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Evolving the GSICS Approach to Inter-Satellite Calibration

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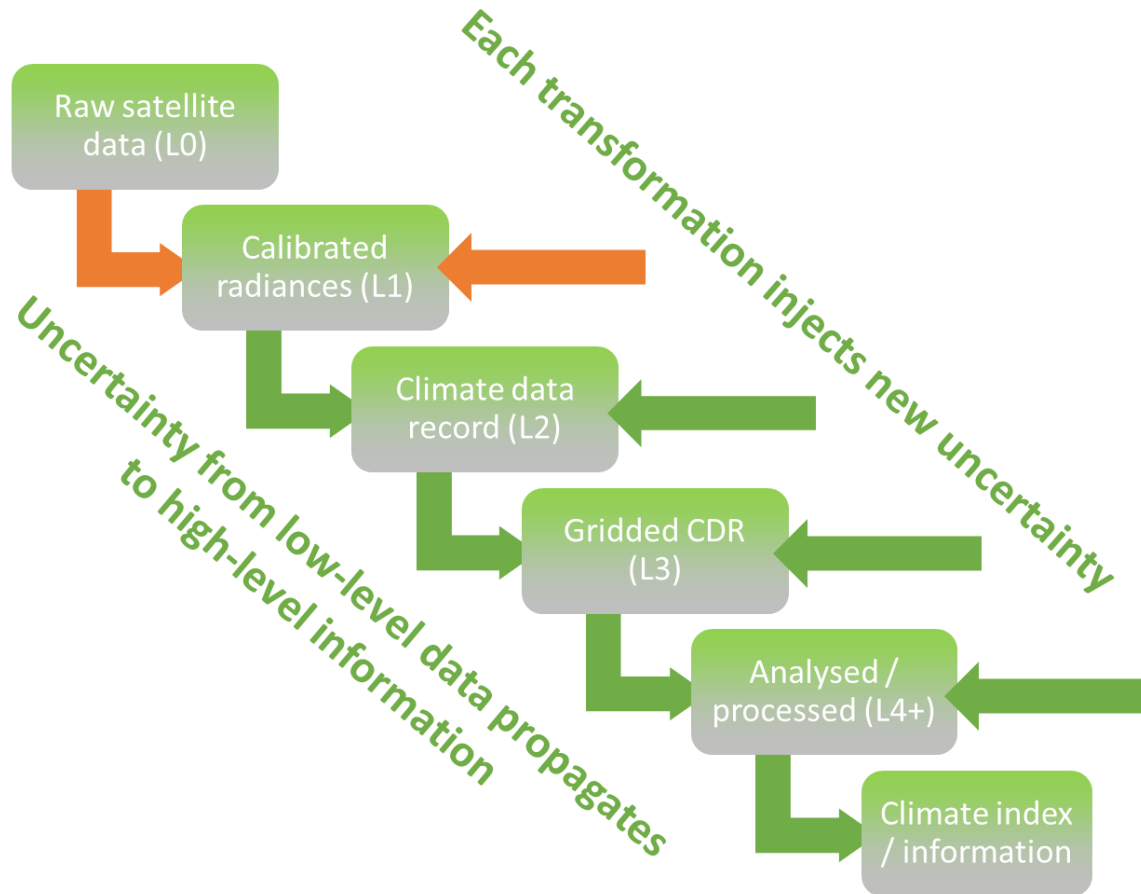
Outline

- Brief introduction to FIDUCEO and the aims of WP6.
- Describe the evolution of inter-satellite calibration in relation to GSICS approach and current progress.
- Outline future work.

FIDUCEO

- Fidelity and **U**ncertainty in **C**limate data records from **E**arth **O**bservation.
- **Ambition**: develop a widely applicable metrology framework for Earth observation (EO).
- **Motivation**: establish **defensible, uncertainty-quantified** climate data records (CDRs) for climate and environmental change from the satellite record.
- **Limitation of the status quo**: if uncertainty in fundamental climate data records (FCDR) is not characterised it cannot be propagated to the CDR level.

