

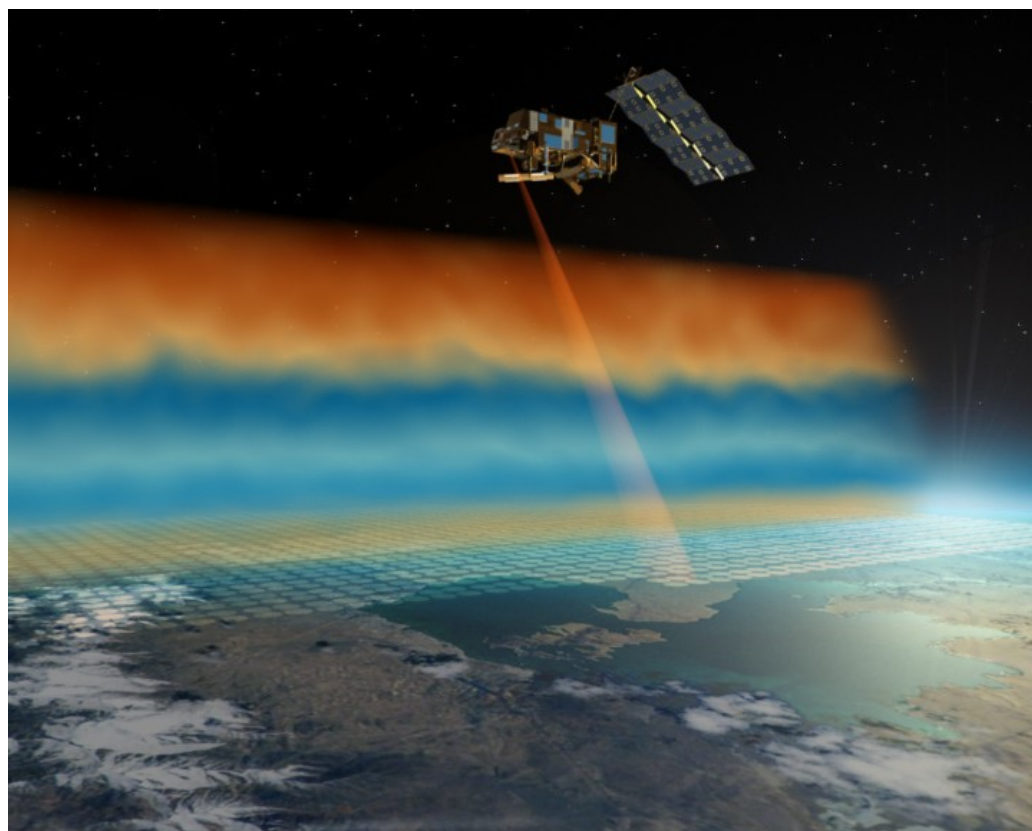
Latest results of variational bias correction in LAM DA systems

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DAWD, Ljubljana

19.9.2017



Bias correction scheme is essential for a satellite data assimilation into NWP models.

Observation bias:

Bias prediction model:

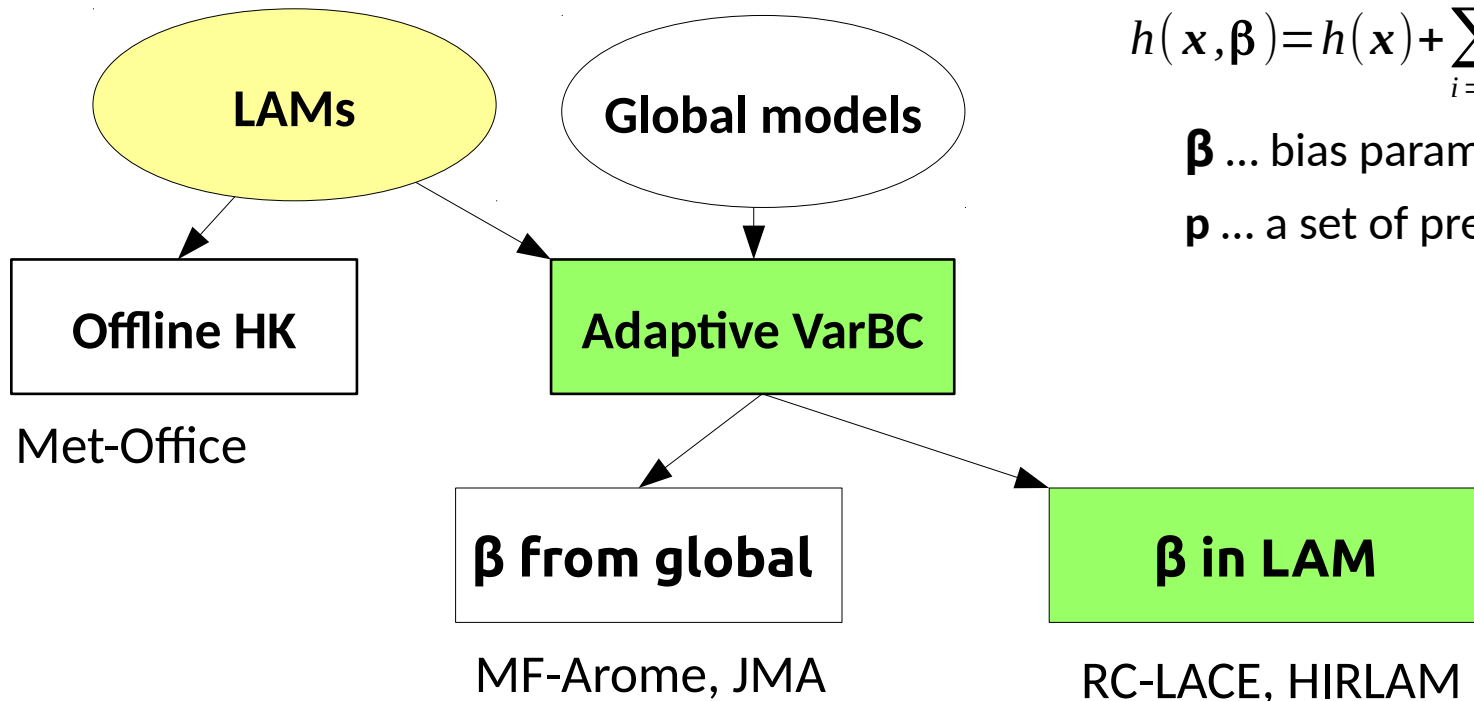
$$b_0 = \langle \mathbf{y} - h(\mathbf{x}) \rangle$$

$$b_0 = \langle \mathbf{y} - h(\mathbf{x}, \boldsymbol{\beta}) \rangle = 0$$

$$h(\mathbf{x}, \boldsymbol{\beta}) = h(\mathbf{x}) + \sum_{i=1}^{N_p} \beta_i p_i$$

$\boldsymbol{\beta}$... bias parameters

\mathbf{p} ... a set of predictors



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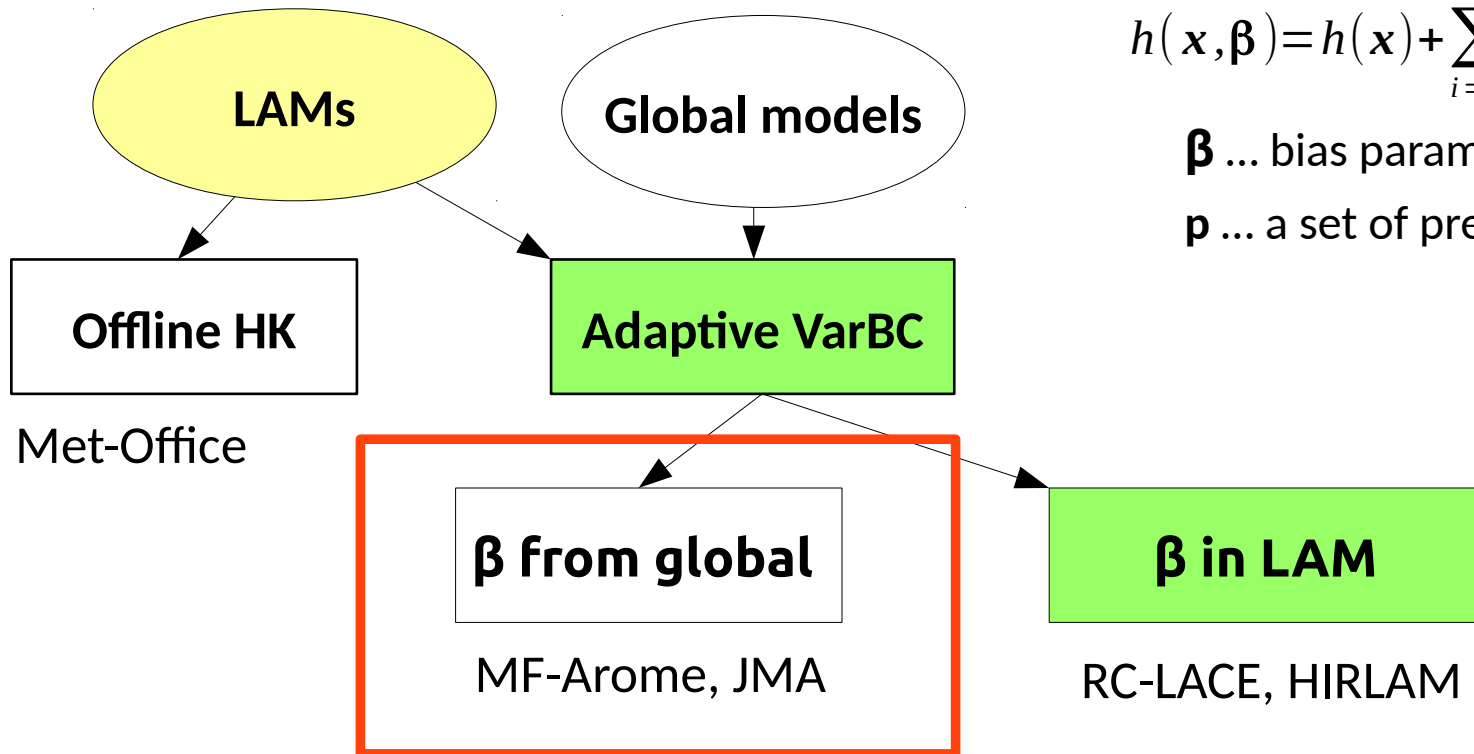
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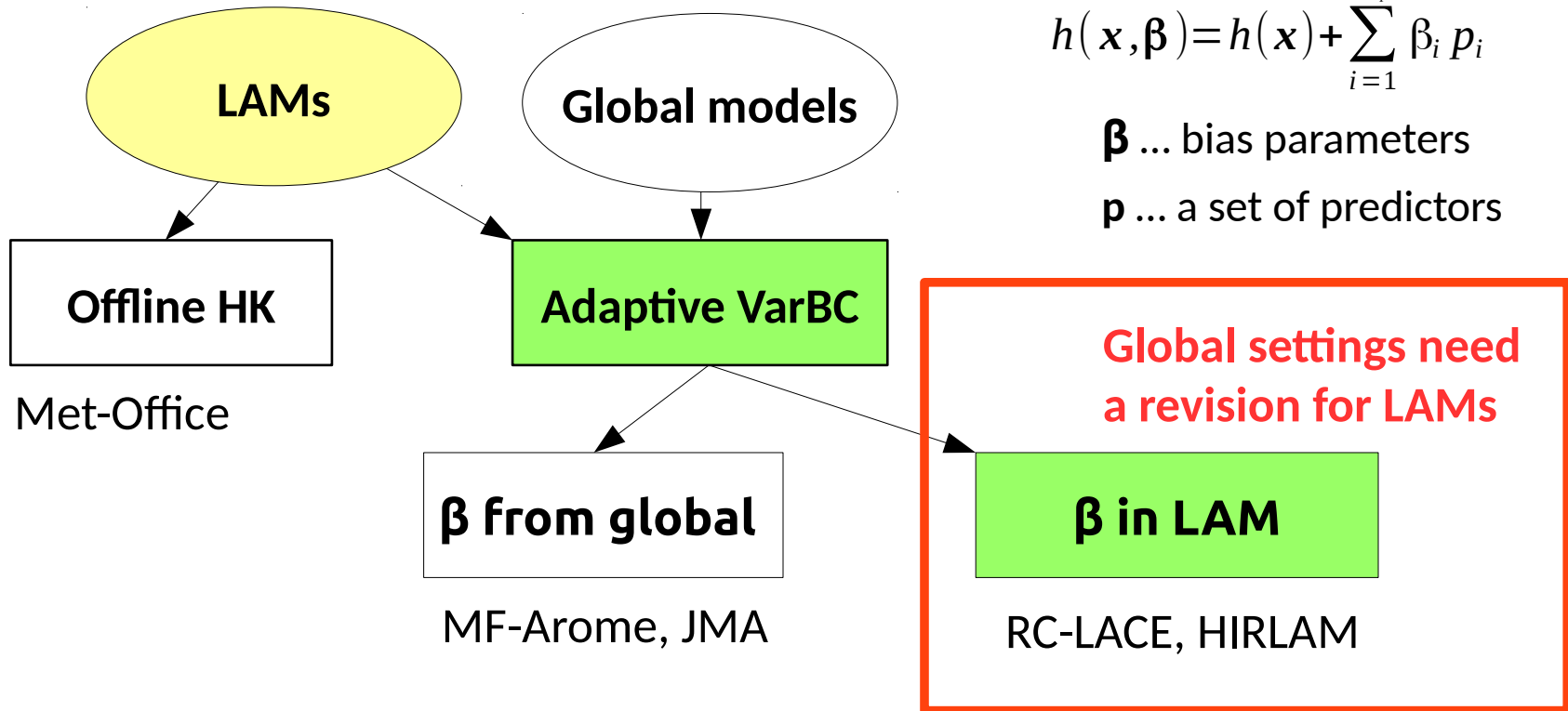
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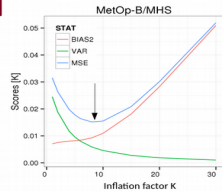


