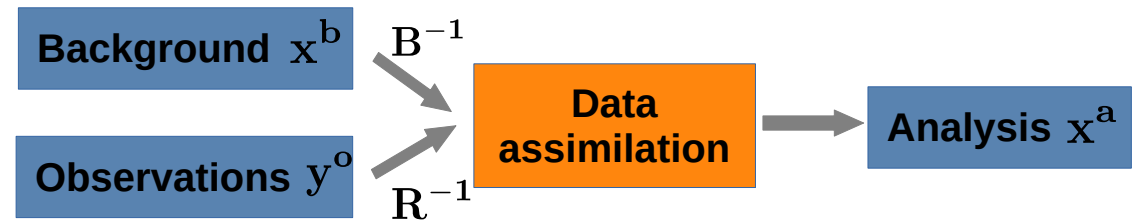
**Number of available observations before
quality control and cloud detection : 8323**

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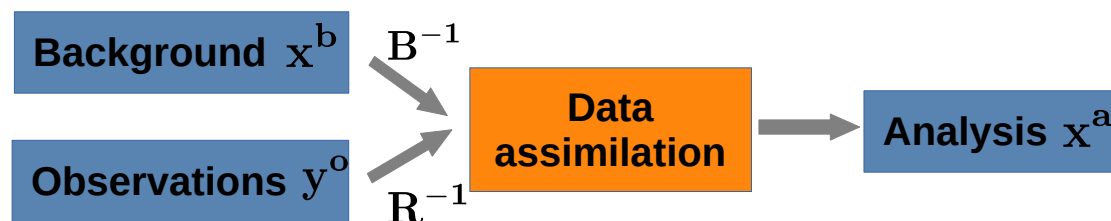
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- Essential for computing the analysis
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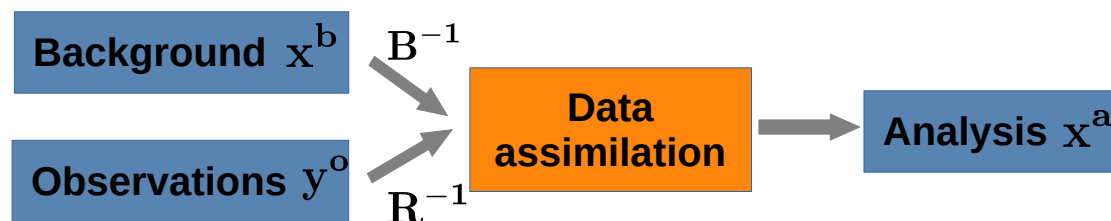


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- Come from various sources (instrument, modeling, representativeness)
- Can exhibit strong correlations

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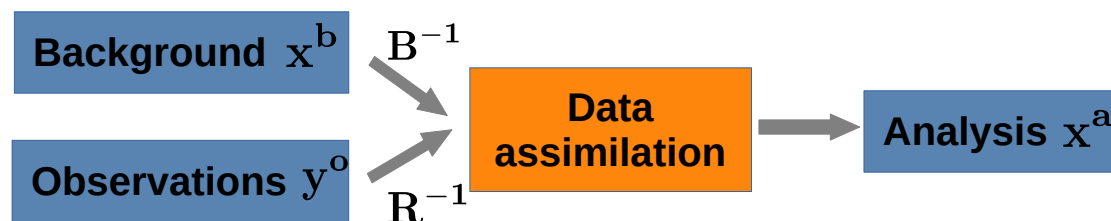
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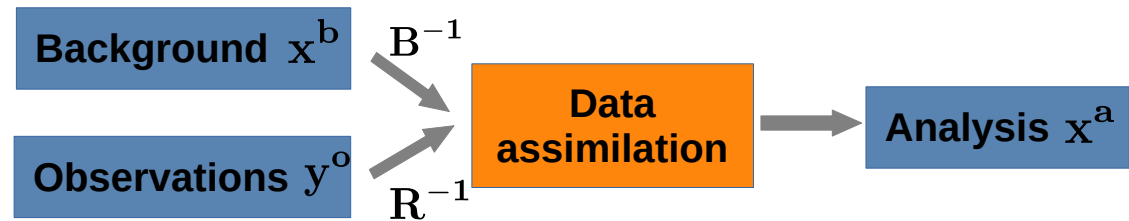
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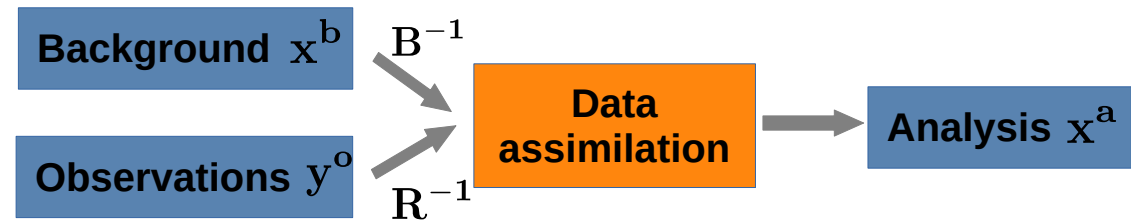
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- Interchannel correlations $\longrightarrow$ neglected for SEVIRI in AROME

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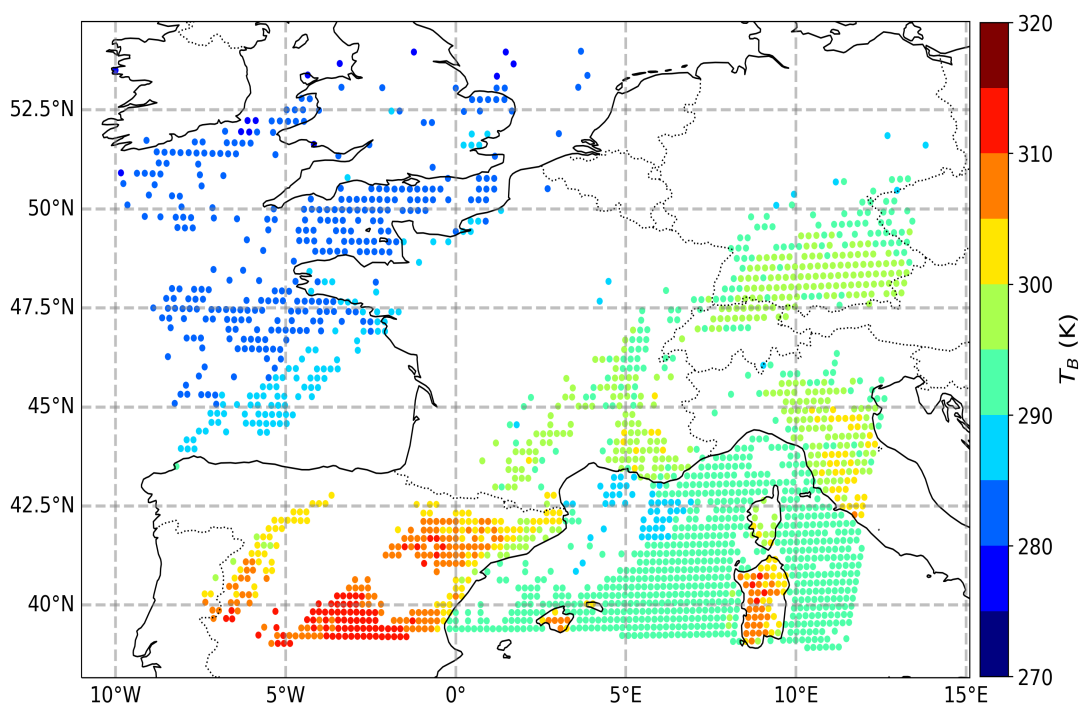
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Correlations of observation errors

- Temporal correlations $\longrightarrow$ neglected
- Interchannel correlations $\longrightarrow$ neglected for SEVIRI in AROME
- Horizontal correlations $\longrightarrow$ implicit treatment (thinning & variance inflation)

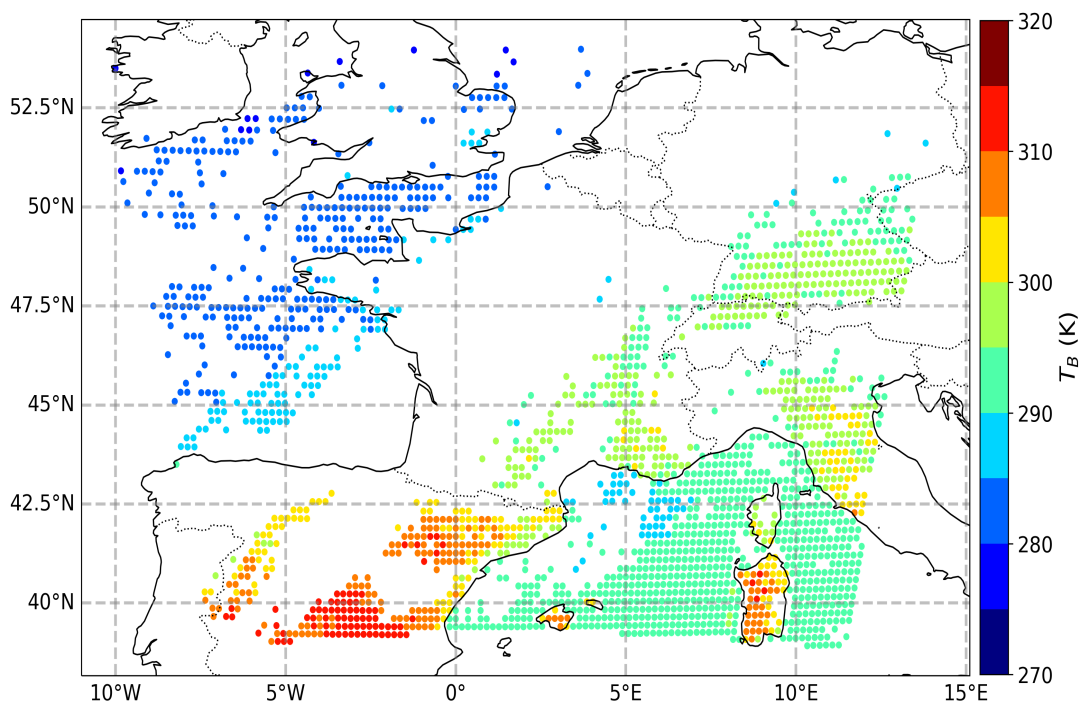
Impact of 70 km thinning in AROME for SEVIRI channel 8.7 μm