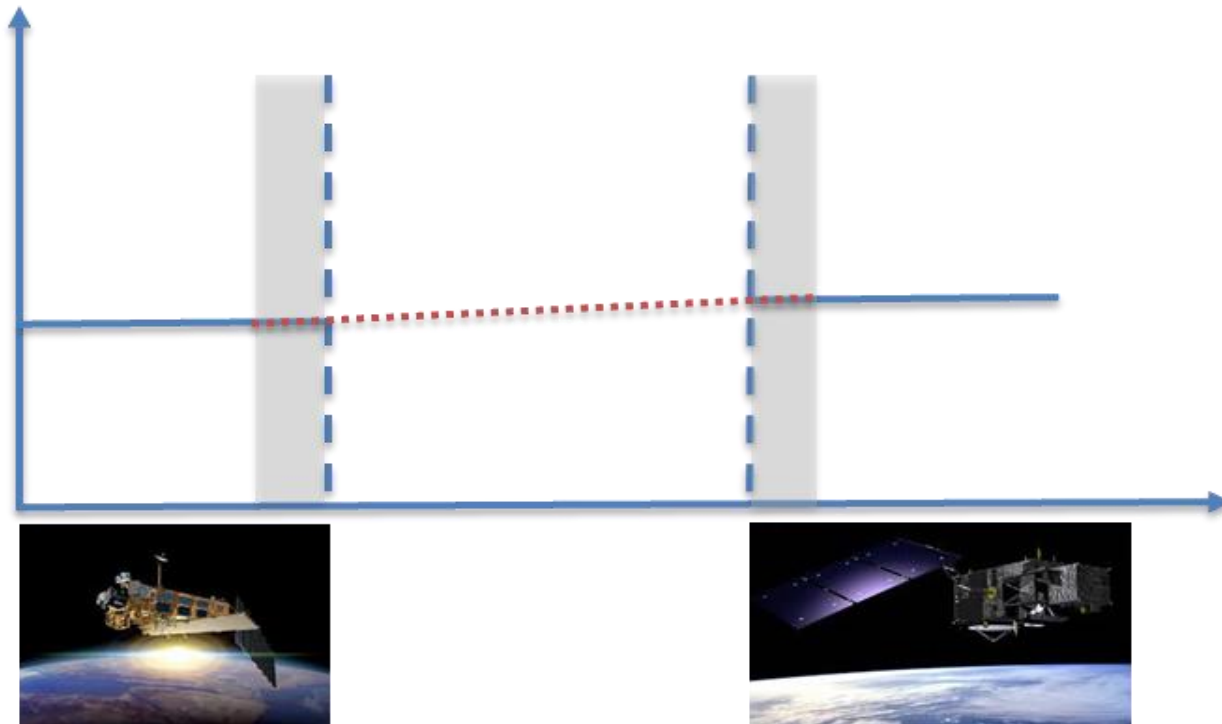
FIDUCEO



- Further information can be found at <http://www.fiduceo.eu/blogs>

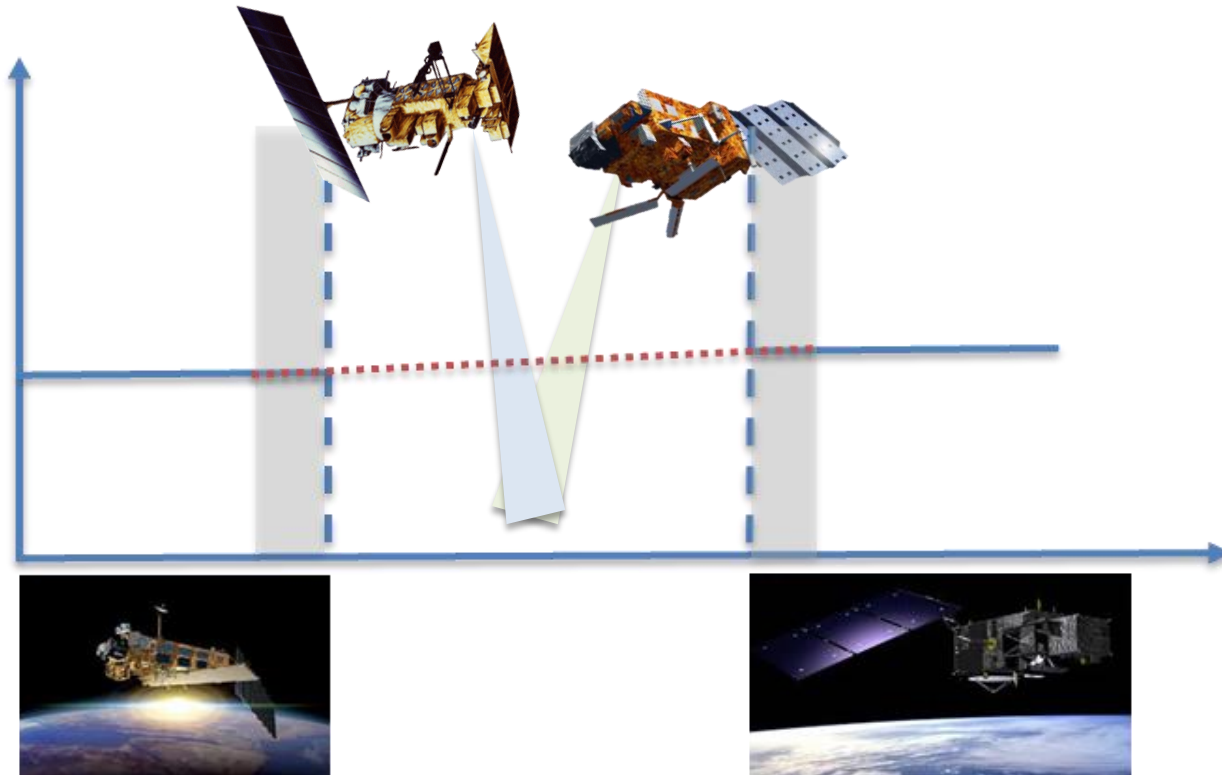
Aims of WP6

- Evaluation of FIDUCEO FCDRs during (A)ATSR-SLSTR gap.
- Goal is to build a generic framework that can handle different intercomparison types: (i) satellite-to-satellite and (ii) satellite-to-in situ



