

A Consortium for CONvection-scale modelling  
Research and Development

## Evaluation of some HARMONIE-AROME cy46 data assimilation tunable settings

Jana Sanchez Arriola, Magnus Lindskog,  
Eoin Whelan, Joan Campins, David Schönach, Reima Eresmaa

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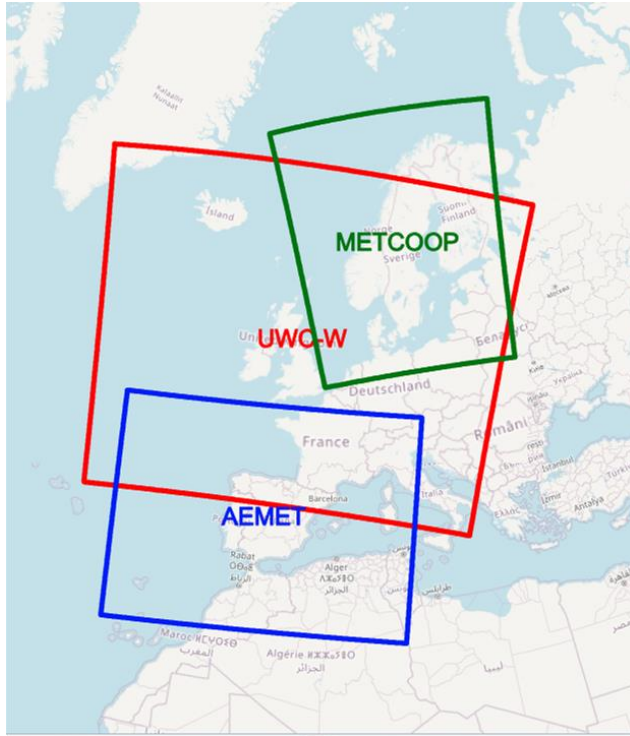
# Outline

- CY46 evaluation runs
- Description of tunable settings
- Diagnostic methods
- Recommended adjustments
- Limitations revealed by BGOS and how to alleviate
- Summary and conclusions

Full report available here:

[https://opensource.umr-cnrm.fr/attachments/download/6142/Tuning\\_report\\_jan25.pdf](https://opensource.umr-cnrm.fr/attachments/download/6142/Tuning_report_jan25.pdf)

# Harmonie-AROME Cycle 46 evaluation runs



Extensive DA experiments have been carried out over 3 domains and 4 different seasons to evaluate the performance of the Cycle 46 as compared to the currently operational Cycle 43.

They have been run over a month period , with a 15 days spin-up period.

Observation usage was the same in both model versions and roughly corresponding to operational use in each domain.

For this tuning we use the results from the cy46 runs

- Description of tunable settings

# Description of tunable settings

## First guess Check


Tunable parameter L

First-guess check to reject observations due to Gross errors

$$\{[H(x_b)]_i - y_i\}^2 / \sigma_{b,i}^2 > L \times \lambda, \quad (1)$$

where  $\lambda = 1 + \sigma_{o,i}^2 / \sigma_{b,i}^2$  rejection limit and

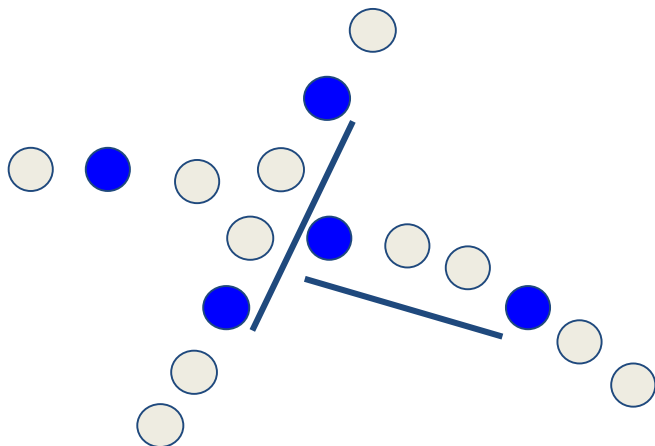
$\sigma_{o,i}$  and  $\sigma_{b,i}$  - error standard deviations ( $\sigma_{b,i}$  from errgrib-file)

By default L=25  reject obs if  $|y - Hx_b| > 5 \text{ times } \sqrt{\sigma_{b,i}^2 + \sigma_{o,i}^2}$

