

Evolving the GSICS Approach to Inter-Satellite Calibration

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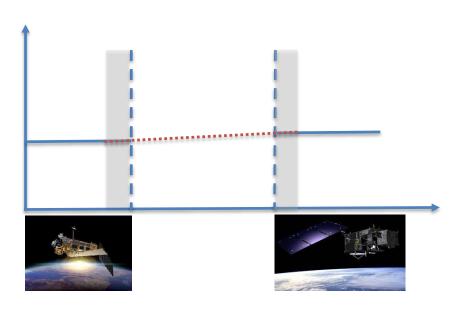


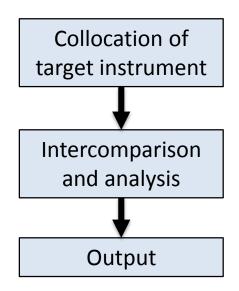




Aims

 Discuss framework for inter-comparisons during (A)ATSR-SLSTR gap.





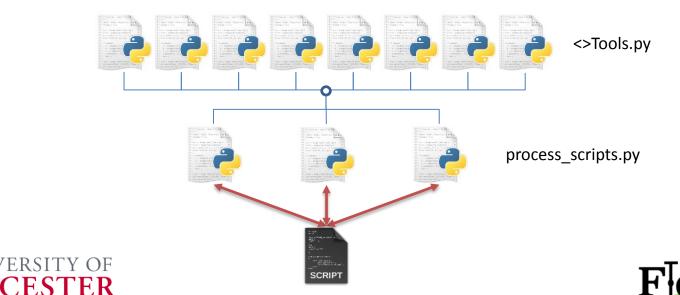
 Goal is to build a generic framework that can handle different intercomparison types: (i) satellite-tosatellite and (ii) satellite-to-in situ





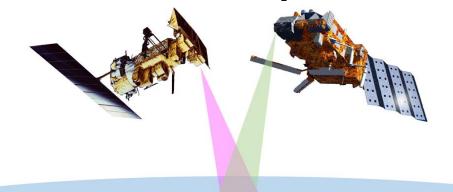
Aims

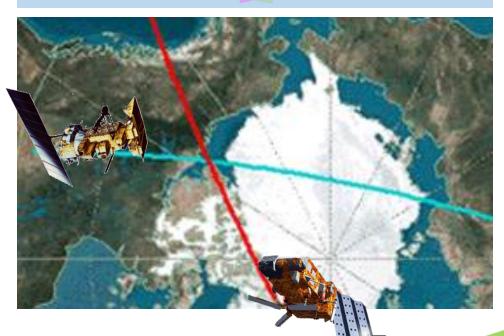
- Introduce current GSICS approach to inter-satellite calibration through simultaneous nadir overpasses (SNOs)
- Discuss briefly some of the issues of using this approach within FIDUCEO
- Present an evolved approach within the GSICS framework
 - Software tools: flat pack approach, "some assembly required"



Simultaneous Nadir Overpasses

- A simultaneous nadir overpass is a common technique used to compare observations from 2 satellites.
- Global Space-Based Inter-Calibration System (GSICS) conduct routine calibration of NWP satellites using SNOs.
- The idea is to use a reference standard satellite observation (e.g. IASI) to calibrate a target instrument (e.g. HIRS).



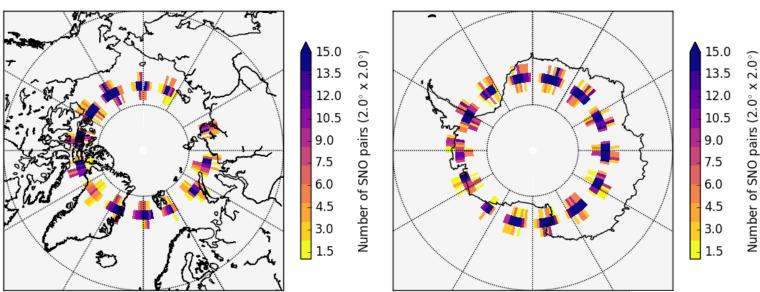




Simultaneous Nadir Overpasses

- 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.

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



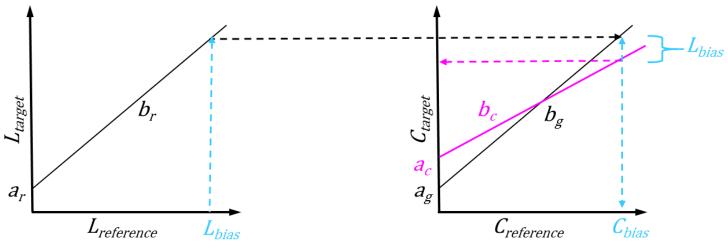




GSICS Inter-Calibration Method

- 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

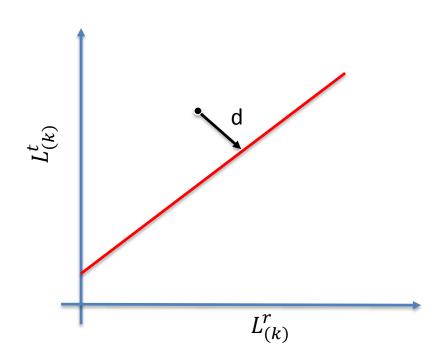
$$- \hat{L}_{target} = \left(\frac{1}{b_r}\right) L_{target} - \left(\frac{a_r}{b_r}\right)$$







- GSICS 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 $(L^r_{i(k)}, L^t_{i(k)})$ and the straight line $L^t_{(k)} = a^r_{(k)} + b^r_{(k)}L^r_{(k)}$ where the distance is specified in multiples of the error standard deviations of each measurement?"



