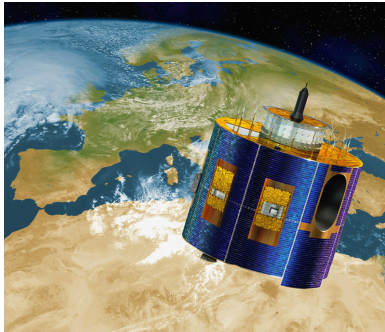


# Satellite bias correction for MSG



Patrik Benáček

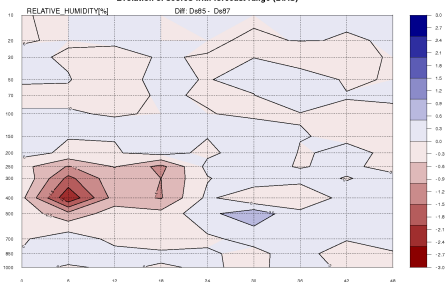


Czech Hydrometeorological Institute  
Department of Meteorology and Environment Protection



## Impact of MSG-WV channels (RH BIAS)

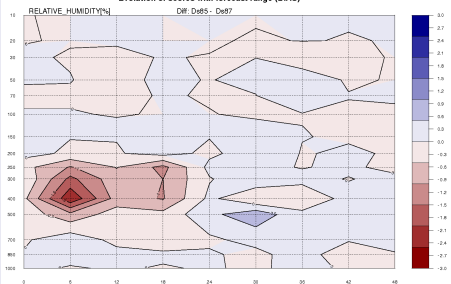
Evolution of scores with forecast range (BIAS)



- more humidity/clouds
- period (20.-30.3.2011)

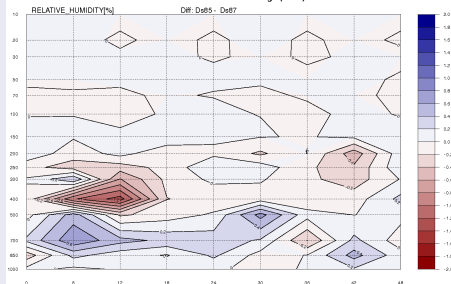
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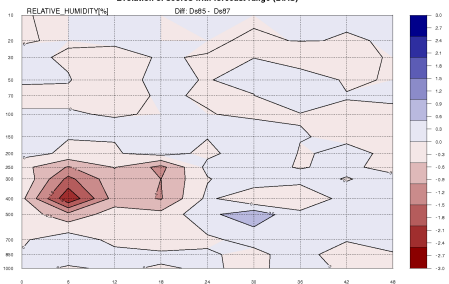
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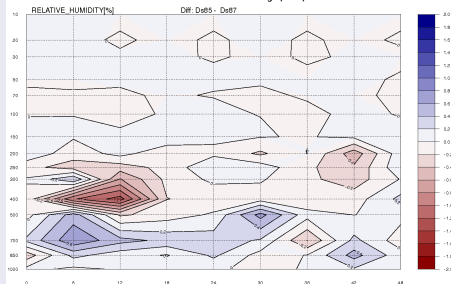
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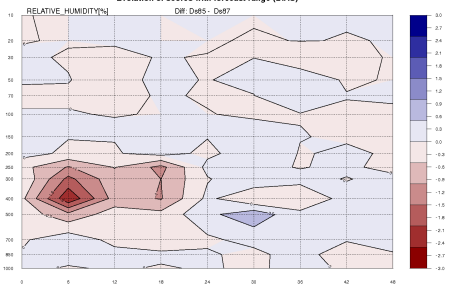


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Why we get the negative impact for WV-channels?

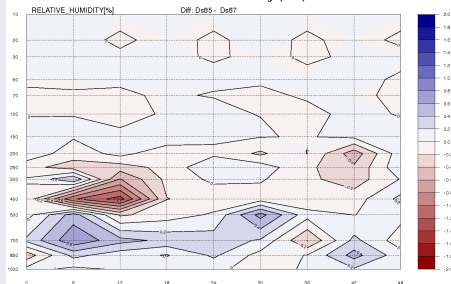
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Why the impact depends on the period?