# Description of tunable settings

## Observation thinning

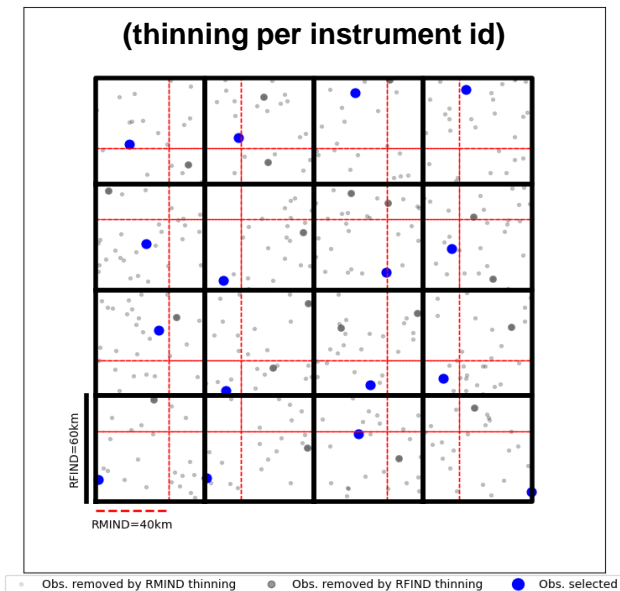
AMDAR and SHIP  
(thinning per station id)



Tunable parameters RMIND (AMDAR) and RSHIDIS (SHIP)

Satellite radiances

(thinning per instrument id)



Tunable parameters RFIND, RMIND

# Description of tunable settings

## Background and Observation error scaling

Relative weight of background and observations in minimisation of penalty function J:

$$J = J_b + J_o = \frac{1}{2} \delta \mathbf{x}^T \mathbf{B}^{-1} \delta \mathbf{x} + \frac{1}{2} (\mathbf{H} \mathbf{x}^b + \mathbf{H} \delta \mathbf{x} - \mathbf{y})^T \mathbf{R}^{-1} (\mathbf{H} \mathbf{x}^b + \mathbf{H} \delta \mathbf{x} - \mathbf{y})$$



Tunable parameters  
REDNMC, REDNMC\_Q

stabal\_cv.dat  
stabal\_bal.dat

SIGMAO\_COEFF(OBSTYPE)



but note that  $\sigma_{o,i}$  and  $\sigma_{b,i}$  in different spaces (unbalanced temperature relative to brightness temperature)

# Background Errors in Observation Space (BGOS)

Application developed in the OOPS framework to compute background error standard deviation in observation space:

$$\sigma_b^o = \sqrt{\frac{1}{N} \sum_{i=1}^N (HU\chi_i)^2}$$

where

$\sigma_b^o$  bg error std in ob space

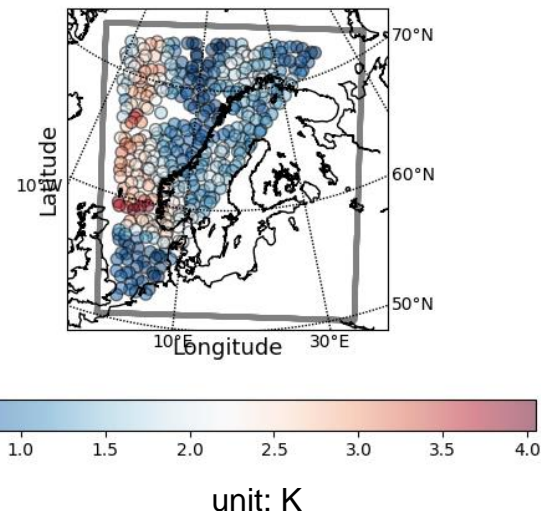
N sample size

U is the series of transform applied to get a unit B matrix in minimization, H is the observation operator (tl or nl) and  $\chi_i$  is the control vector (containing Gaussian errors) for the individual member i.

$$\left( \delta x = U\chi \quad J_b = \frac{1}{2} \delta x^T B^{-1} \delta x = \frac{1}{2} \chi^T \chi \right)$$

## Example

MHS channel 3 background error standard deviations in observation space 15 July 2022, 00 UTC. MetCoOp domain.