This presentation introduces a behavior of the VarBC scheme in LAMs and suggests a new adaptivity setting.

The optimum adaptivity of VarBC is estimated by a minimization of the MSE of bias parameters.

$$N_{hg} \approx \max(N_{avg}, N_{min}) \left[\frac{2nW \left(\frac{4N_{avg}n^2b_{max}^2}{V^2\sigma_{obs}^2} \right)^{-1}}{K} \right]$$

- W ... Lambert-W function (non-complex)
- N_{min} ... minimum number of observation
- b_{max} ... the maximum observation bias
- n ... spin-up period
- K ... stiffness parameter inflation factor

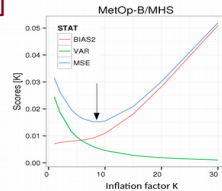


Sampling issues in LAMs

The optimum adaptivity of VarBC is estimated by a minimization of the MSE of bias parameters.

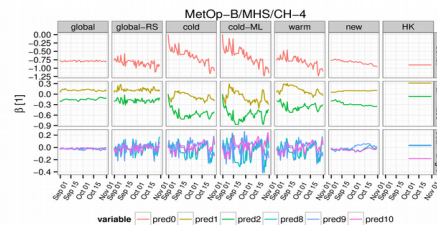
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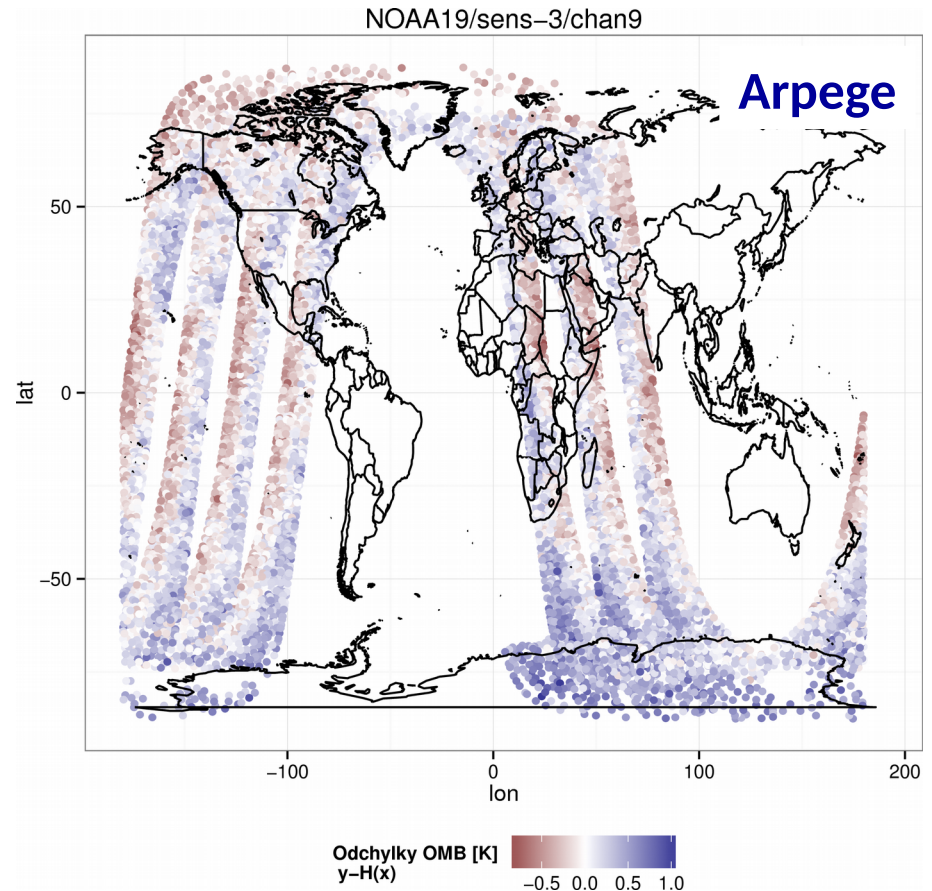
The VarBC scheme and a new adaptivity setting

Initialization of VarBC: new adaptivity approach is able to reduce a contribution of NWP model biases for MHS channels.



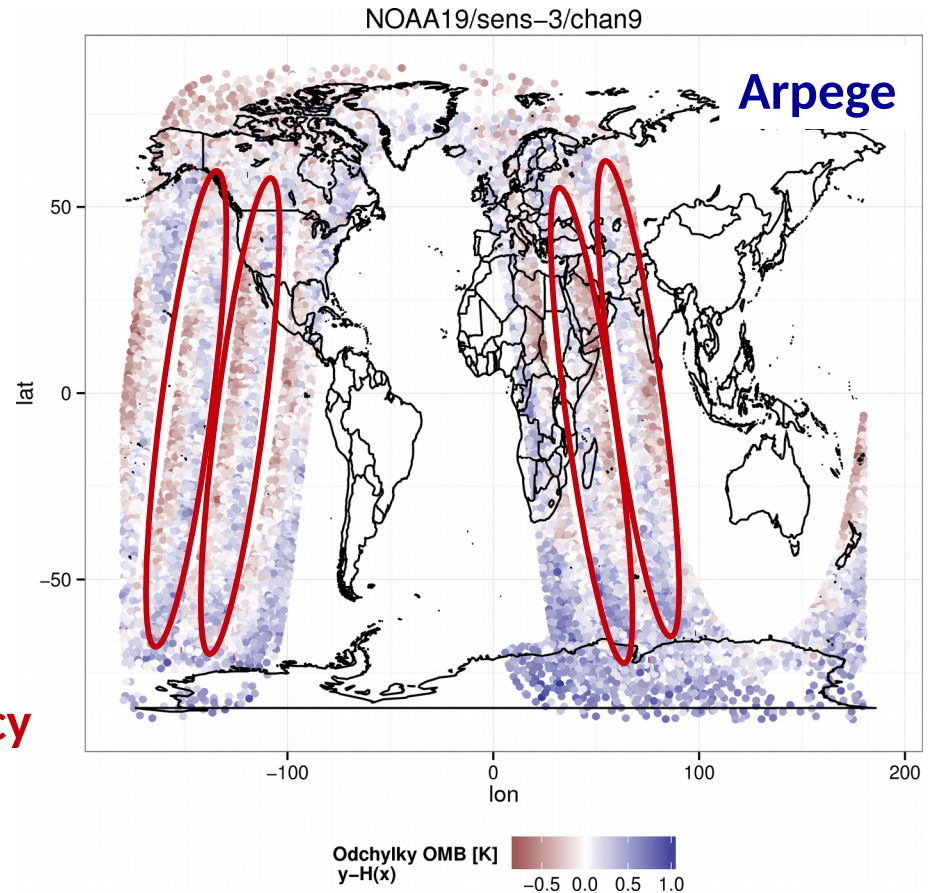
Evaluation of bias correction

Satellite bias detection is statistically meaningful provided normally distributed random sample of observations.