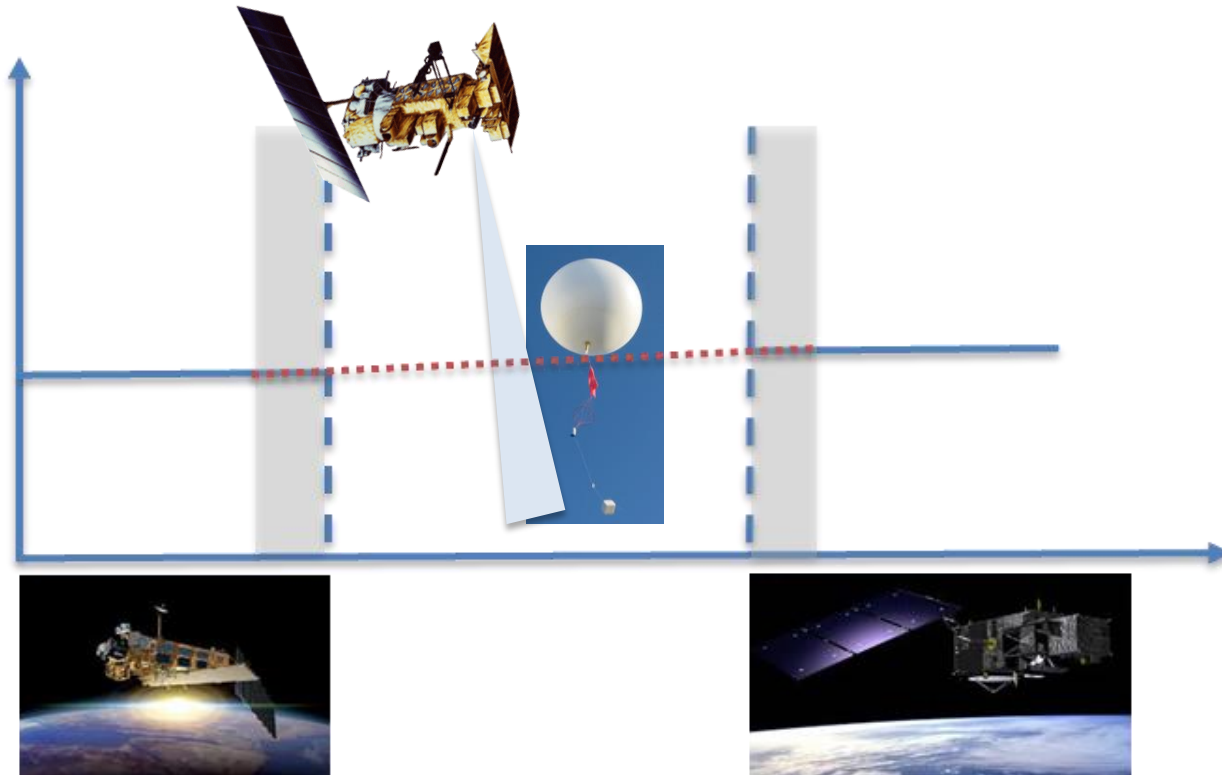
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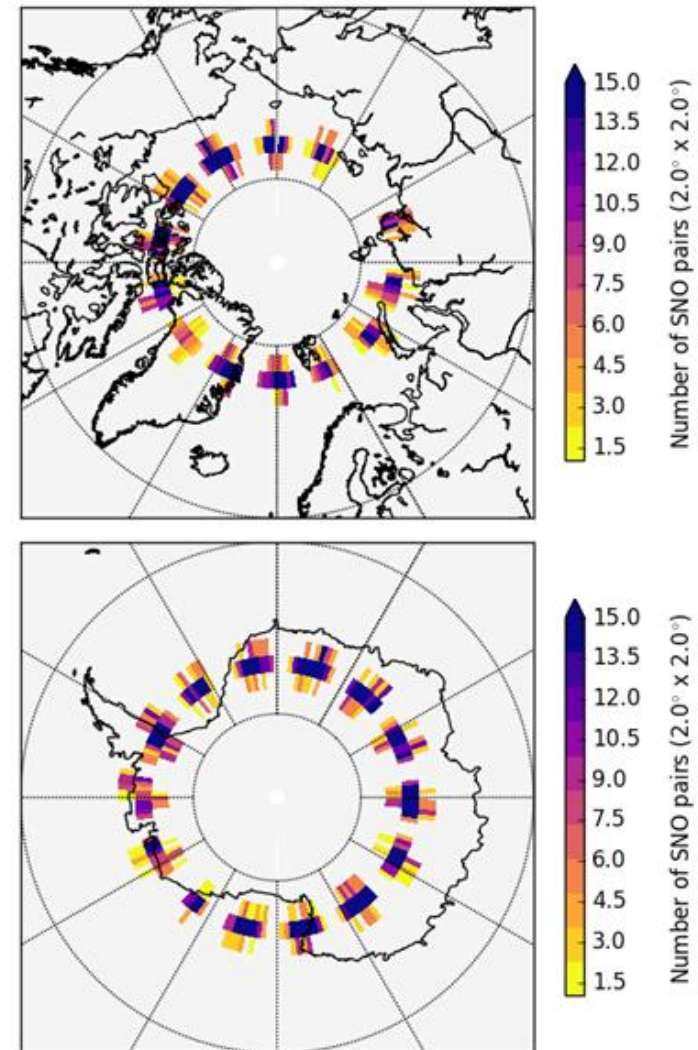
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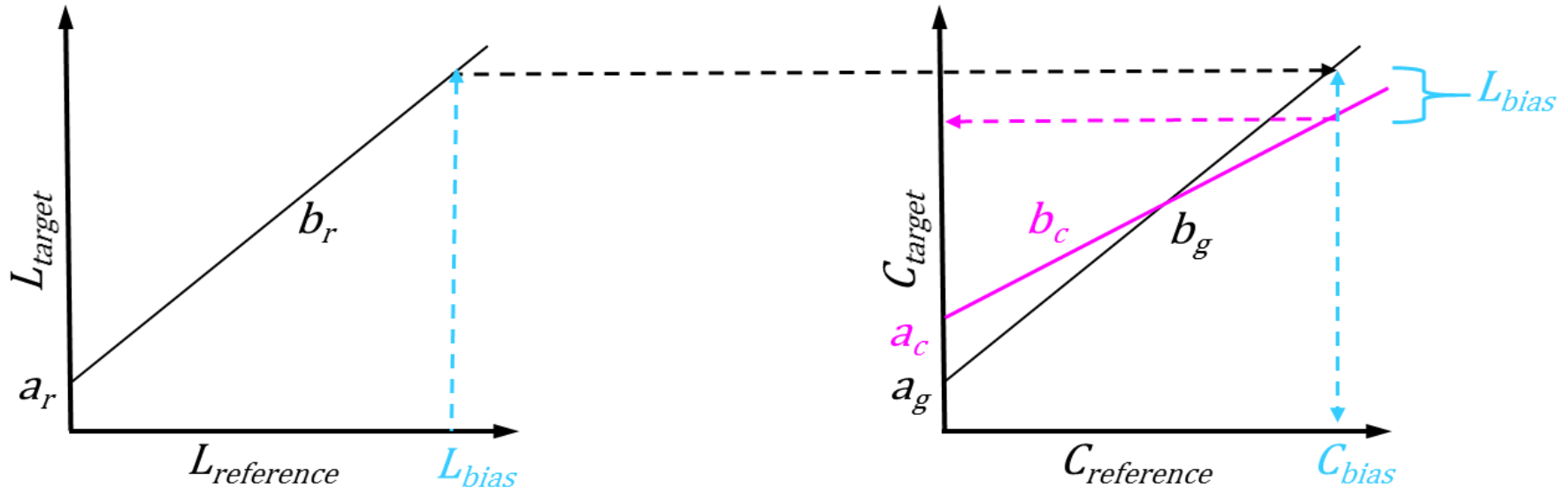
Simultaneous Nadir Overpasses

