

Harmonisation of microwave humidity sounder time series for climate applications

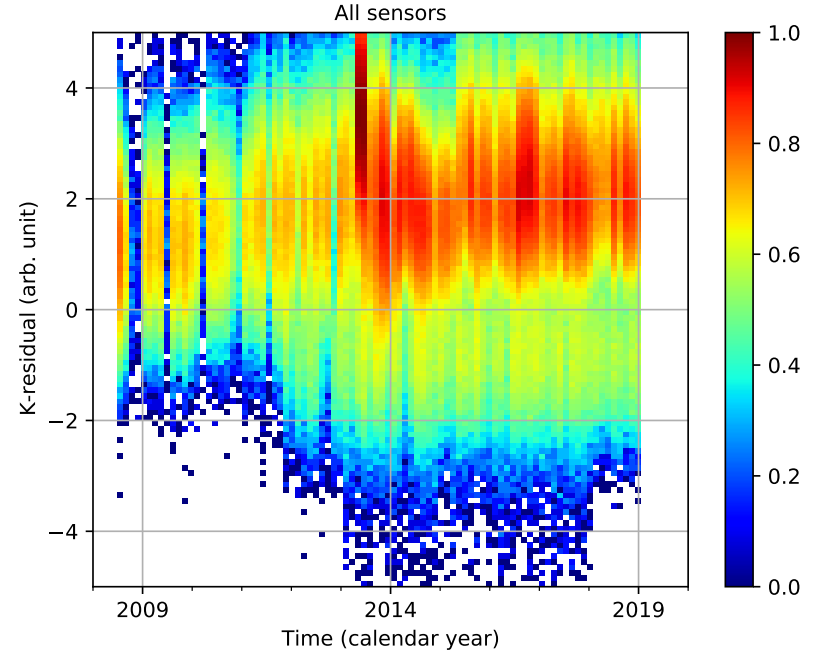
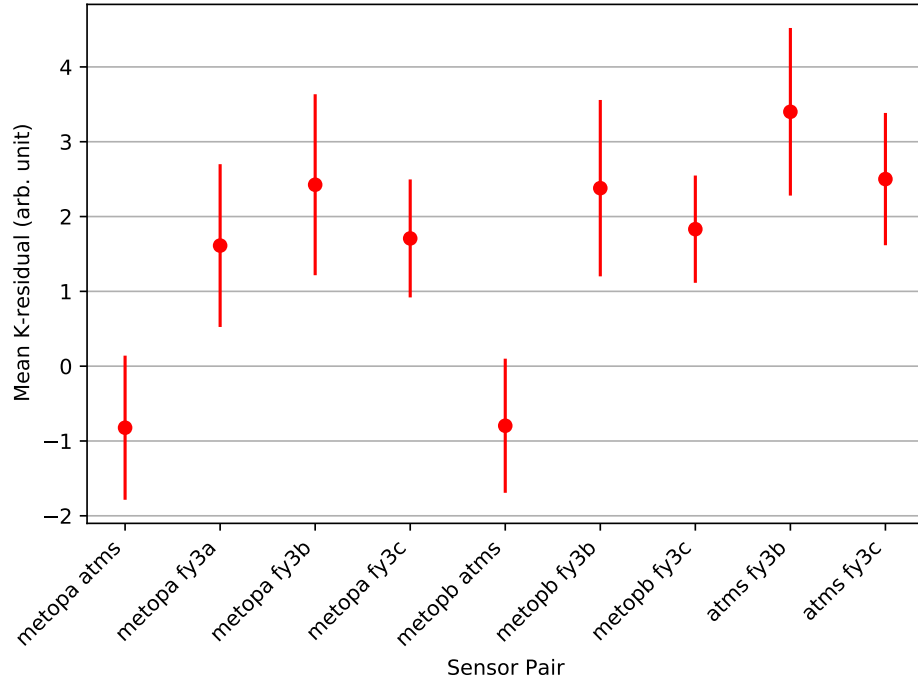
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Motivation: microwave sensor differences