OmG of AMSU-A channel 9 for a particular day

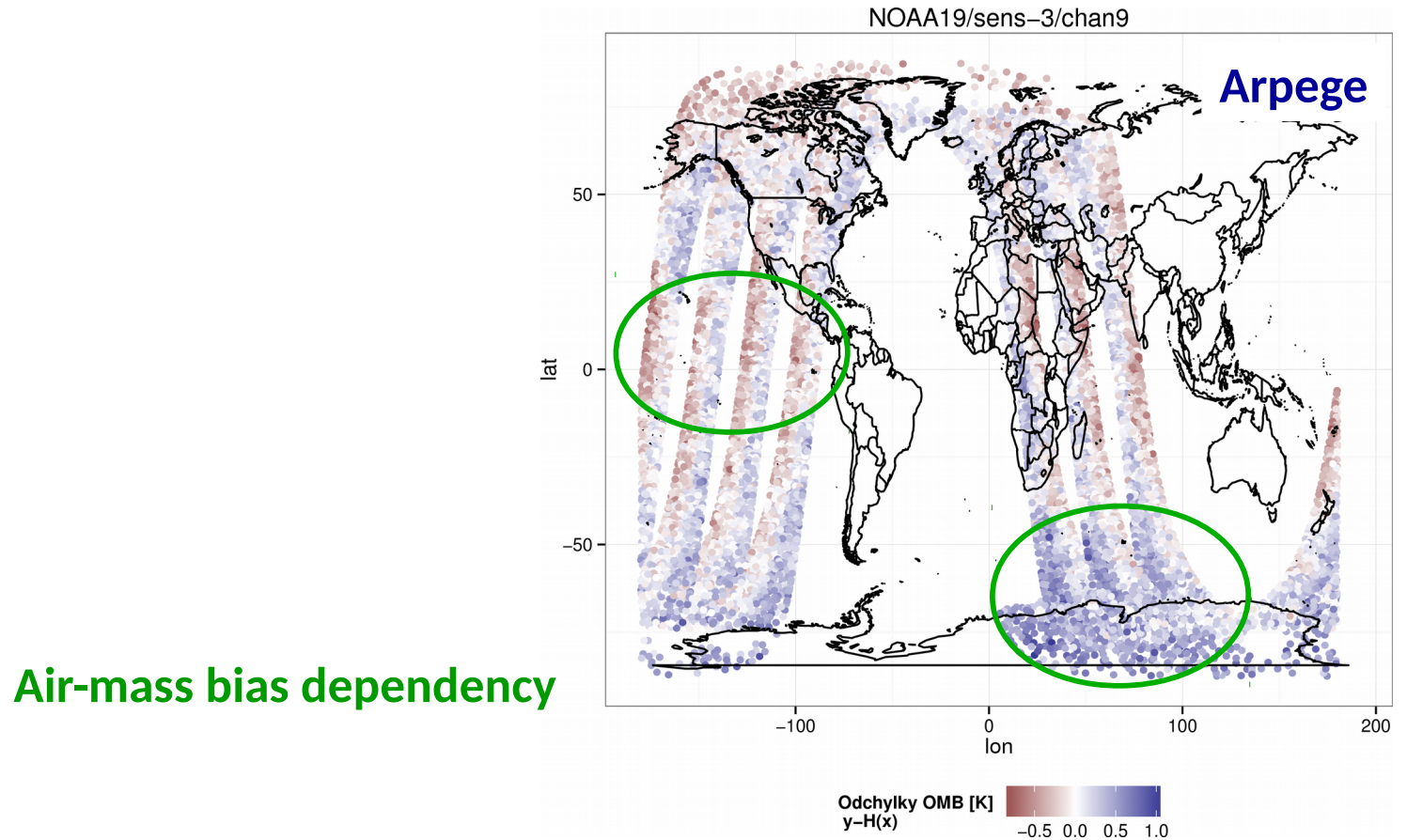
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Satellite scan angle bias dependency

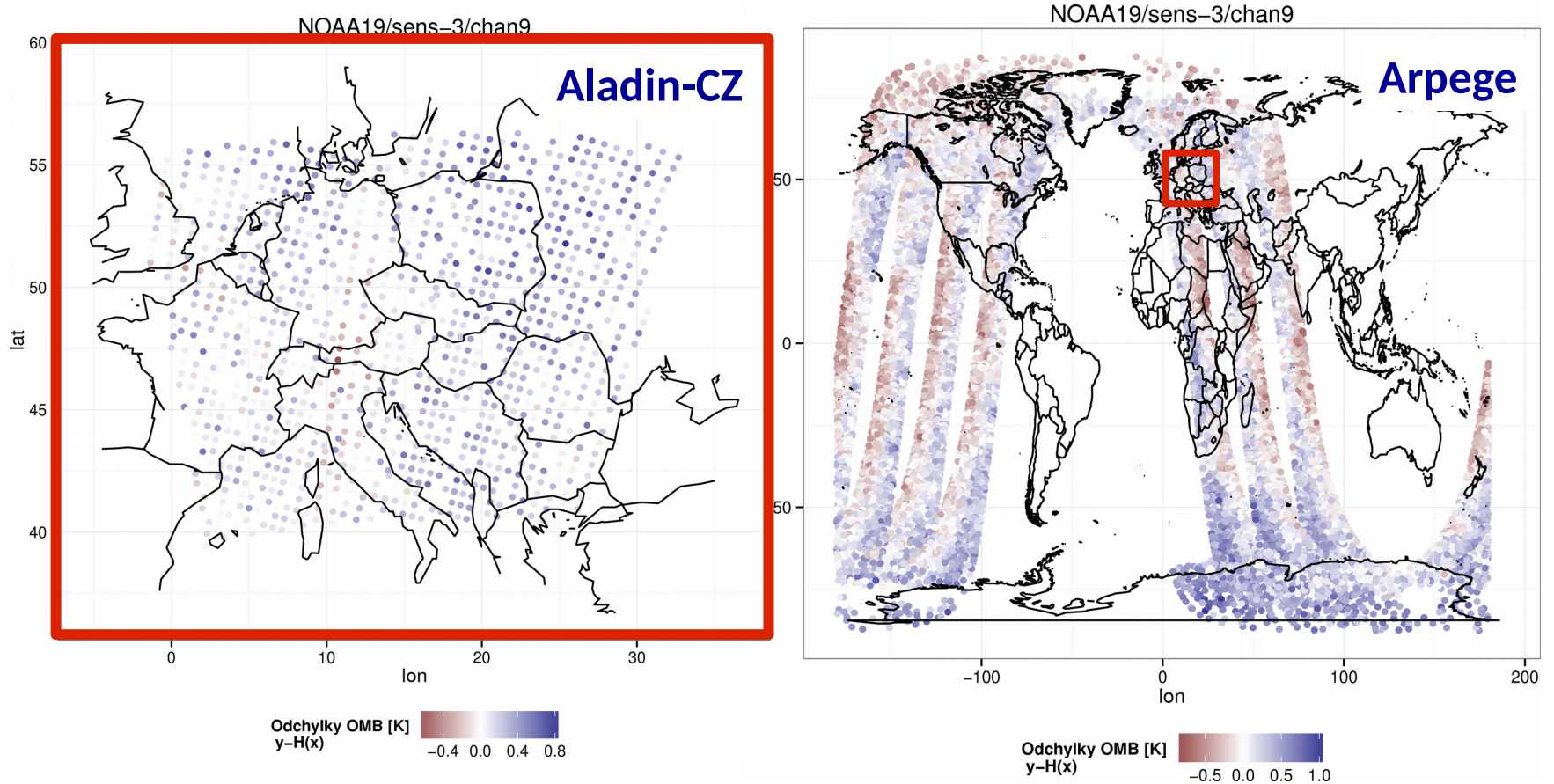
OmG of AMSU-A channel 9 for a particular day

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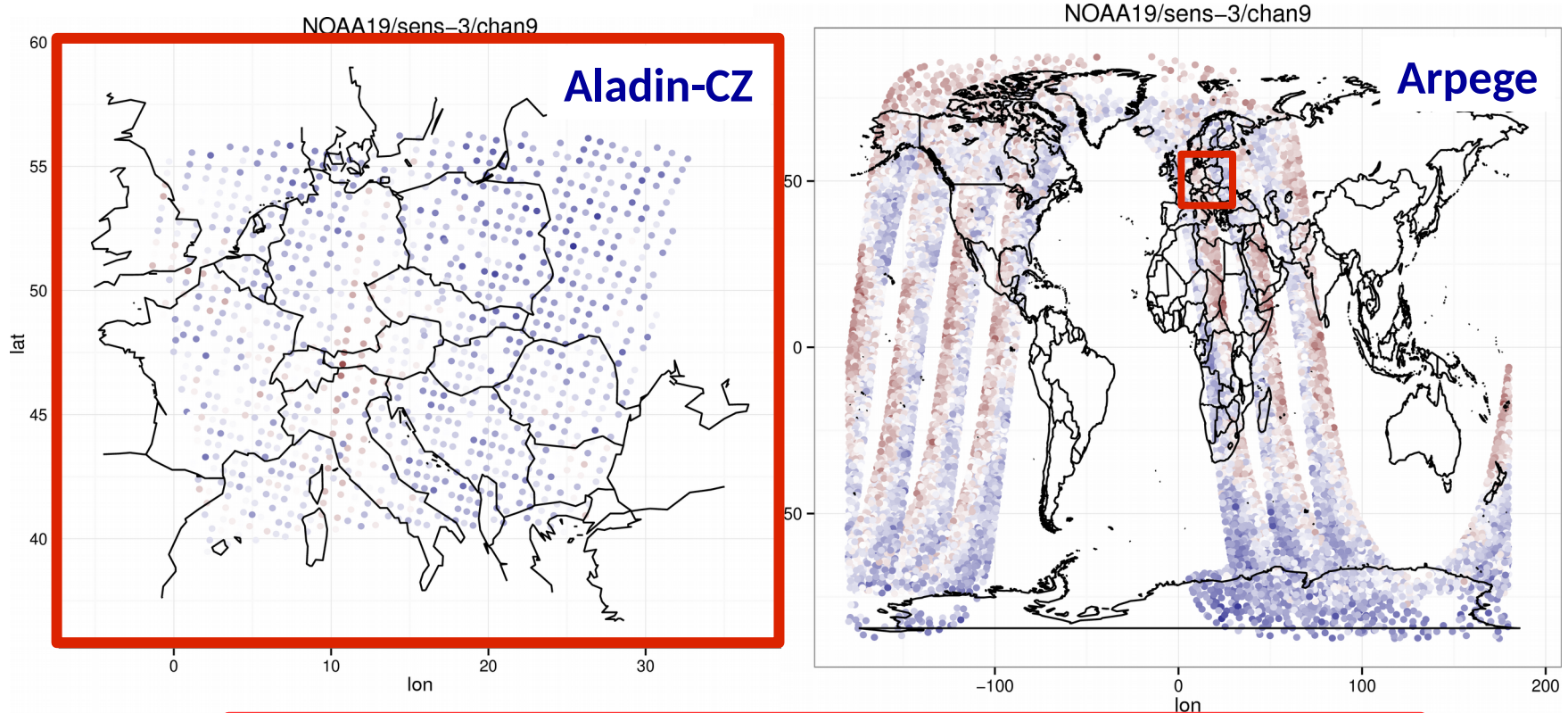


OmG of AMSU-A channel 9 for a particular day

The observation sample is non-uniform, spatially and time dependent in LAMs.



