

Activities of the FIDUCEO project: www.fiduceo.eu

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ray**ference** 👤









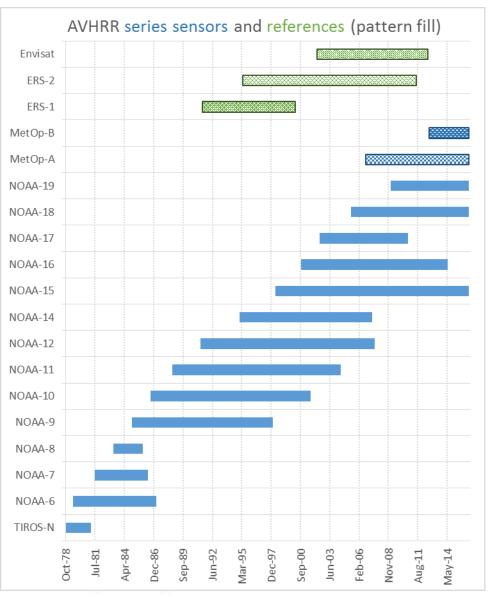








Information in historical sensors



- 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





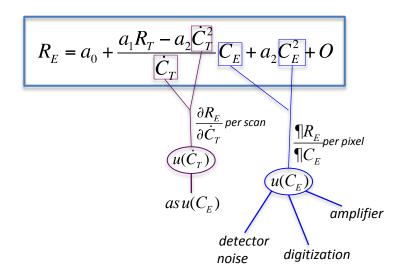
The measurement equation defines the relationship between counts and radiance (or reflectance)

$$R_E = a_0 + \frac{a_1 R_T - a_2 \dot{C}_T^2}{\dot{C}_T} C_E + a_2 C_E^2 + O$$



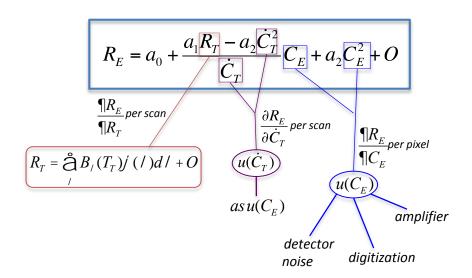


Each term in the measurement equation has associated uncertainty from one or more effects