(corresponding observation error standard deviation is 1.8 K)

Now we can compare bg errors with obs errors

- Diagnostic methods

# Diagnostic methods

## 1- Andersson&Järvinen Q. J . R. Meteorol. SOC. (1999). 125, pp. 691-122

**AIM:** To select appropriate check limits (FgLim) for background check. Assumption is that observations with errors outside Gaussian distribution are affected by Gross errors and should be removed prior to the data assimilation.

$$([H(\mathbf{x}_b)]_i - y_i)^2 / \sigma_{b,i}^2 > \text{FgLim} \times \lambda$$

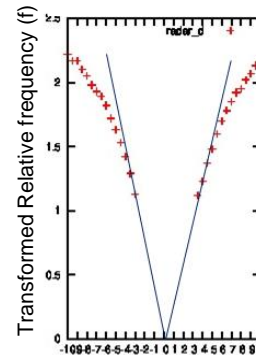
**HOW:** Plots histograms and transformed histograms of innovations to identify when distribution starts to deviate from Gaussian and where to put rejection limit.

where  $\lambda = 1 + \sigma_{o,i}^2 / \sigma_{b,i}^2$ , FgLim is the rejection limit and  $[H(\mathbf{x}_b)]_i$  denotes the projection of the model state on  $y_i$  observation, where the potential observation bias has been accounted for.  $\sigma_{o,i}$  and  $\sigma_{b,i}$  are the standard deviation of the observation error and background error equivalent, respectively.

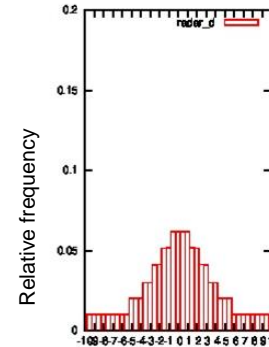
Transformation formula:

$$\hat{f} = \sqrt{-2 \ln[f / \max(f)]}$$

**Example for radial winds:**



Innovation departure (m/s)



Innovation departure (m/s)

# Diagnostic methods

## 2- Obstool (based on P. Benachecks developments)

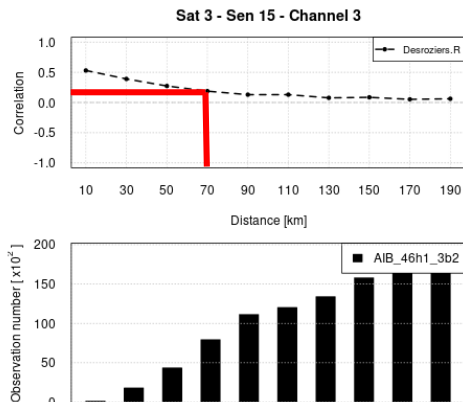
**AIM:** To set the thinning distances applied to high spatial density data in accordance with estimated observation error correlation length scales. The spatial thinning is applied both to limit data amounts and to compensate for our current lack of representation of spatial observation error correlations.

**HOW:** Based on DA feedback statistics files, innovations are separated into observation error correlations and background error correlations. From plots of the observation error correlation part, appropriate thinning distance is estimated with distance when the observation correlation drop to 0.2.

### Example for satellite MHS channel 3 data

Derived observation error correlation as function of distance between data pairs.

Number of data in each bin as function of distance between data pairs.