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Two possibilities how to detected a meaningful satellite bias:

- 1) gathering the data over a long-time (the offline method)**
- 2) cycling satellite bias information in time (the VarBC method)**

Sampling variance of the observation bias is negligible in a global model but inflated in LAMs because of the spatially dependent observation sample.

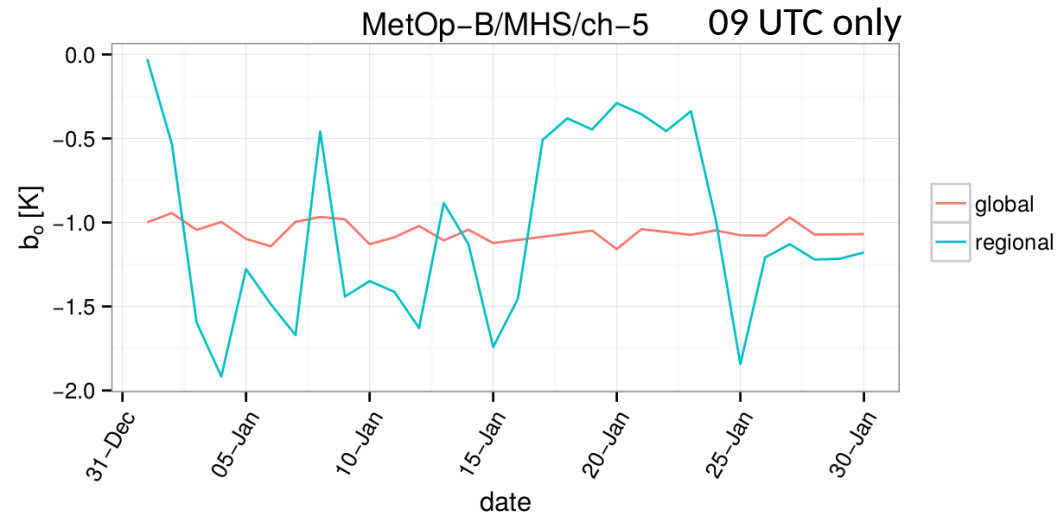
Sampling variance of the mean of observation bias:

Global

$$\sigma_{b_o}^2 = \frac{\sigma_o^2}{N_{avg}}$$

LAM

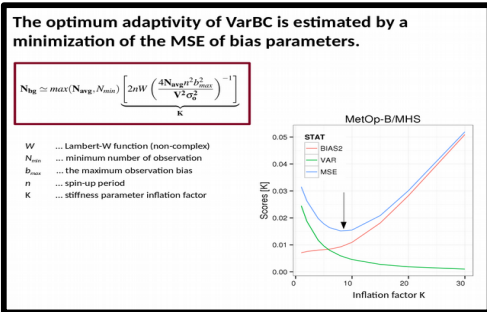
$$\sigma_{b_o}^2 = \frac{\sigma_o^2 \cdot V^2}{N_{avg}}$$



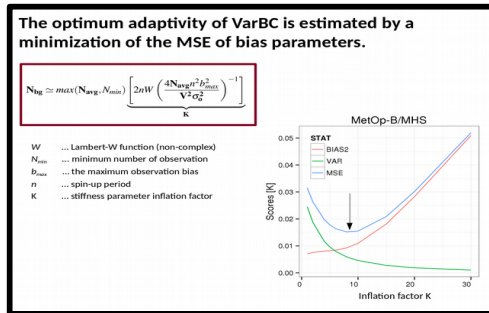
The inflation factor V corresponds to a contribution of NWP model biases:

$$V = \sqrt{\frac{N_{avg}}{\sigma_o^2} \frac{1}{n} \sum_{i=1}^n (b_{oi} - \bar{b}_o)^2},$$

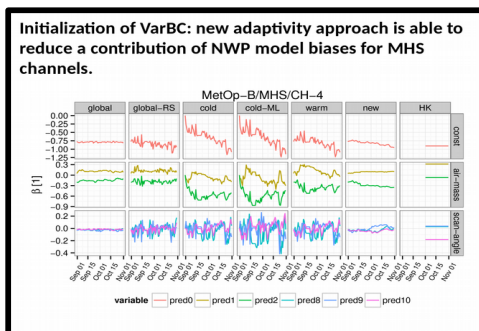
This presentation introduces a behavior of the VarBC scheme in LAMs and suggests a new adaptivity setting.



Sampling issues in LAMs



The VarBC scheme and a new adaptivity setting



Evaluation of bias correction

The VarBC scheme advantages: maintenance, an adaptive bias correction WRT time, observation, analysis.

The VarBC scheme implementation:

$$\begin{aligned} J(\mathbf{x}, \boldsymbol{\beta}) = & (\mathbf{x} - \mathbf{x}_b)^T \mathbf{B}^{-1} (\mathbf{x} - \mathbf{x}_b) \\ & + (\boldsymbol{\beta} - \boldsymbol{\beta}_b)^T \mathbf{B}_\beta^{-1} (\boldsymbol{\beta} - \boldsymbol{\beta}_b) \\ & + (\mathbf{y} - h(\mathbf{x}, \boldsymbol{\beta}))^T \mathbf{R}^{-1} (\mathbf{y} - h(\mathbf{x}, \boldsymbol{\beta})), \end{aligned}$$

$\boldsymbol{\beta}_b$... background bias parameters

\mathbf{B}_β ... background bias parameter error covariance matrix

Diagonal \mathbf{B}_β with elements: $\sigma_{\beta_b}^2 = \frac{\sigma_o^2}{N_{bg}}$ \longrightarrow $\sigma_{\beta_b}^2 = \frac{\sigma_o^2}{N_{bg} V^2}$

N_{bg} ... stiffness parameter (5000) ~ adaptivity of VarBC

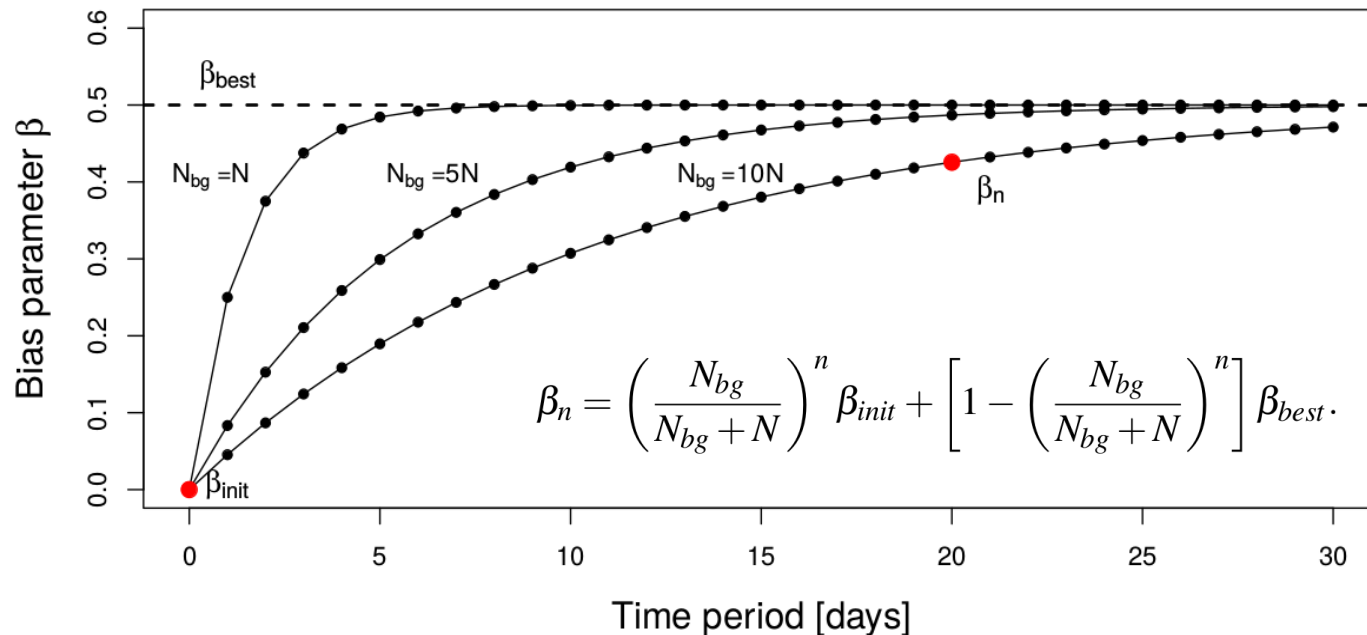
How set optimal stiffness parameters in LAMs ??

The optimum adaptivity of the VarBC is a trade-off between the variance and bias of bias parameters.

The concept of a variance-bias trade-off

$$MSE(\beta_n) = var(\beta_n) + bias(\beta_{best}, \beta_n)^2$$

See Cameron and Bell (2016)



The optimum adaptivity of VarBC is a trade-off between the variance and bias of bias parameters.

The concept of a variance-bias trade-off

$$MSE(\beta_n) = var(\beta_n) + bias(\beta_{best}, \beta_n)^2$$

$$bias(\beta_{best}, \beta_n) = \beta_{best} \left(\frac{N_{bg}}{N_{bg} + N} \right)^n$$

$$var(\beta_n) = var(\beta_{best}) \left[1 - \left(\frac{N_{bg}}{N_{bg} + N} \right) \right]^2$$