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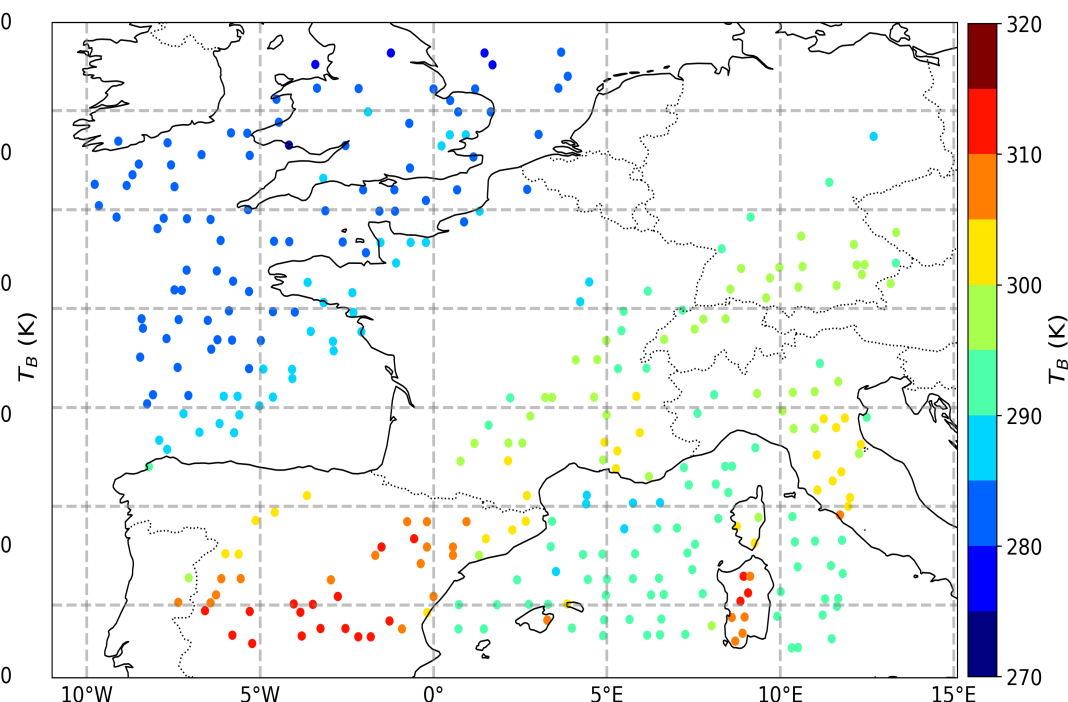


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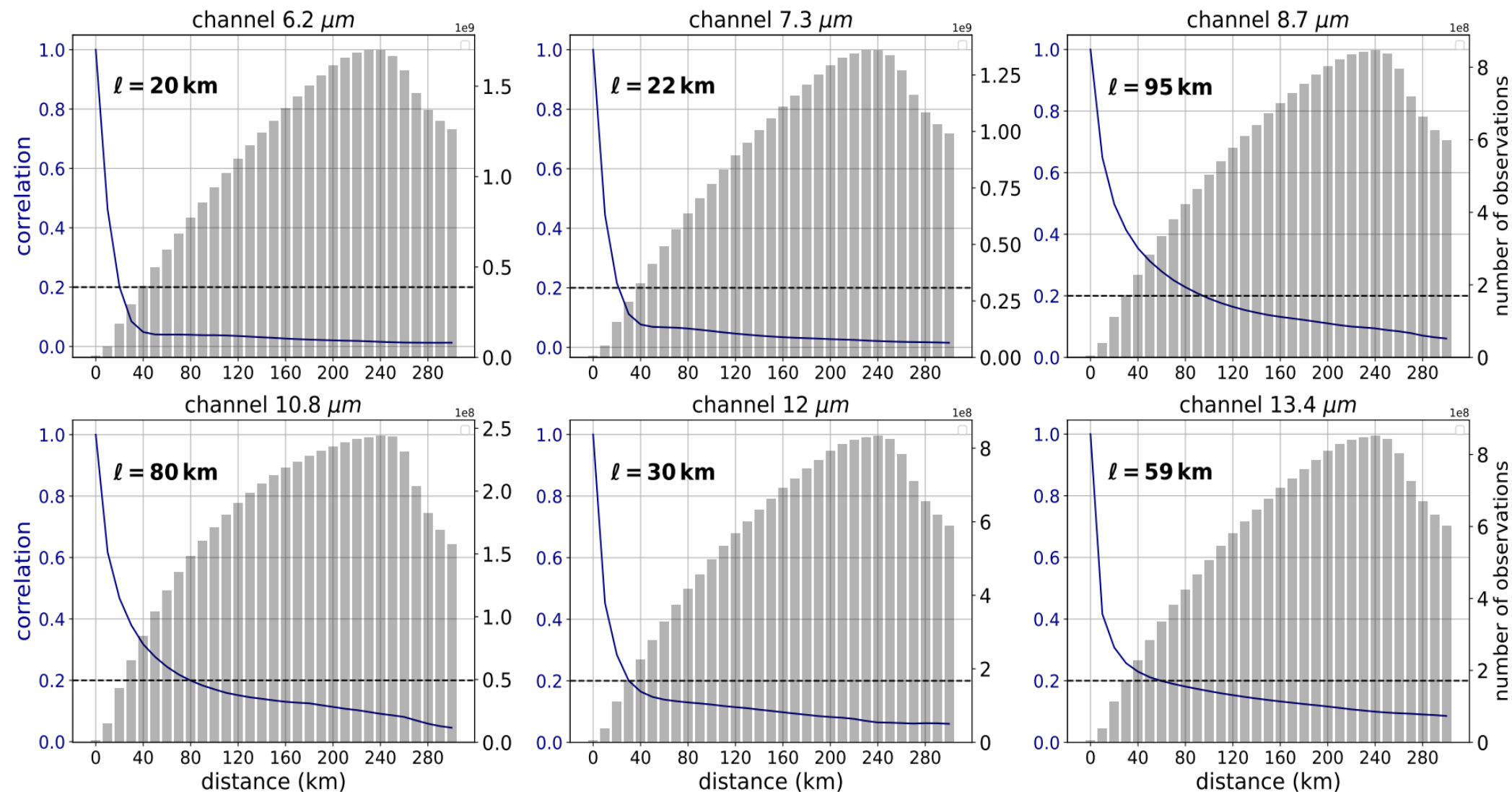


310 obs assimilated



85% of observations discarded

Horizontal correlations diagnosed for SEVIRI in AROME



How to introduce horizontal correlations of SEVIRI observation errors into the data assimilation system of the model AROME ?

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Method

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- Represent its inverse $\mathbf{R}^{-1}$ and incorporate it into the AROME DA system

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Scientific challenges

- Large-scale matrix-vector products
- Distribution of observations in a parallel computing environment

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Evaluation

- Numerical performance (convergence of the inner loop)
- Quality of the analysis

Horizontal correlations modeling

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Diffusion-based method (Guillet et al. 2019)

- Specify R^{-1} (sparse matrix)
- Saves computing time and memory
- Models only one class of correlation function

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Exponential model (Simonin et al., 2019, Hu et Dance., 2024)

- Specify R (full matrix)
- More costly in computing time and memory
- Allows use of multiple correlation lengths

$$C_{i,j} = e^{-\frac{\Delta_{i,j}}{L_d}}$$

$\Delta_{i,j}$ distance between 2 observations
 L_d correlation length

Horizontal correlations modeling

$$C_{ij} = \alpha e^{-\frac{d_{i,j}}{L_1}} + (1 - \alpha) e^{-\frac{d_{i,j}}{L_2}}$$

L_1, L_2 **correlation lengths**

$d_{i,j}$ **distance between
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α **weight**

Horizontal correlations modeling

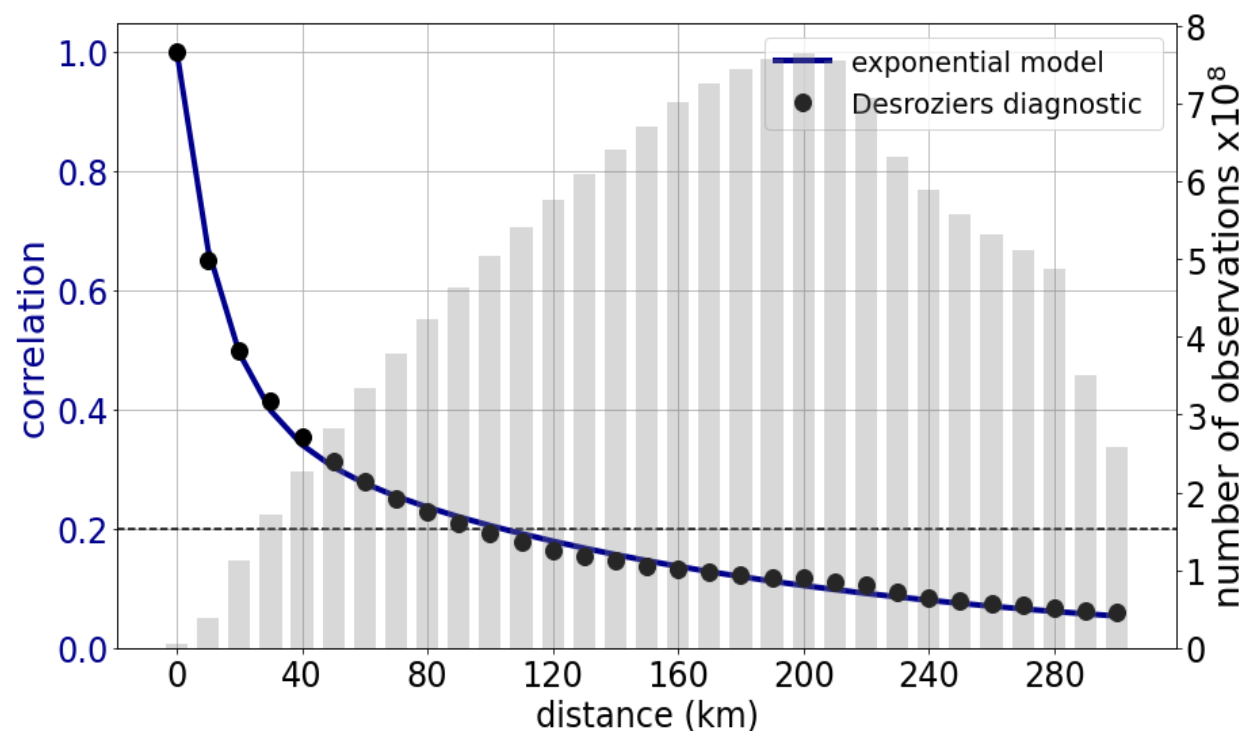
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SEVIRI channel 8.7 μm