Channel 183.31 ± 1 GHz



- Large offsets
- Some differences change with time

Inter-satellite re-calibration

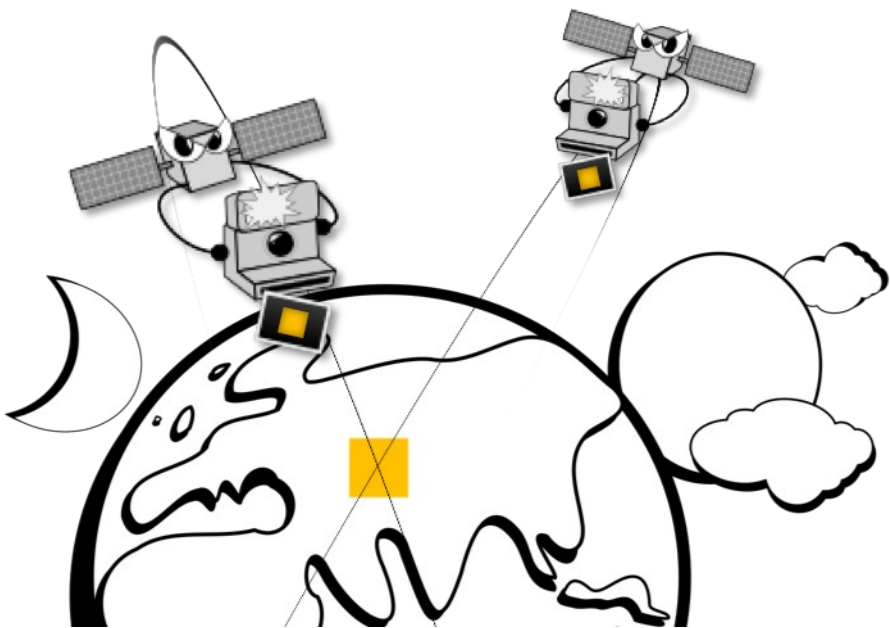
Motivation:

- long term Climate Data Records (CDR's) need to combine observations from several sensors
- **but** there are unexpected **sensor to sensor differences** even after careful calibration

Methods:

- bias and trend correction is applied after calibration on >Level-1b data
- homogenisation does not take differences in Spectral Response Function (SRF) into account
- harmonisation (inter-satellite re-calibration)
 - based on measurement equations
 - given SRF differences are taken into account
 - uses fundamental measurements, i.e. counts in level-1a data
 - considers all uncertainties, **metrological** based approach
 - uses simultaneous overpasses or **match-ups**

Match-ups



Sensors on satellites looking

- at the same **area** on earth
- at the same **time**
- from the same **distance**
- with the same **viewing angle**
- with the same photon **sensitivity**

shall measure the same radiance

Here we assume

- largely overlapping **area**
- small **time** difference
- small **viewing angle** differences
- differences in Spectral Response Functions (SRF)

Due to the spatial and temporal variability of the radiances these intervals add a match-up uncertainty

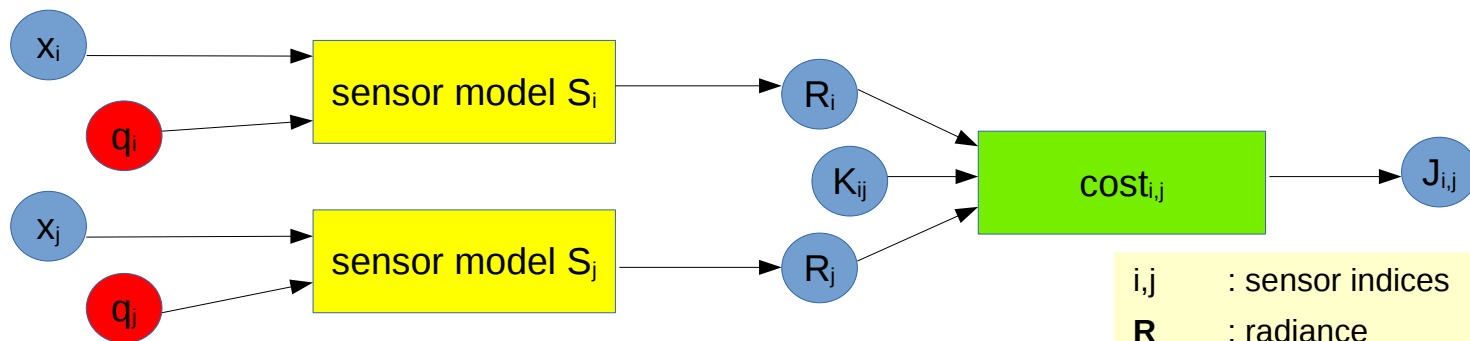
Harmonisation

Solution:

- From simultaneous overpasses of satellites match-up data sets are derived.
- Additional 'warm' match-ups of targets varying slowly based on geostationary satellites are used.
- Uncertainty of each fundamental datum is assessed.
- Expected differences in radiance due to different SRF are calculated.
Uncertainty of expected differences is assessed.
- Uncertainty in radiance difference due to match-up is assessed. Zero mean difference is assumed.
- A Bayesian based cost function (J) is defined to quantify the sensor to sensor misfit.
- The cost function is minimized by changing parameters of the sensors measurement equation.
- Parameters (x) can be polynomial coefficients and/or calibration coefficients determined pre-launch or in-orbit.
- Certain other parameters (**sensor state variables** q) of the measurement equation have uncertainties to be considered
- All derivative code generated by Automatic Differentiation (AD)

- **Marginalised Error In Variables (MEIV) method:**
 - does not optimize sensor state variables (q)
 - substantially reduced memory and computing time resources
 - takes all uncertainties into account (including correlation)

Harmonisation Framework



$$J = \sum_{i,j,i < j} J_{i,j}$$

$$J_{i,j} = \frac{1}{2} (R_i - R_j - K_{ij})^t (C_i + C_j + C_{kij})^{-1} (R_i - R_j - K_{ij})$$

$$R_i = S_i(x_i, q_i)$$

$$C_i = \left(\frac{\partial S_i}{\partial q_i} \right) W C_q W^t \left(\frac{\partial S_i}{\partial q_i} \right)^t$$

- i,j** : sensor indices
- R** : radiance
- K** : expected difference in radiance
- J** : cost function
- X** : calibration parameters of sensor model
- q** : sensor state variables
- C** : error covariance matrix
- C_{kij}** : uncertainty of match and expected difference in radiance
- C_i** : uncertainty of sensor radiance
- W** : averaging matrix
- C_q** : uncertainty of sensor state variables

Harmonisation Framework MEIV

C_i is **huge**, handling and inverting a huge matrix is expensive
 ==> using matrix free algorithms

$$J_{i,j} = \frac{1}{2} (R_i - R_j - K_{ij})^t (C_i + C_j + C_{kij})^{-1} (R_i - R_j - K_{ij})$$

$(C_i + C_j + C_{kij})^{-1} (R_i - R_j - K_{ij})$ by solving $(C_i + C_j + C_{kij}) x = (R_i - R_j - K_{ij})$ using conjugate gradient solver

$$C_i x = \left(\frac{\partial S_i}{\partial q_i} \right) W C_q W^t \left(\frac{\partial S_i}{\partial q_i} \right)^t x$$

$\left(\frac{\partial S_i}{\partial q_i} \right)^t x$ matrix free by reverse/adjoint mode Automatic Differentiation

$\frac{\partial S_i}{\partial q_i} v$ matrix free by forward/tangent mode AD

minimization of \mathbf{J} by limited memory quasi Newton BFGS algorithm using:

$\frac{\partial J}{\partial x}$ gradient by reverse/adjoint mode AD

measurement equations METOP-A,B (MHS)

Antenna temperature by Planck law

$$R_{OBCT} = \text{Planck}(T_{OBCT})$$

$$R_{DSV} = \text{Planck}(T_{DSV})$$

Two-point calibration

$$a = \frac{R_{OBCT} - R_{DSV}}{C_{OBCT} - C_{DSV}}$$

$$R_{ME} = R_{OBCT} + a * (C_E - C_{OBCT}) + u * a^2 * (C_E - C_{DSV}) * (C_E - C_{OBCT})$$

Antenna Pattern Correction (APC)

$$R = \frac{1}{g_E} * [R_{ME} - (1 - g_E) * R_{DSV}]$$

Brightness temperature by inverse Planck law

$$T = \text{Planck}^{-1}(R)$$

R : radiance
T : brightness temperature
v : frequency

sensor state variables:

C_{DSV} : Deep Space View counts

C_E : Earth counts

C_{OBCT} : On Board Calibration Target counts

u : non-linearity factor

g_E : portion from Earth (APC)

 3 varied parameters

Planck law $R = \frac{(2hc^2 * v^3)}{(e^{(hc/k_b * v)/T} - 1)}$

inverse Planck law $T = \frac{hc / k_b * v}{\log\left(\frac{2hc^2 v^3}{R} + 1\right)}$

measurement equations Fengyun-3A (MWHS)

Antenna temperature by Planck law

$$R_{OBCT} = Planck(T_{OBCT})$$

$$R_{DSV} = Planck(T_{DSV})$$

Two-point calibration

$$a = \frac{R_{OBCT} - R_{DSV}}{C_{OBCT} - C_{DSV}}$$

$$R_{ME} = R_{OBCT} + a * (C_E - C_{OBCT})$$

Brightness temperature by inverse Planck law

$$T_{ME} = Planck^{-1}(R_{ME})$$

Non linear correction

$$T_{nl} = u_1 + u_2 * T_{ME} + u_3 * T_{ME}^2$$

sensor state variables:

C_{DSV} : Deep Space View counts

C_E : Earth counts

C_{OBCT} : On Board Calibration Target counts

R_{OBCT} : On Board Calibration Target radiance

u_i : coefficients of APC

APC_1 : pixel dependent Antenna Pattern Correction



3 varied parameters

Antenna Pattern Correction

$$T = APC_1 + APC_2 * T_{nl}$$

measurement equations Fengyun-3B/C (MWHS/MWHS2)

Antenna temperature by Planck law

$$R_{OBCT} = \text{Planck}(T_{OBCT})$$

$$R_{DSV} = \text{Planck}(T_{DSV})$$

Two-point calibration

$$a = \frac{R_{OBCT} - R_{DSV}}{C_{OBCT} - C_{DSV}}$$

$$R_{ME} = R_{OBCT} + a * (C_E - C_{OBCT}) + u * a^2 * (C_E - C_{DSV}) * (C_E - C_{OBCT})$$

Brightness temperature by inverse Planck law

$$T_{ME} = \text{Planck}^{-1}(R_{ME})$$

Antenna Pattern Correction

$$T = APC_1 + APC_2 * T_{ME}$$

sensor state variables:

C_{DSV} : Deep Space View counts

C_E : Earth counts

C_{OBCT} : On Board Calibration Target counts

R_{OBCT} : On Board Calibration Target radiance

u : non-linearity factor

APC_1 : pixel dependent Antenna Pattern Correction



3 varied parameters

measurement equations SNPP (ATMS)

Cold and warm target temperatures

$$T_C = A + B(T_{COSMIC} + dT_C)$$

$$T_W = A + B(T_{OBCT} + dT_W)$$

Antenna temperature

$$T_{lin} = T_W + (C_E - C_{OBCT}) * \frac{T_W - T_C}{C_{OBCT} - C_{DSV}}$$

Non-linear correction

$$T_{nonlin} = T_{lin} + u * \left(1 - 4 \left(\frac{T_{lin} - T_E}{T_W - T_C} - 0.5 \right)^2 \right)$$

Antenna Pattern Correction

$$T = T_{nonlin} - APC$$

sensor state variables:

C_{DSV} : Deep Space View counts

C_E : Earth counts

C_{OBCT} : On Board Calibration Target counts

u : non-linearity factor

APC : pixel dependent Antenna Pattern Correction



3 varied parameters

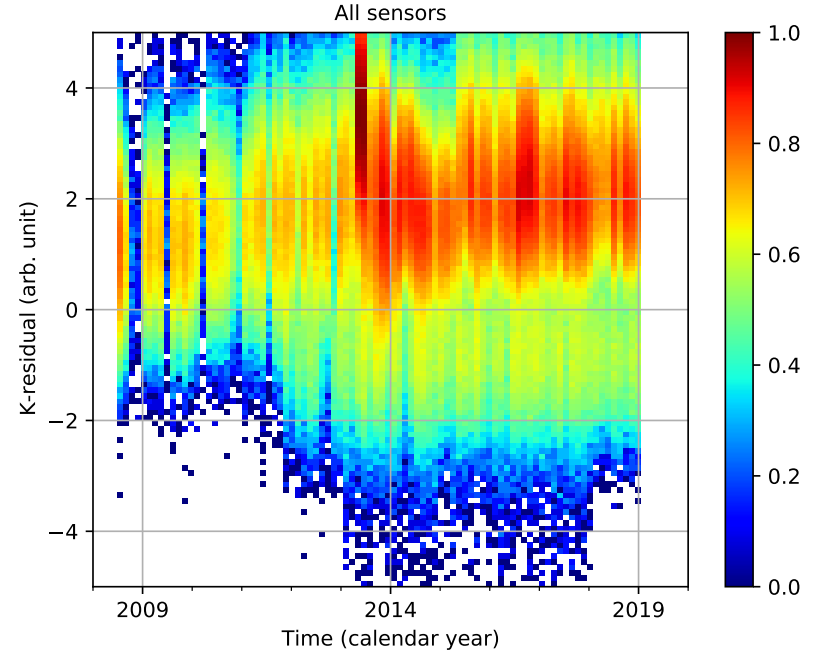
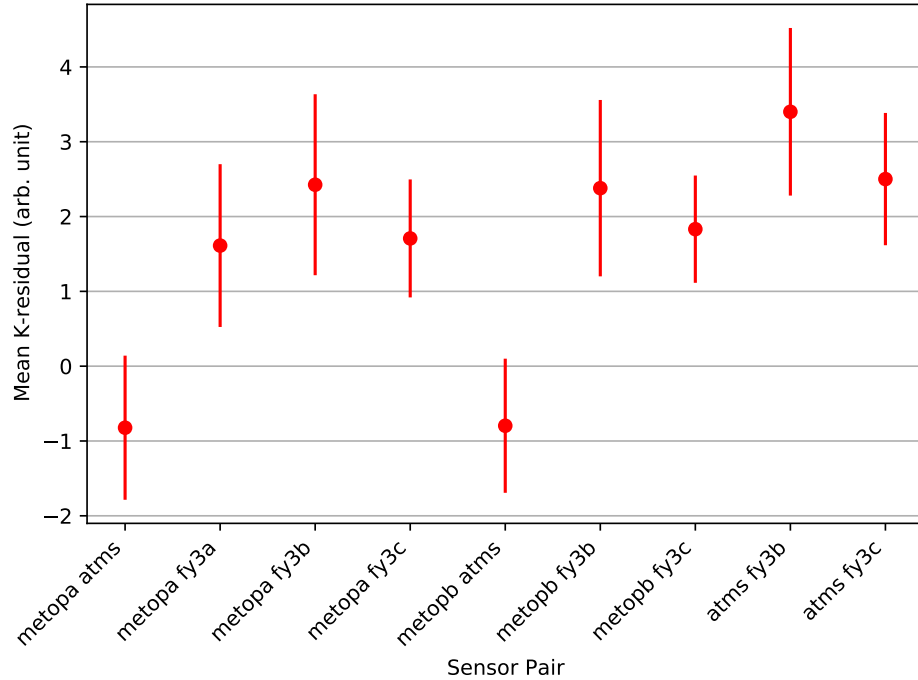
Harmonisation setup for microwave

Sensors: **MetOp-A**, MetOp-B, ATMS, Fengyun-3A, -3B, -3C

- Cold match-ups:
 - difference in space: 10 km
 - difference in time: **5 min**
 - difference in viewing angle: ratio of path length < 0.05
- Intermediate warm match-ups, LEO-GEO:
 - MVIRI homogeneous 3x3pixel scene, std.dev. < 1 K
 - MVIRI $\Delta T < 1$ K
 - difference in space: 4 km
 - difference in time: 5 min
 - difference in viewing angle: ratio of path length < 0.05
- Warm match-ups, LEO-LEO:
 - difference in space: 10 km
 - difference in time: **6 hours**
 - difference in viewing angle: ratio of path length < 0.05
- 1.5 million match-ups by 9 sensor pairs
- Channel 183.31 ± 1 GHz, also available are ± 3 GHz, and ± 7 GHz
- 15 parameters to optimise: 5 sensors a 3 parameters
- one sensor serves as **reference**, i.e. only radiances with uncertainties are used (not necessary, can be replaced by additional constraint on parameters)

