

RESPONSE FUNCTION-BASED CALIBRATION OF IMAGING SPECTROMETERS

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DLR Earth Observation Center (EOC)



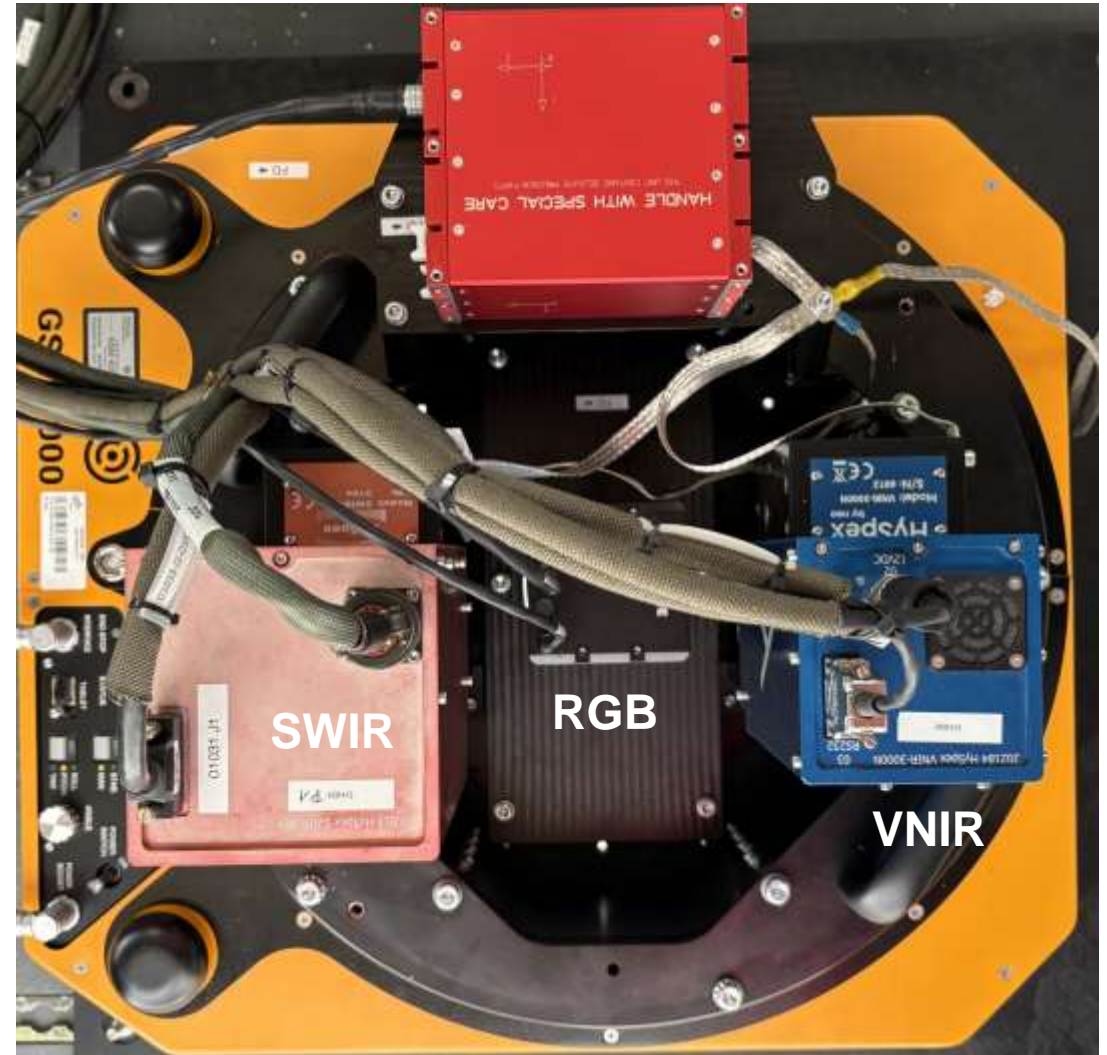
HySpex Sensor System

HySpex VNIR-3000N

- $N \rightarrow$ Nyquist (sampled)
- 700 bands, 3408 pixels
- Spectral Range: 400 – 994 nm
- FWHM: 3 – 7 nm

HySpex SWIR-384

- 288 bands, 384 pixels
- Spectral range: 952 – 2516 nm
- FWHM: 5.5 – 8.5 nm



Calibration Home Base (CHB)