Adequate fit between modeling and diagnostic

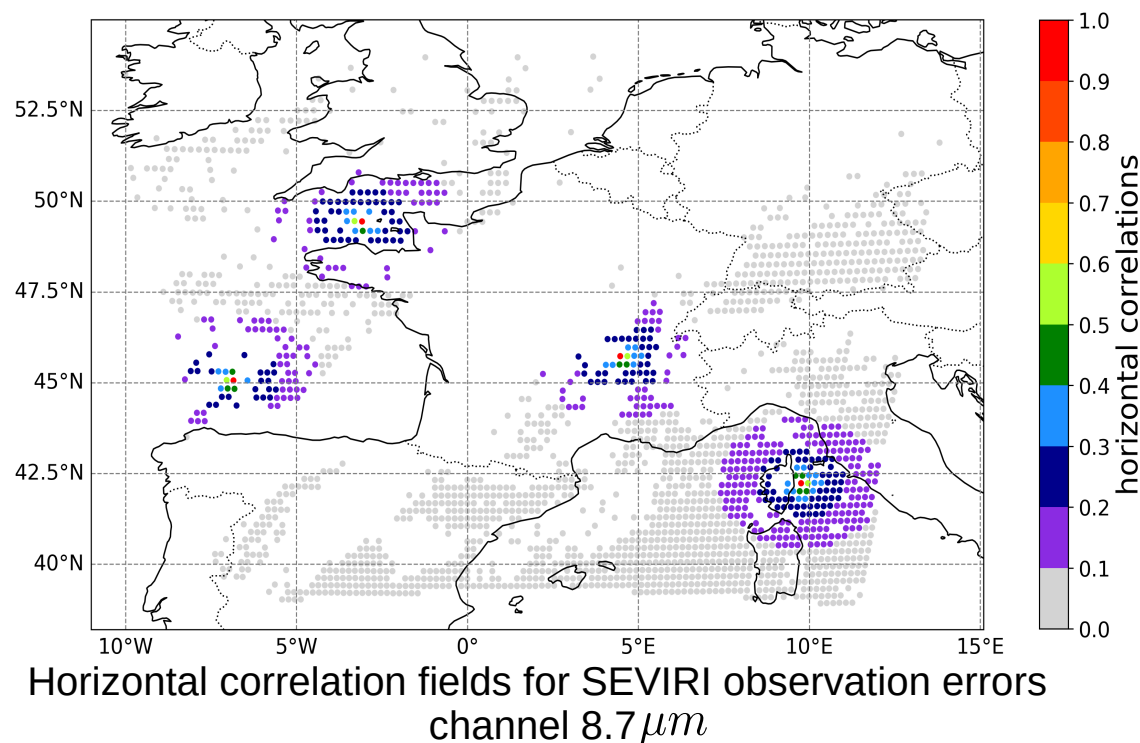
Simplified minimization environment

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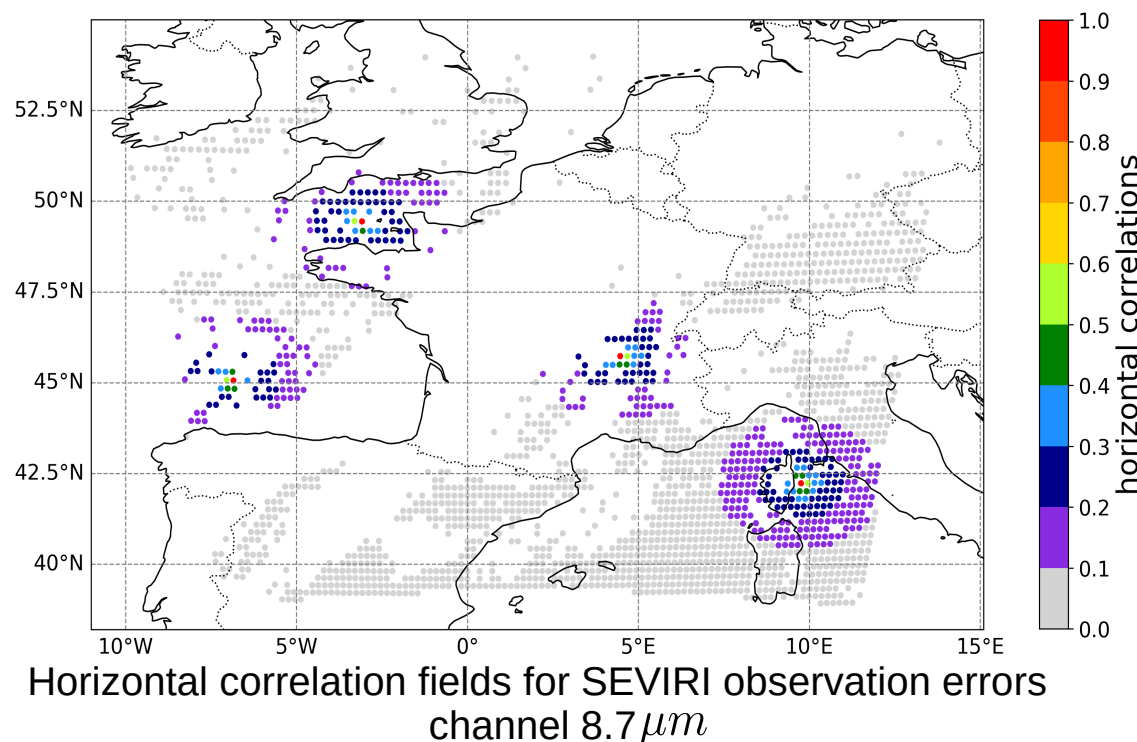


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Minimization experiments

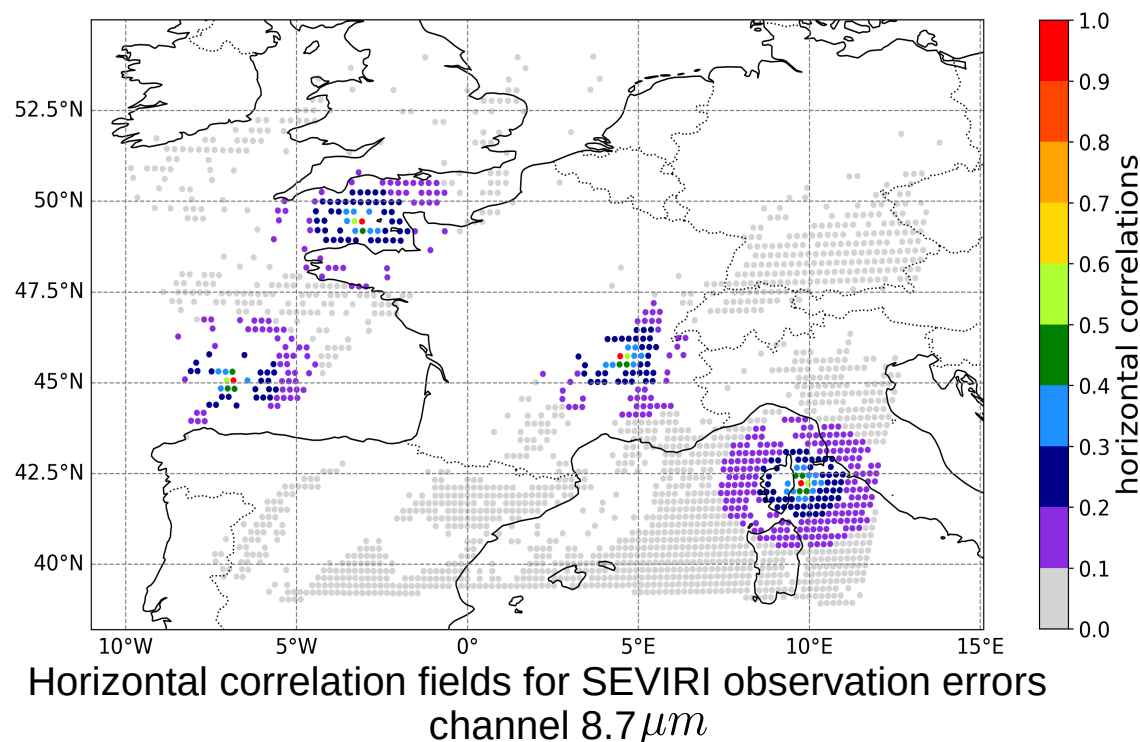
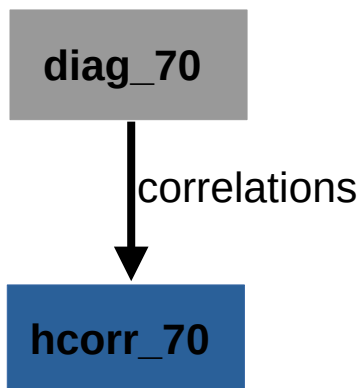
diag_70