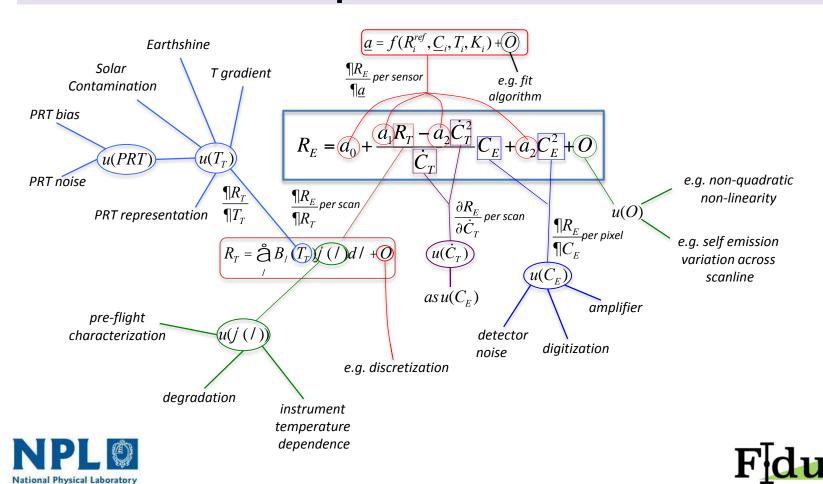




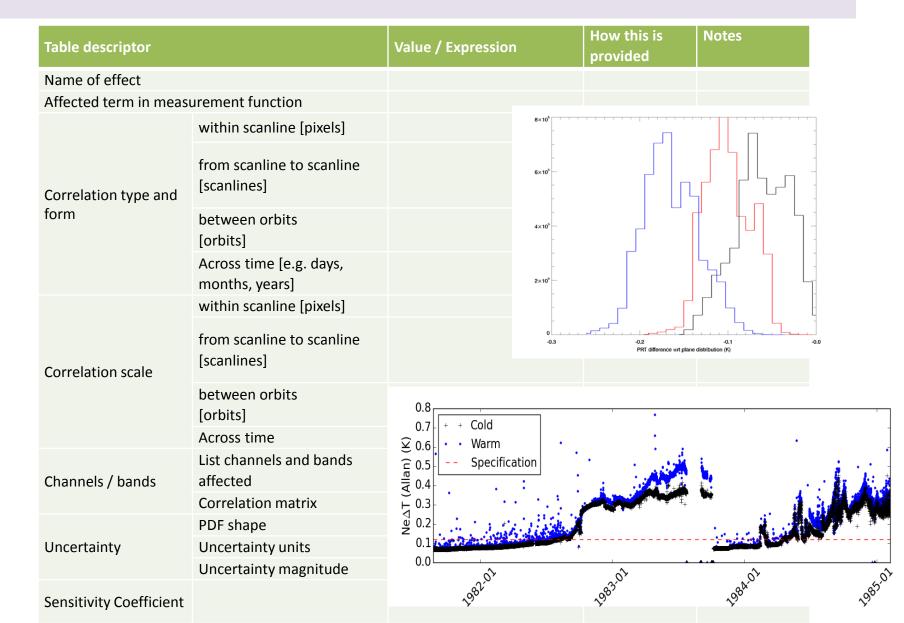




We include a +0 to relate to errors due to approximations in the equation form



Capture in an effects table



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

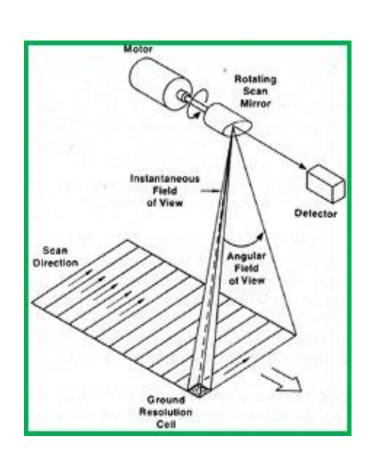
- Is different from effect correlation
- (Metrologists often forget to say "error")

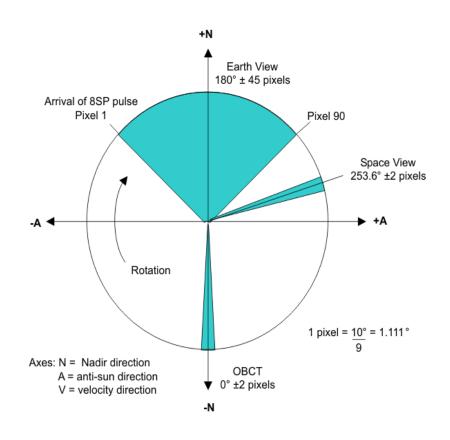
- Matters in higher level processing:
 - Combining values from different channels
 - Combining values from different pixels





Error correlation: something in common









When it can be described explicitly

 Error between bands due to common blackbody calibration target

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

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

$$u\left(\tilde{L}_{\text{ICWT,A}}, \tilde{L}_{\text{ICWT,B}}\right) = \frac{\partial \tilde{L}_{\text{ICWT,A}}}{\partial T} \frac{\partial \tilde{L}_{\text{ICWT,B}}}{\partial T} u^{2} \left(T\right)$$

