

End-user driven optimization of the SWIR spectral sampling for a future hyperspectral sensor using end-to-end simulations

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Context

Project needs:

- Use a single SWIR detector for cost savings and HSI design complexity reduction
 - How to reduce the number of SWIR bands compared to the initial 10-nm sampling strategy?

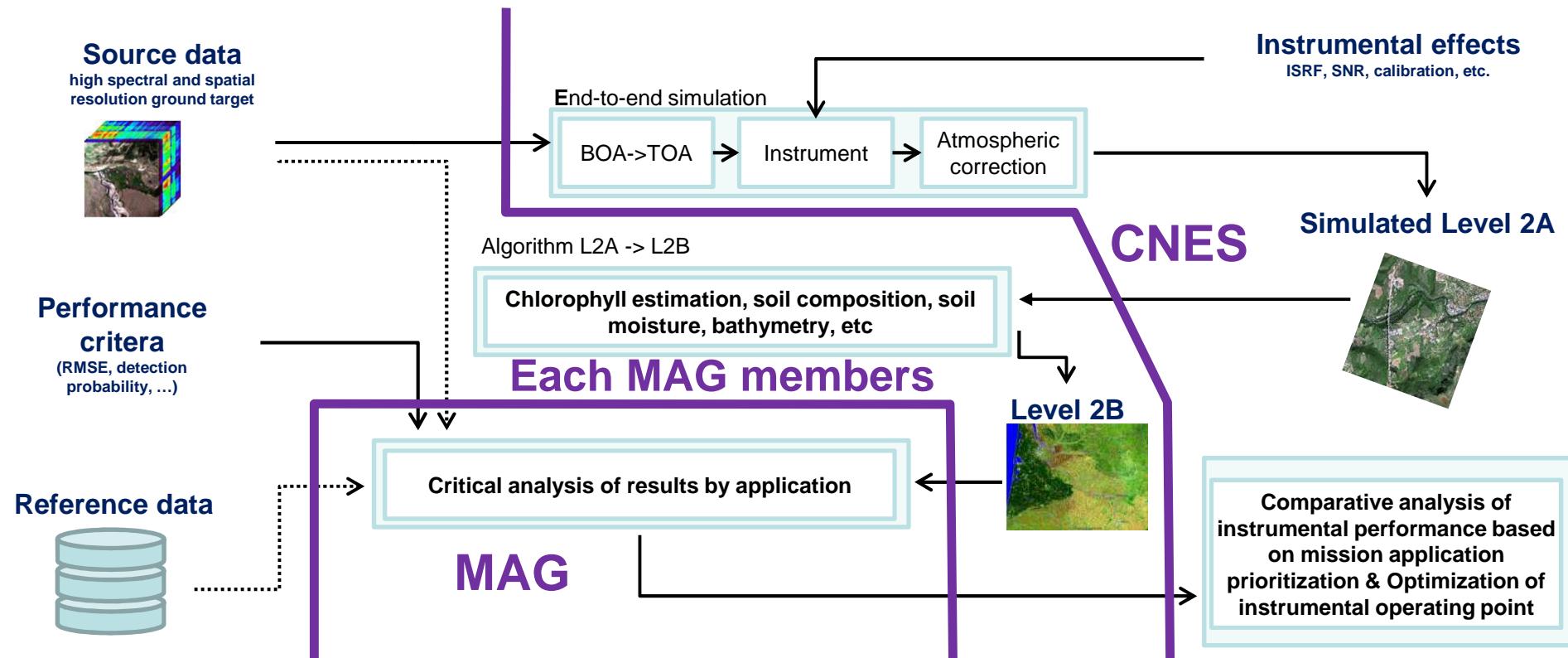
Several options to reduce the number of spectral bands in the SWIR:

- Remove spectral range intervals, increase spectral sampling, increase spectral resolution

MAG study goal was to recommend a spectral sampling strategy in the SWIR taking into account SNR and image quality suitable for different applications:

- Geosciences: mineralogy, soil moisture content
- Natural Vegetation: tree species classification, mediterranean forest EBV (Chl_{ab} , Car, LMA, EWT) at tree level
- Coastal and continental waters: bathymetry, bottom classification of shallow water, classification of coastal habitats
- Urban area: urban land cover
- Industrial site: aerosols (PM1), methane, carbon dioxide
- Cryosphere: specific surface area, black carbon rate
- Atmosphere: water vapor, carbon dioxide and aerosol