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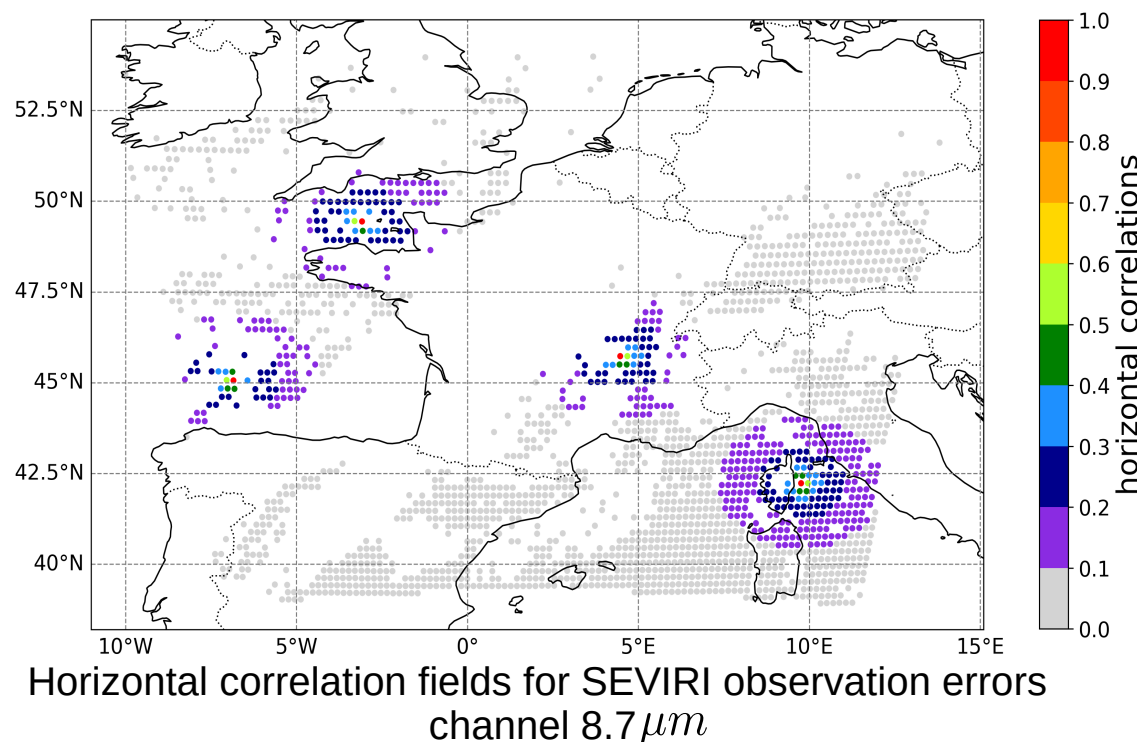
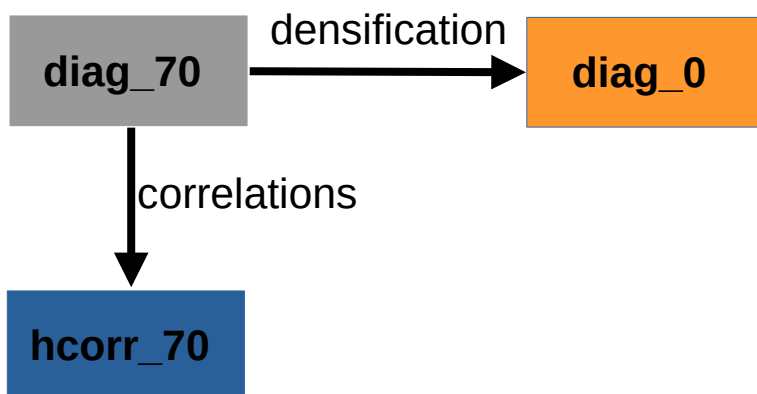
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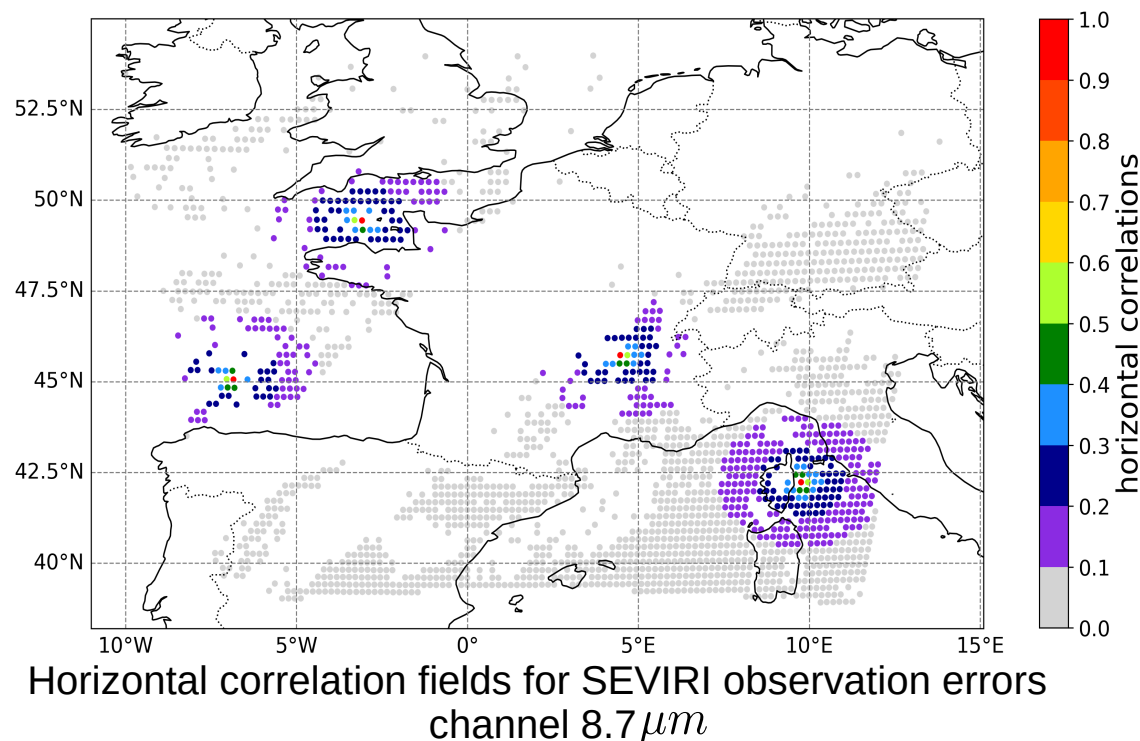
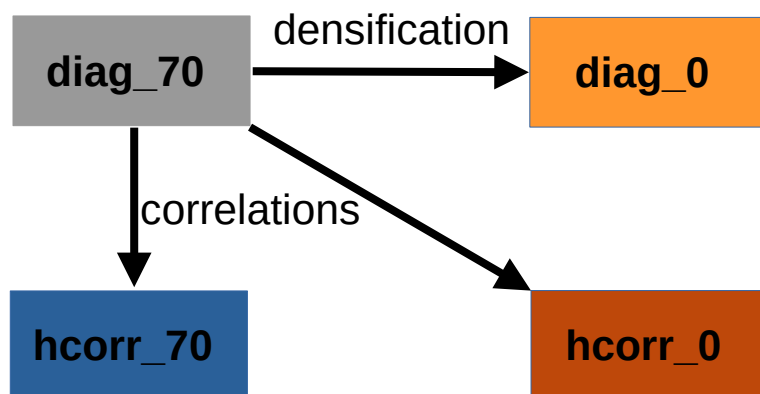
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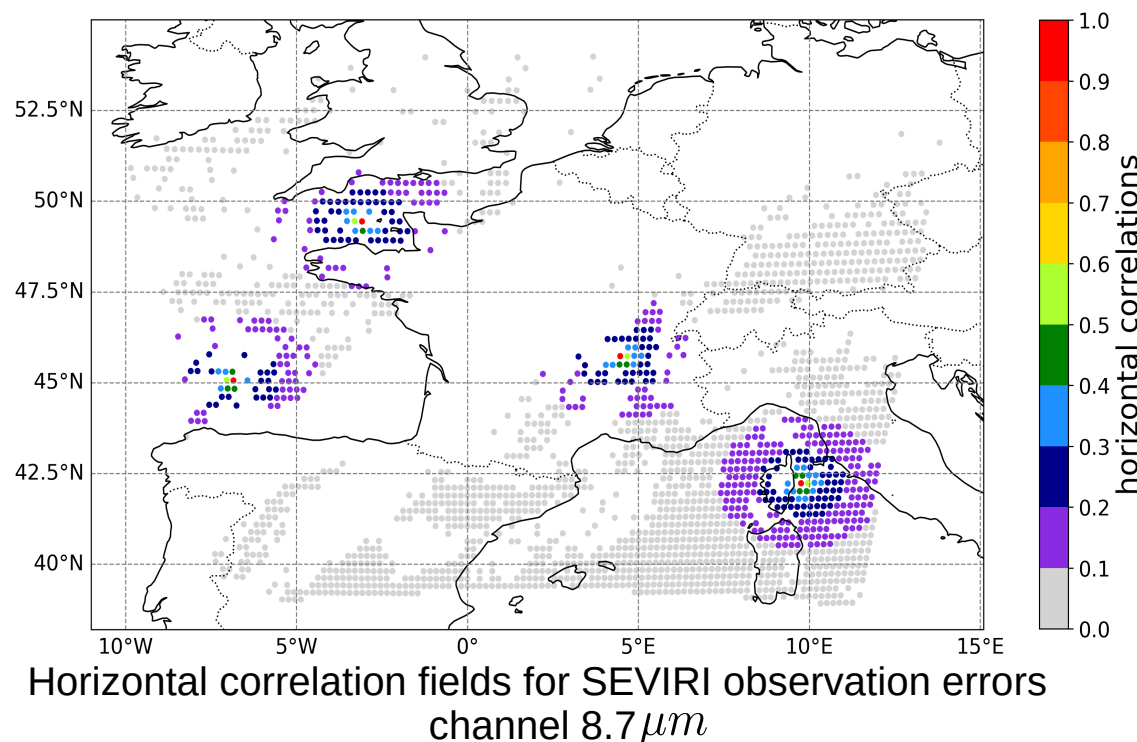
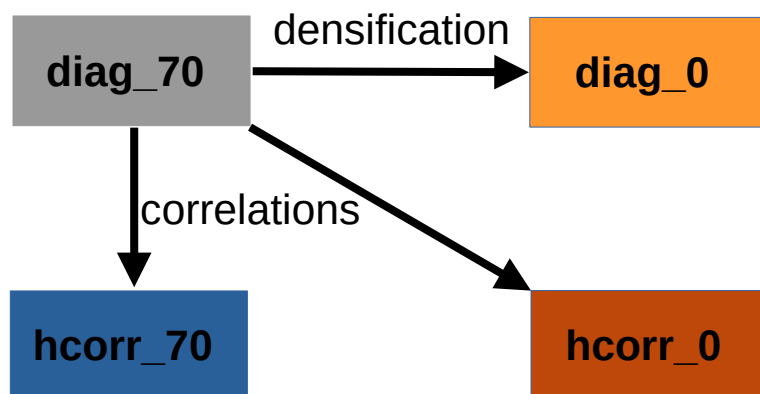
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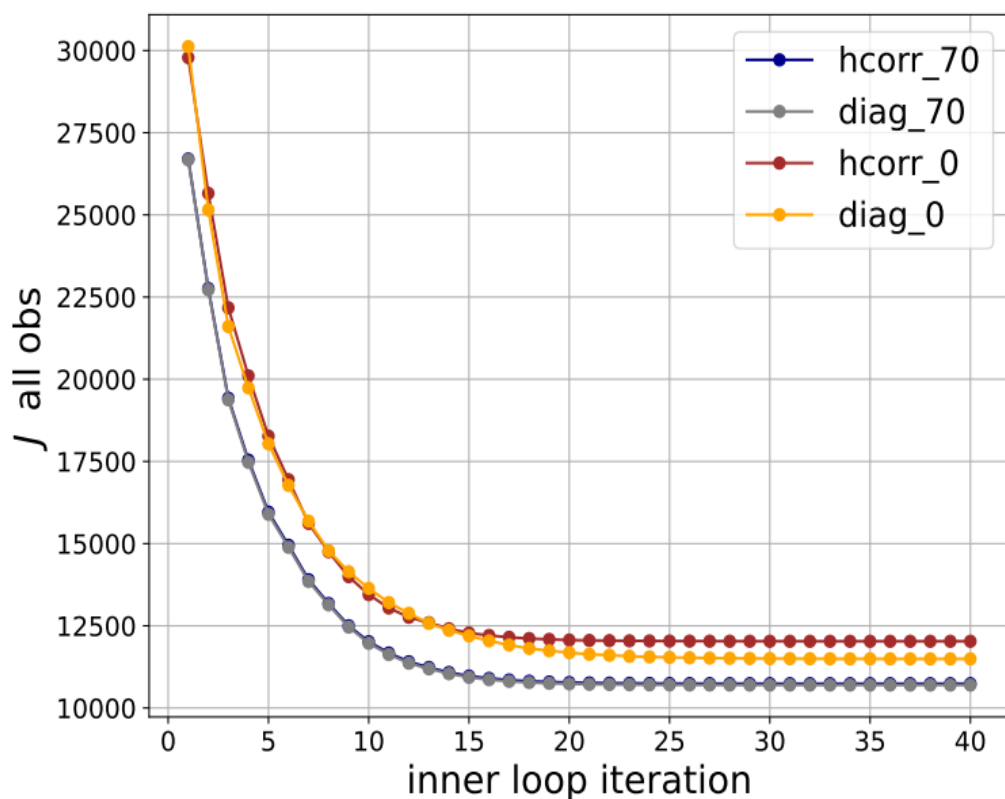


Evaluation of impact

- Minimization algorithm (inner loop)
- Analysis increments

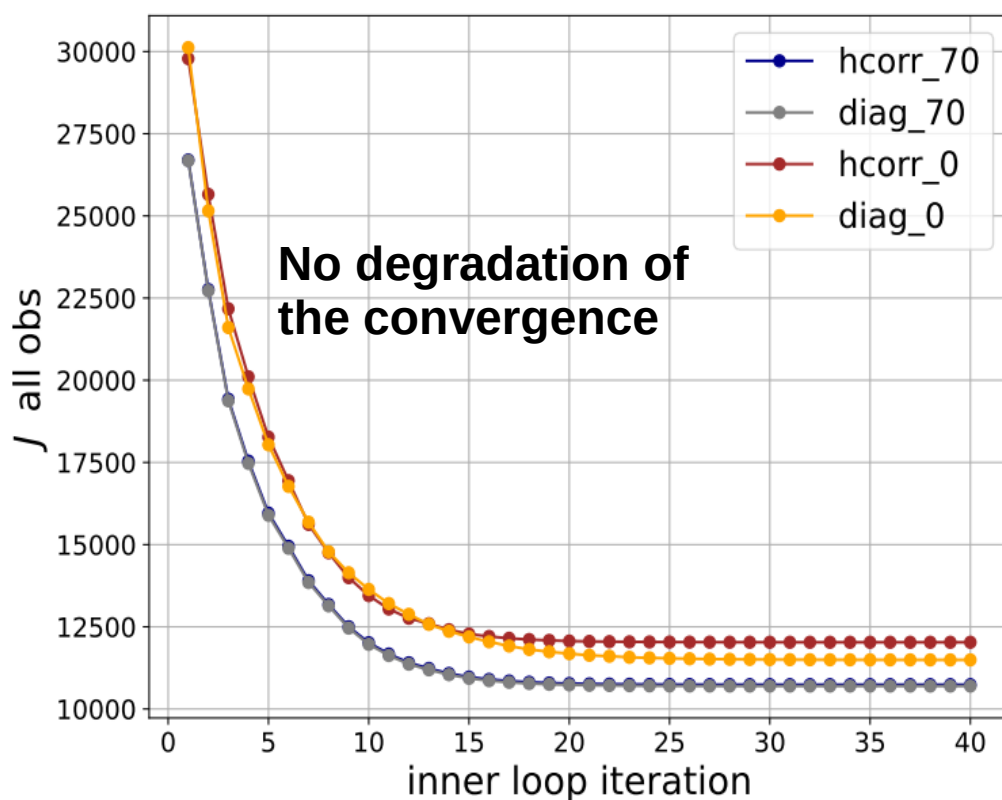
Minimization algorithm (inner loop) convergence