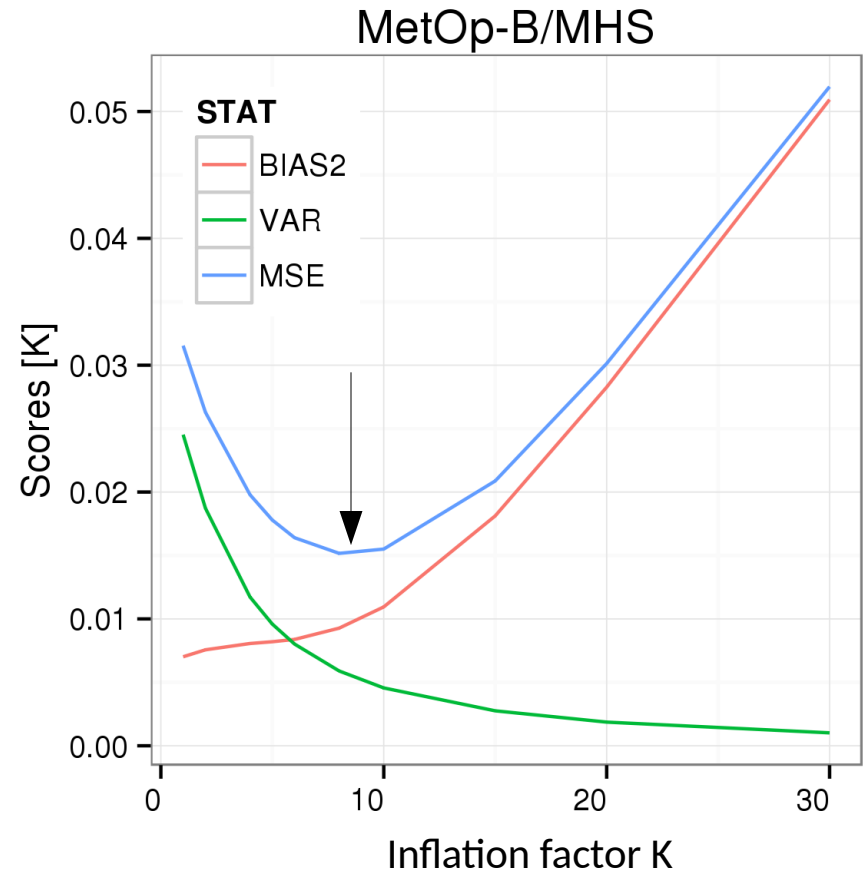
Optimum N_{bg} is estimated by minimizing the MSE:

$$\frac{dMSE}{dN_{bg}} = 0$$

The optimum adaptivity of VarBC is estimated by minimizing of the MSE of bias parameters.

$$\mathbf{N}_{bg} \simeq \max(\mathbf{N}_{avg}, N_{min}) \underbrace{\left[2nW \left(\frac{4\mathbf{N}_{avg}n^2b_{max}^2}{V^2\sigma_0^2} \right)^{-1} \right]}_K$$

- W ... the Lambert-W function (non-complex)
- N_{min} ... the minimum number of observation
- b_{max} ... the maximum observation bias
- n ... the length of spin-up period
- K ... a stiffness parameter inflation factor

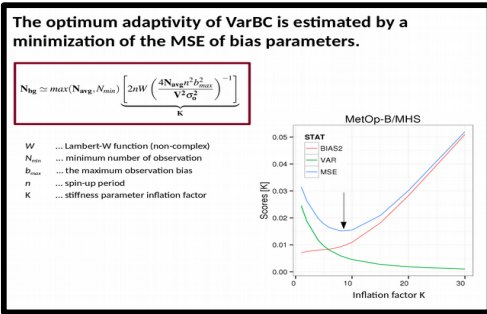


Estimating of K for the ATOVS in the LAM model Aladin-CZ.

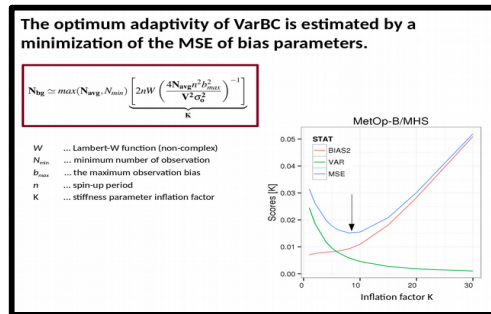
<i>Instrument</i>	σ_o [K]	<i>V</i>	passive			active		
			N_{avg}	K_{30}	$N_{bg} [\times 10^3]$	N_{avg}	K_{120}	K_{10}
AMSUA-5	0.25	4	380	6	2	60	20	3
AMSUA-6	0.18	4	460	6	3	70	20	3
AMSUA-7	0.22	4	480	6	3	75	20	3
AMSUA-8	0.22	4	920	6	6	130	20	3
AMSUA-9	0.22	6	920	6	6	130	20	3
AMSUA-10	0.26	9	920	7	6	130	20	3
AMSUA-11	0.30	14	920	8	7	130	30	5
AMSUA-12	0.66	16	920	10	9	130	40	8
AMSUA-13	1.44	22	910	14	13	130	50	15
MHS-3	3.0	6	6970	9	63	570	30	7
MHS-4	2.5	6	6690	8	54	540	30	7
MHS-5	2.0	11	6460	9	58	520	30	7

The factor V is estimated based on the period in October 2015. The remaining parameters are set to $b_{max} = 0.3 K$, $N_{min} = 100$ and $n = 10, 30$ and 120 days.

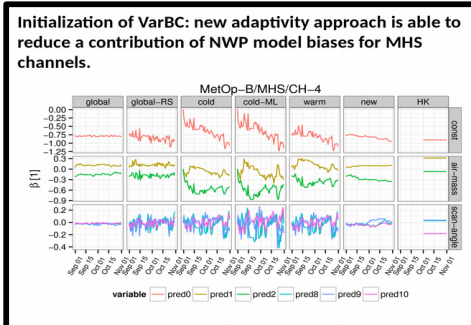
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Sampling issues in LAMs



The VarBC scheme and a new adaptivity setting



Evaluation of bias correction

The setting of evaluation

Model

The LAM Aladin-CZ with 3D-Var upper-air DA at 9 UTC only; the ATOVS on MetOp-A and -B

Period Training (spin-up): September 2015
Testing: October 2015

Passive DA experiments

Method	Description	N_{bg}
Global	Raw global β	---
Global-RS	Global β adopted to LAM at each analysis time	5000
Cold	Initial β set to zero	5000
Cold-ML	Tuned cold proposed by Lindskog et al. (2012)	2500
Warm	Initial global β + cycling β in LAM	5000
New	Initial global β + cycling β in LAM	new formulation
HK	Offline method based on Harris & Kelly (2001)	---

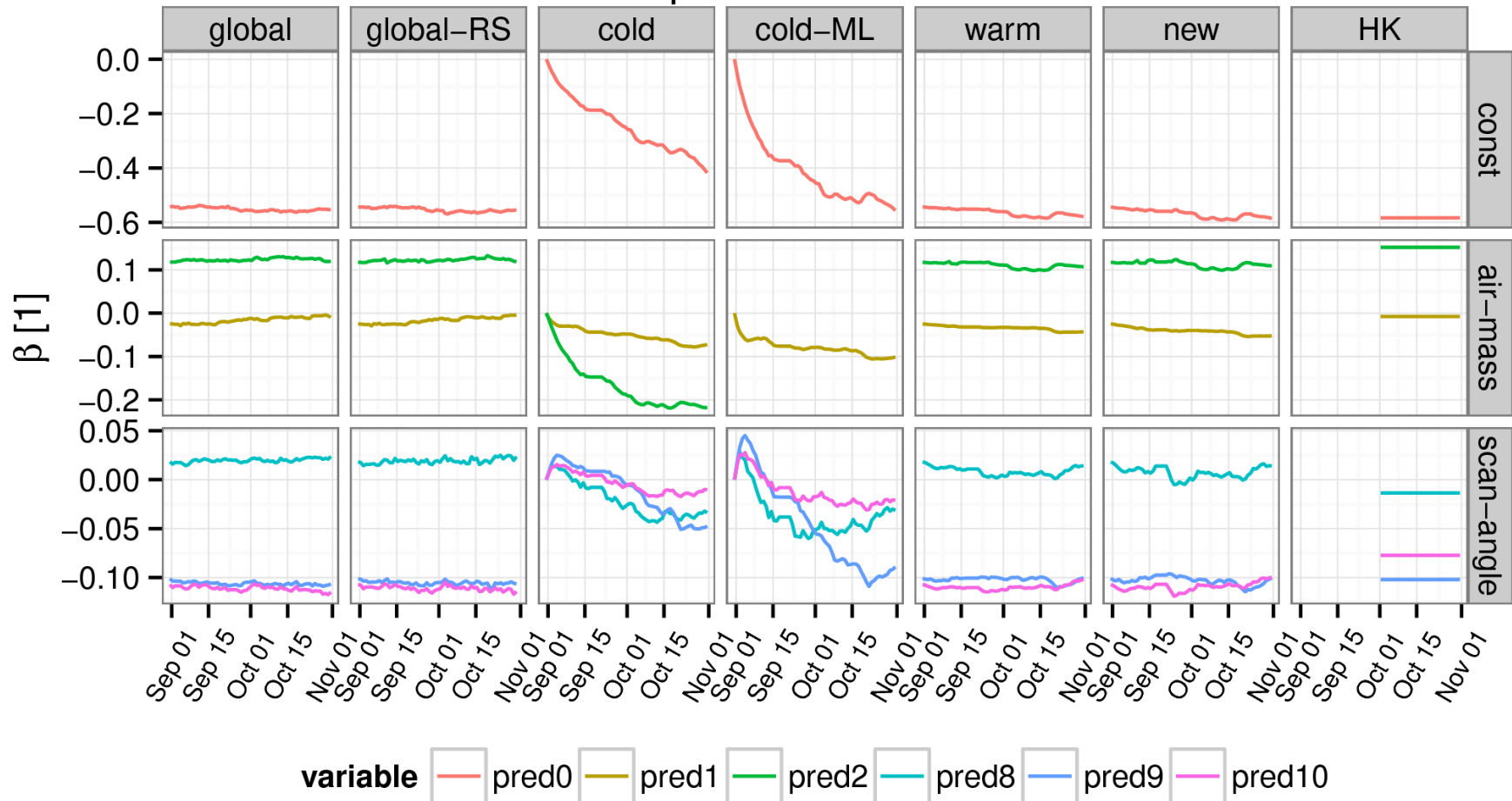
Active DA experiments

Warm_ACT70 default N_{bg} set to 5000

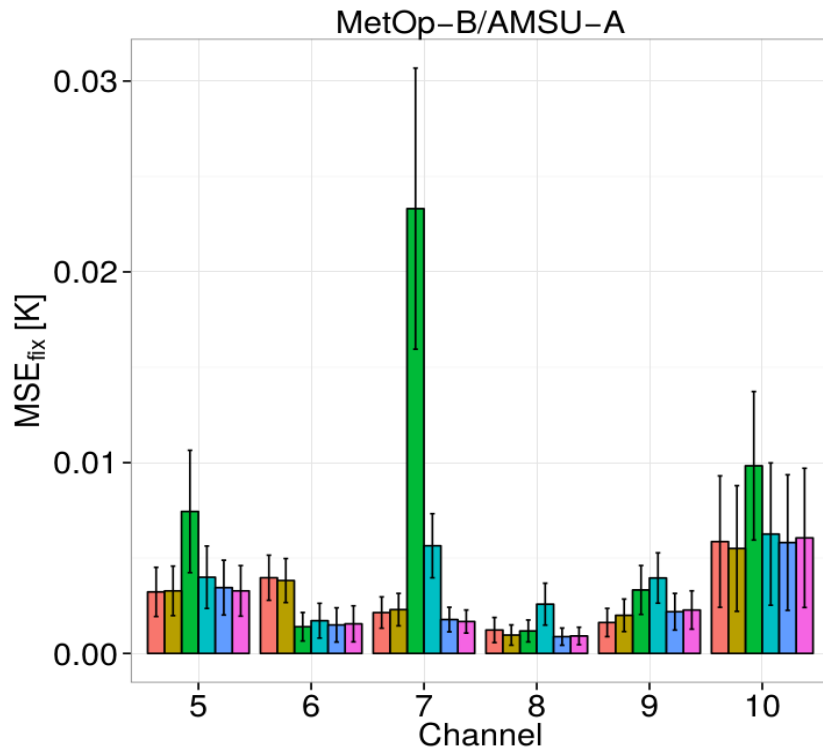
New_ACT70 new formulation

Initialization of VarBC: coldstart methods require excessive spin-up period to initialize bias dependencies for particular AMSU-A channels.