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- NOT constant – vary with time, geographical position, scan position of the instrument or air-mass dependent
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## Mixture of model and satellite bias

$$BIAS = BIAS_{sat} + BIAS_{model}$$



# Identifying/separating sources of bias

- NOT to separate leads to:
  - re-enforce the model bias
  - degrade the analysis fit to other observation
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## Outline/main goals:

- study model and satellite bias separately
- examine new parameters for satellite bias correction
- application of the improve bias correction in the assimilation system (3DVar)

- 1 Model bias
  - Radiosonde bias
  - Model bias
- 2 Satellite bias
  - Pre-processing
  - Air-mass dependency
- 3 Results
  - Impact of WV channels
- 4 Conclusion

# How to separate model/satellite bias?

- cross validation method against radiosonde (RS) data

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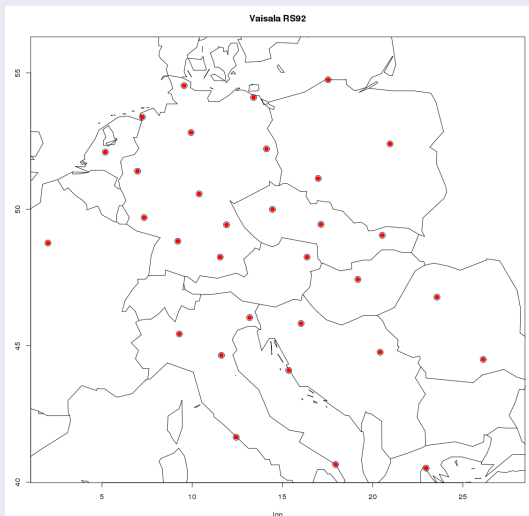
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$$BIAS_{sat+model} = \langle y_{SAT} - h(x) \rangle$$

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<b>Atmosphere</b>	<b>Nighttime</b>	<b>Daytime</b>
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Detection using selected RS data:

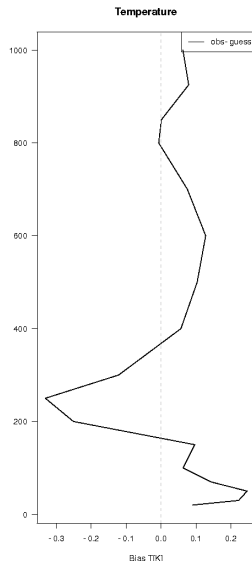
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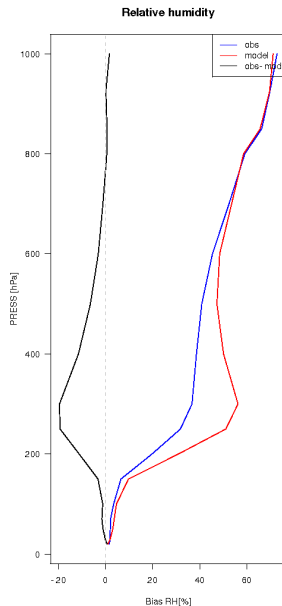
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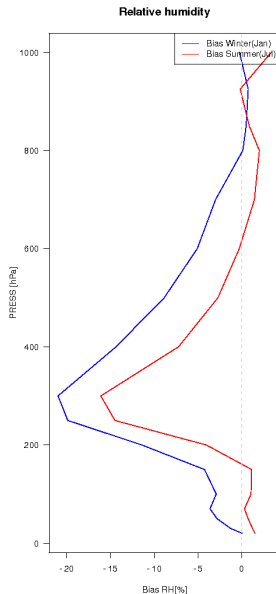
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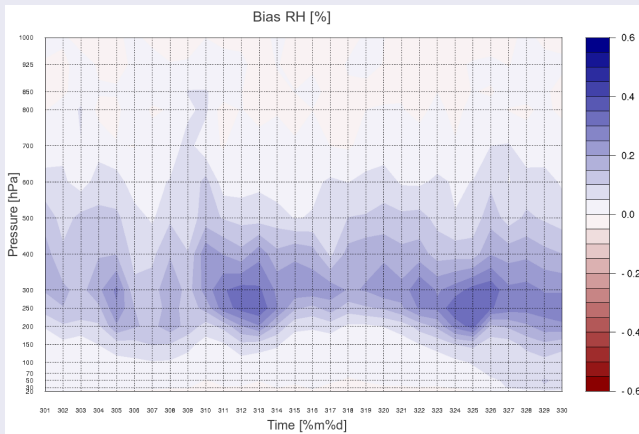