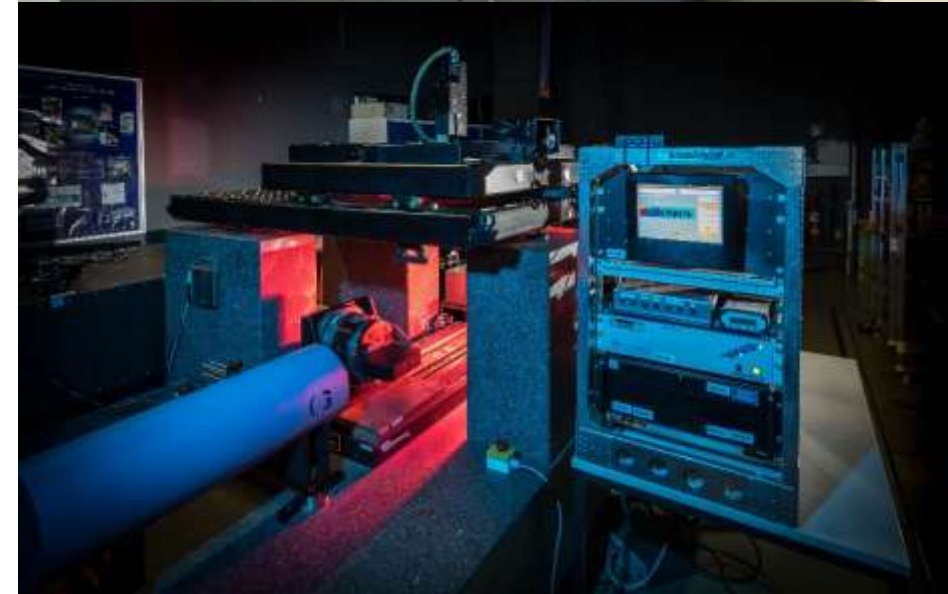
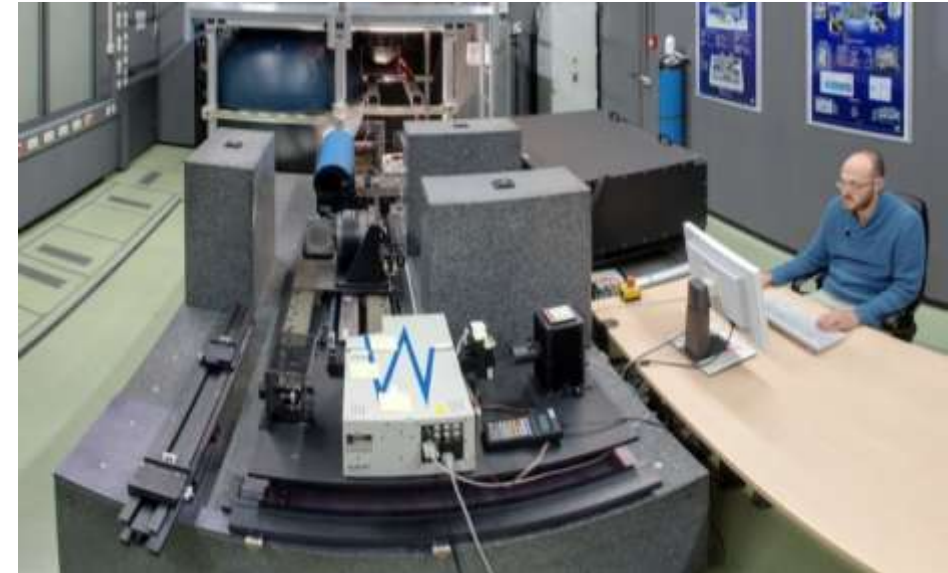
Operational Setups

- Geometric Response Function
- Spectral Response Function
- Radiometric Response & Polarization