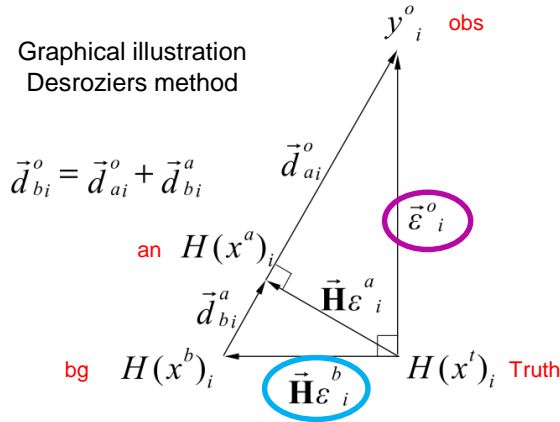
# Diagnostic methods

## 3- Desroziers

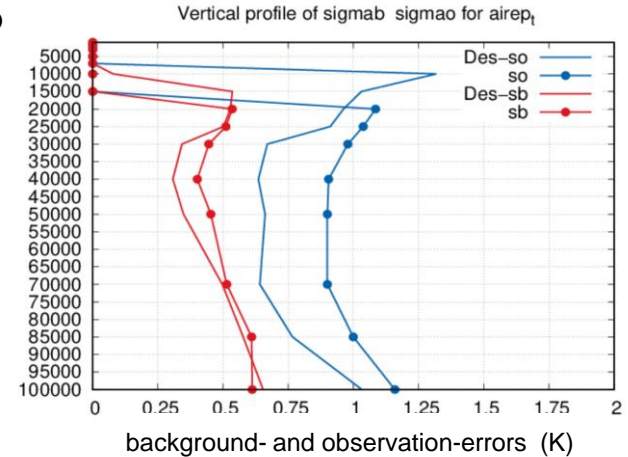
Q. J. R. Meteorol. Soc. (2005), 999, pp. 1–999

**AIM:** To compare used background- and observation-errors with theoretical ones calculated by Desroziers method and exploit if revisions of error standard deviation specifications needed.

**HOW:** Use DA feedback statistics of residuals and innovations from parallel cy46 evaluation experiments. Investigate plots of the current prescribe and the by Desroziers method suggested observation and background error values.



Example for ABO  
temperatures



- Recommended adjustments

# Recommended adjustments

## Background check rejection limits

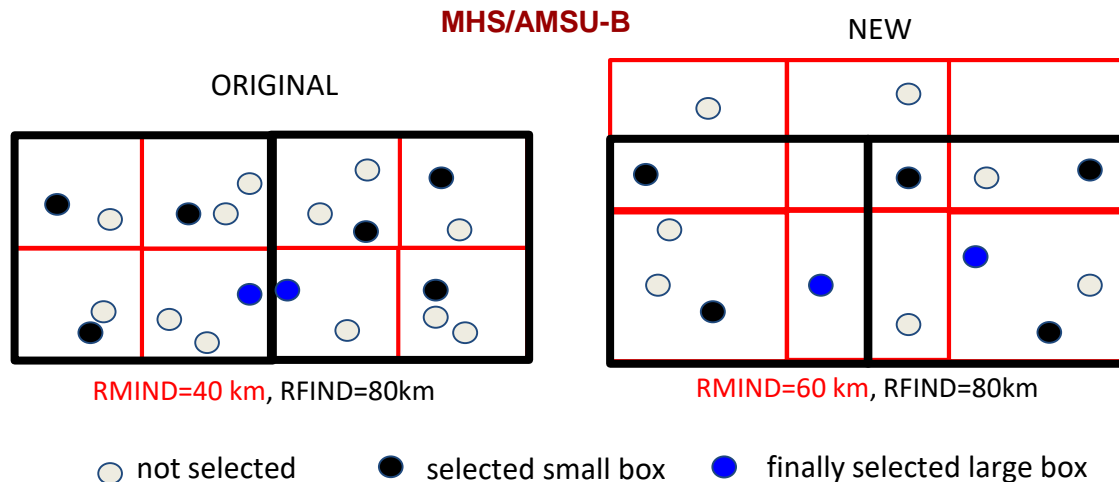
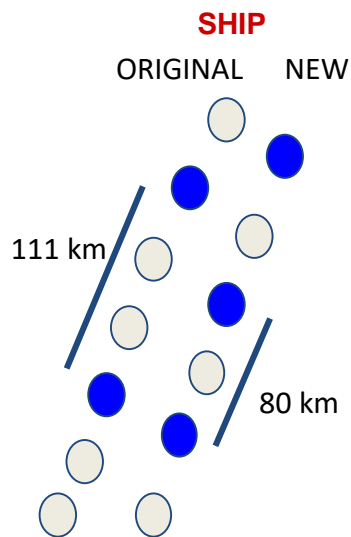
default is 5 times  $\sqrt{\sigma_{b,i}^2 + \sigma_{o,i}^2}$

PARAMETER /CONFIGURATION	MetCoOp		UWC-W		AEMET	
	Before	After	Before	After	Before	After
GNSS ZTD	4	3	4	3	4	3
RADAR RH	4	3	4	3	4	3
MHS	4	3	4	3	4	3
ATMS	4	3	4	3	4	3
MWHS-2	4	3	4	3	4	3

General reduction of L value for these observations

# Recommended adjustments

## Thinning distances



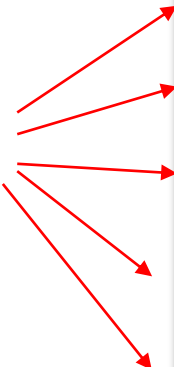
Note that likelihood two close observations less with 60 km/80 km (overlapping boxes)!