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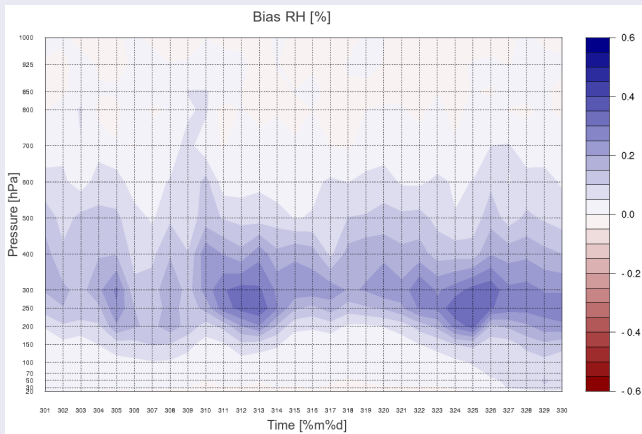
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  - **daily**: synoptic situations

## Daily varies of model RH bias





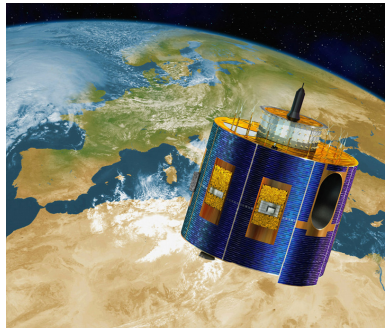
## Daily varies of model RH bias



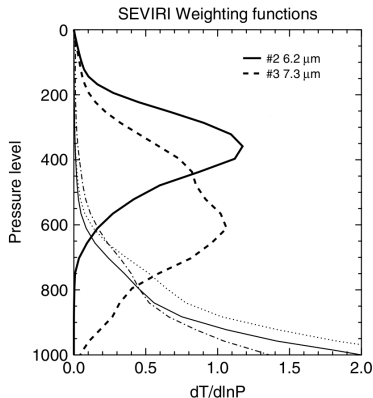
- increasing model RH bias in MT/UT (up to 20%)
- seasonal/daily/local RH bias variation

# Satellite data

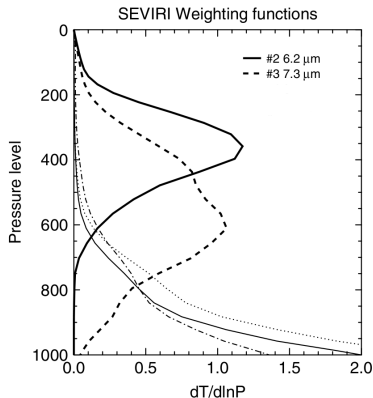
- the channels WV6.2 and WV7.3 from SEVIRI on board MSG
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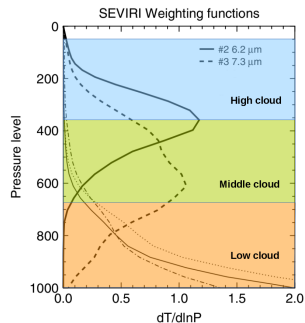
## Bias sources:

$$BIAS_{sat} = \sigma_{instr} + \sigma_{rttov} + \sigma_{cloud} + \sigma_{airmass}$$

- $\sigma_{instr}$  - satellite instrument error (calibration, environmental effects, etc.)
- $\sigma_{rttov}$  - radiative transfer model error (spectroscopy, non-modelled processes)
- $\sigma_{cloud}$  - data pre-processing error (cloud/precipitation detection)
- $\sigma_{airmass}$  - air-mass characteristics dependency (T, Q, etc.)