- Collocation in space and time between to satellite instruments using a tight set of criteria
- Due to time criteria most SNOs occur at the poles.
- SNOs collect in clusters (x,y,t)
 - basis for processing multiple years/platforms
 - can use same definition for radiosonde comparisons.
- The idea is to use a reference standard satellite observation (e.g. IASI) to calibrate a target instrument (e.g. HIRS).

HIRS-N18 IASI-MA SNOS between
2011-04-15 to 2011-04-21



GSICS Inter-Calibration Method



- Adaptation of figure 17 from ATBD showing the relationship between radiances observed by the target instrument (L_{target}) and those observed by reference instrument ($L_{reference}$). The left hand plot shows how the difference between the operation target radiance and corrected radiance (L_{bias}) is related to the newly calculated calibration coefficients (a_g , b_g) that are applied to the target instrument digital counts (C_{target}).

GSICS Inter-Calibration Method

- The GSICS inter-calibration process (Hewison et al, 2013a) uses thousands of SNO pairs from a 14 day window in order to allow comparison of the target and reference instruments via ordinary linear regression (OLS):
 - $L_{target} = a_r + b_r L_{reference}$
- Where a_r and b_r are the correction coefficients and L_{target} is the target instrument radiance. The target radiance is then converted to a consistent reference radiance (\hat{L}_{target}) by inverting the linear relationship in equation 1
 - $\hat{L}_{target} = \left(\frac{1}{b_r}\right) L_{target} - \left(\frac{a_r}{b_r}\right)$

GSICS Inter-Calibration Method

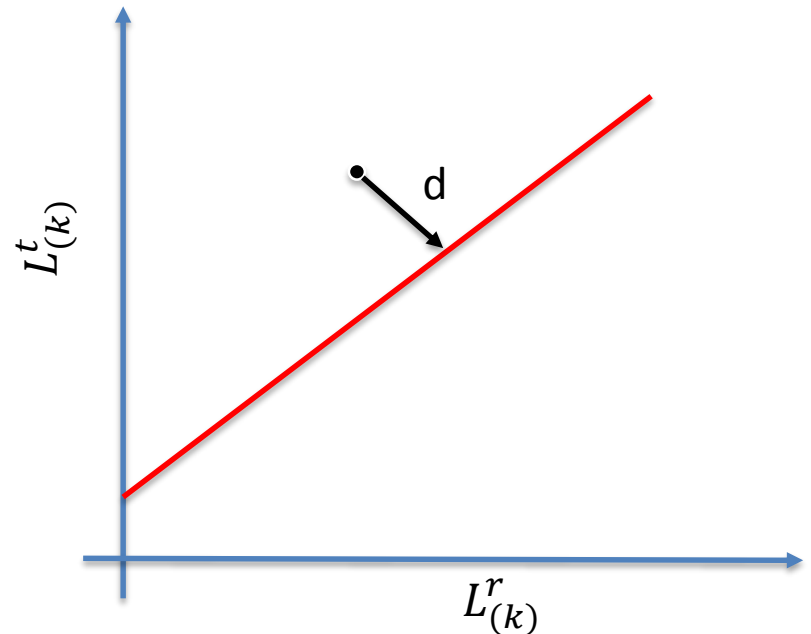
- The operational radiance of the target instrument is calculated from the digital counts (C_{target}):
 - $L_{target} = a_c + b_c C_{target}$
- Where a_c and b_c are the operational offset and slope calibration coefficients respectively (EUMETSAT, 2016). The corrected GSICS radiance can then be calculated directly from the raw counts through the substitution of equation 3 into 2:

$$– \hat{L}_{target} = \left(\frac{a_c - a_r}{b_r} \right) + \left(\frac{b_c}{b_r} \right) C_{target}$$

- The terms $\left(\frac{a_c - a_r}{b_r} \right)$ and $\left(\frac{b_c}{b_r} \right)$ are also referred to as a_g and b_g respectively.

