Summary of the ECMWF/EUMETSAT NWP SAF Workshop on the Treatment of Random and Systematic Errors in Satellite Data Assimilation of NWP

Virtual workshop, 2 – 5 November 2020 Niels Bormann, Bill Bell, Patrick Laloyaux, Karen Clarke



Context: Role of systematic and random errors in NWP

• Systematic errors:

- Present in observations, observation operators (e.g., radiative transfer), forecast models, etc
- Need to estimate/remove prior to or as part of the assimilation

• Random errors:

 $TB^{-1}(x, y) = (x - H[x])^{T}R^{-1}(y - H[x])$

- Estimates of random errors in background fields and observations are used to determine the weighting of both to produce an analysis
- Observation-related random errors arise due to measurement errors, observation operators (e.g., radiative transfer), representativeness differences between observations/model fields, etc

Workshop overview

Three main themes, three sessions:

- 1. Estimating uncertainties
- 2. Correction of observational and model biases in data assimilation
- 3. Representing random observation errors in data assimilation

23 oral presentations; 12 posters

Panel discussions and informal discussion groups at the end of each day. Working groups on treatment of systematic and random errors, respectively.

Almost **200 experts** from NWP centres, space agencies, and academia.

Recordings, posters, etc, see: https://events.ecmwf.int/event/170/

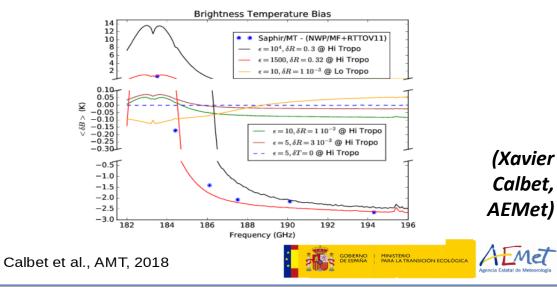


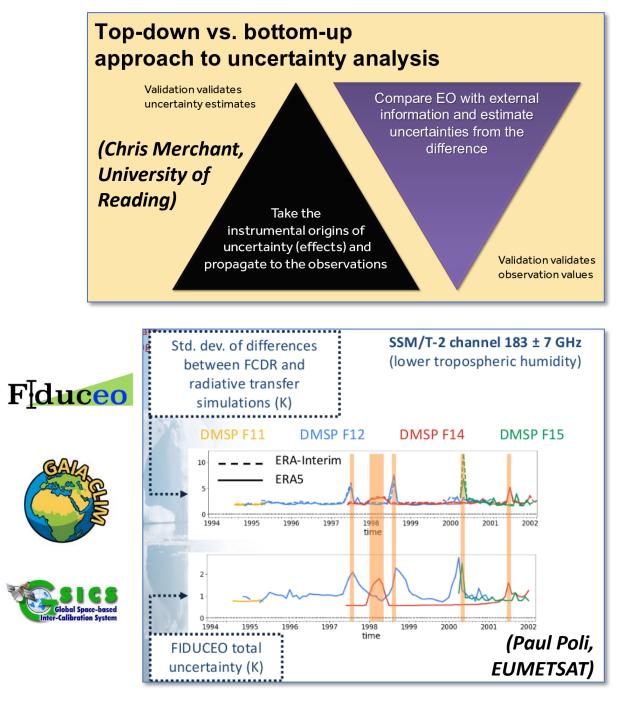
Estimating uncertainty: Main topics

- Uncertainty estimates from instrument characterisation
- Metrological approaches
- Estimation of radiative transfer and representation errors
- Uncertainty estimation and diagnosis in NWP
- Uncertainty estimation for historical datasets

Effect of FOV inhomogeneity

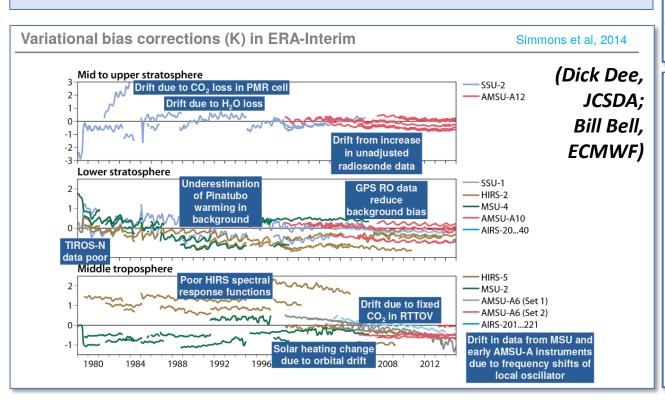
Can turbulence (= inhomogeneity) within the field of view cause significant biases in radiative transfer modelling at the 183 GHz band?



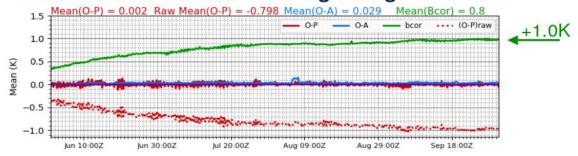


Correction of observational and model biases in data assimilation: Main topics

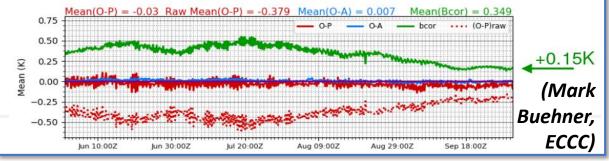
- Treatment of observation biases in reanalyses
- How to estimate both model and observational biases in DA?
 - Variational bias correction; anchor analyses; weak-constraint 4D-Var
- The role of anchor observations and uncertainty constraints
- Machine Learning approaches to bias estimation