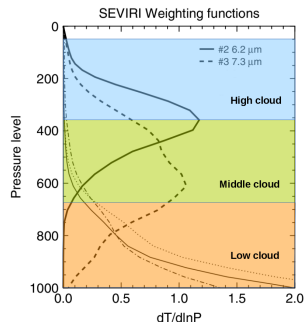
# Cloud contamination of radiation

- absorbing of radiation by high clouds (above the weighting functions)



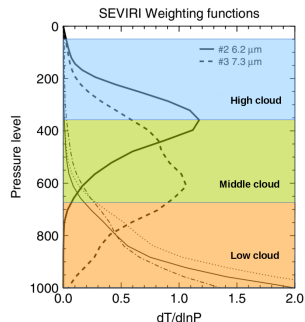
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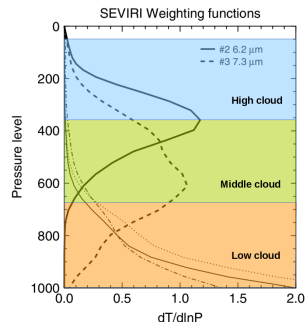
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## Bias contamination by clouds (mean per 4 months)

Chan/CT	Cloud free 1	Very low 6	Low 8	Middle 10	High 12	Very high 14
<b>WV6.2</b>	-3.1 ± 1.5	-2.7 ± 1.7	-2.8 ± 1.7	-2.5 ± 1.7	-6 ± 3	-14 ± 3
<b>WV7.3</b>	-1.1 ± 1.2	-1.2 ± 1.3	-1.9 ± 1.4	-4.3 ± 2.4	-15 ± 5	-30 ± 4

CT selection: used(green), rejected(yellow/red), investigated(yellow)

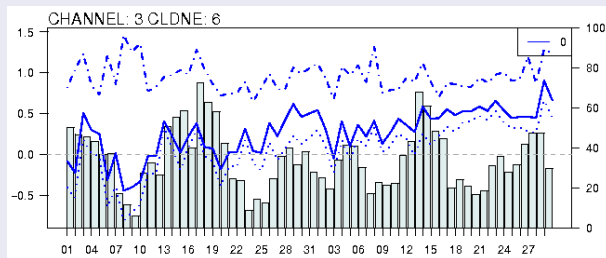
## Satellite/model bias summary (in WV radiation spectra)

Chan/Bias[K]	<b>BIAS<sub>sat+model</sub></b>	<b>BIAS<sub>sat</sub></b>	<b>BIAS<sub>model</sub></b>
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- monitoring of  $BIAS_{sat+model}$  for 03/2011



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## Detecting method

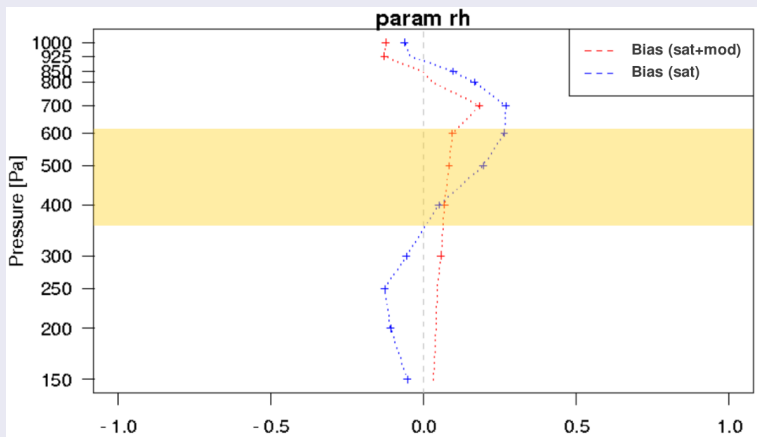
- using Spearman correlation method (not sensitive to outliers/non-normality):