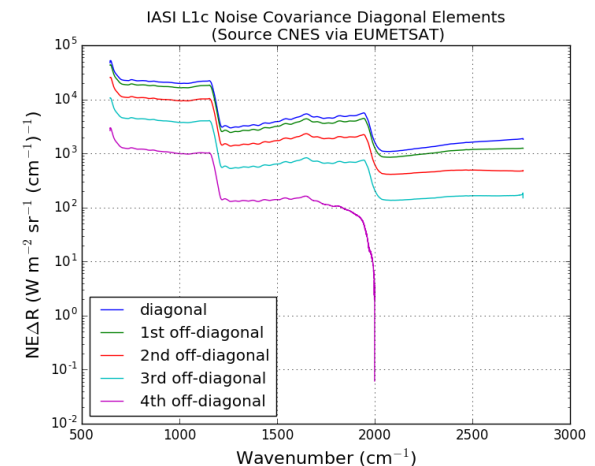
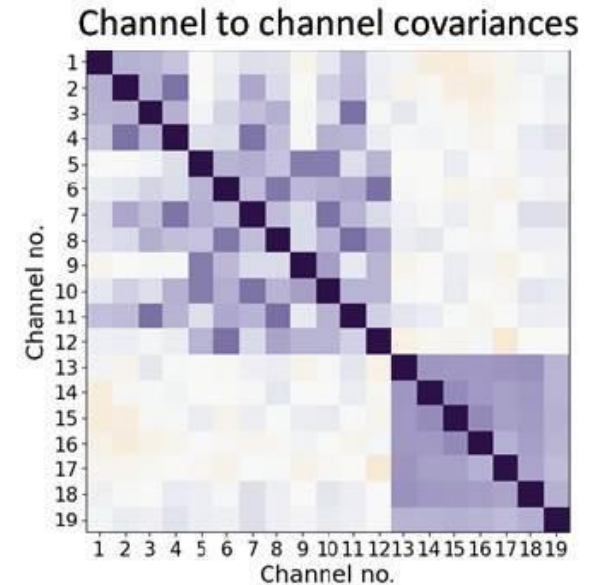
Evolution of the Inter-Calibration Method

- Current approach assumes no uncertainty in either measurement **or** account of the uncertainty in the collocation.
- Need a system that can account for heterogeneity and measurement uncertainties and their correlations.
- 1st we need to repose the problem so we consider uncertainties in both observations.
 - *“What is the distance between a measurement pair $(L_{i(k)}^r, L_{i(k)}^t)$ and the straight line $L_{(k)}^t = a_{(k)}^r + b_{(k)}^r L_{(k)}^r$ where the distance is specified in multiples of the error standard deviations of each measurement?”*



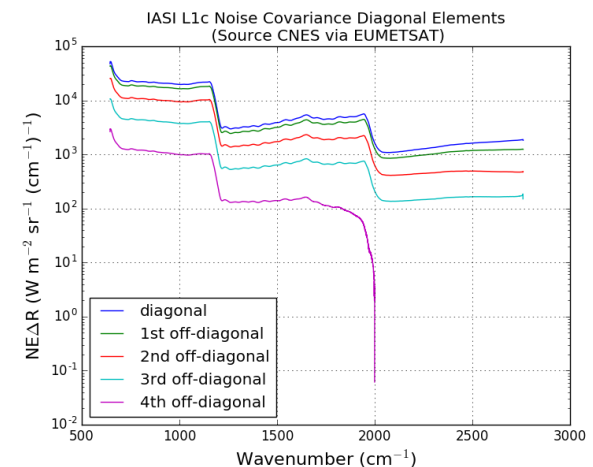
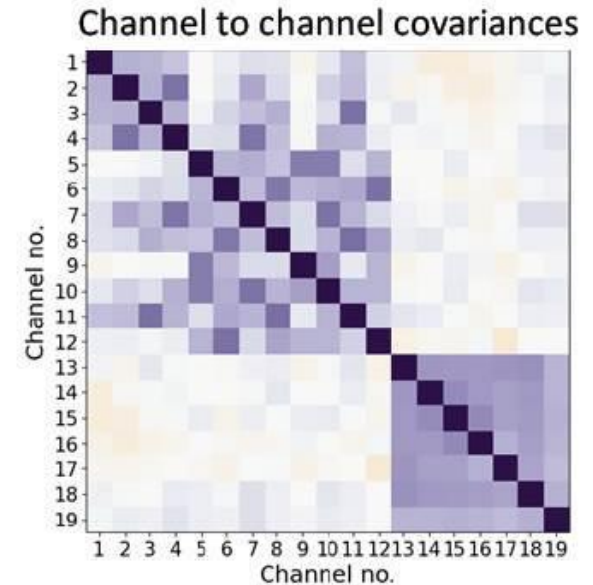
Evolution of the Inter-Calibration Method

- As FCDR products provide detailed covariance's which account for correlations between channels
 - we need to solve for all channels (k) at once (observational packet = 1 spectra).
- For each MMS file there will be m observational packets (i = obs. packet index).



Evolution of the Inter-Calibration Method

- Therefore we can define the following objects:
 - (i) l^r - observation packet for the reference instrument.
 - (ii) l^t - observation packet for the target instrument
 - (iii) a - a k element vector of $a_{(k)}^r$ values.
 - (iv) B – a $k \times k$ matrix whose diagonal values are a vector (b) made up of $b_{(k)}^r$ values.
 - (v) R^r, R^t - $k \times k$ observation error covarinces
- The regression model relating the 2 obs. packets:
 - $l^t = a + Bl^r$



Evolution of the Inter-Calibration Method

- Scaled distance:

$$- d_{i(k)}^2 = (l^r - l_i^r)^T \mathbf{R}^{r-1} (l^r - l_i^r) + \frac{1}{2} (l^t - l_i^t)^T \mathbf{R}^{t-1} (l^t - l_i^t)$$

- Substituting in for l^t (eq. 8):