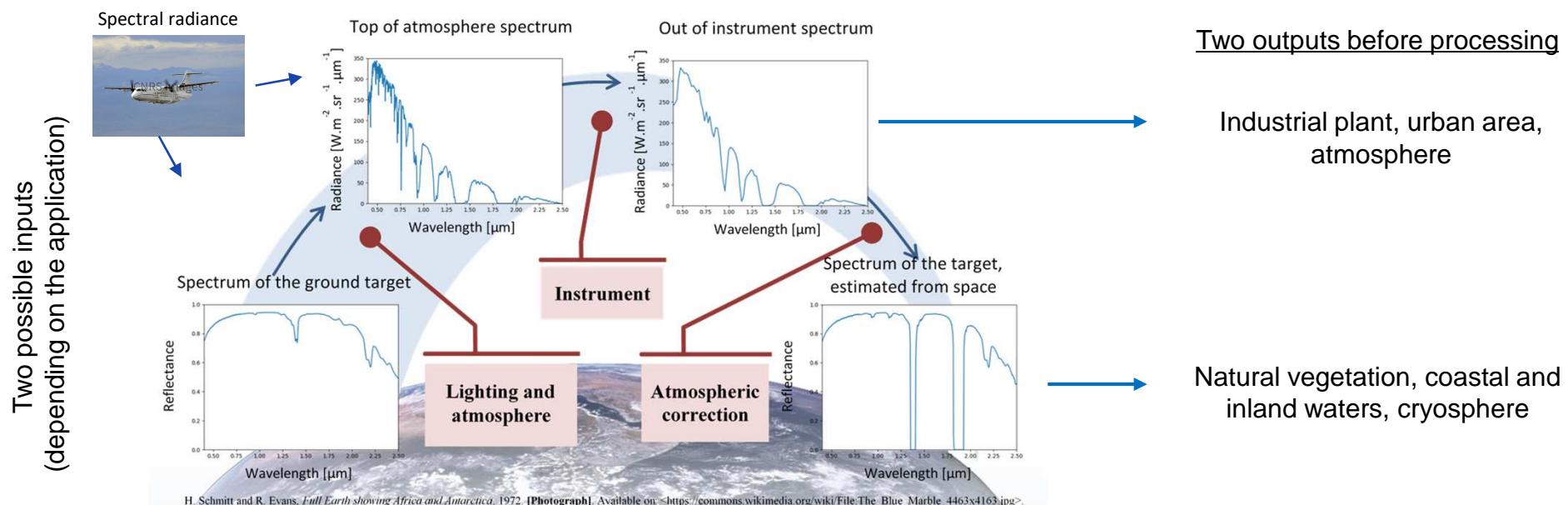
Mission Advisory Group: Work breakdown - 2022 activity



Method: IHS images simulation

Inputs: 22 images + 3 spectral libraries

All the applications were processed with the same EtoE chain



Method: instrument

IHS Instrument operating points:

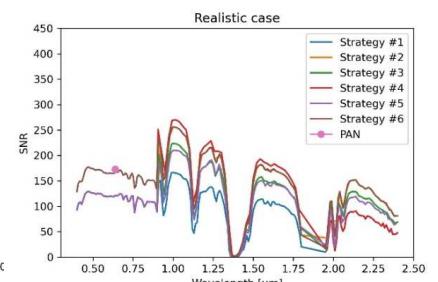
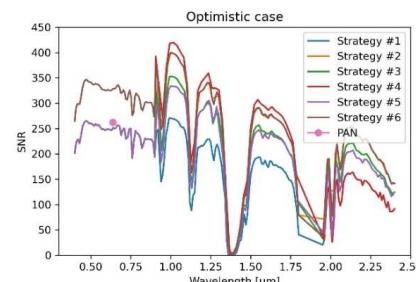
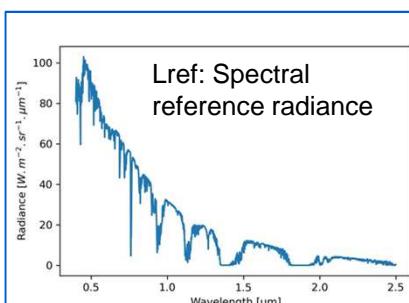
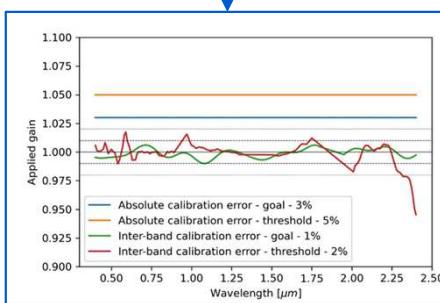
- 6 spectral sampling strategies:

24 simulations

- 2 SNR:
Realistic: [50, 250] @ Lref
Optimistic: [100, 400] @ Lref

- 2 calibration performances:
Target: 3% absolute / 1 % interband
Threshold: 5% absolute / 2 % interband

Strategy	VNIR [400 – 900]		SWIR]900 – 1 800] \cup [1 950 – 2 400]	
	Spectral step [nm]	Spectral width [nm]	Spectral step [nm]	Spectral width [nm]
#1	10	10	10	10
#2	10	10	20	20
#3	10	10	16	16
#4	10	10	22 for $\lambda \leq 1,95$ 10 for $\lambda > 2,05$	22 for $\lambda \leq 1,95$ 10 for $\lambda > 2,05$
#5	10	10	12	Linear increase: from ~14 nm to ~17 nm over [0,9 – 1,3], [1,3 – 1,8] and [1,95 – 2,4].
#6	8	16	10	20



Results: spectral sampling

Most of the applications tested are not dependent on the spectral strategy except for

Geosciences: Mineralogy

Tested minerals : kaolinite, calcite, Gypsum, Alunite, Hématite, Géolithe, Rare Earth Element (REE)

- Best strategies: #1, 4, acceptable strategies: #5, 6

Bathymetry: #6 non recommended

Industrial Plants:

Aerosols (PM1): Foucher et al., 2019, Calassou et al., 2020

- If aerosol model not known: only the realistic-Threshold, strategy #6 is acceptable

Gas (detection and quantification of methane, carbon dioxide) Nesme et al., 2021

- Realistic #5 does not able to detect methane
- (with an accuracy < 1500 ppm.m)
- and carbon dioxide (< 150 000 ppm.m)

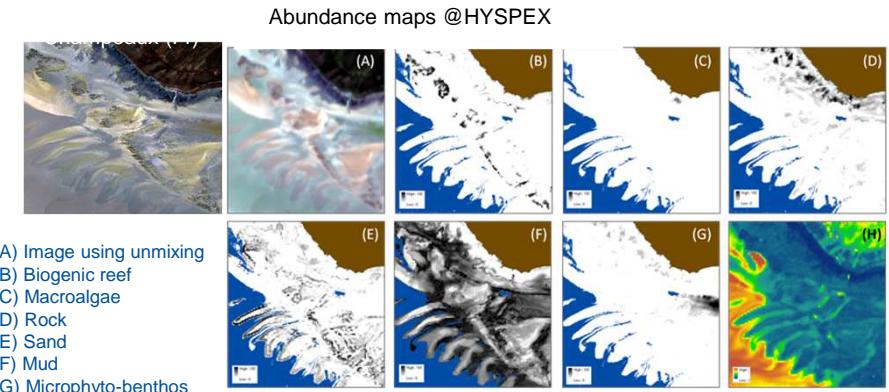
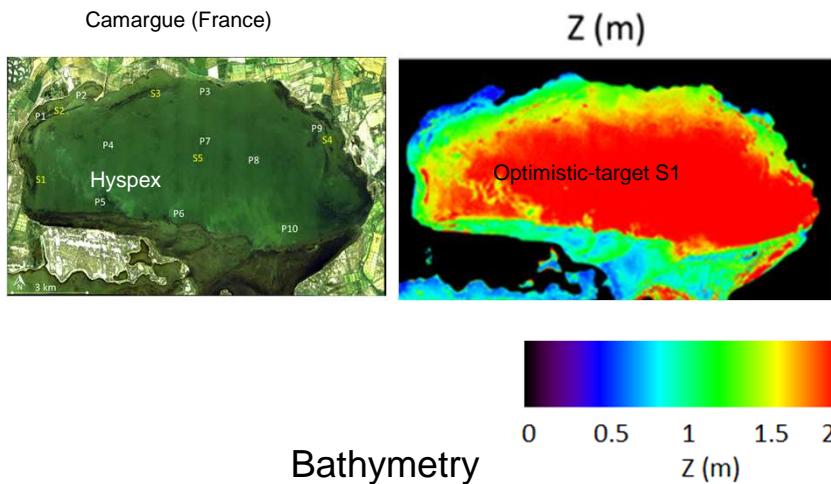
Atmosphere : Gas (H_2O , CO_2) #2 non recommended,
Aerosols: do not depend on the spectral strategy