# Recommended adjustments

## Background and observation error standard deviations in minimization

Decrease sigmao

PARAMETER /CONFIGURATION	MetCoOp Before/After		UWC-W Before/After		AEMET Before/After	
REDNMC	0.6	0.63	0.9	0.85	0.9	0.85
SIGMAO_COEF TEMP T	0.90	0.81	0.90	0.81	0.90	0.81
SIGMAO_COEF TEMP u/v	0.90	0.81	0.90	0.81	0.90	0.81
SIGMAO_COEF SYNOP Z	1.0	0.9	1.0	0.9	1.0	0.9
SIGMAO_COEF ASCAT	0.9	0.86	0.9	0.86	0.9	0.86
SIGMAO_COEFF RADAR WIND	1.0	0.95	1.0	-	1.0	-

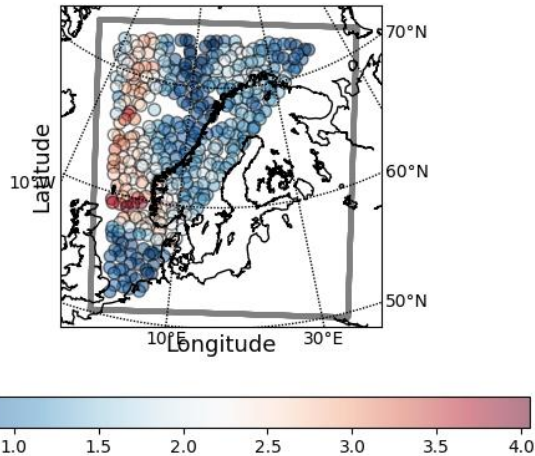


- Limitations revealed by BGOS and how to alleviate

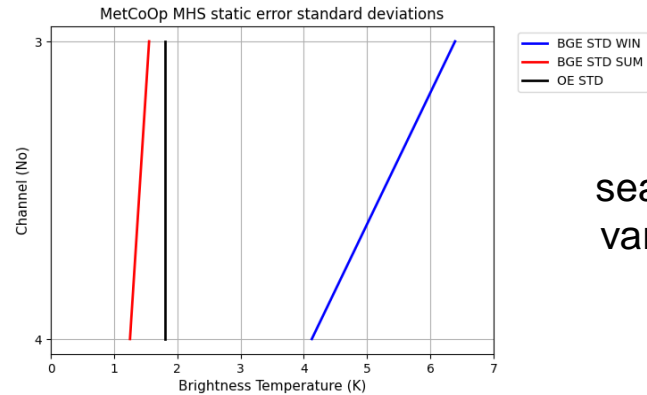
# Limitations revealed by BGOS and how to alleviate

Application of BGOS revealed **strong spatial and seasonal variation** for **humidity sensitive satellite radiances** and **radar relative humidity** background errors in observation space (despite background errors in model space constant). See example below for MHS radiances.

spatial  
variation



MHS channel 3 background error standard deviations in observation space 15 July 2022, 00 UTC. MetCoOp domain.



seasonal  
variation

*Background errors in MHS channel 3-4 observation space with standard operational MetCoOp B matrix*

*Black curve observation error, Blue winter background error in MHS ch 3-4 space and Red summer background error in MHS ch 3,4 space.*

# Limitations revealed by BGOS and how to alleviate

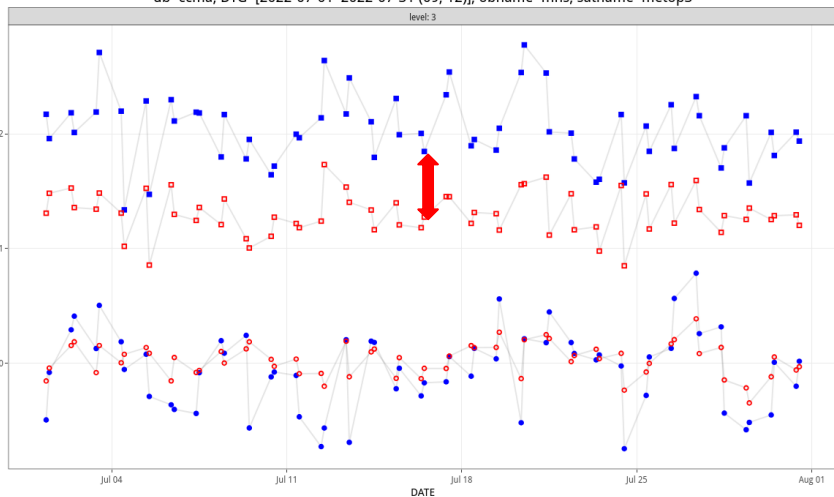
## MetCoOp operational observation monitoring