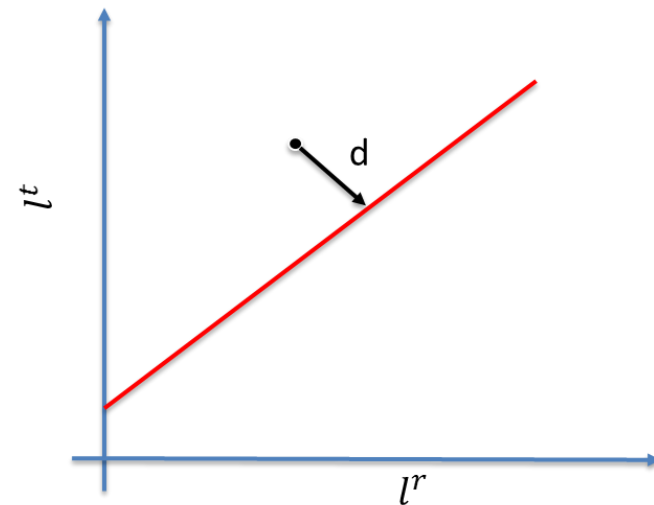
$$- d_{i(k)}^2 = (l^r - l_i^r)^T \mathbf{R}^{r-1} (l^r - l_i^r) + \frac{1}{2} (a + \mathbf{B}l^r - l_i^t)^T \mathbf{R}^{t-1} (a + \mathbf{B}l^r - l_i^t)$$

- By making this stationary we can then find the value of l^r that minimises the distance:

$$- \nabla_{l^r} d_i^2 = \mathbf{R}^{r-1} (l^r - l_i^r) + \mathbf{B} \mathbf{R}^{t-1} (a + \mathbf{B}l^r - l_i^t) = 0$$

- This happens when:

$$- l^r = (\mathbf{R}^{r-1} + \mathbf{B} \mathbf{R}^{t-1} \mathbf{B})^{-1} [\mathbf{R}^{r-1} l_i^r + \mathbf{B} \mathbf{R}^{t-1} (a - l_i^t)]$$

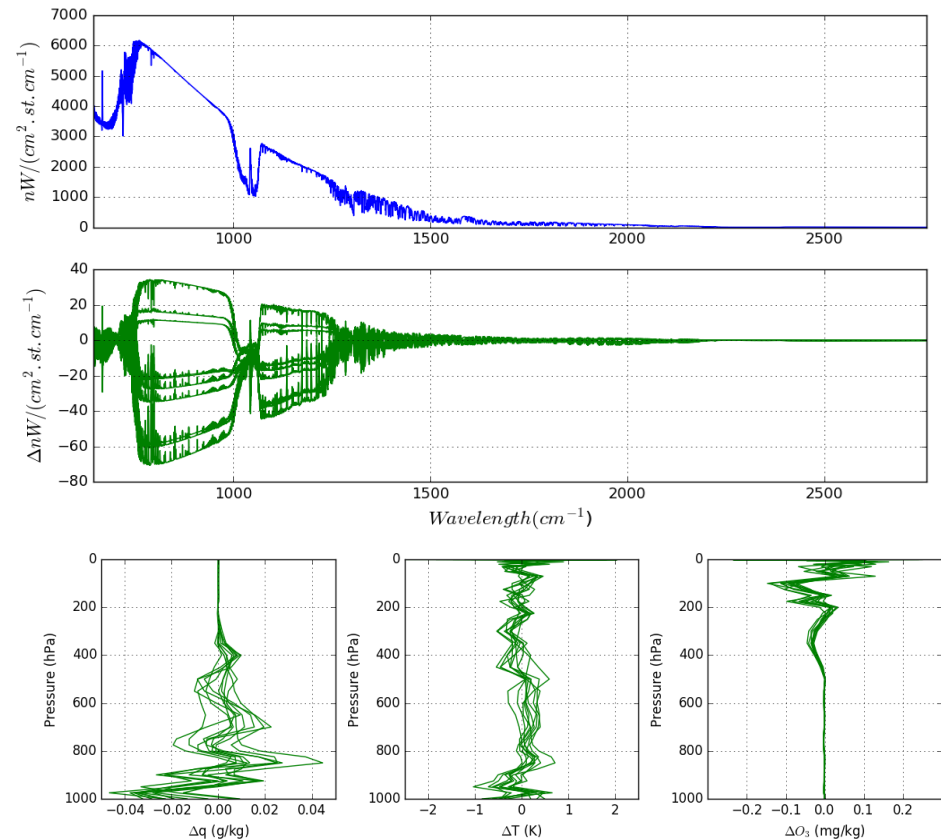


Evolution of the Inter-Calibration Method

- After substitution and factorisation the cost function which a and b minimises is expressed as:
 - $J[a, b] = \frac{1}{2} \sum_i d_i^2 = \frac{1}{2} \sum_{i=0}^m [l_i^t - a - \mathbf{B}l_i^r]^T (\mathbf{R}^t + \mathbf{B}\mathbf{R}^r\mathbf{B})^{-1} [l_i^t - a - \mathbf{B}l_i^r]$
- Still some issues to consider ...