$$BIAS_{sat} \sim (G, T, q, RH, u, v, \dots)$$

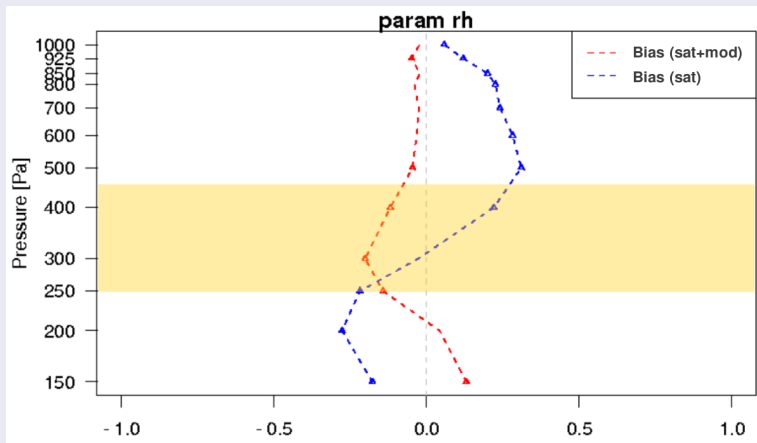
$$BIAS_{sat+model} \sim (G, T, q, RH, u, v, \dots)$$



## Relative humidity $\sim$ WV7.3



## Relative humidity $\sim$ WV6.2



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## Correction method

$$BIAS = \beta_0 P_0 + \beta_1 P_1 + \beta_2 P_2 + \beta_3 P_3$$

- $P_0$  – constant
- $P_1$  – thickness of layer 1000-300hPa
- $P_2$  – thickness of layer 200-50hPa
- $P_3$  – the total column water vapor

# Air-mass correction results

## Set of bias parameters for WV channels:

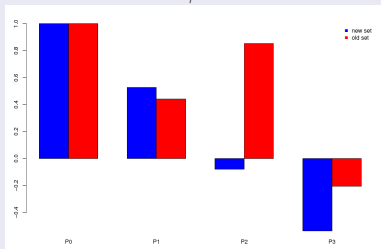
Exp	Chan	$\beta_0$	$\beta_1$	$\beta_2$	$\beta_3$
<b>BIAS<sub>sat</sub></b> (new set)	WV6.2	-2.90	-1.53	0.23	1.56
	WV7.3	-0.83	-0.42	0.19	0.98
<b>BIAS<sub>sat+model</sub></b> (old set)	WV6.2	-0.34	-0.15	-0.29	0.07
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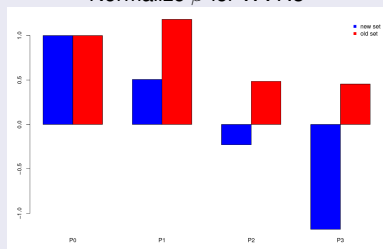
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Normalize  $\beta$  for WV6.2



Normalize  $\beta$  for WV7.3

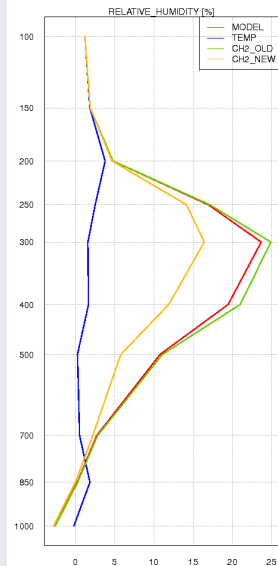




Testing for 20/03/2011:

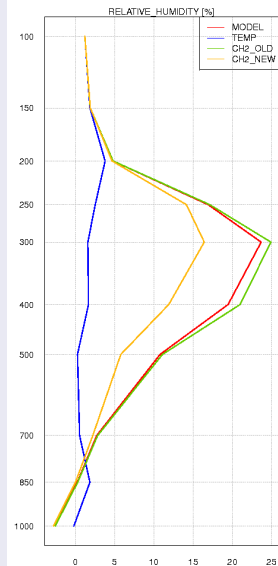
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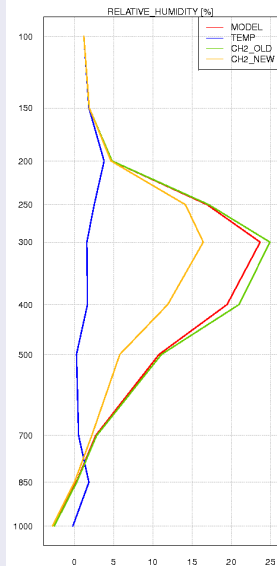
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- testing for 20/03/2011:
  - spurious oscillation in  $Q, T$  analysis (not detected)
  - spin-up oscillation of  $p_s$  tendency (not detected)



## Experiment settings:

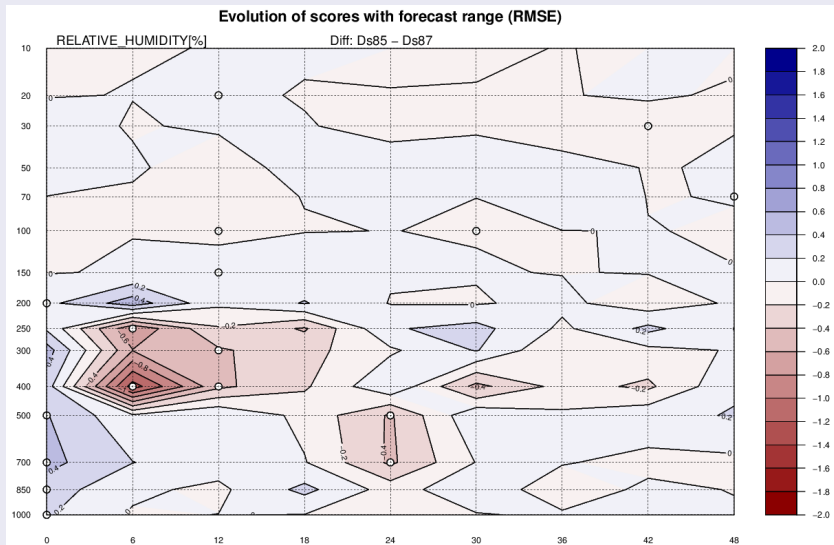
- testing periods 1.3.-30.3.2011 at 00UTC
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- ensemble B-matrix; REDNMC=2.
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  - **TEMP** (s85) – reference
  - **TEMP+MSG(2,3)\_new** (s86) – new set of  $\beta$  for WV6.2/WV7.3
  - **TEMP+MSG(2,3)\_old** (s87) – old set of  $\beta$  for WV6.2/WV7.3