### Metop C MHS channel 3 observation fit statistics time-series

#### Summer (July 2022)

MEPS-mbr000: ObsFit

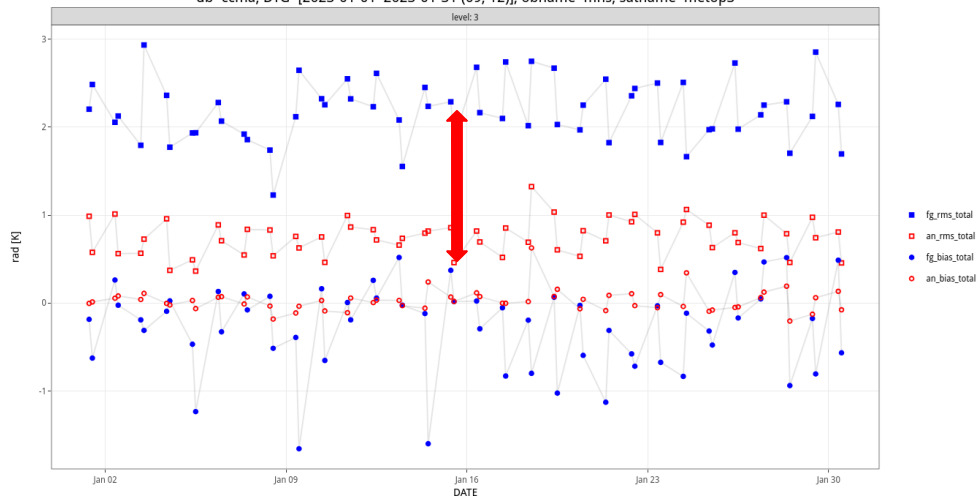
db=ccma, DTG=[2022-07-01-2022-07-31 (09, 12)], obname=mhs, satname=metop3



#### Winter (January 2023)

MEPS-mbr000: ObsFit

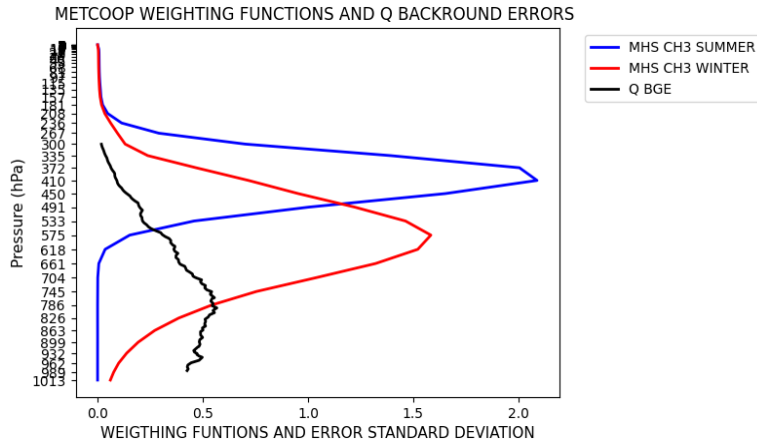
db=ccma, DTG=[2023-01-01-2023-01-31 (09, 12)], obname=mhs, satname=metop3



More weight to MHS observations during winter than during summer!

# Limitations revealed by BGOS and how to alleviate

## The reason for spatial and seasonal variation for MHS



**MHS channel 3 summer condition weighting functions**  
**MHS channel 3 winter condition weighting functions**  
**MetCoOp vertical profile of seasonally and spatially averaged used vertical profile of unbalanced background error specific humidity profile.**

In cold and dry conditions (winter-time) the radiative transfer weighting function peak lower in the atmosphere and where the seasonally averaged humidity standard deviations are larger. Then, in cold conditions/winter-time we have larger background errors in observation space and therefore larger impact of observations.