Co-funded by ESA for APEX calibration

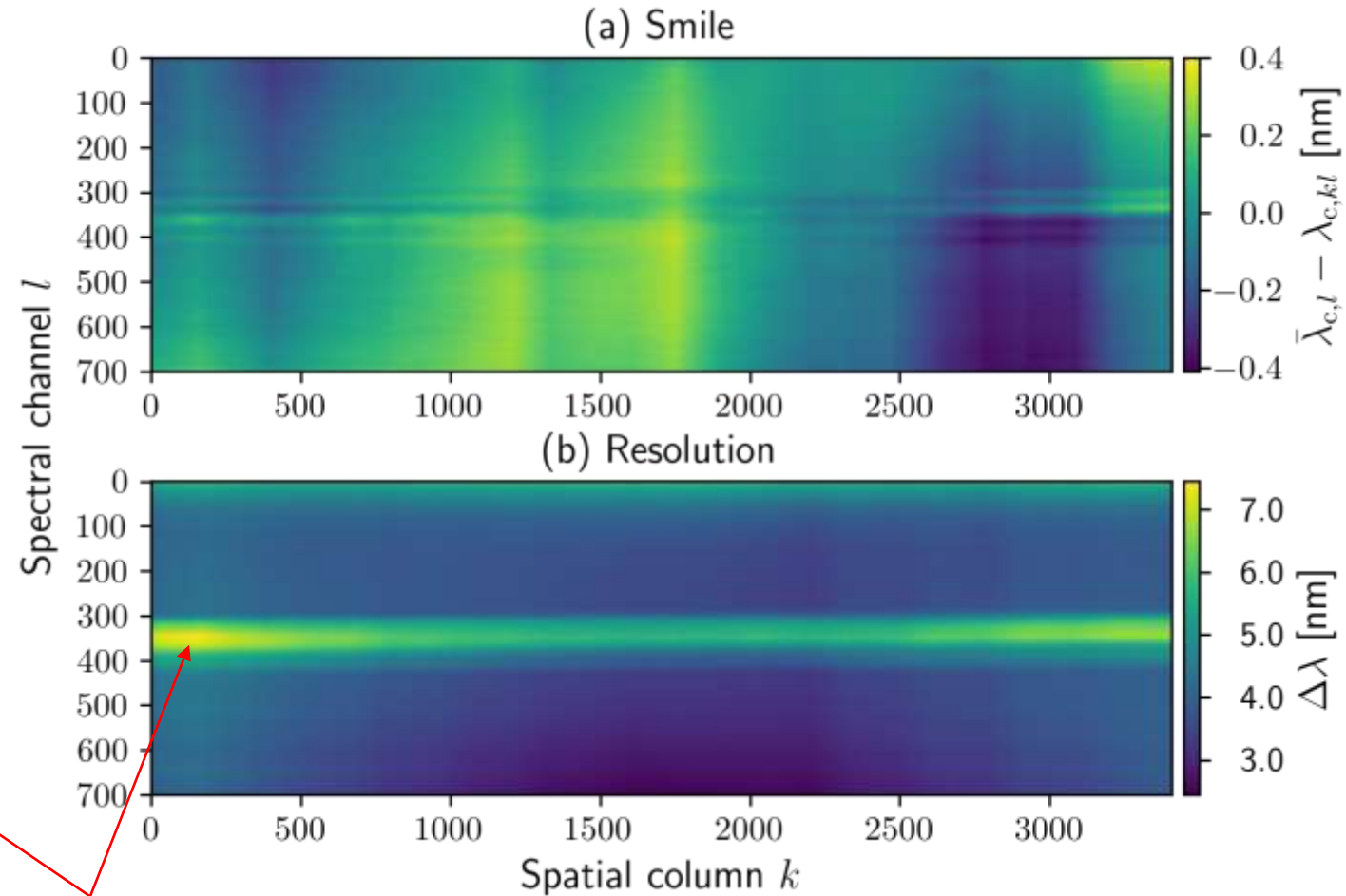
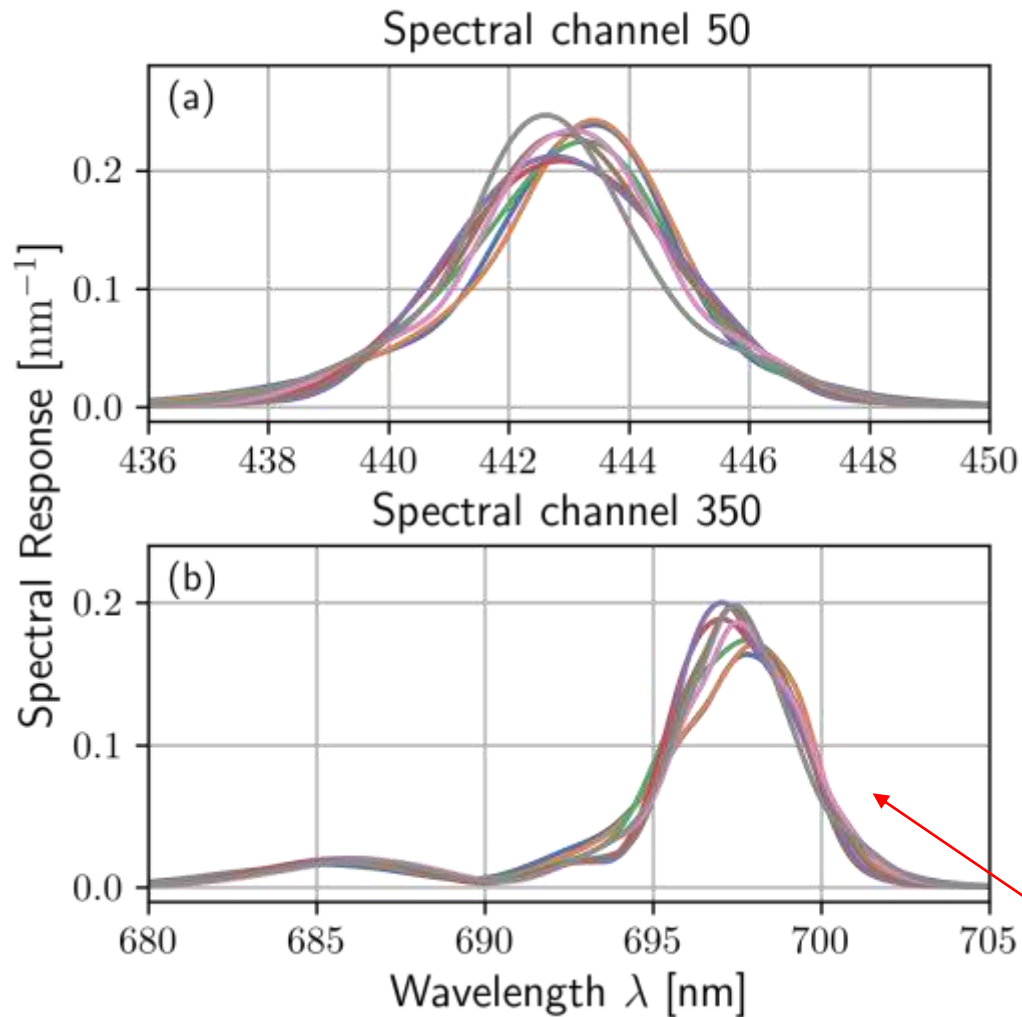
Research & Development

- Non-linearity correction
- Straylight correction
- Temperature dependence



Exemplary Results for VNIR-3000N

Similar information is available since 2013 for the original HySpex System



spectral order filter

Nice ...