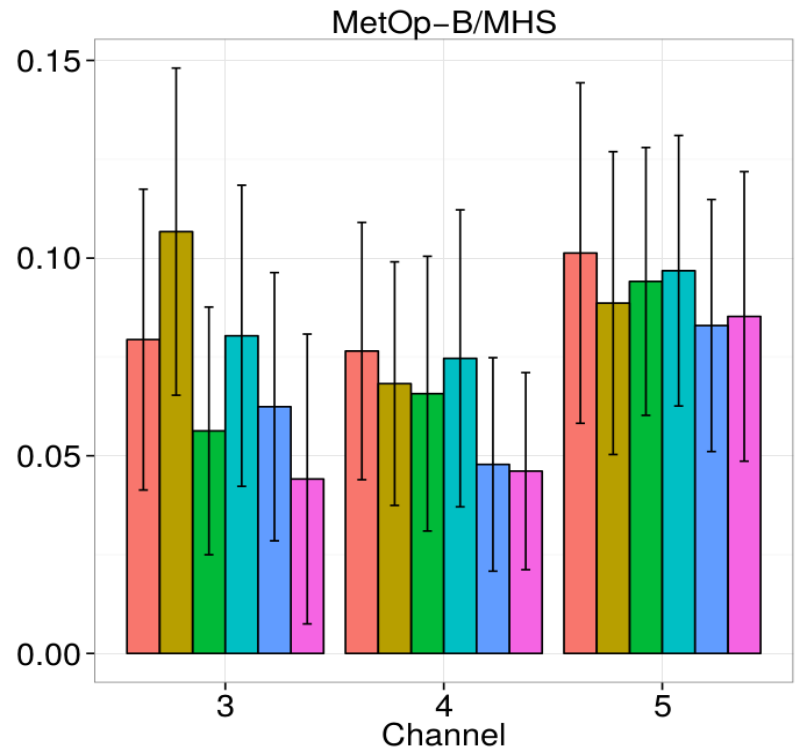
MetOp-B/AMSU-A/CH-7



The coldstart and global-RS methods provide worse quality of bias correction (MSE) than the warmstart and new methods.

