

FIDUCEO has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no. 638822

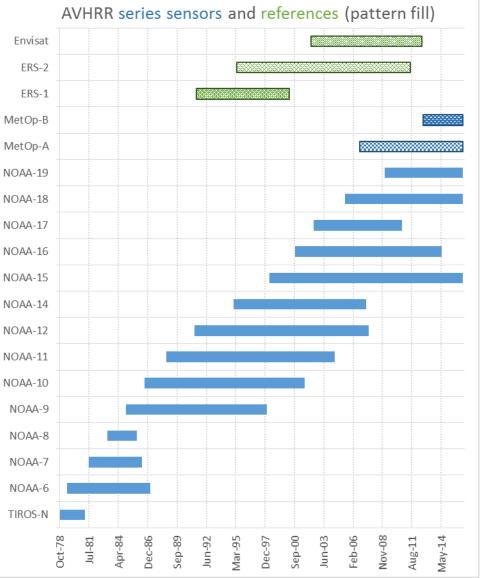


Activities of the FIDUCEO project: www.fiduceo.eu

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Information in historical sensors



- How to get meaningful trend analysis?
- Recalibration (traceably) of historical sensors



FIDUCEO FCDRs (L1)

FCDR: fundamental climate data record (calibrated radiances) from which climate data can be derived

DATASET	NATURE	POSSIBLE USES
AVHRR FCDR	Harmonised infra-red radiances and best available reflectance radiances, 1982 - 2016	SST, LSWT , aerosol , LST, phenology, cloud properties, surface reflectance
HIRS FCDR	Harmonised infra-red radiances, 1982 - 2016	Atmospheric humidity, NWP re-analysis, stratospheric aerosol
MW Sounder FCDR	Harmonised microwave BTs for AMSU-B and equivalent channels, 1992 – 2016	Atmospheric humidity, NWP re-analysis
Meteosat VIS FCDR	Improved visible spectral response functions and radiance 1982 to 2016	Albedo, aerosol, NWP re- analysis, cloud, wind motion vectors,





How do we get metrological rigour in historical sensors?

Start from the measurement equation

Understand and quantify correlation

Use harmonisation approaches to recalibrate sensors





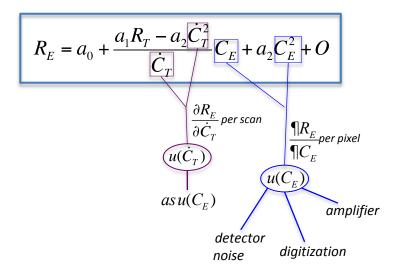
The measurement equation defines the relationship between counts and radiance (or reflectance)

$$R_{E} = a_{0} + \frac{a_{1}R_{T} - a_{2}\dot{C}_{T}^{2}}{\dot{C}_{T}}C_{E} + a_{2}C_{E}^{2} + O$$



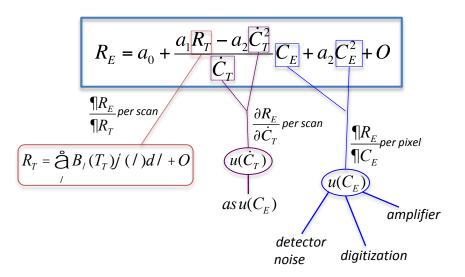


Each term in the measurement equation has associated uncertainty from one or more effects