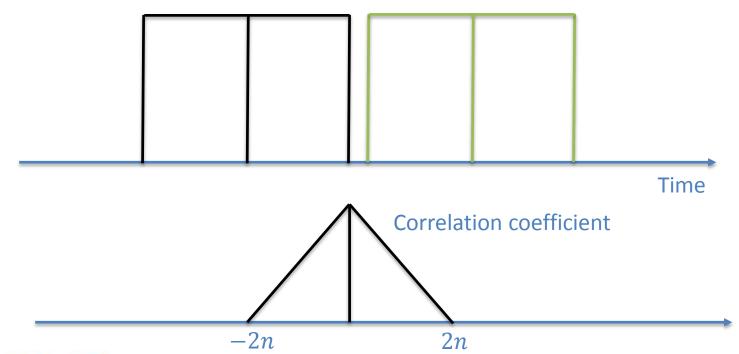




Rolling averages

$$\overline{C} = \frac{1}{2n+1} \sum_{i=-n}^{n} C_i$$

Moving simple average

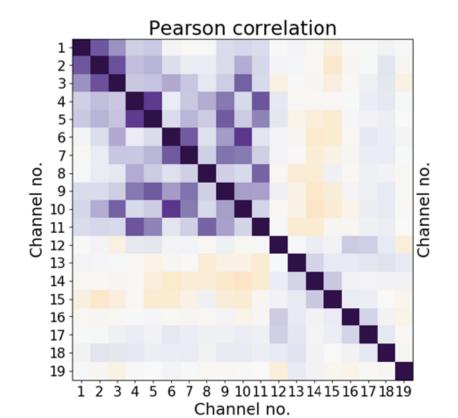






Numerical approach to correlation analysis

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



Correlation between noise in different channels for HIRS



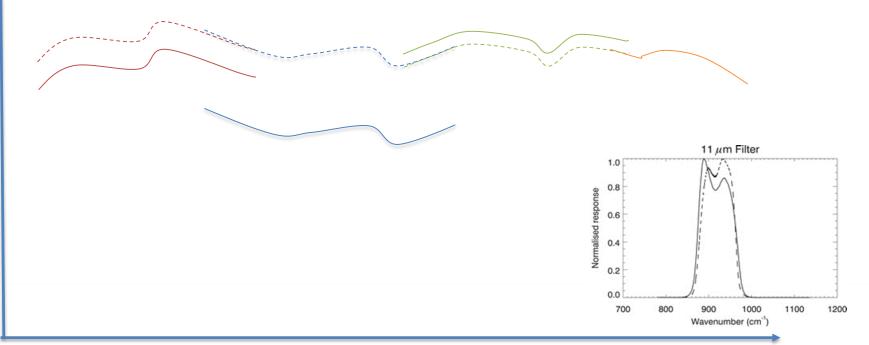


Capture in an effects table

| Table descriptor | | Value / Expression | How this is provided | Notes | |
|---------------------------------------|----------------------------------------|-----------------------------------------------------------------------------|----------------------|----------|--|
| Name of effect | | | | | |
| Affected term in measurement function | | | | | |
| Correlation type and form | within scanline [pixels] | | | | |
| | from scanline to scanline [scanlines] | Correlatio • Random | n forms: | | |
| | between orbits [orbits] | Systematic / Rectangular | | | |
| | Across time [e.g. days, months, years] | absolute • Triangula | r (simple a | verage) | |
| Correlation scale | within scanline [pixels] | _ | • | . | |
| | from scanline to scanline [scanlines] | Truncated Gaussian (weighted average and other effects) | | | |
| | between orbits [orbits] | Repeating truncated Gaussian (orbital effects) | | | |
| | Across time | () | , | | |
| Channels / bands | List channels and bands affected | | | | |
| | Correlation matrix | | | | |
| Uncertainty | PDF shape | | | | |
| | Uncertainty units | | | | |
| | Uncertainty magnitude | | | | |
| Sensitivity Coefficient | | | | | |

Harmonisation

$$R_E = a_0 + \frac{a_1 R_T - a_2 \dot{C}_T^2}{\dot{C}_T} C_E + a_2 C_E^2 + 0$$



Time (years)

Harmonisation model

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

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

a (unknown) sensor calibration parameters

x, y, ... stimulus variables earth counts, temperatures, ...

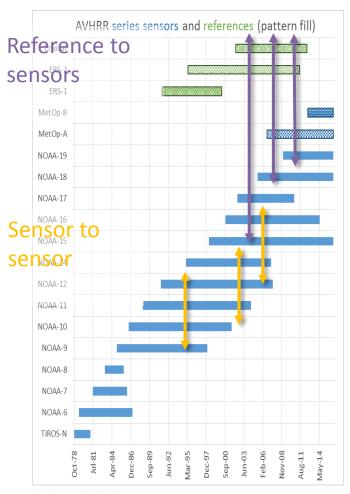
 $L_{\rm ref}$ radiances from single reference sensor

K adjustment factor

Data with uncertainty information



Match-ups



- Reference radiance, or sensor-to-sensor
- Many (50 million +)
- Correlated!





Solving the harmonisation problem

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

Approaches:

- Orthogonal distance regression + Monte
- • Full errors-in-variables approach, takin





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Sharing the FCDR

- Full FCDR:
 - Uncertainty data by correlation structure

$$u^{2}(R_{E,ijk}) = c_{a_{0}}^{2}u^{2}(a_{0}) + c_{C_{E,ijk}}^{2}u^{2}(C_{E,ijk})$$

$$+c_{R_{ICT,jk}}^{2}u^{2}(R_{ICT,jk})$$

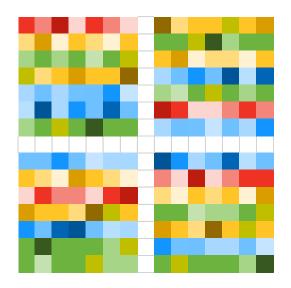
$$+c_{\delta R_{ICT,0}}^{2}u^{2}(\delta R_{ICT,0})$$

$$+c_{\delta R_{ICT,0,grad,jk}}^{2}u^{2}(\delta R_{ICT,0,grad,jk})$$

$$+c_{C_{ICT,jk}}^{2}u^{2}(C_{ICT,jk})$$

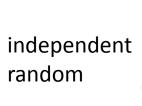


 Ensemble of realisations



 "Easy FCDR" with guidance

random



systematic and structured random







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Thank you!

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