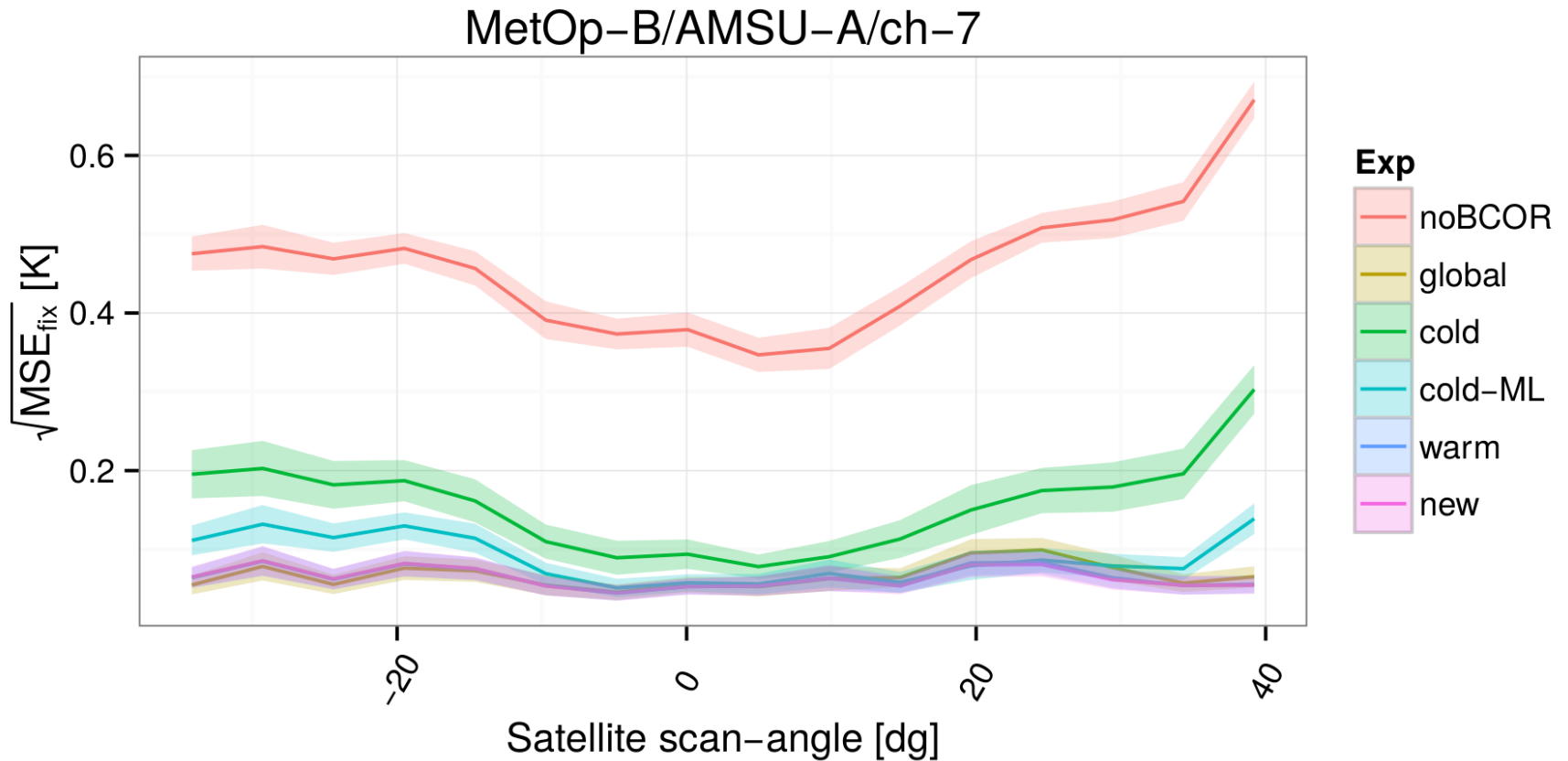


global global-RS cold cold-ML warm new

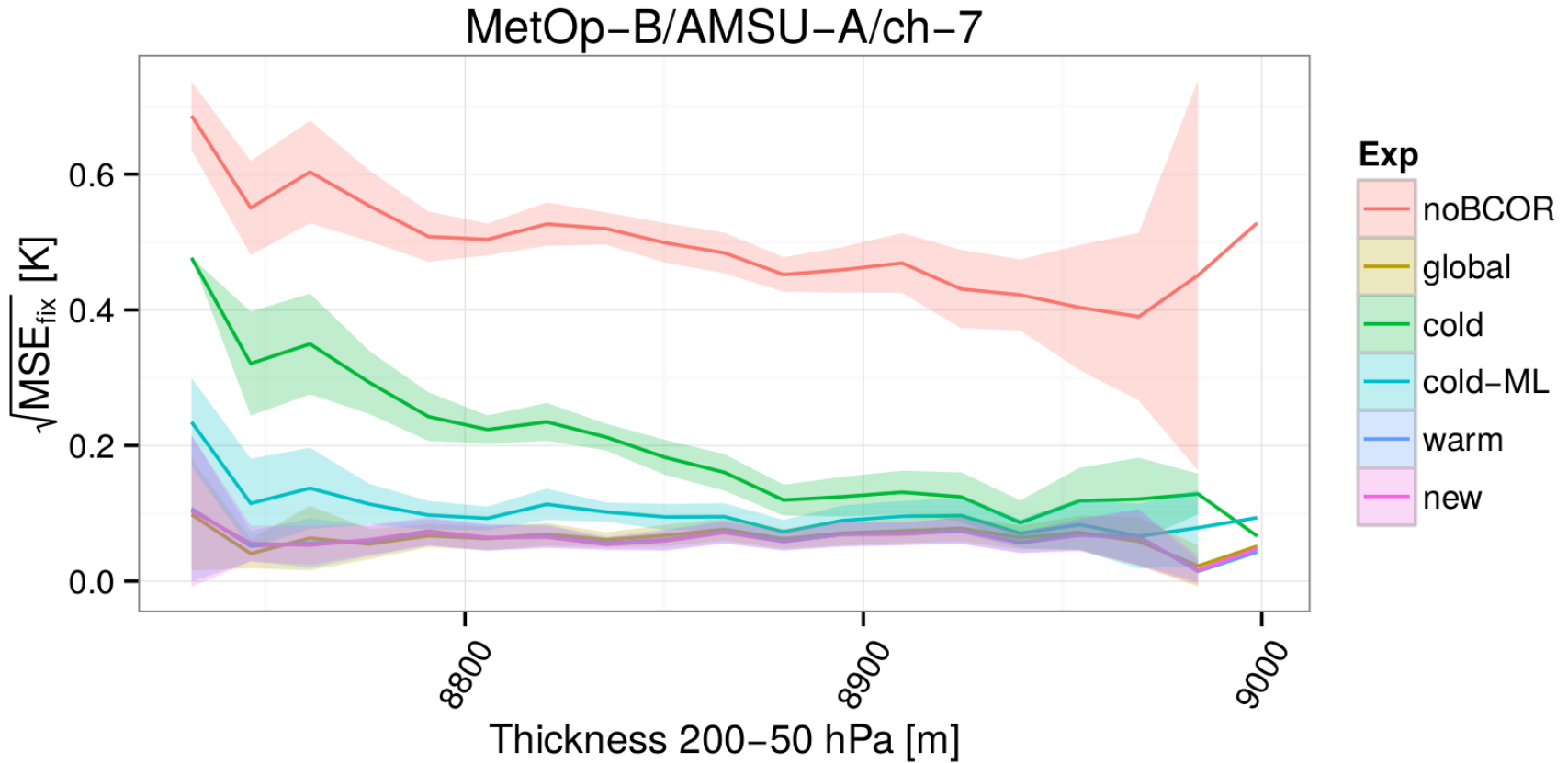


global global-RS cold cold-ML warm new

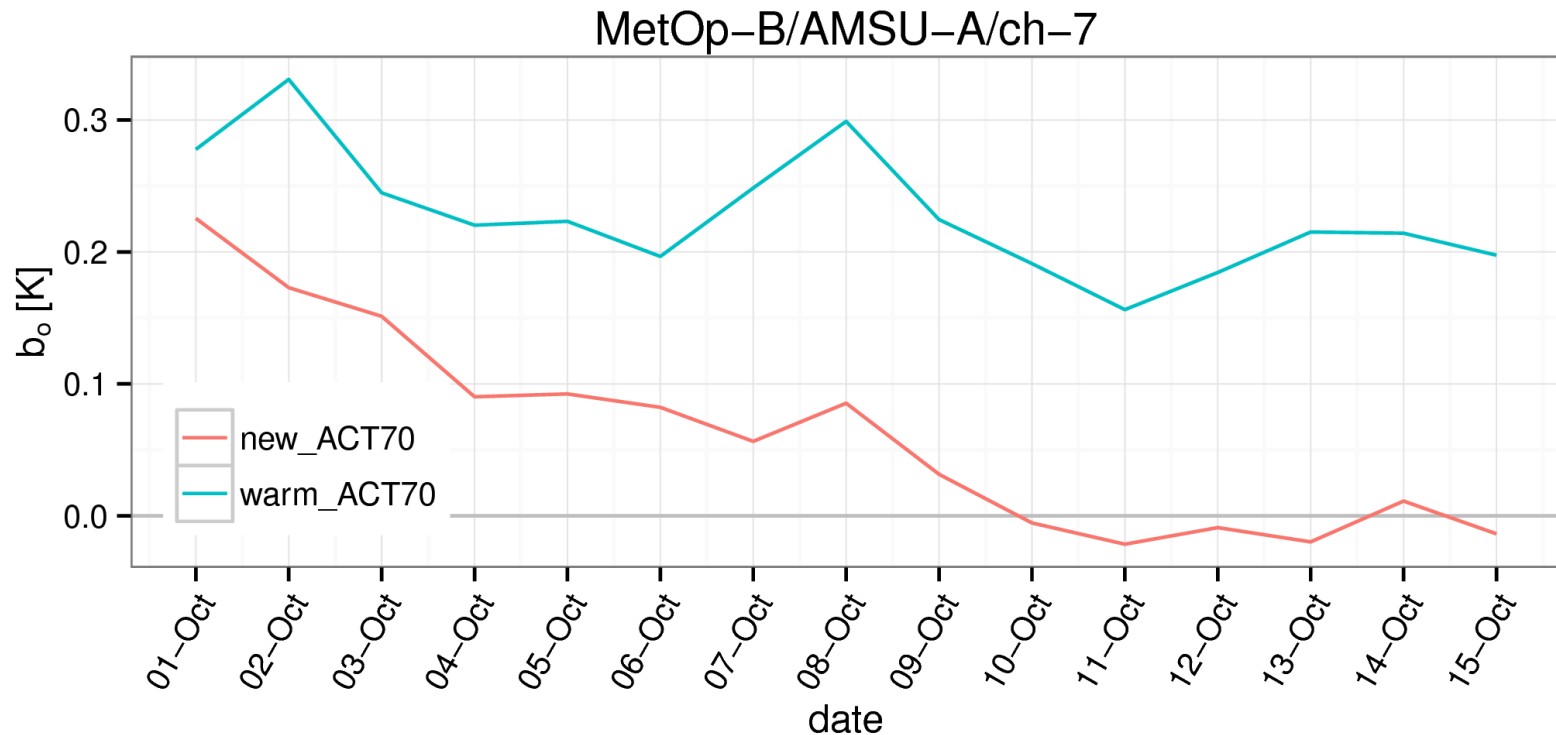
The initialization methods based on global bias parameters provide a better satellite scan-angle bias correction.



The initialization methods based on global bias parameters provide a better air-mass bias correction.



In active DA, the default setting of stiffness parameter has too slow response to satellite bias changes for particular AMSU-A channels.



The VarBC response to an artificial bias (0.3 K) during October 2015.

Summary

- VarBC advantages: a maintenance, an adaptive bias correction WRT time, analysis and available observations.
- **Initialization of VarBC:**
 - use global-RS if observation bias differences between global and LAM models are not large,
 - use warmstart instead of the coldstart methods,
 - use the new formulation of N_{bg} to estimate the adaptivity of VarBC → especially for the active data assimilation!!
- **Preliminary results on RC-LACE webpage:**
 - *Mate Mile, Patrik Benacek, 2016: Comparison of different VARBC initialization approaches*
- **Latest results – submitted manuscript to the MWR journal:**
 - *in a peer review process (08/2017)*

Future plans

- Forecast impact studies
- How to deal with marginal data coverage
- Testing 6h/24h cycling
- Summarize the up to now results concerning:
 - channel selection (IASI, ATOVS, SEVIRI),
 - VarBC scheme,
 - thinning distance,
 - observation error setting.

Thank you for your attention.

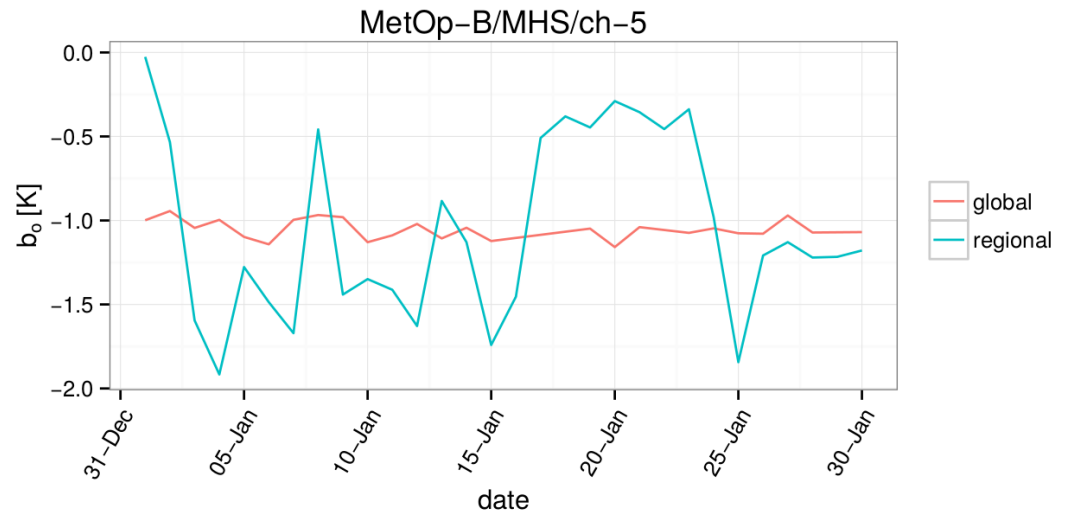
Setting of the stiffness parameter determines a response of VarBC to satellite bias changes as well as NWP model biases.

$$\sigma_{\beta_b}^2 = \sigma_{b_o}^2 = \frac{\sigma_o^2}{N_{bg}}$$

How set N_{bg} in LAMs??

“A statistically meaningful estimate of the observation bias is obtained if the bias parameters vary slowly in time and space.” Dee (2005).

$$\sigma_{\beta_b}^2 = \frac{\sigma_o^2}{N_{bg} V^2}$$



Initialization of bias parameters from a global model requires a consistency between global and LAM systems.