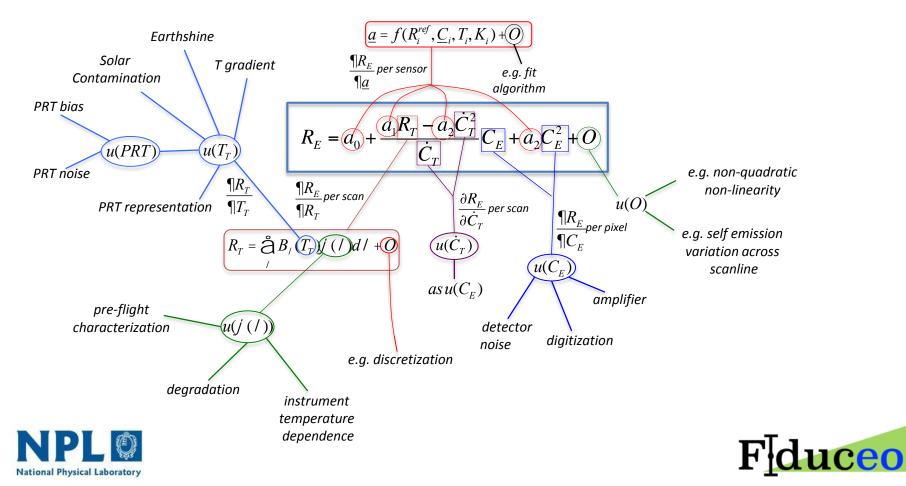








We include a +0 to relate to errors due to approximations in the equation form



Capture in an effects table

Table descriptor		Value / Expression		How this is provided	Notes
Name of effect					
Affected term in measurement function					
Correlation type and form	within scanline [pixels]		8×10 ⁵		
	from scanline to scanline [scanlines]		6×10 ⁵	Г Г Г	
	between orbits [orbits]		- 4×10 ⁹ — -		
	Across time [e.g. days, months, years]		2×10 ⁵		
Correlation scale	within scanline [pixels]		-		
	from scanline to scanline [scanlines]		-0.3	0.2 PRT difference wit plane	-0.1 -0.0 distribution (K)
	between orbits [orbits]	0.8 0.7 + + Cold	7	· ·	
Channels / bands	Across time List channels and bands affected	(¥) 0.6 Generation Warm (Generation 0.5 (H) 0.4 (H) 0.3 (H) 0	on		
	Correlation matrix				
Uncertainty	PDF shape	⊎ 0.2 0.1	Controller of		
	Uncertainty units	0.0	+	, 	
Sensitivity Coefficient	Uncertainty magnitude	1982.01		1983.01	1984.01

Traceable uncertainty

- Traceability diagram, measurement centred
 - to organise
 - to document
- Branching structure reflects the nature of the problem
- Standardised "effects table" per "twig"
 - systematic documentation
 - this is codified into FCDR format
- Same for deriving higher-order products (CDRs)
 - uncertainty from L1 is simply one of the effects in L2





Error correlation

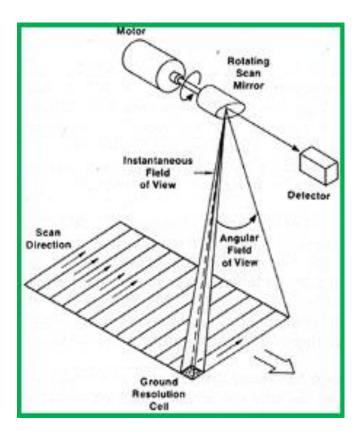
- Is different from effect correlation
- (Metrologists often forget to say "error")

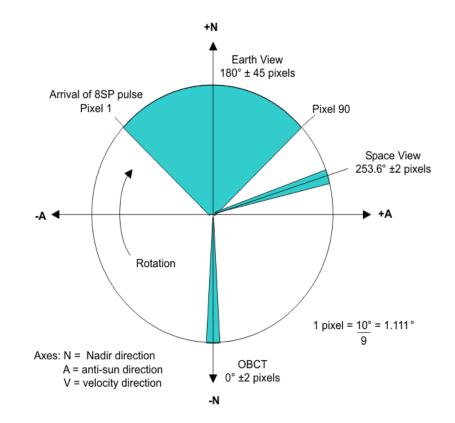
- Matters in higher level processing:
 - Combining values from different channels
 - Combining values from different pixels