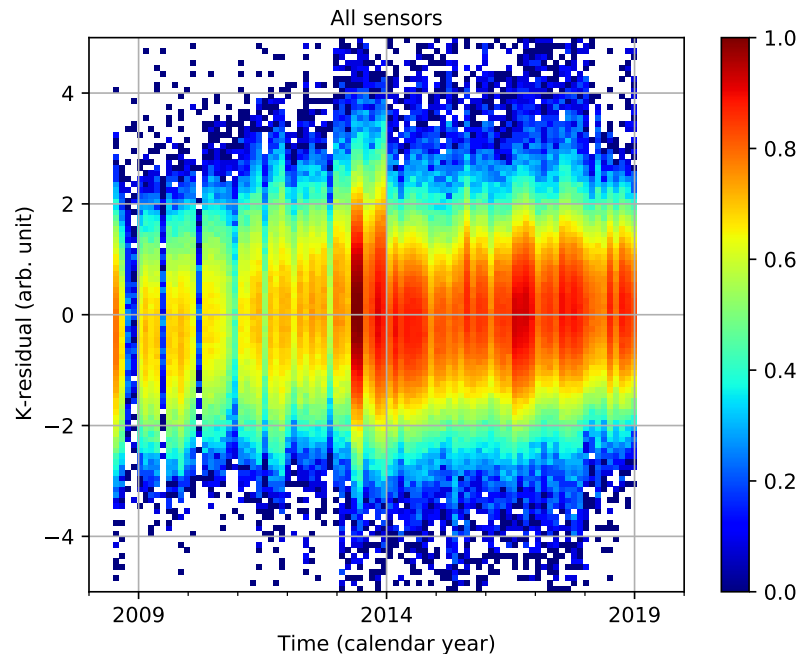
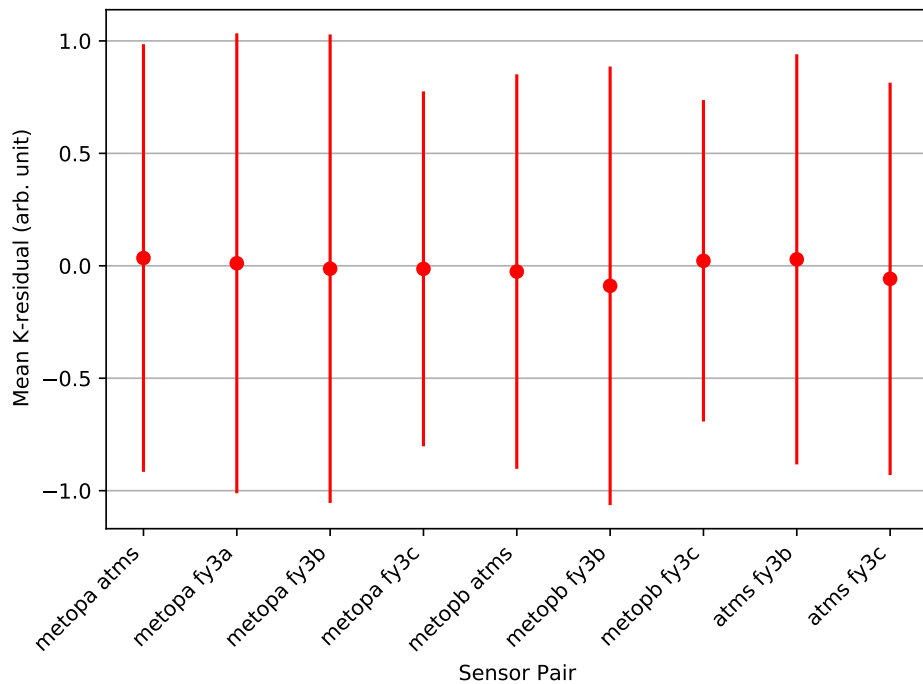
Microwave sensor differences