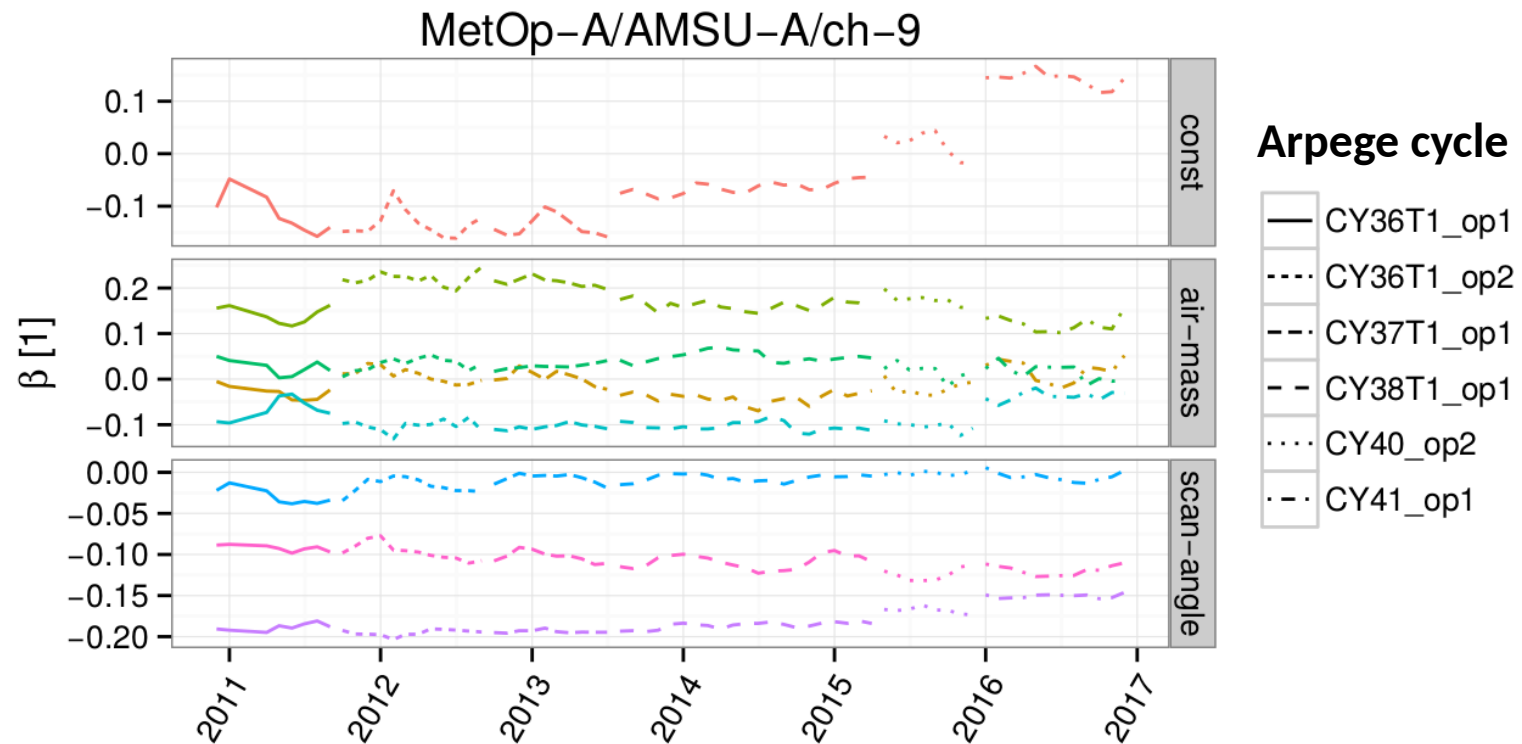
set of predictors,



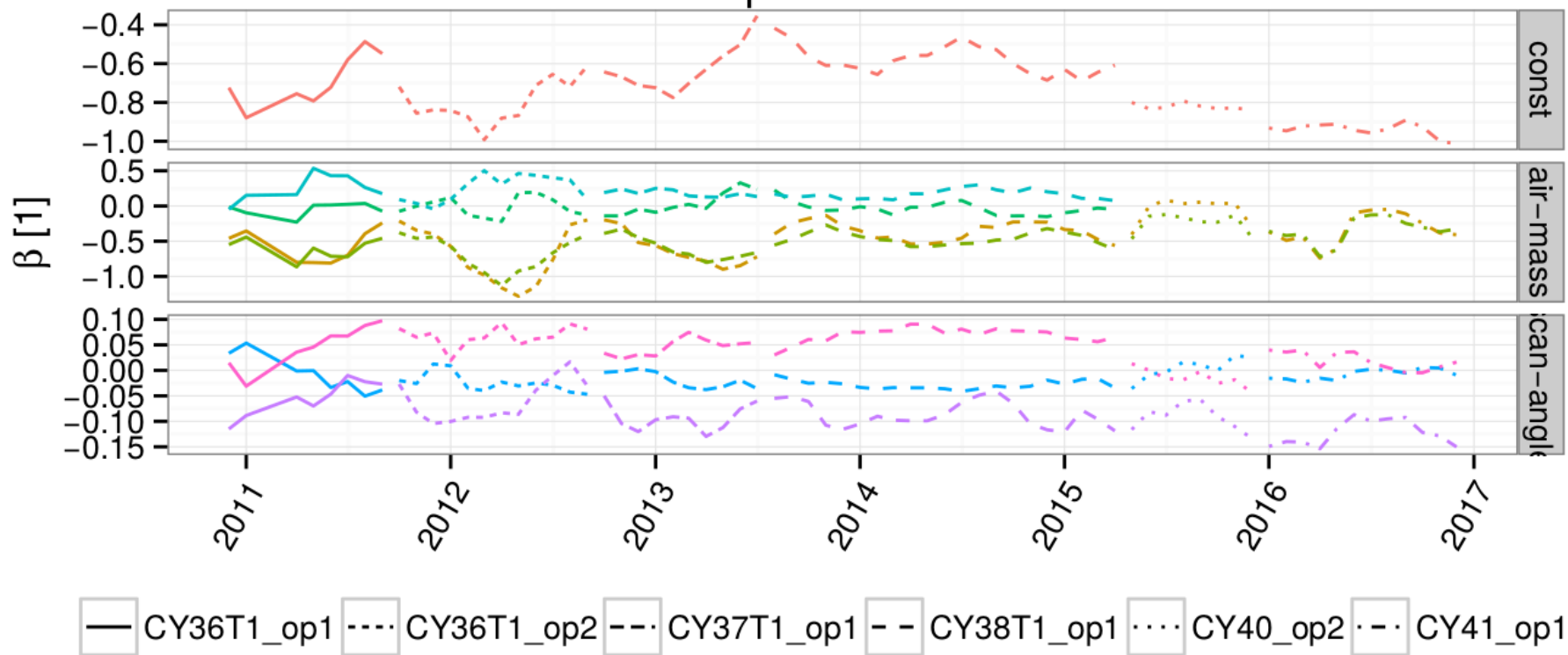
the RTTOV and satellite data pre-processing,



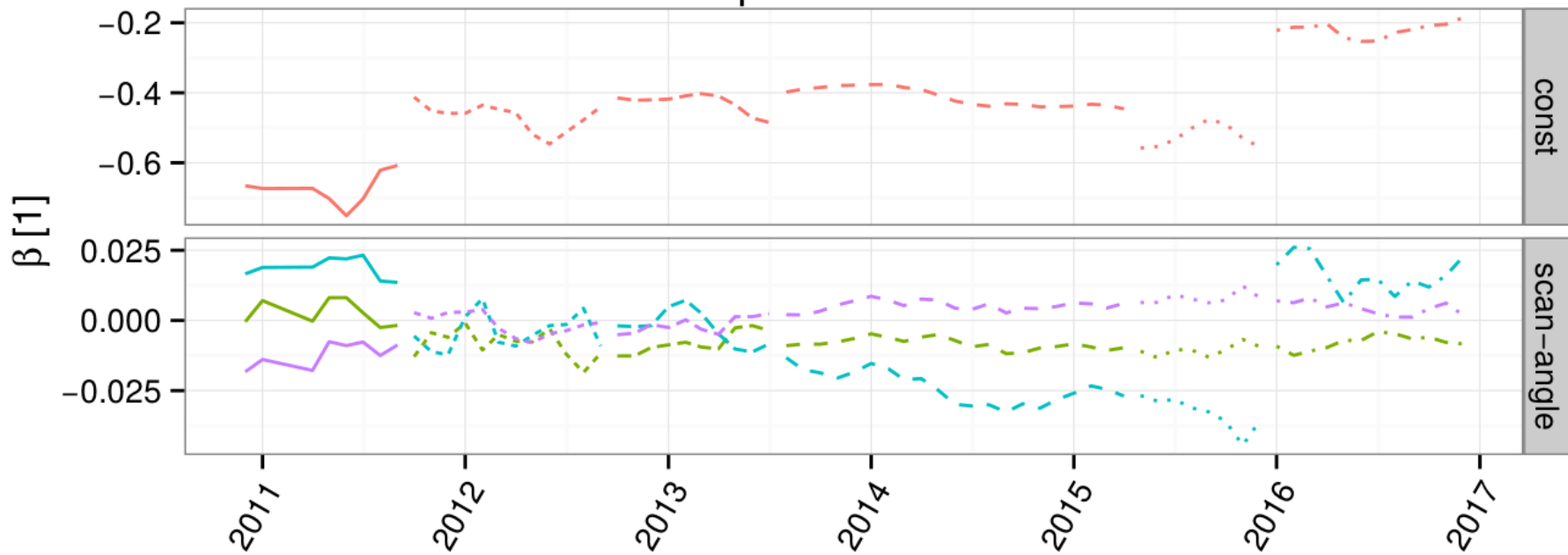
differences between the radiance bias are not large.



MetOp-A/MHS/ch-5

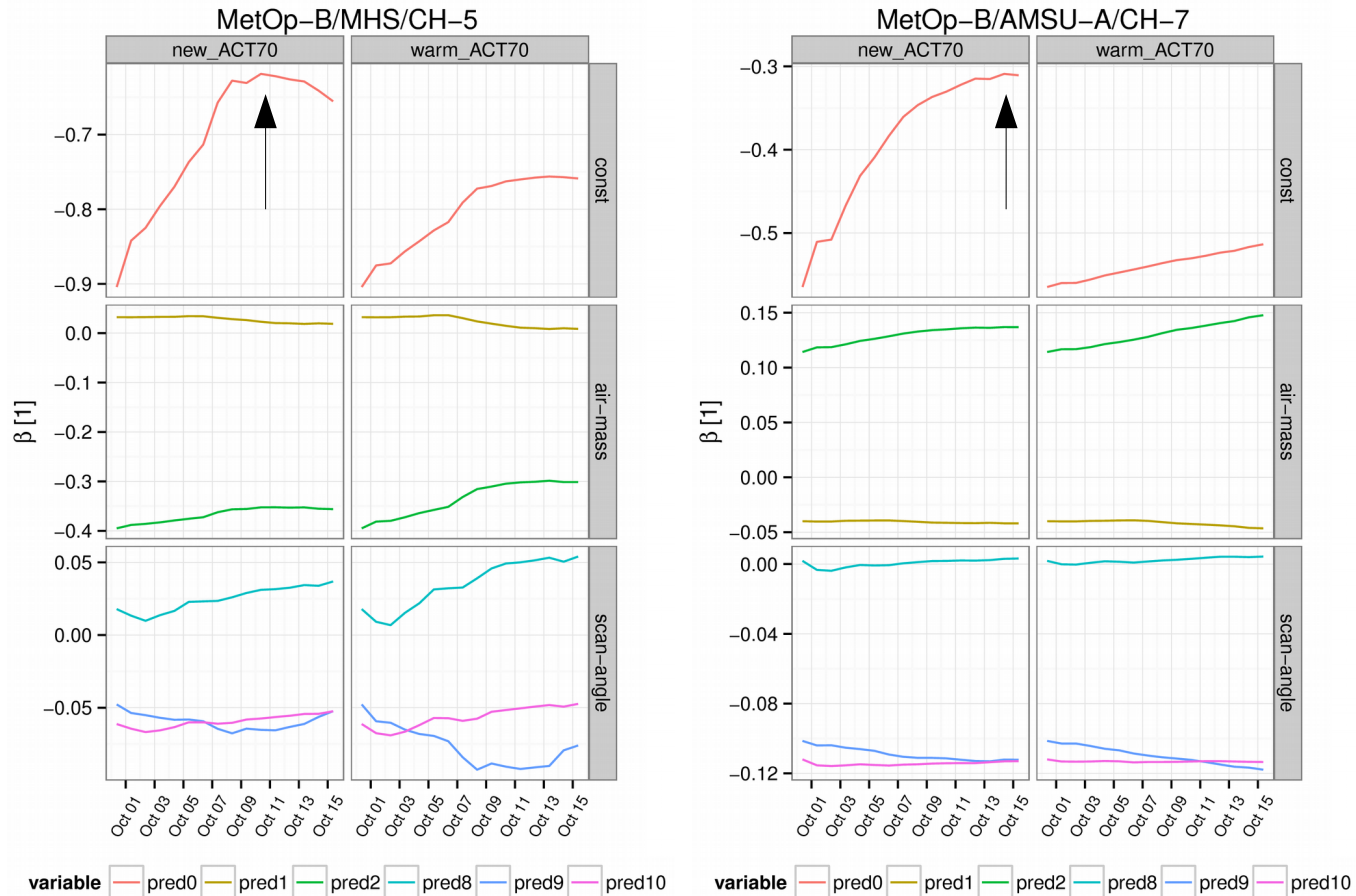


MetOp-A/IASI/ch-434



— CY36T1_op1 CY36T1_op2 -.- CY37T1_op1 - - CY38T1_op1 CY40_op2 -.- CY41_op1

In the active DA, the default setting of the VarBC adaptivity has too slow response to satellite bias changes for particular AMSU-A channels.



The VarBC response to an artificial bias 0.3 K in terms of bias parameters during October 2015.