Error correlation: something in common









When it can be described explicitly

 Error between bands due to common blackbody calibration target

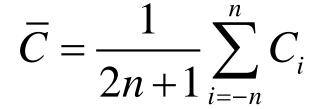
$$\tilde{L}_{\text{ICWT,A}} = \frac{\varepsilon_{\text{A}} c_{1,L}}{\lambda_{\text{A}}^{5} \left(\exp\left[c_{2}/\lambda_{\text{A}}T\right] - 1 \right)}$$
$$\tilde{L}_{\text{ICWT,B}} = \frac{\varepsilon_{\text{B}} c_{1,L}}{\lambda_{\text{B}}^{5} \left(\exp\left[c_{2}/\lambda_{\text{B}}T\right] - 1 \right)}$$
$$\left(\tilde{L}_{\text{ICWT,A}}, \tilde{L}_{\text{ICWT,B}} \right) = \frac{\partial \tilde{L}_{\text{ICWT,A}}}{\partial T} \frac{\partial \tilde{L}_{\text{ICWT,B}}}{\partial T} u^{2} \left(T \right)$$



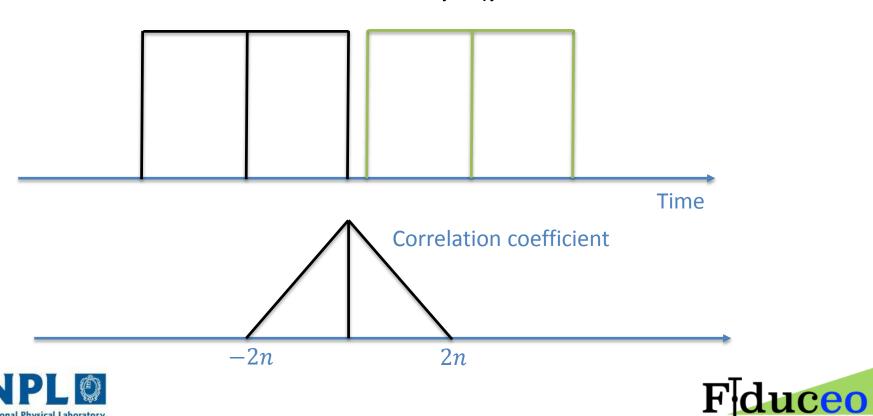
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Rolling averages



Moving simple average



National Physical Laboratory

Numerical approach to correlation analysis

$$r(x,y) = \frac{\sum_{i=1}^{n} (x_i - \overline{x})(y_i - \overline{y})}{\sqrt{\sum_{i=1}^{n} (x_i - \overline{x})^2} \sqrt{\sum_{i=1}^{n} (y_i - \overline{y})^2}} = \frac{1}{n-1} \sum_{i=1}^{n} \left(\frac{x_i - \overline{x}}{s_x}\right) \left(\frac{y_i - \overline{y}}{s_y}\right)$$

Pearson correlation Channel no. Channel no 9 1011 1213 14 15 16 17 18 19 Channel no.