State-of-the-art sensor system with extensive key calibration data

... but

A lot of these measurements may never find their way into derived products!

Question

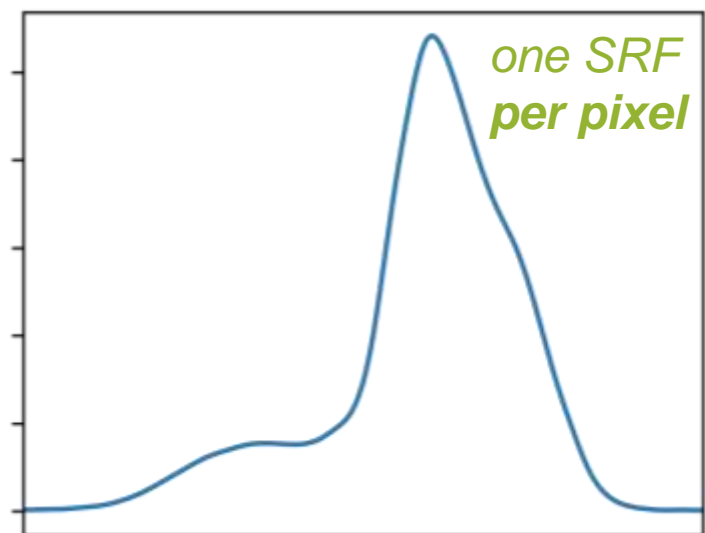
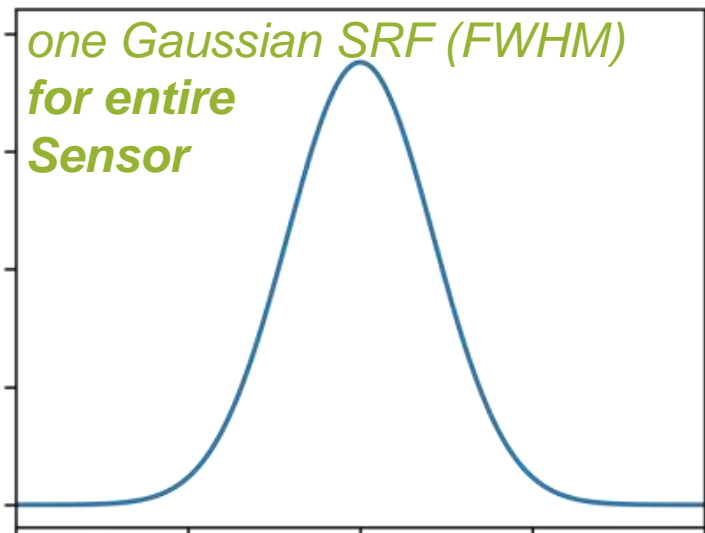
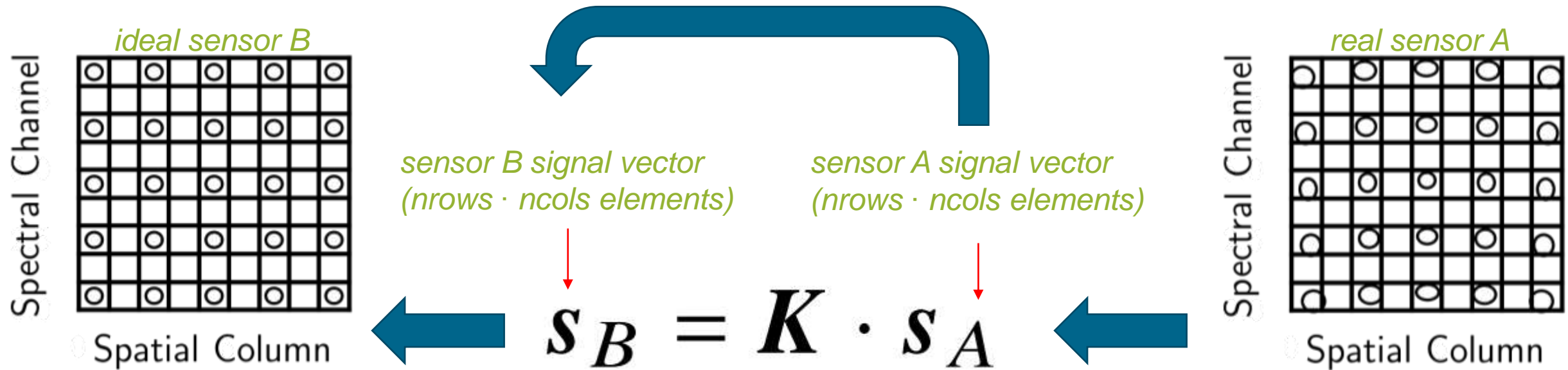


Is it possible to create a L1C product, which ...