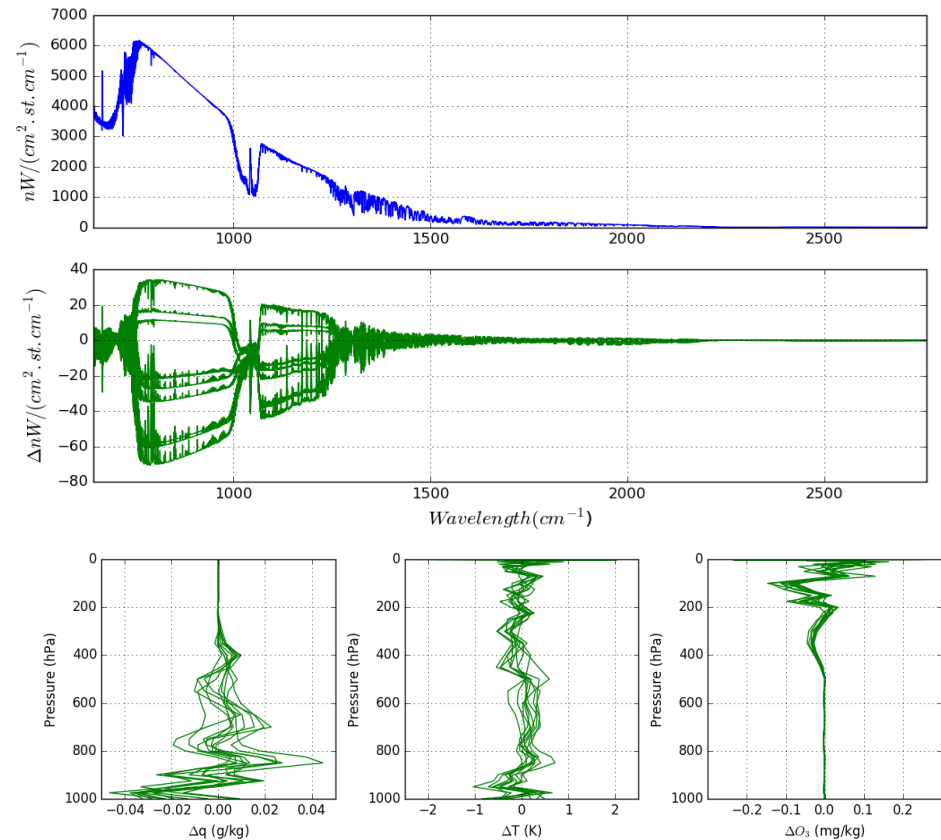
Evolution of the Inter-Calibration Method

- Approach still needs allow for non-coincidental scenes.
- Therefore we apply the following assumptions:
 1. Clear skies (for now)
 2. Access to additional information about the scene from (a) accurate model geophysical variables, and (b) a RTM capable of simulating both instruments.
 3. Both instruments share the same RTM



Evolution of the Inter-Calibration Method

- **Assumption 1:** Employ the IASI L1c cloud flag
- **Assumption 2:** ERA 5 analysis + 10 member ensemble fields
- **Assumption 3:** Reference Forward Model (RFM), line-by-line RTM, can vary all inputs including spectroscopy, emissivity, atmospheric state vector and cloud properties



Evolution of the Inter-Calibration Method

- Using these assumptions we can continue to adapt the method to account for non-coincidence.
- Expand the list of defined variables to include terms based on new assumptions.
- Define a set of relationships that link them.

Variable	Description	Variable	Description
x^r	Atmospheric state vector for the reference instrument	x^t	Atmospheric state vector for the target instrument
δx^r	Uncertainty on the reference instrument state vector	δx^t	Uncertainty on the target instrument state vector
θ^r	Viewing angle of the reference instrument	θ^t	Viewing angle of the target instrument
x_{true}^r	True version of x^r	x_{true}^t	True version of x^t
l^r	Observation packet from the reference instrument	l^t	Observation packet from the target instrument
δl^r	Uncertainty on reference observational packet	δl^t	Uncertainty on target observational packet
l_{true}^r	Noise-free observation packet that a perfect reference instrument would observe	l_{true}^t	Noise-free observation packet that a perfect target instrument would observe
$h(x, \theta)$	The RTM output	ϵ^h	Uncertainty of RTM output
δx^{rt}	Difference in reference and target state vectors	$\delta \theta^{rt}$	Difference in viewing angles of reference and target instruments.

Evolution of the Inter-Calibration Method

- We want to regress l^t to something close to l_{true}^t
- Using the defined relationships we can develop an expression for l_{true}^t
 - $l_{true}^t = h(x_{true}^t, \theta^t) - \epsilon_2^h$
 - $l_{true}^t = l^r - \delta l^r + \epsilon_1^h + \mathbf{H}_{x^r, \theta^r} \delta x^r + \mathbf{H}_{x^r, \theta^r} \delta x^{rt} + \mathbf{H}_{\theta^r, x^r} \delta \theta^{rt} - \mathbf{H}_{x^t, \theta^t} \delta x^t - \epsilon_2^h$
- This allows us to relate the l^r to the l_{true}^t , rearranging allows us to collect the terms:

$$l_{true}^t = \underbrace{l^r + \mathbf{H}_{x^r, \theta^r} \delta x^{rt} + \mathbf{H}_{\theta^r, x^r} \delta \theta^{rt}}_{\text{known/calculable}} + \underbrace{-\delta l^r + \mathbf{H}_{x^r, \theta^r} \delta x^r - \mathbf{H}_{x^t, \theta^t} \delta x^t}_{\text{statistics known}} + \underbrace{\epsilon_1^h - \epsilon_2^h}_{?}$$

Reference measurement (l_i^r)

Reference Error covariance (R^r)