Correlation between noise in different channels for HIRS



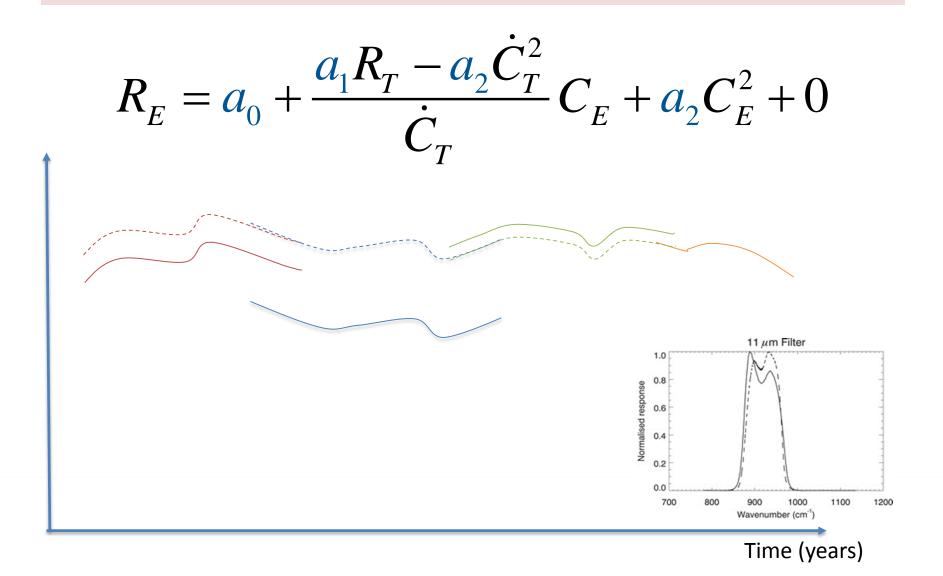


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	between orbits [orbits]	•	ic / Rectan	gular
	Across time [e.g. days, months, years]	absolute • Triangular	r (simple a	verage)
	within scanline [pixels] from scanline to scanline [scanlines]	Truncated	、 1	(weighted
	between orbits [orbits] Across time	 Repeating (orbital ef 	-	d Gaussian
Channels / bands	List channels and bands affected Correlation matrix			
Uncertainty	PDF shape Uncertainty units Uncertainty magnitude			
Soncitivity Coofficient				

Sensitivity Coefficient

Harmonisation



Harmonisation model

- Model for spectral radiance measured by each sensor L = f(a; x, y, ...)
- Model for adjustment between pairs of sensors

$$K = h[f(\boldsymbol{a}_{s}; x_{s}, y_{s}, ...)] - \begin{cases} h[f(\boldsymbol{a}_{t}; x_{t}, y_{t}, ...)] \\ h[L_{\text{ref}}] \end{cases}$$

- *a* (unknown) sensor calibration parameters
- x, y, ... stimulus variables

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earth counts, temperatures, ...

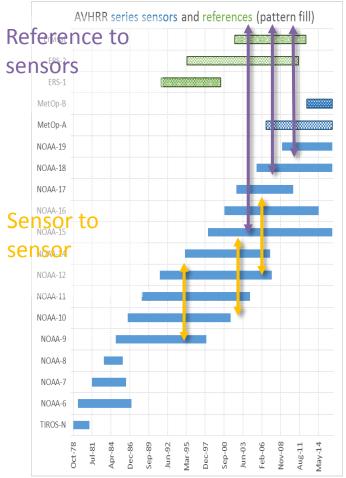
*L*_{ref} radiances from single reference sensor

adjustment factor

Data with uncertainty information

F<u>duce</u>

Match-ups



- sensor-to-sensor
 Many (50 million +)
 - Correlated!

• Reference radiance, or





Solving the harmonisation problem

Harmonisation problem is a non-linear regression, with correlated data and millions of match-ups.

Approaches:

- Orthogonal distance regression + Monte C
- wonte cests-in-variables approach, takin oor advantage of sparsity in covariance in Province Norkin Province Full errors-in-variables approach, taking



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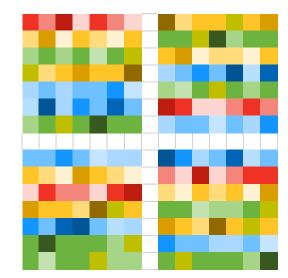


Sharing the FCDR

- Full FCDR:
 - Uncertainty data by correlation structure

$$u^{2}(R_{E,ijk}) = c_{a_{0}}^{2}u^{2}(a_{0}) + c_{C_{E,ijk}}^{2}u^{2}(C_{E,ijk}) + c_{R_{ICT,jk}}^{2}u^{2}(R_{ICT,jk}) + c_{\delta R_{ICT,0}}^{2}u^{2}(\delta R_{ICT,0}) + c_{\delta R_{ICT,0,grad,jk}}^{2}u^{2}(\delta R_{ICT,0,grad,jk}) + c_{C_{ICT,jk}}^{2}u^{2}(C_{ICT,jk})$$

• Ensemble of realisations



uceo

• "Easy FCDR" with guidance

 \prec



independent random systematic and structured random



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Thank you!

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