- ... enables users to profit from **all** our calibration measurements (e.g. full SRF instead of just center wavelength & FWHM)
- ... simplifies working with L1 data for non-experts by reducing complexity

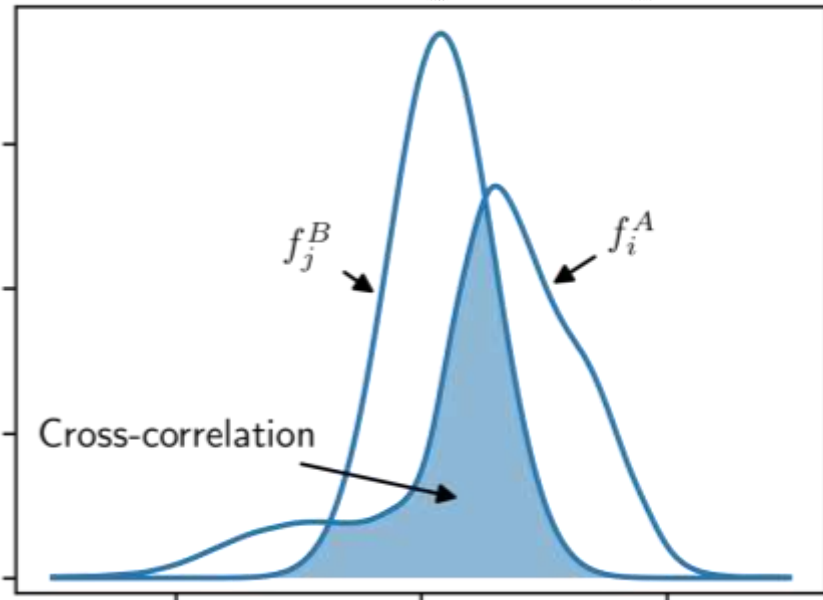


Idea

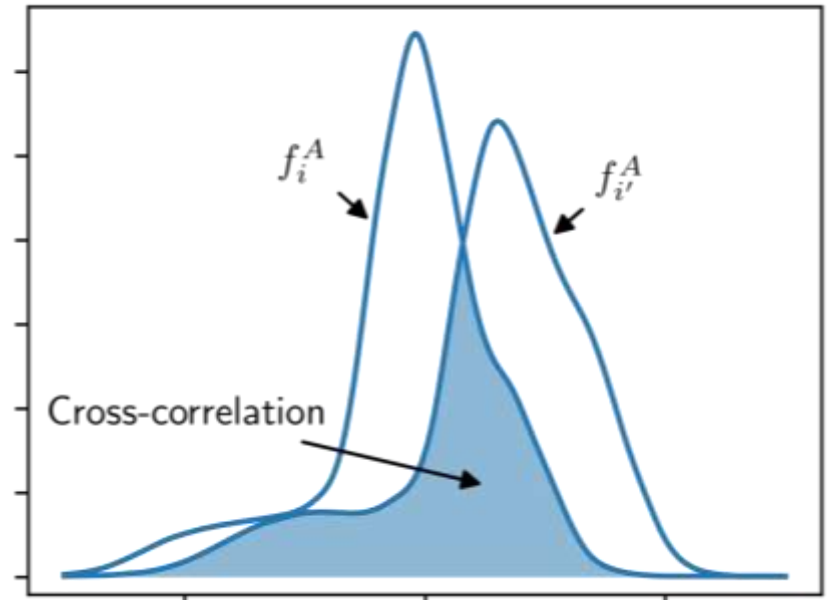


The Convolution Kernel

$$C_{ij}^{BA} := \langle f_i^B, f_j^A \rangle$$



$$C_{ij}^{AA} := \langle f_i^A, f_j^A \rangle$$



$$C^{BA} = K C^{AA}$$

*Kernel can be computed from
Cross-correlation matrices*

$$\hat{K} = \arg \min_K \left\{ \left\| K C^{AA} - C^{BA} \right\|_2^2 + \gamma^2 \left\| K \Gamma \right\|_2^2 \right\}$$

Tikhonov solution



Applications



Sensor A	Sensor B	Application
HySpex	“Better” HySpex	<i>homogeneous L1C product smart binning stray-light correction</i>
HySpex	EnMAP	<i>EnMAP validation</i>
CHIME	Sentinel-2	<i>cross-validation</i>
CHIME	Sentinel-5p	<i>wishful thinking!</i>

HySpex VNIR-3000N Processing Settings



Parameter	Original	Transformed
Spatial columns	3408	1337 <i>smart binning</i>
Spectral channels	700	295 <i>smart binning</i>
Smile [nm]	<0.8	0
Keystone [mrad]	<0.14	0
Angular resolution [mrad]	0.09 – 0.43	0.44
Spectral resolution [nm]	2.5 – 7.5	4 <i>super resolution</i>
SRF and ARF shape	Spline model (complicated)	Gaussian