Evolution of the Inter-Calibration Method

$$\langle \delta \mathbf{l}^r \delta \mathbf{l}^{rT} \rangle$$

$$J[\mathbf{a}, \mathbf{b}] = \frac{1}{2} \sum_i d_i^2 = \frac{1}{2} \sum_{i=1}^m [\mathbf{l}_i^t - \mathbf{a} - \mathbf{B} \mathbf{l}_i^r]^T (\mathbf{R}^t + \mathbf{B} \mathbf{R}^r \mathbf{B})^{-1} [\mathbf{l}_i^t - \mathbf{a} - \mathbf{B} \mathbf{l}_i^r]$$

$$\mathbf{l}_{\text{true}}^t = \underbrace{\mathbf{l}^r + \mathbf{H}_{\mathbf{x}^r, \theta^r} \delta \mathbf{x}^{rt} + \mathbf{H}_{\theta^r, \mathbf{x}^r} \delta \theta^{rt}}_{\text{known/calculable}} + \underbrace{-\delta \mathbf{l}^r + \mathbf{H}_{\mathbf{x}^r, \theta^r} \delta \mathbf{x}^r - \mathbf{H}_{\mathbf{x}^t, \theta^t} \delta \mathbf{x}^t}_{\text{statistics known}} + \underbrace{\epsilon_1^h - \epsilon_2^h}_{?}$$

Reference measurement (l_i^r)

Reference Error covariance (R^r)

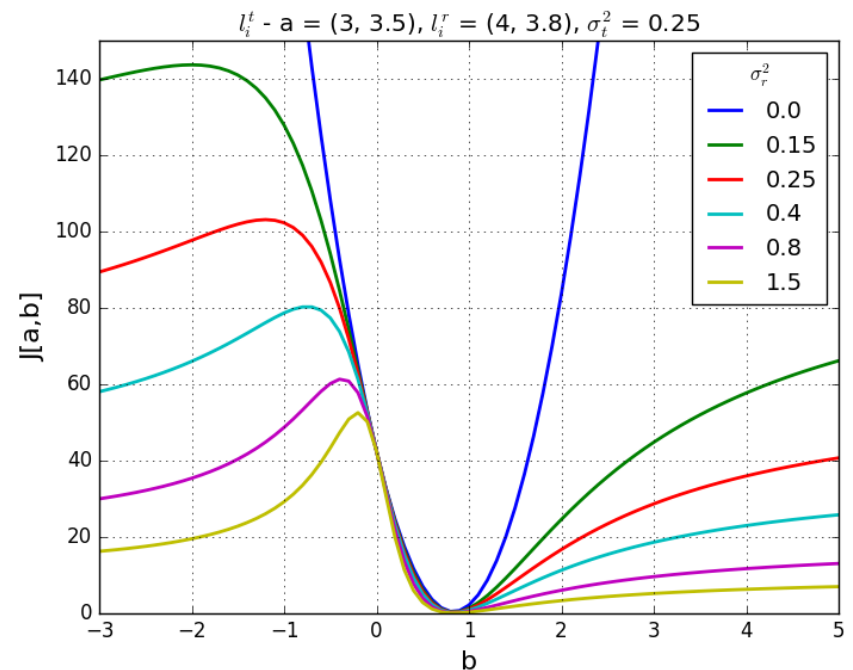
$$\mathbf{R}^r = \langle \delta \mathbf{l}^r \delta \mathbf{l}^{rT} \rangle + \langle (\mathbf{H}_{\mathbf{x}^r, \theta^r} \delta \mathbf{x}^r) (\mathbf{H}_{\mathbf{x}^r, \theta^r} \delta \mathbf{x}^r)^T \rangle + \langle (\mathbf{H}_{\mathbf{x}^t, \theta^t} \delta \mathbf{x}^t) (\mathbf{H}_{\mathbf{x}^t, \theta^t} \delta \mathbf{x}^t)^T \rangle + \langle (\mathbf{H}_{\mathbf{x}^r, \theta^r} \delta \mathbf{x}^r) (\mathbf{H}_{\mathbf{x}^t, \theta^t} \delta \mathbf{x}^t)^T \rangle + \langle (\mathbf{H}_{\mathbf{x}^t, \theta^t} \delta \mathbf{x}^t) (\mathbf{H}_{\mathbf{x}^r, \theta^r} \delta \mathbf{x}^r)^T \rangle$$

Evolution of the Inter-Calibration Method

- Final thing to consider is how to minimise the cost function.

$$J[\mathbf{a}, \mathbf{b}] = \frac{1}{2} \sum_i d_i^2 = \frac{1}{2} \sum_{i=1}^m [\mathbf{l}_i^t - \mathbf{a} - \mathbf{B}\mathbf{l}_i^r]^T (\mathbf{R}^t + \mathbf{B}\mathbf{R}^r\mathbf{B})^{-1} [\mathbf{l}_i^t - \mathbf{a} - \mathbf{B}\mathbf{l}_i^r]$$

- The problem is not quadratic. Example: for a case of 2 pairs, single channel and R_r diagonal elements of possible values range from 0-1.5. This results in non-quadratic cost function with a less well-defined minimum, an asymmetric profile around the minimum, and a nearby maximum.
- Currently looking at methods to solve this, potential to use block gradient descent approach.



Outlook

- Functional pre-processing stage, which includes:
 - Scripts for downloading ERA5 data based on cluster information
 - User defined RTM functionality regarding Emissivity (λ), trace gases and clouds
- Creation of regression tool box
- Initial setup for RFM complete and tested for optimisation. Each SNO case requires 24 RFM runs jobs are submitted as an array of all cases in a cluster x 24.
- Currently working with 1 test cluster – can run others when happy with RFM setup (no FCDR data needed at this stage).
- Create cluster meta data
- Build in ability to process collocations with GRUAN
 - additional complexity
 - GRUAN Processor (GAIA-CLIM)
- Inter-comparison between FIDUCEO methodologies
 - (see <http://www.fiduceo.eu/content/gracious-wobbling-new-dancing>)

Thank you for listening