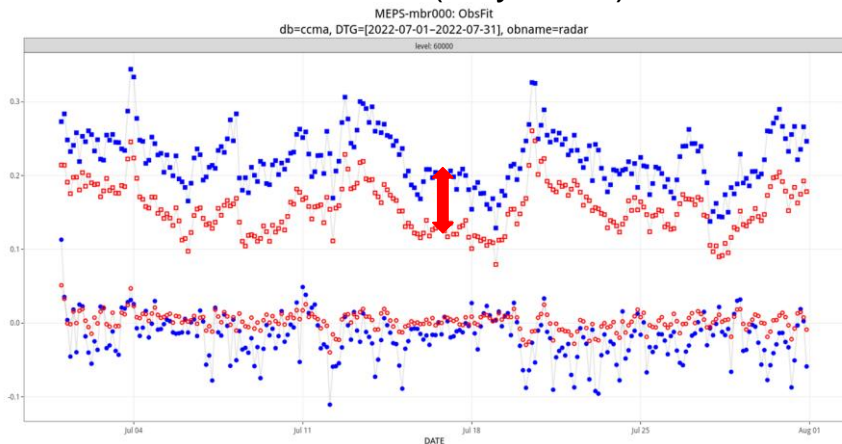
# Limitations revealed by BGOS and how to alleviate

And same for Radar RH

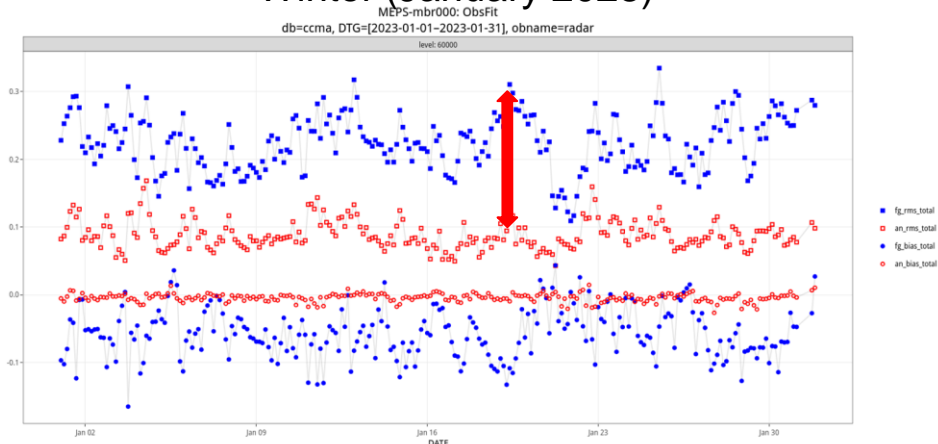
MetCoOp operational observation monitoring

RADAR relative humidity observation fit statistics time-series

Summer (July 2022)



Winter (January 2023)



Now the reason is again that bg error standard deviations are specified in terms of climatological profiles for unbalanced specific humidity. When they are transformed into errors in the space of Relative Humidity, the errors will be larger when the temperature is high in cold areas and during wintertime (them saturation water vapor smaller).