Cost function for all observations



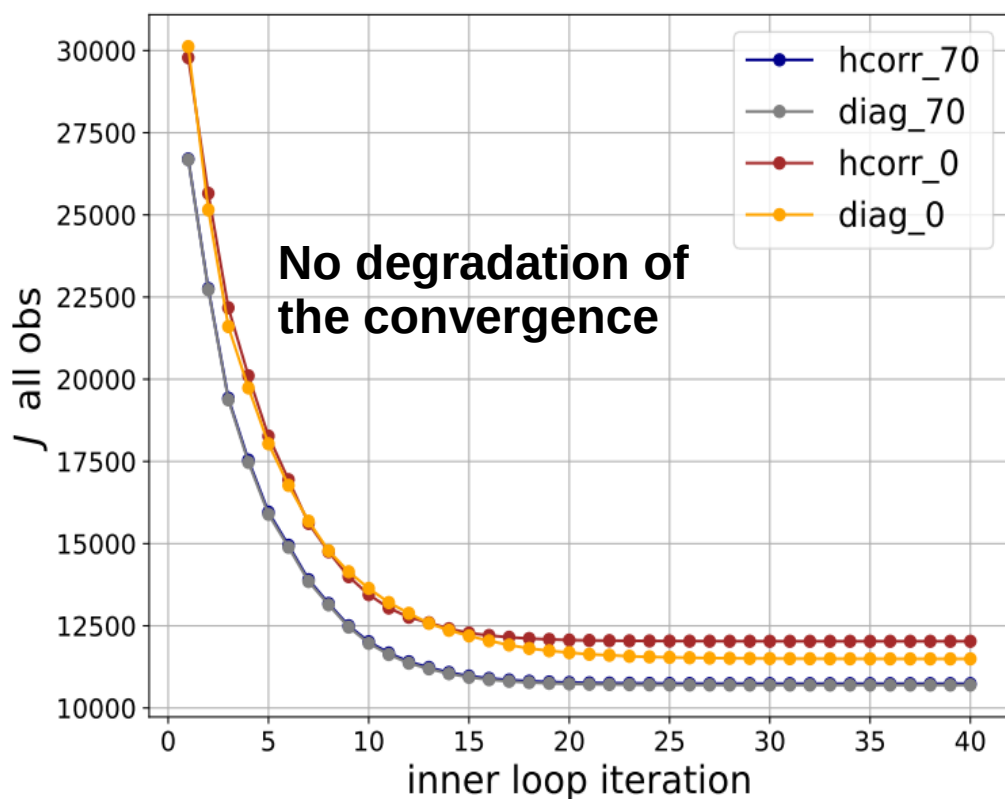
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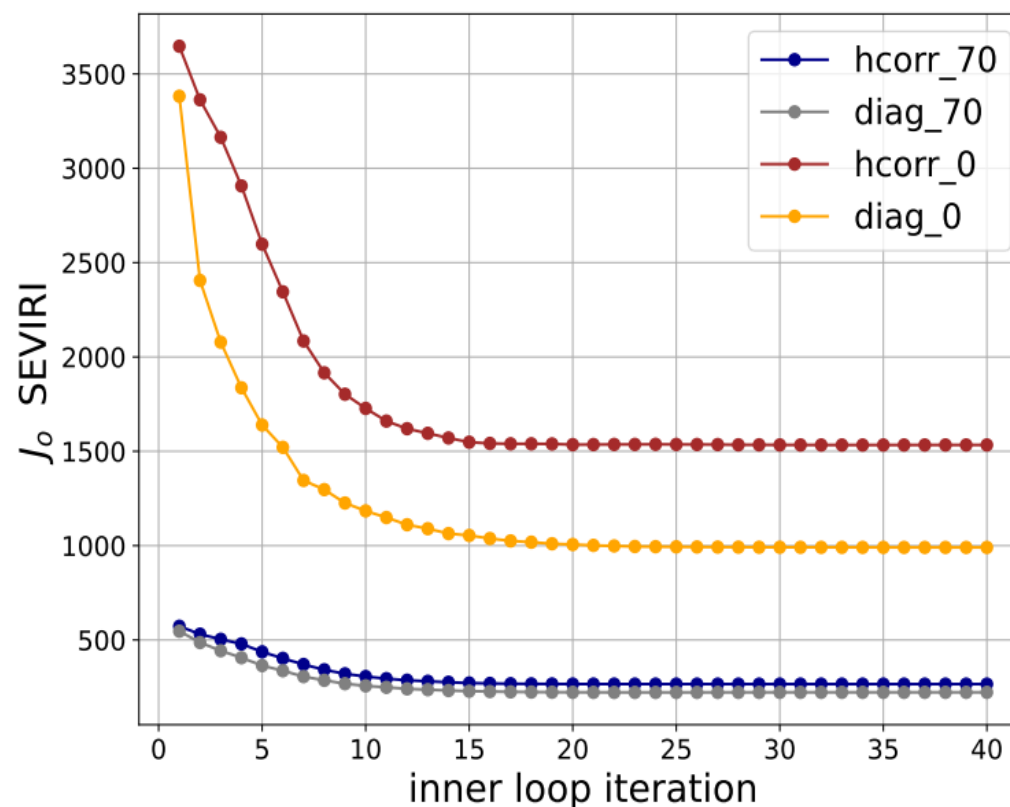


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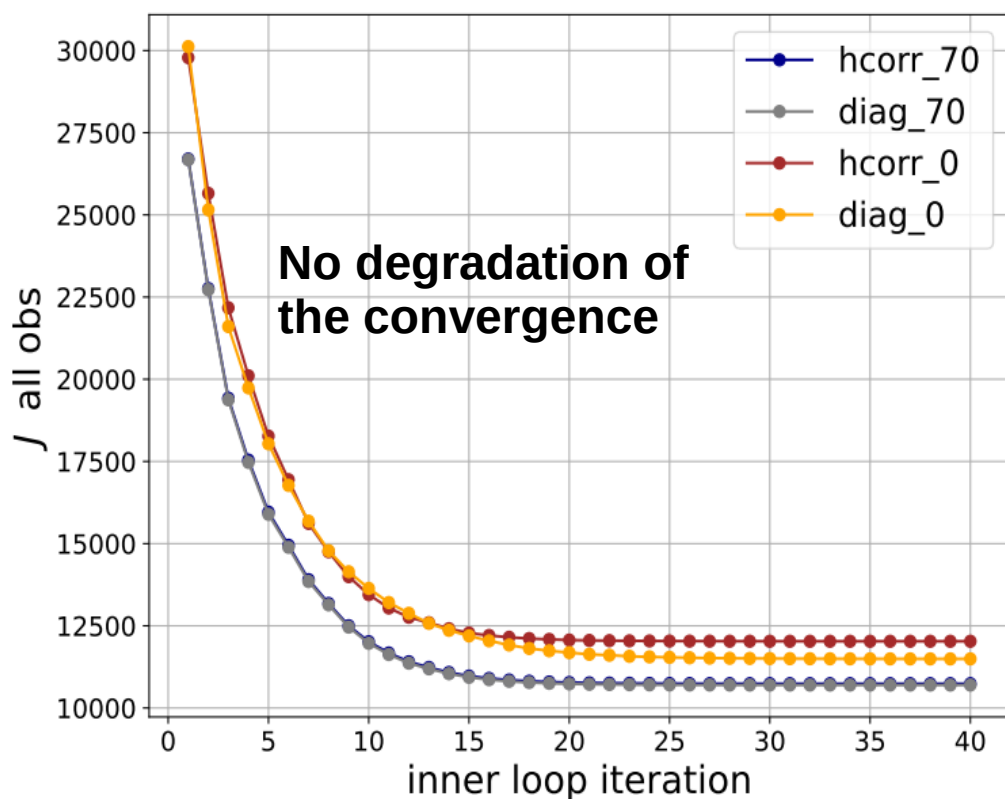