Channel 183.31 ± 1 GHz



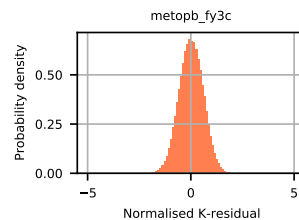
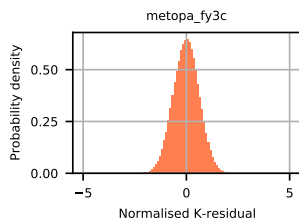
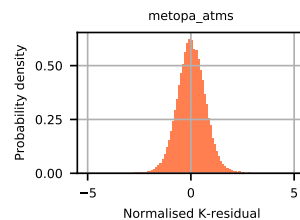
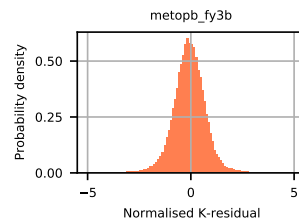
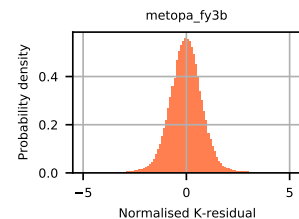
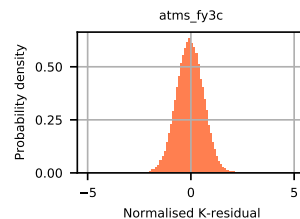
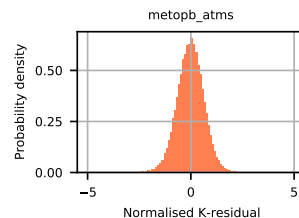
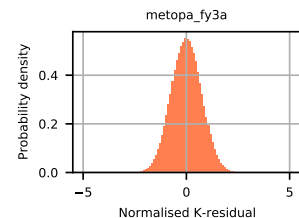
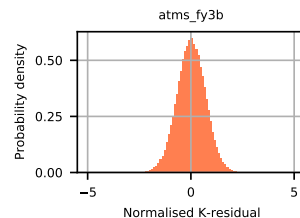
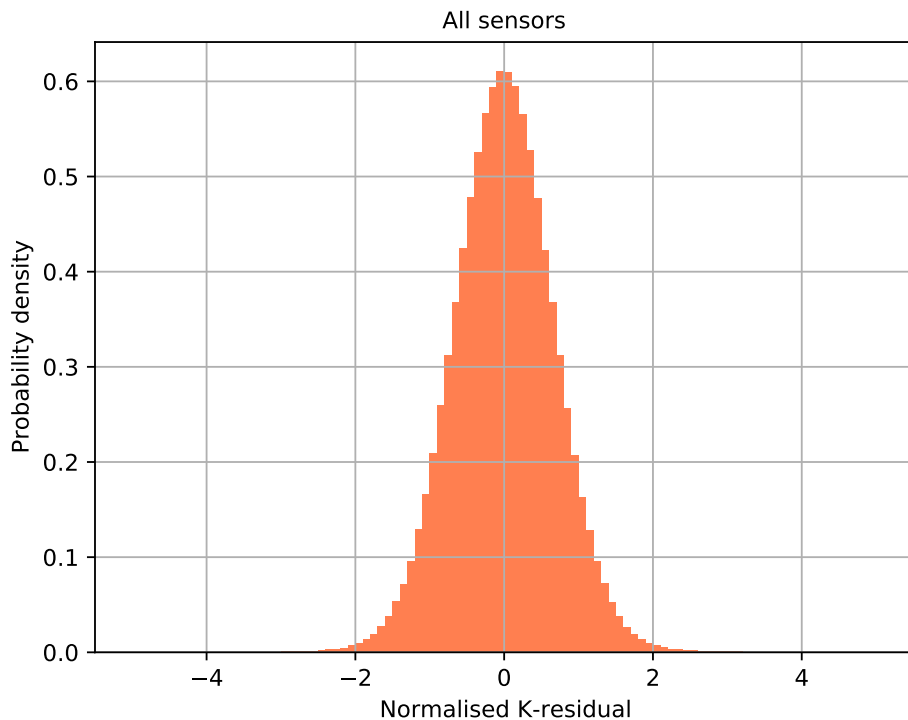
Residuals after harmonisation

Channel 183.31 ± 1 GHz



- mean residuals are close to zero
- distribution is similar for all pairs
- a problematic sensor would stand out

Distribution of residuals



Final residuals are almost Gaussian distributed.