Bias correction estimated using background state

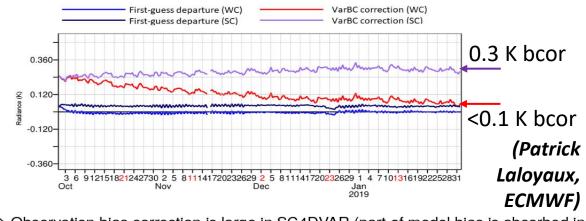


Bias correction estimated using 3D-Var with anchor obs



Interaction of weak-constraint 4D-Var with VarBC

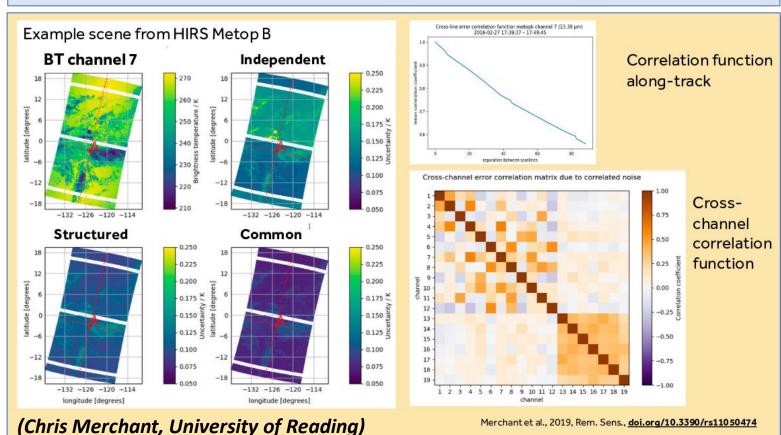
First-guess departure and observation bias correction in AMSU-A channel 10

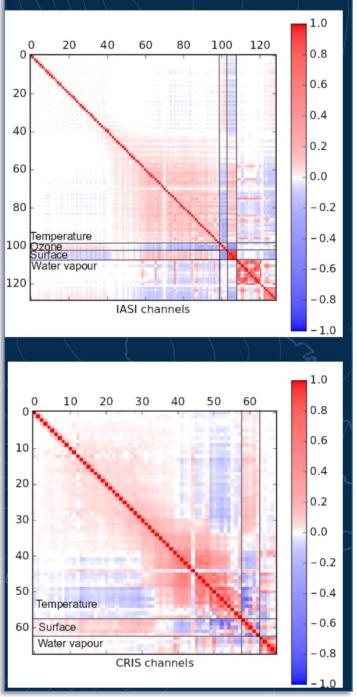


→ Observation bias correction is large in SC4DVAR (part of model bias is absorbed in VARBC). VarBC is much smaller in WC4DVAR

Representing random observation errors in data assimilation: Main topics

- Treatment of situation-dependent observational uncertainties
- Observation error correlations (incl from representation error):
 - Inter-channel, spatial
 - How to estimate them?
 - Accounting for them in the assimilation





(Fiona Smith, BoM)

Main outcomes: Overarching points

- More work needed regarding metrological/physical understanding of systematic as well as random observation-related errors.
 - Includes instrument-related errors, as well as representation errors, arising, for instance from observation operators (e.g., radiative transfer) or spatial representation errors (e.g., linked to turbulent scales).
 - Seen as fundamental in informing the treatment of random and systematic errors in data assimilation.
 - Particularly relevant in the context of increasing diversity of observations (incl. emergence of data from constellations of small satellites).
- More effort is also needed to utilize available information from metrological/ physical error analysis in the treatment of random and systematic errors in data assimilation.
- Support cross-community efforts, aimed at identifying errors/biases at source:
 - E.g., inter-comparison of bias corrections from different centres (incl. bias models used)
 - Dialogue between instrument experts, NWP centres, GSICS, etc.

 $(y - H[x])^{T}R^{-1}(y - H[x])$

Main outcomes: treatment of systematic errors

• Need for better reference observations or fuller error characterization of observations:

- to anchor bias corrections (for observations and models),
- to better identify model biases, particularly for tropospheric humidity, high-altitude (mesosphere) and higher-depth (deep ocean).
- Could hyperspectral IR become an anchor?

 $(x-x) + (x - H[x])^{T} R^{-1} (y - \Lambda)$

- Establish traceable uncertainties for GNSSRO and hyperspectral IR?
- Recommend NWP centres to revisit bias models used in adaptive observation-bias corrections to reflect advances in reducing systematic errors.
 - Potential to aid disentangling estimating observation bias and model bias.

• Possibilities for better characterization of uncertainties in climate reanalyses:

- E.g., observation denials, model parameter perturbations, ensemble approaches
- Producing unbiased analyses is a higher priority for reanalyses compared to NWP, calling for dedicated developments in certain areas (e.g., more extensive use of CVarBC).

Main outcomes: treatment of random errors

 $TB^{-1}(x, x) + (x - H[x])^{T}R^{-1}(y - H[x])$

- Assignment of random errors is often based on diagnostic or ad-hoc methods.
 - Recommendation to better understand these diagnostic uncertainty estimations.
 - Cross-comparison of results from different methods is recommended, as well as comparison to metrological/physical estimates.
- Benefits expected from further refinements in the treatment of situationdependent (representation) errors in observations.
- Recommend to NWP centres to increase efforts to overcome the technical challenges that currently limit taking into account horizontal observation error correlations.