Note larger impact in winter! rms of ob-an is much smaller in winter

# Limitations revealed by BGOS and how to alleviate

## How to alleviate this artefact on humidity analysis??

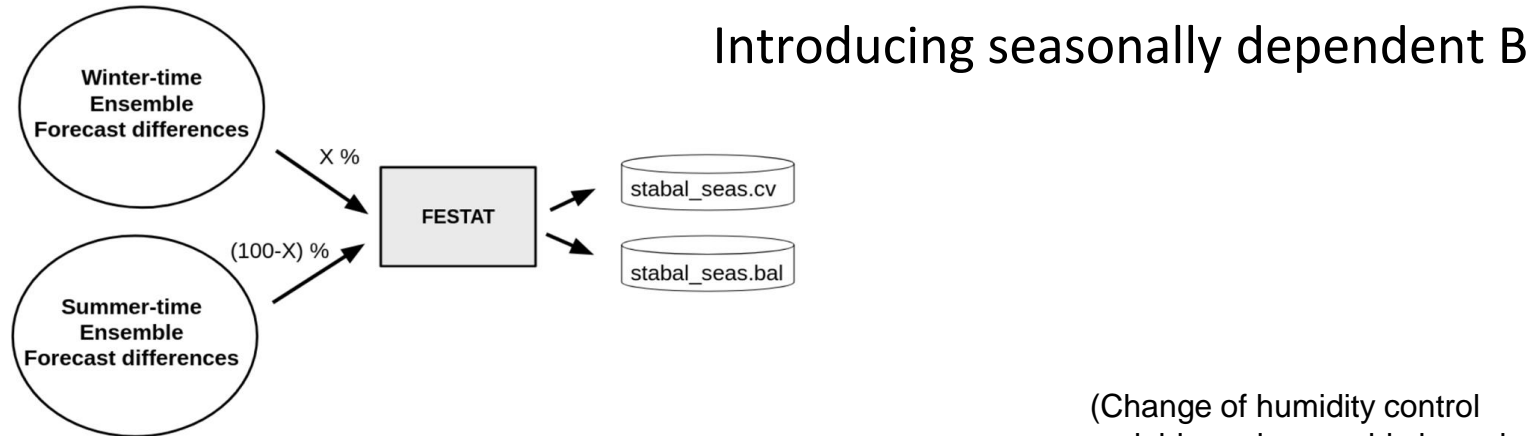
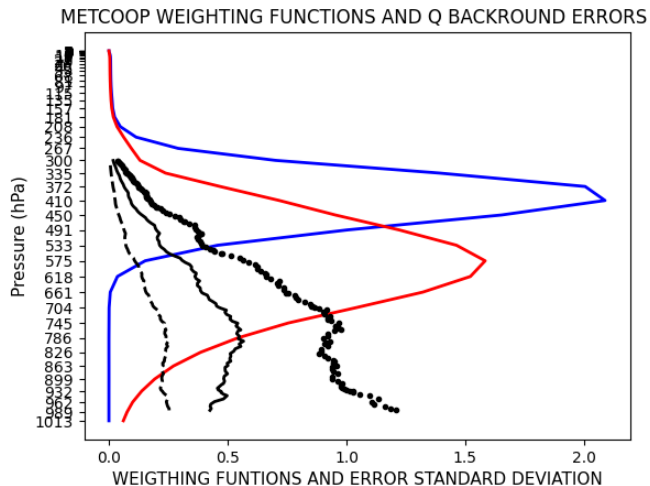


Figure C1. Illustration of proposed mix of seasonal input data for FESTAT. Winter:  $X=100$  and  $seas=win$ , Summer  $X=0$  and  $seas=sum$ , Autumn, Spring:  $X=50$  and  $seas=aut, spr$ .

(Change of humidity control variable and ensemble based background errors also solve)

# Limitations revealed by BGOS and how to alleviate

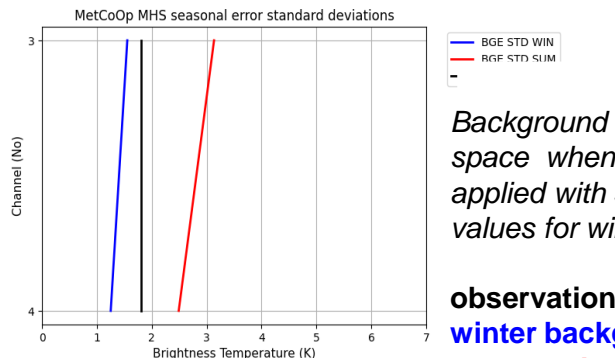
## First illustration effect of introducing seasonally dependent B



**MHS channel 3 summer condition weighting functions**

**MHS channel 3 winter condition weighting functions**

- All season humidity background error
- Winter humidity background error
- Summer humidity background error



*Background errors in MHS channel 3-4 observation space when a seasonally dependent B matrix was applied with summer errors for summer case and winter values for winter case.*

**observation error**

**winter background error in MHS ch 3-4 space**  
**summer background error in MHS ch 3-4 space.**

# Summary and conclusions

- HARMONIE-AROME Cycle 46 has been subject to an extensive evaluation of tunable settings for the three operational domains.
- After a detailed analysis, we conclude that everything looked good but just some minor revisions were proposed.
- The application of BGOS revealed a weaknesses due to our current handling of humidity in data assimilation.
- This work resulted in preparation of tuning tools to be available on ACCORD DA tools.

Thank you for your attention!

[jsancheza@aemet.es](mailto:jsancheza@aemet.es)