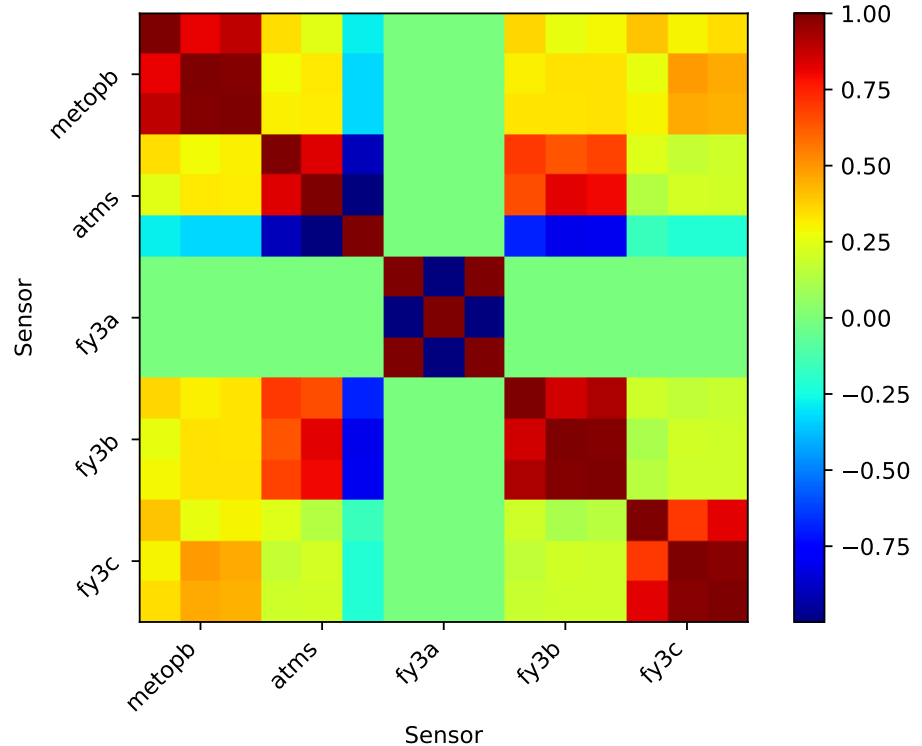
Posterior uncertainty covariance matrix

$$C_i = \left(\frac{\partial J^2}{\partial x^2} \right)^{-1}$$

posterior uncertainty covariance
is inverse Hessian

$$\frac{\partial J^2}{\partial x^2}$$

Full Hessian by vector forward
over scalar reverse mode AD



correlation matrix

Please note, correlation is covariance divided by variance

Conclusions

- Harmonisation of Microwave sensors based on MEIV successful
 - misfit is reduced substantially
 - posterior statistics are consistent with assumptions
 - full range coverage by including ‘warm’ match-ups from targets changing slowly
- For good results, estimation and propagation of uncertainties is important
- Automatic Differentiation (AD) with TAF (Transformation of Algorithm in Fortran) greatly helped
 - only function code must be written
 - efficient derivative codes generated ‘on the fly’
 - can operate in scalar mode for matrix free uncertainty propagation
or vector mode for full Jacobian/Hessian computation
- Optimal parameters are used to produce new FCDR’s with uncertainties
- Algorithm published in remote sensing 2019, 11(9), 1002; <https://doi.org/10.3390/rs11091002>

Outlook

- Analysis of remaining misfit
 - revisit measurement equation formulation
 - introduce new parameter in measurement equation
- Simultaneous harmonisation of several channels if the misfit is correlated (eg. HIRS)
- MEIV Harmonisation can be extended to use other sources of information
 - simulated radiances
 - known targets