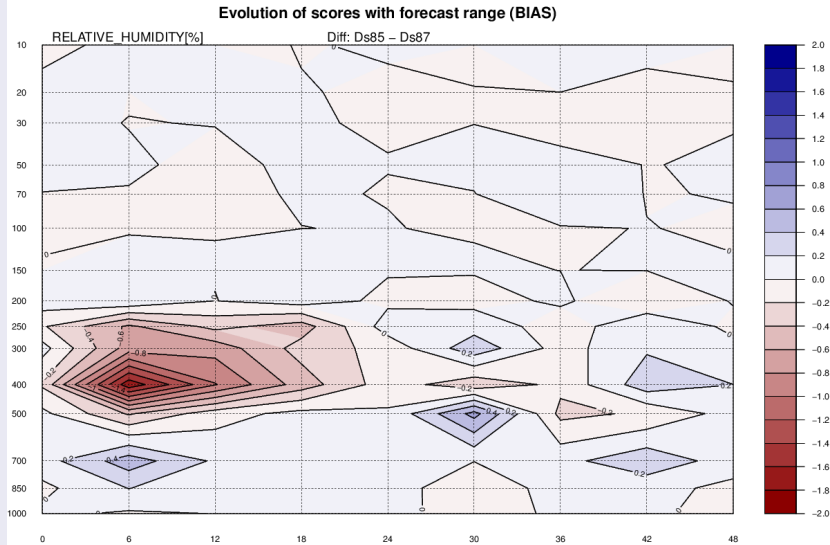
# Impact of new/old scheme on forecast

## Old schema



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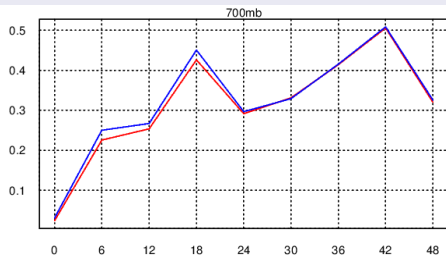
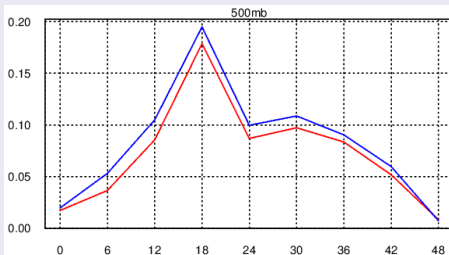
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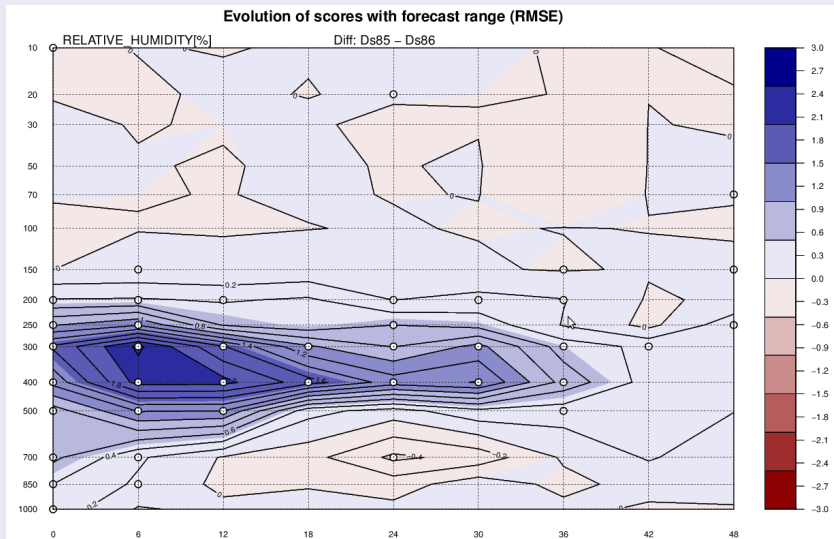


- Temperature bias [K]