HySpex SWIR-384 Processing Settings



Parameter	Original	Transformed
Spatial columns	384	384
Spectral channels	288	288
Smile [nm]	<0.7	0
Keystone [mrad]	<0.12	0
Angular resolution [mrad]	0.5 – 1.1	0.8 <i>super resolution</i>
Spectral resolution [nm]	5.5 – 8.5	6.5 <i>super resolution</i>
SRF and ARF shape	Spline model (complicated)	Gaussian

In-Flight Validation of Transformed L1 Product



- Measurements with Hg pencil lamp and reflectance panel
- Altitude: 3300 m (FL100)
- Expect scalar product of Hg emission peak (δ) and Gaussian SRF
- Fit Intensity + linear baseline to compensate for ambient light
- Isolated narrow lines \rightarrow high frequency stress test

See also Poster by Gabriel Scheib

In-Flight Validation VNIR-3000N

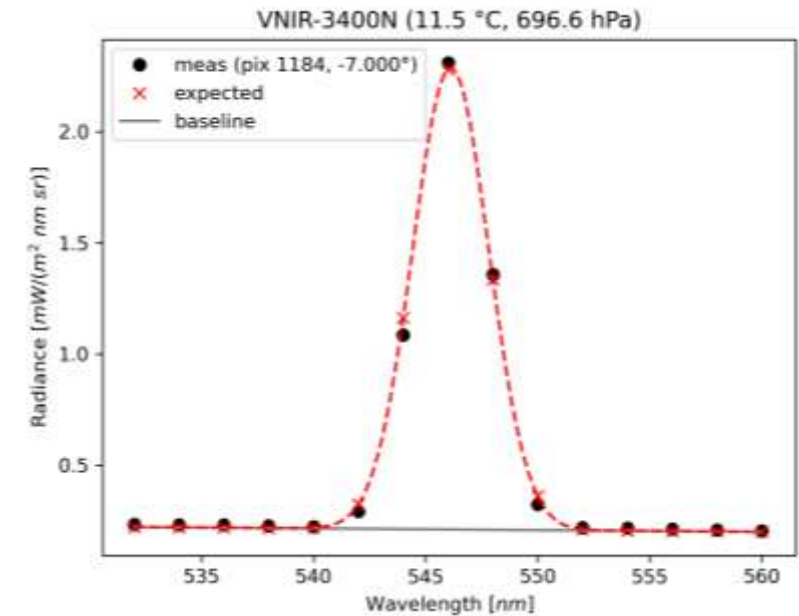
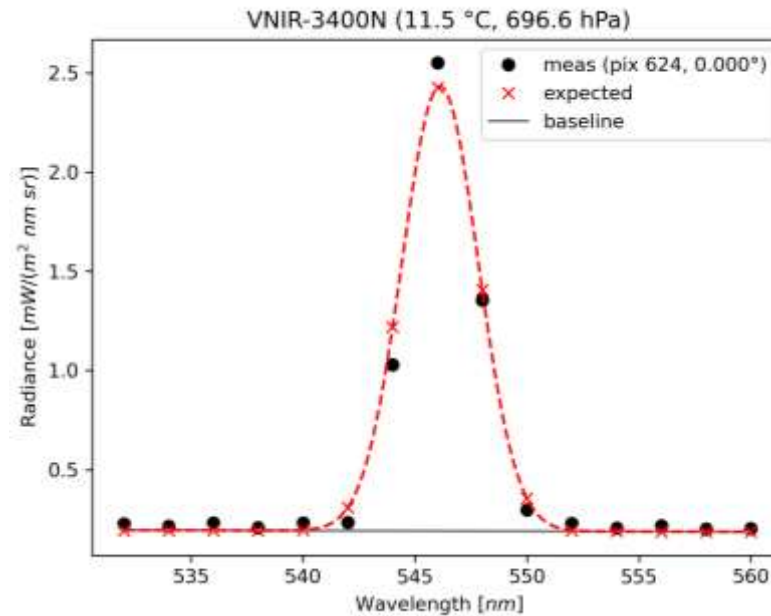
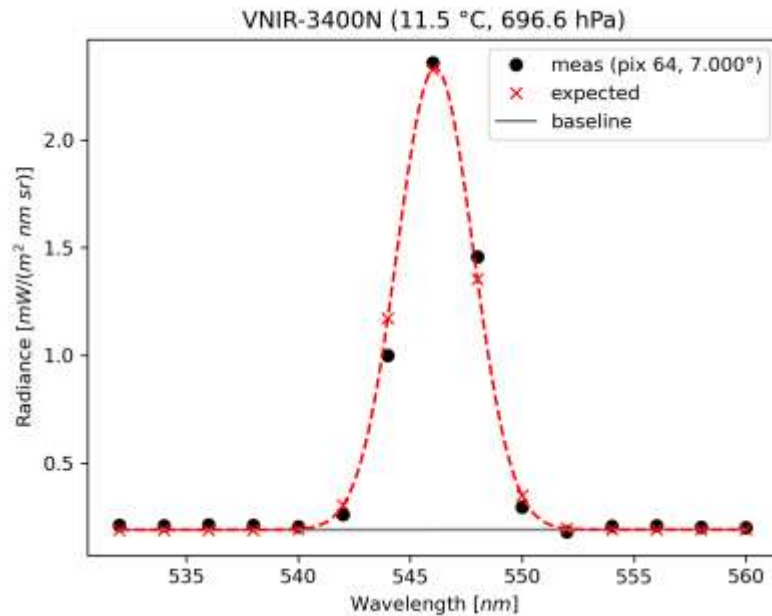