Observations cost function for SEVIRI

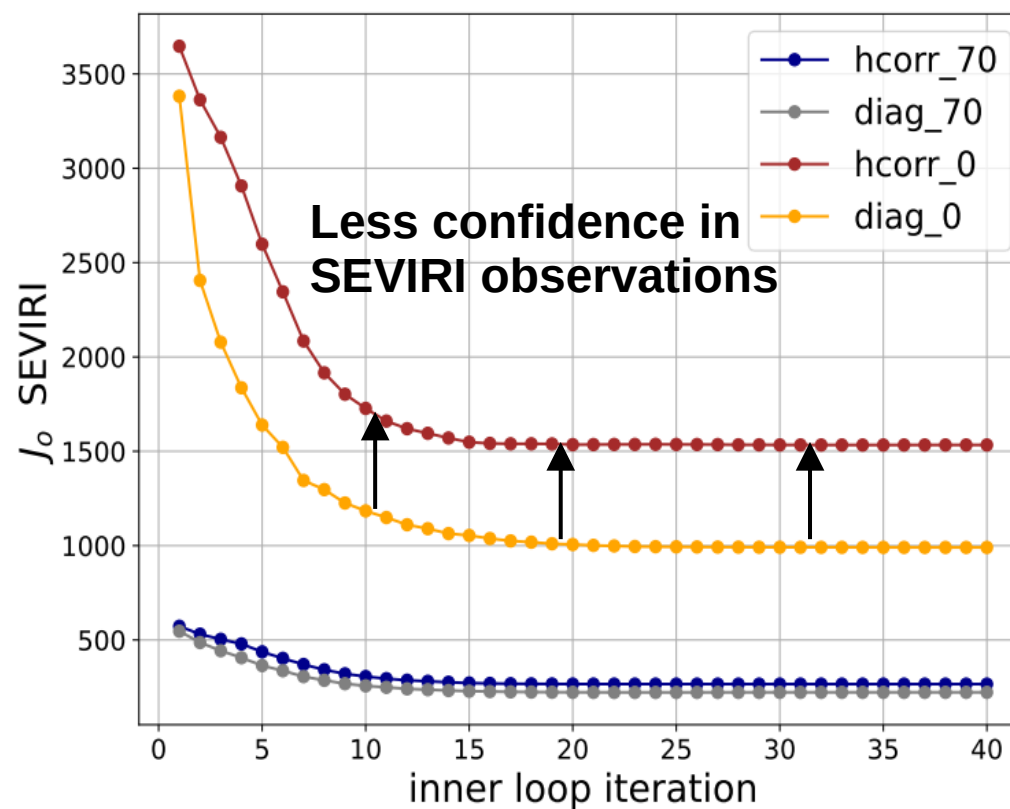


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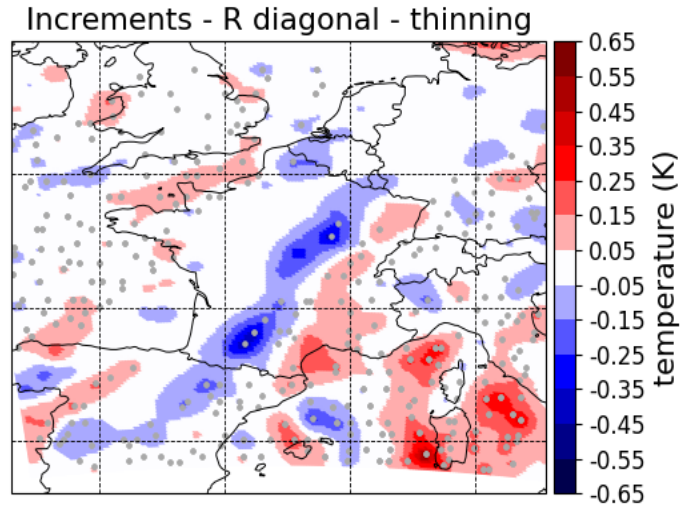
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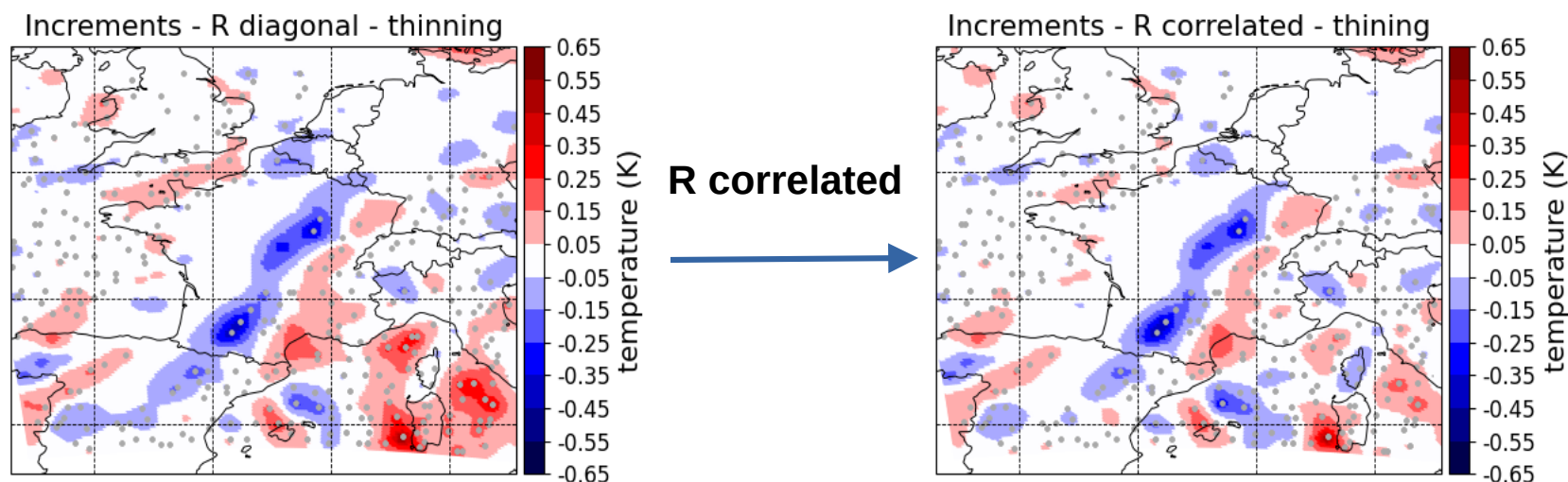
Observations cost function for SEVIRI



Impact of modeling correlations - thinning - temperature - 1012 hPa

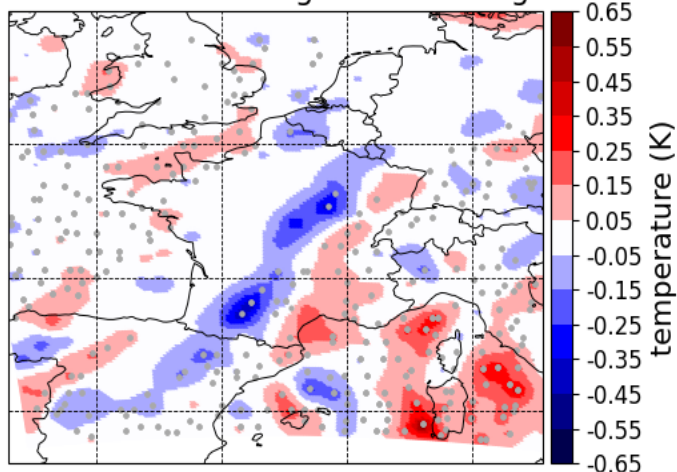


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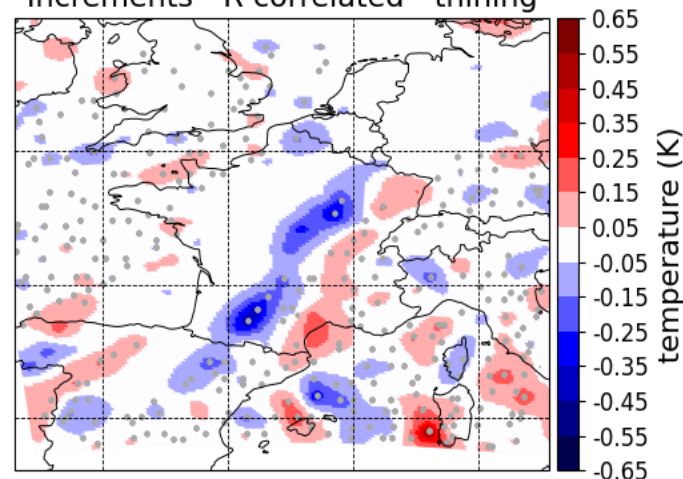
Increments - R diagonal - thinning



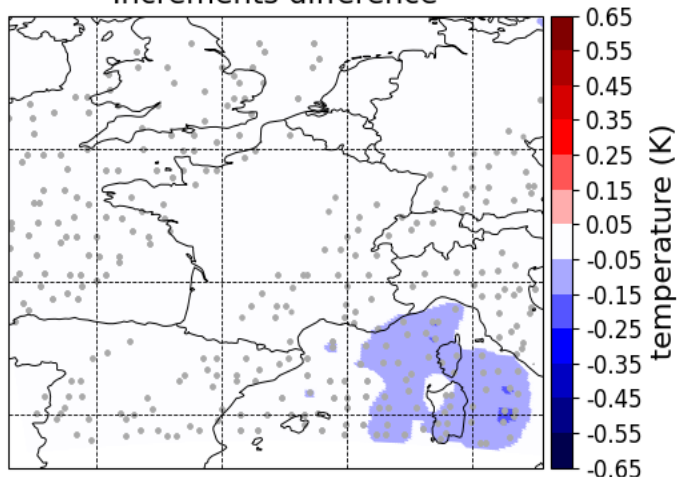
R correlated



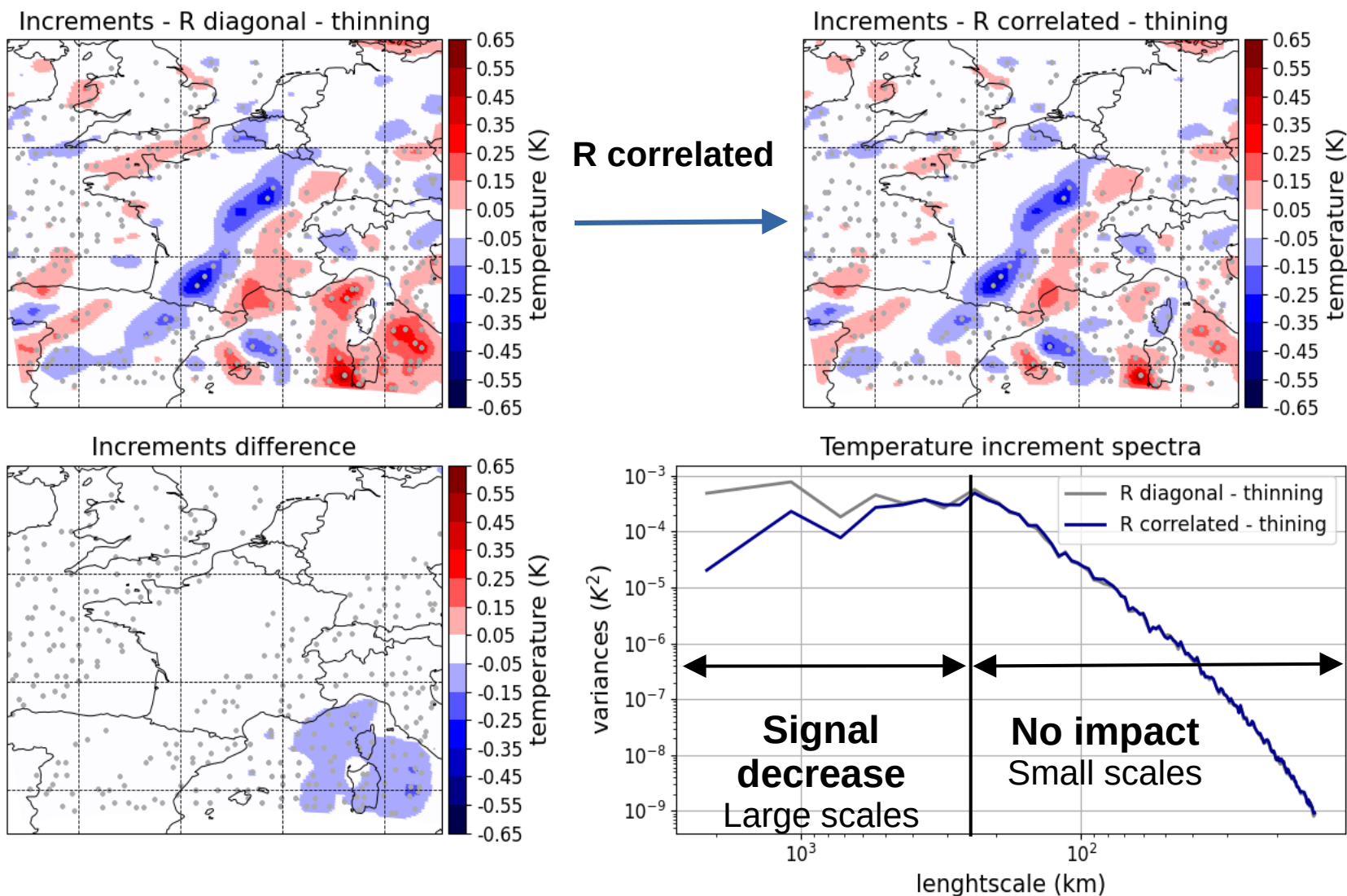
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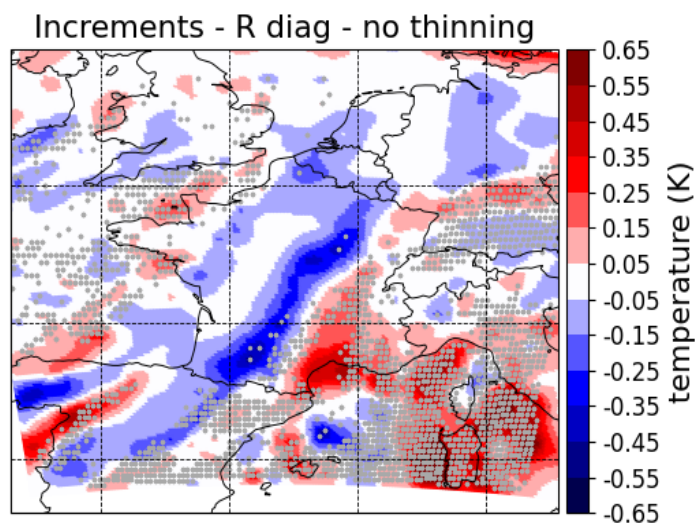
Increments difference