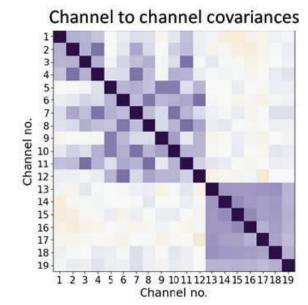


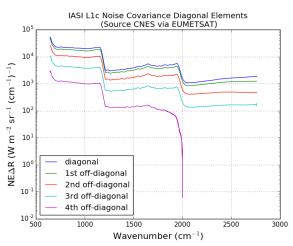


Method

- As FCDR products provide detailed covariances 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).
- 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) \mathbf{B} a k x k matrix whose diagonal values are a vector (b) made up of $b_{(k)}^r$ values.
 - (v) \mathbf{R}^r , \mathbf{R}^t k x k observation error covarinces
- The regression model relating the 2 obs. packets:

$$- l^t = a + Bl^r$$
 [8]









 After substitution and factorisation the cost function which a and b minimises is expressed as:

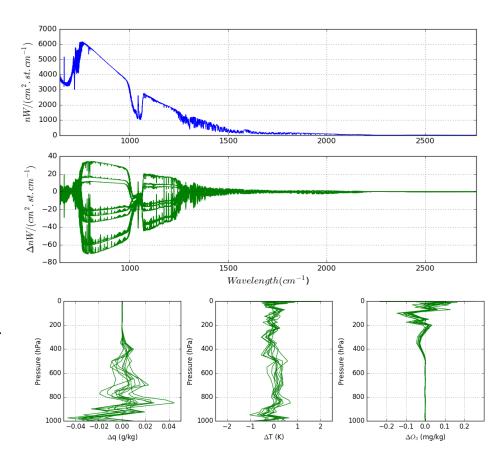
$$- J[a,b] = \frac{1}{2} \sum_{i} d_{i}^{2} = \frac{1}{2} \sum_{i=0}^{m} \left[l_{i}^{t} - a - B l_{i}^{r} \right]^{T} (R^{t} + B R^{r} B)^{-1} \left[l_{i}^{t} - a - B l_{i}^{r} \right]$$

• Still some issues to consider ...





- Approach still needs allow for noncoincidental 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
- 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







- Using these assumptions we can continue to adapt the method to account for non-coincidence.
- Expand the list of defined variables to include terms base don new assumptions.

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





Evolution of the Inter-Calibration Method (\delta \text{I}^{\dagger \text{I}})

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

$$\begin{aligned} \mathbf{l}_{\text{true}}^{\text{t}} = & \underbrace{\mathbf{I}^{\text{r}} + \mathbf{H}_{\mathbf{x}^{\text{r}},\theta^{\text{r}}} \delta \mathbf{x}^{\text{rt}} + \mathbf{H}_{\theta^{\text{r}},\mathbf{x}^{\text{r}}} \delta \theta^{\text{r}}}_{\text{known/calculable}} + \underbrace{-\delta \mathbf{l}^{\text{r}} + \mathbf{H}_{\mathbf{x}^{\text{r}},\theta^{\text{r}}} \delta \mathbf{x}^{\text{r}} - \mathbf{H}_{\mathbf{x}^{\text{t}},\theta^{\text{t}}} \delta \mathbf{x}^{\text{t}}}_{?} + \underbrace{\epsilon_{1}^{\mathbf{h}} - \epsilon_{2}^{\mathbf{h}}}_{?} \end{aligned}$$

$$\mathbf{Reference\ measurement}\ (l_{i}^{r}) \qquad \mathbf{Reference\ Error\ covariance}\ (\mathbf{\textit{R}}^{r})$$

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

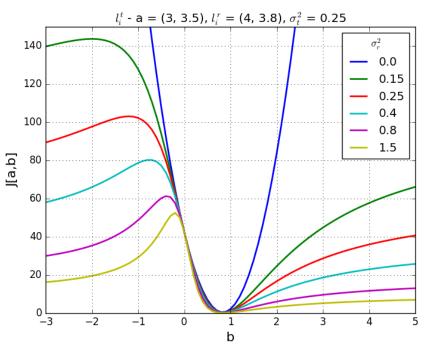




Final thing to consider is how to minimise the cost function.

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

- The problem is not quadratic. Example: for a case of 2 pairs, single channel and Rr 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.







Thank you for listening