left edge

center

right edge

... of field of view



*Overall good agreement between transformed data and simulation
Potentially minor shift and SRF narrowing compared to lab*

In-Flight Validation SWIR-384

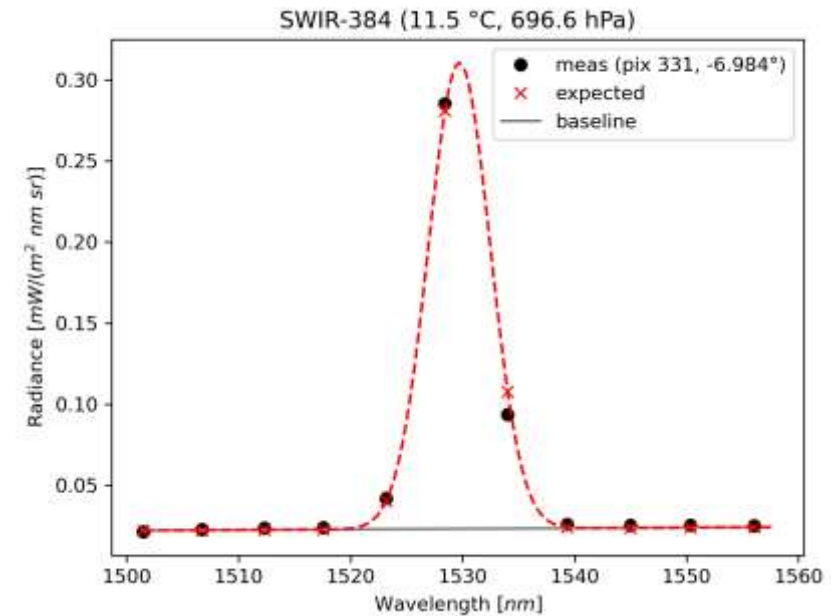
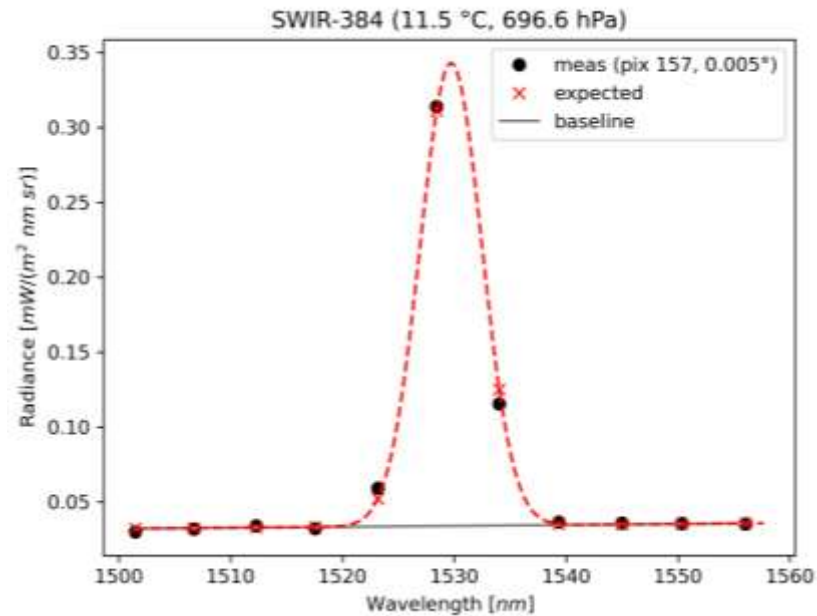
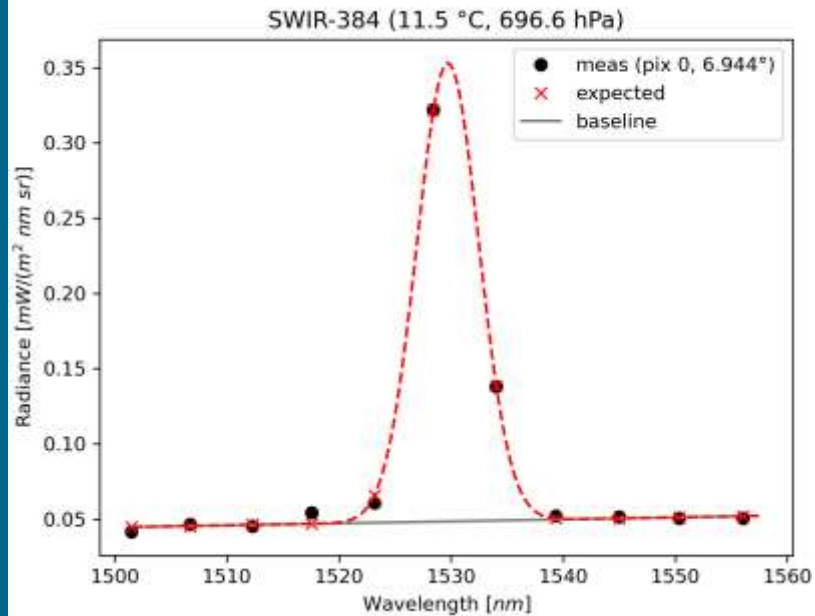