— Ds85  
— Ds87

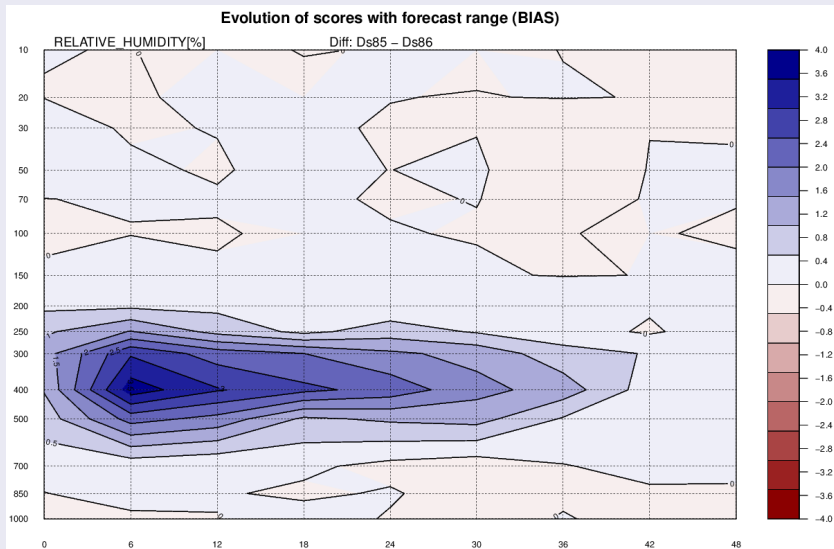
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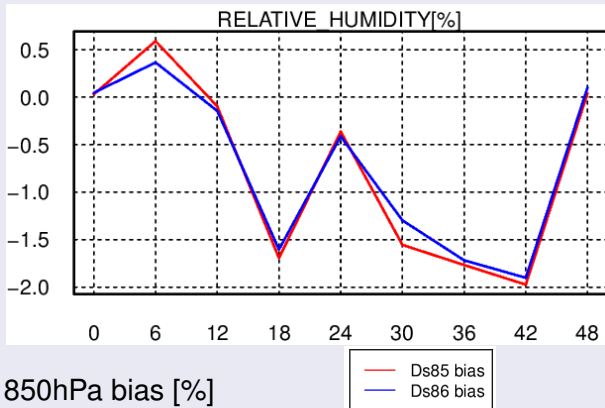
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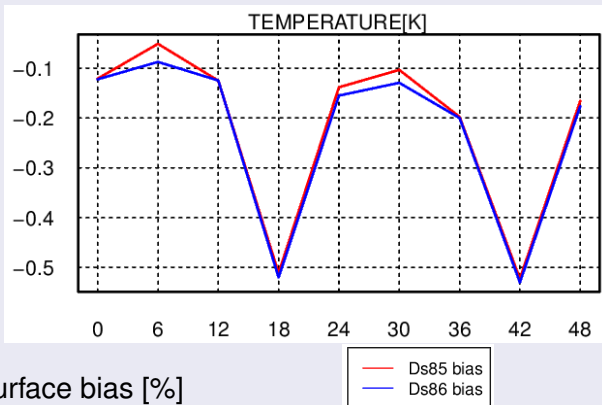


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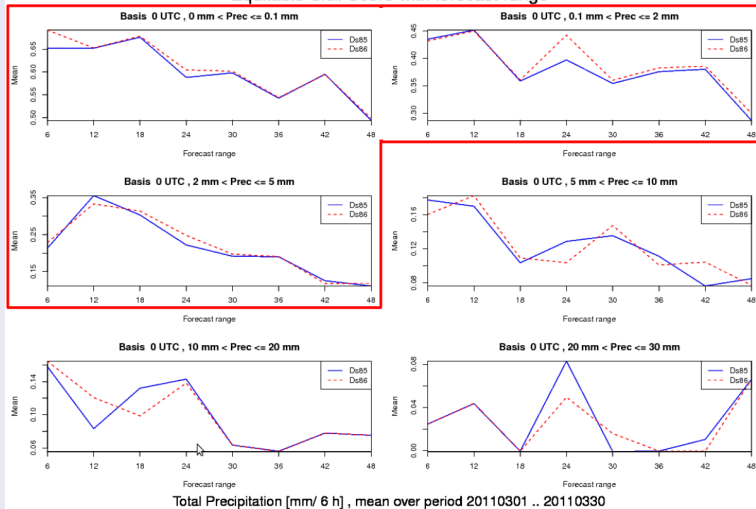


- T surface bias [%]

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### Equitable Skill Score with forecast range



## Summary:

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- application in assimilation system

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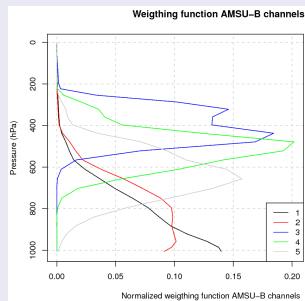
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- more ideas??

## Consequences:

- potential problem with the high peaking  $H_2O$  absorption bands channels (sensor AMSU-B, MHS, IASI)



-  *M. Miloschevich, Holger Vommel, David N. Whiteman, Thierry Leblanc 2009 Accuracy assessment and correction of Vaisala RS92 radiosonde water vapor measurements.*  
Journal of geophysical research, VOL. 114, D11305
-  *T. McNally 2005 Introduction to bias estimation and correction for satellite data assimilation.*  
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Thank you for your attention.