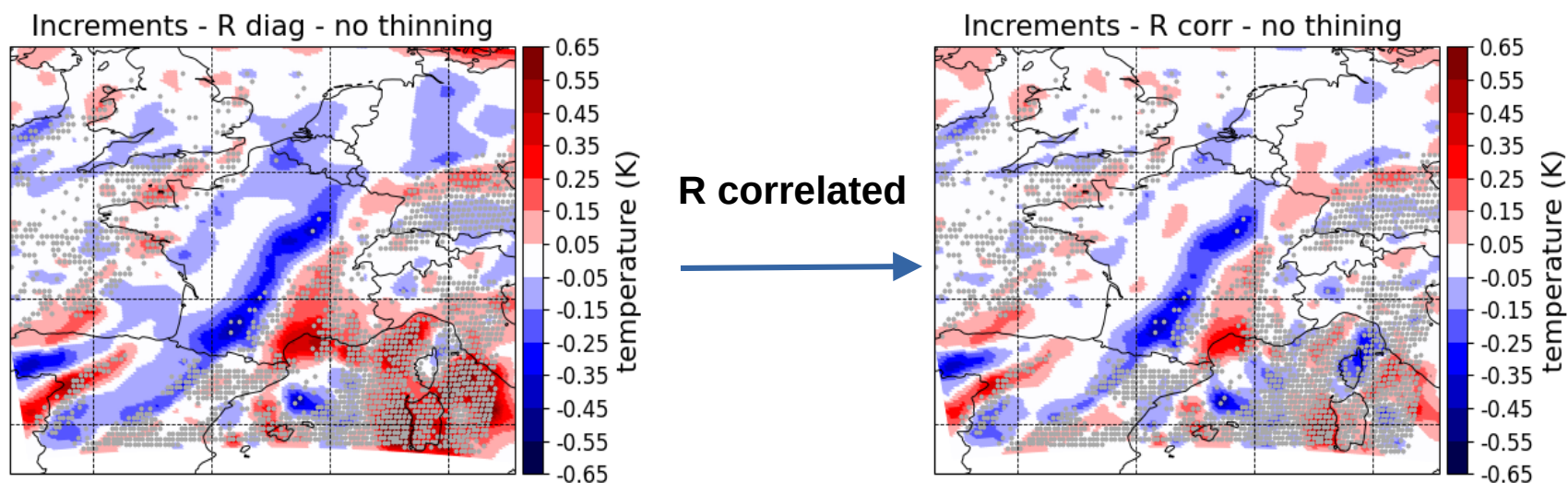
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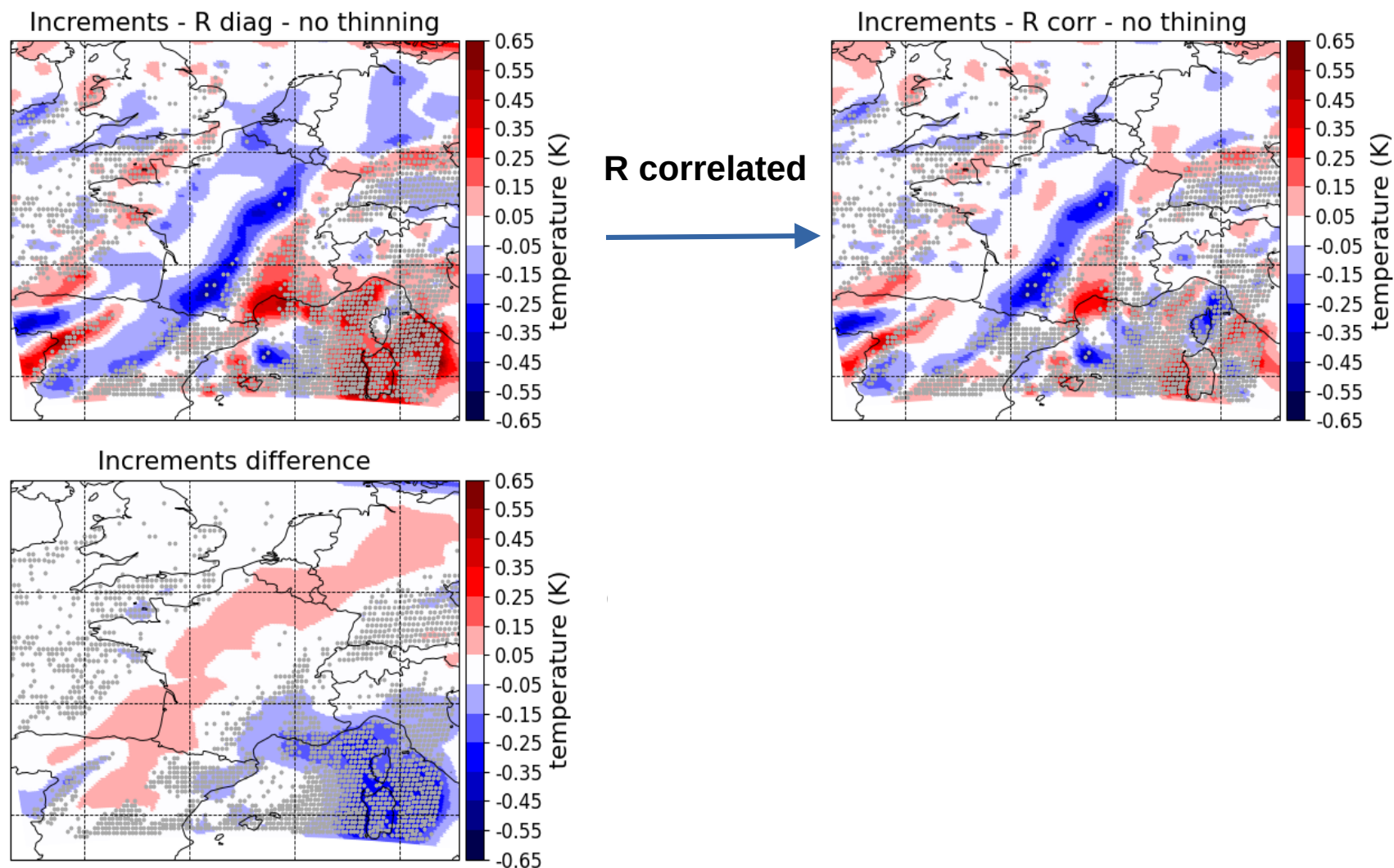
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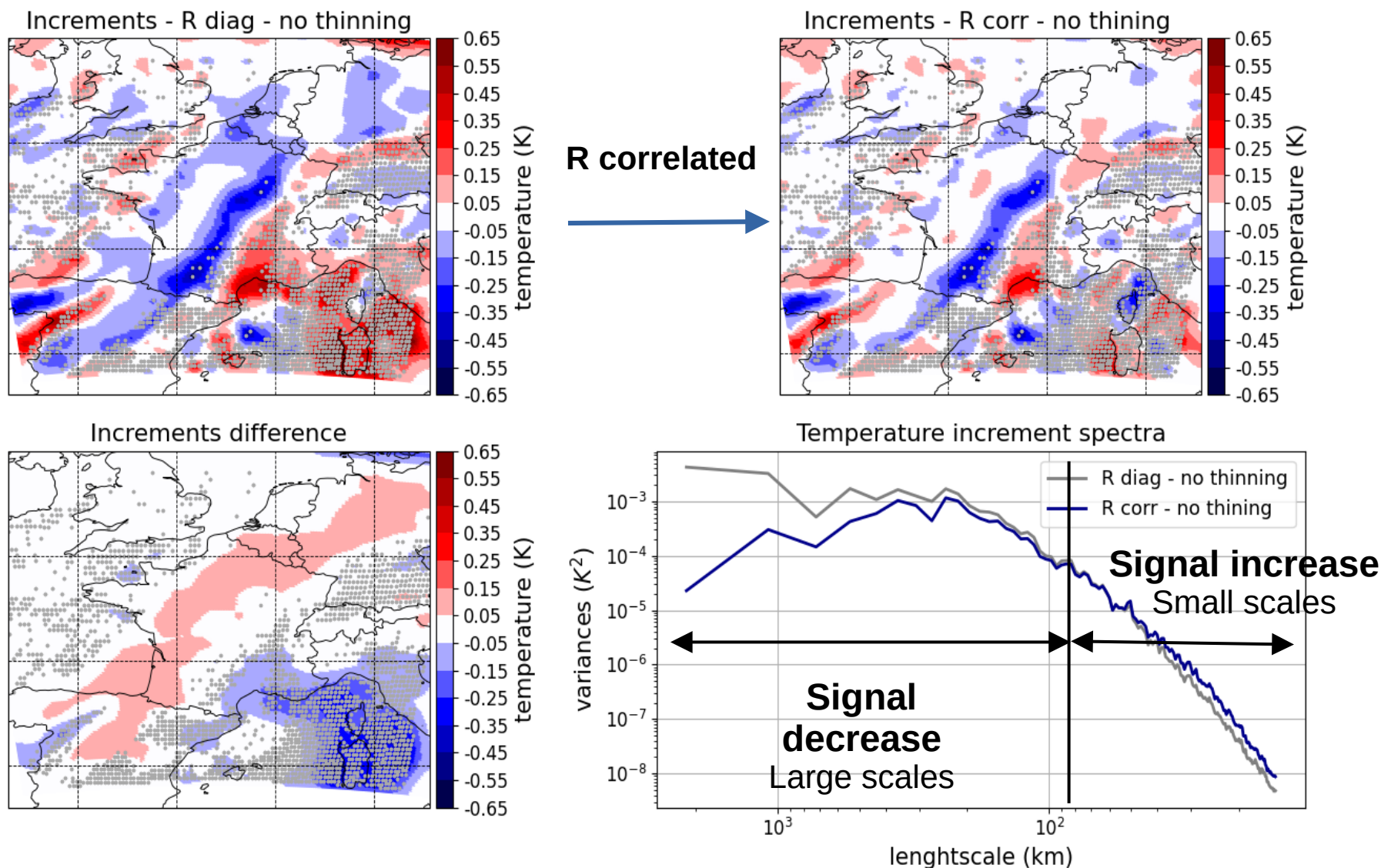
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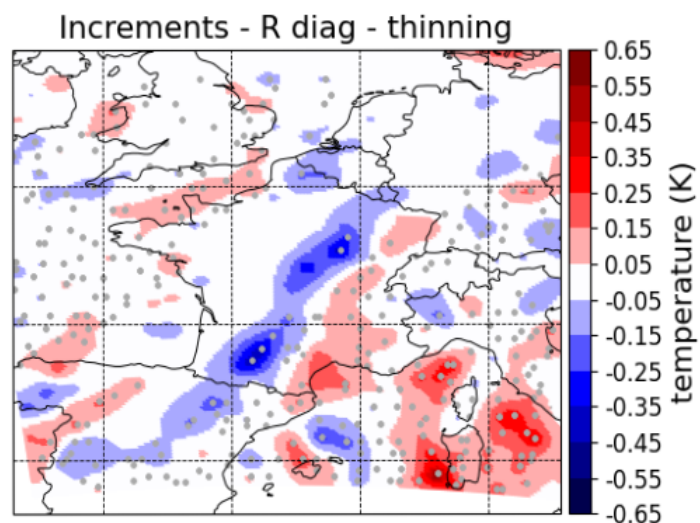
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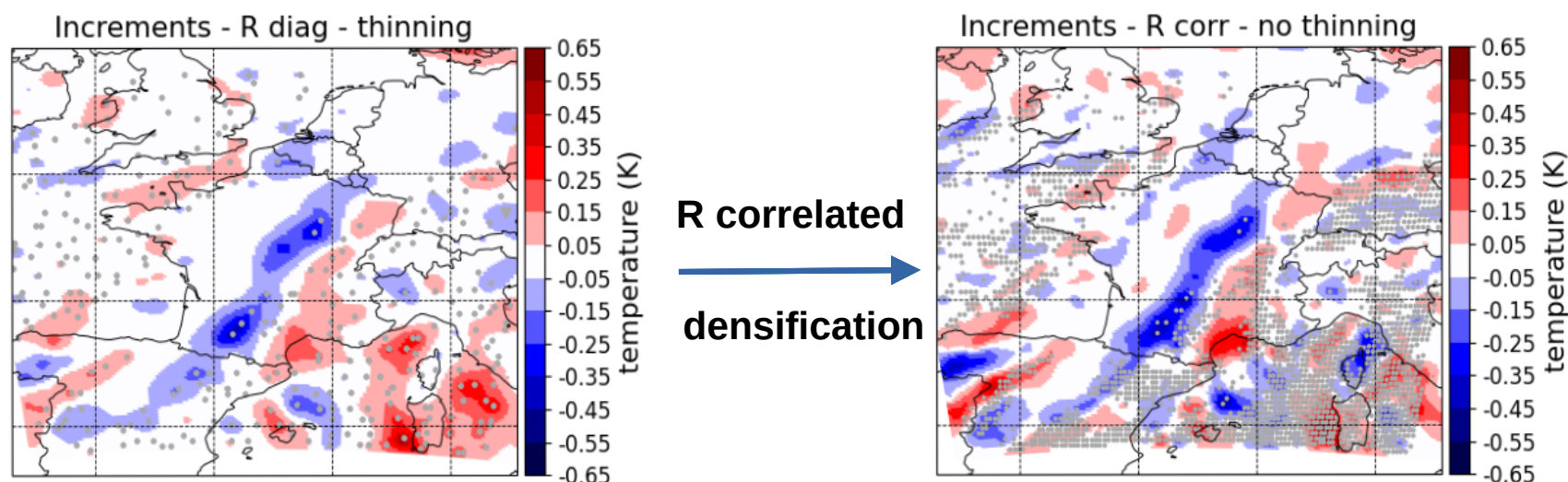
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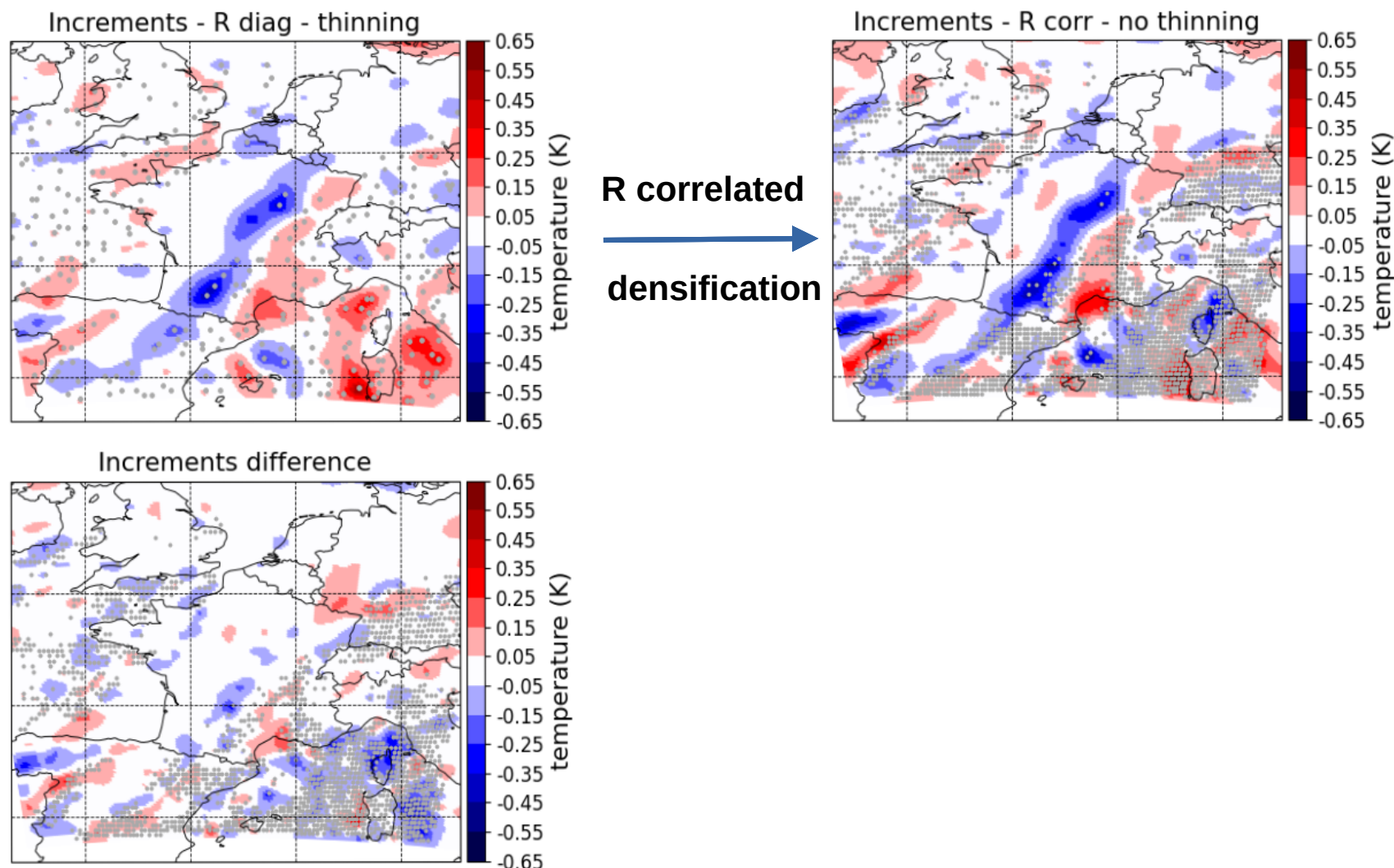
Impact of modeling correlations + thinning - temperature - 1012 hPa