left edge

center

right edge

... of field of view



Very good agreement between transformed data and simulation

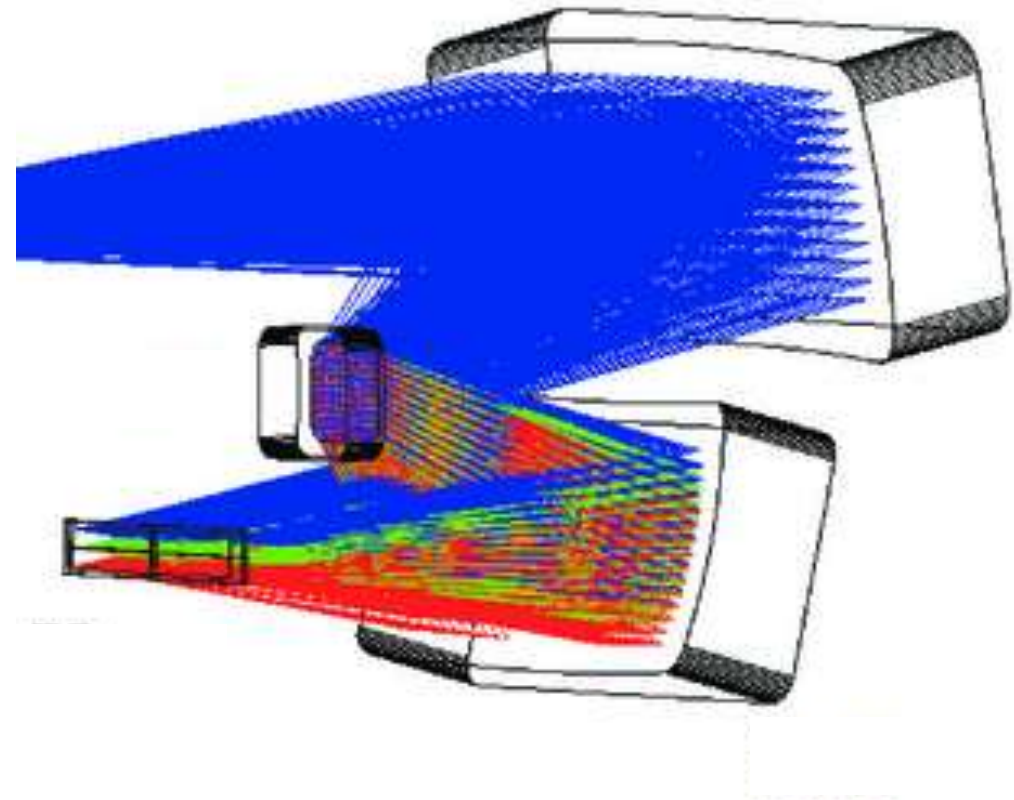
Implications on L1 Processing

- Convolution Kernel is *big data*
 - ~ 941 billion elements for HySpex VNIR
 - 7.5 TB at full resolution
 - 5.5 GB in practice (sparse matrix)
- Processing Time
 - Moderately optimized python code
 - VM with 16 CPUs @ 2.5 GHz, 128 GB RAM
 - 3-4 frames per second HySpex VNIR
 - ~50 frames per second HySpex SWIR
- Reduces complexity of all following processing steps
- Only 16 % of HySpex VNIR data points remain after *smart binning*



Implications on Future Sensor Design

- L1 post-processing can compensate for optical design imperfections
- Effectively reduces smile / keystone
- Denser sampling (Nyquist) required compared to typical hyperspectral sensors
- Response Functions have to be available (and stable)



Conclusions



- Kernel-based response function homogenization
 - corrects several optical artifacts
 - reduces data size by smart binning
 - reduces complexity of down-stream processing
 - delivers L1C products which are easier to adopt by the community
- Our findings have implications regarding the design of future sensors
 - Oversampling (spectral & across-track) improves data quality
 - Smile and keystone should not drive the optical design
 - Knowledge of the response functions (including in-orbit changes) is of paramount importance
- We should also take a closer look at along-track oversampling (scene inhomogeneity)

Title: **Response Function Based Calibration of Imaging Spectrometers**

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2 – 4 June 2026, Helsinki, Finland