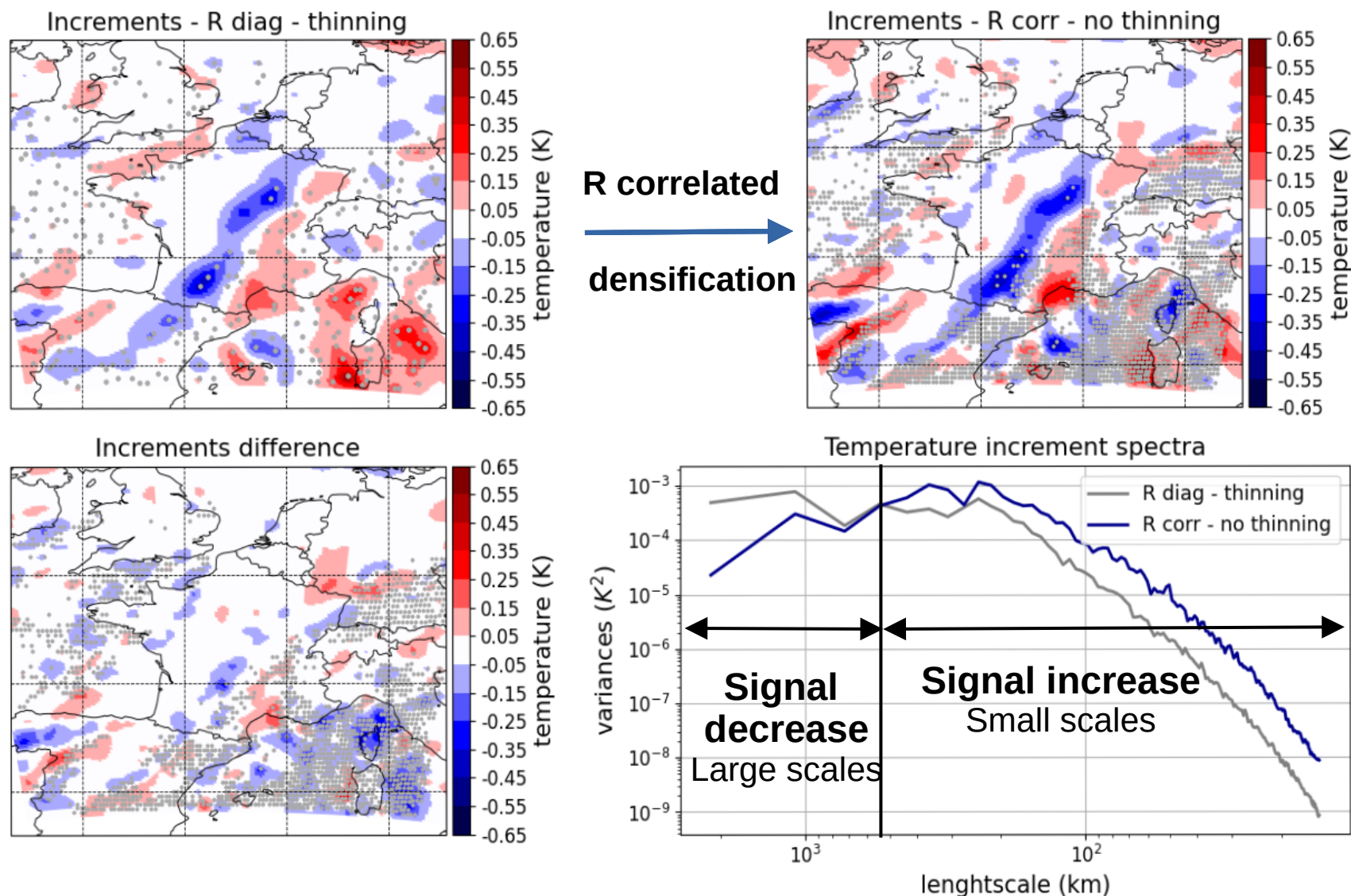
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 article in progress – submission June 2025 (QJRMS)

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Assess the impact of introducing horizontal correlations in terms of forecasts scores

Thank you for listening

Any questions ?