Strategy	VNIR [400 – 900]		SWIR]900 – 1 800] U [1 950 – 2 400]	
	Spectral step [nm]	Spectral width [nm]	Spectral step [nm]	Spectral width [nm]
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#6	8	16	10	

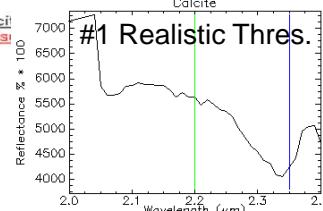
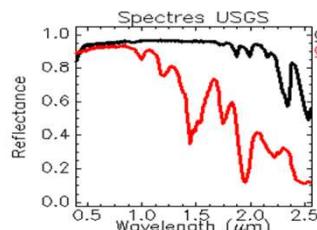
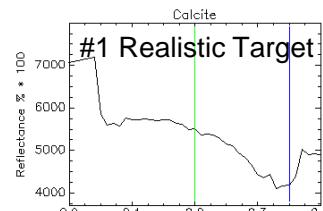
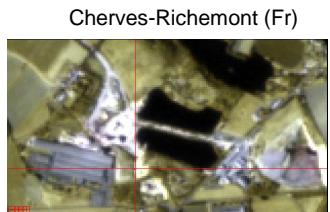


Results: SNR (realistic (SNR@Lref between 50 and 250) and optimistic (SNR@Lref between 100 and 400))

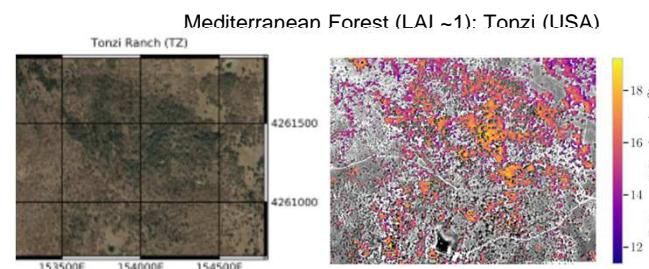
A weak degradation of the estimation is observed from optimistic down to realistic, but most of the applications tested have little or no dependence on SNR except for:



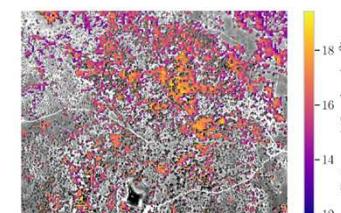
Results: Calibration (Target (3% absolute / 1 % interband) and threshold (5 % absolute / 2 % interband))



Calcite detection



#3 Optimistic Target



#3 Optimistic Thres.

Car quantification of *Quercus Douglasii*

A small but non-significant performances loss is observed from target calibration (3 % absolute / 1 % interband) to threshold target (5% / 2%) except for:

Bathymetry: Not fulfilled for Threshold

Cryosphere: Black Carbon rate not estimated with Threshold

Conclusion (1/1)

- All the products simulated in this exercise use the same end-to-end chain, with the same instrumentation parameters, making it easy to compare different applications
- Applications whose methods are based on the **use of the global signature** have no sampling dependency -> **16/20 nm spectral sampling is sufficient:** Soil Moisture Content, Tree species classification, Forest EBV, Bathymetry, Bottom Classification of shallow water, classification of coastal habitats, urban land cover, snow/ice characterization, spectral aerosol optical thickness, except for Aerosol of industrial Plant.
- Applications whose methods are based on the use of **specific absorption bands** depend on the spectral resolution -> **10 nm spectral sampling is recommended:** Mineralogy, Gas of industrial Plant or of atmosphere
- Generally, a performance loss is observed from target calibration (3 % absolute / 1 % interband) to threshold target (5% / 2%)
- Some applications depend on SNR: bathymetry, bottom classification of shallow water, classification of coastal habitats

Conclusion (1/2)

- A large French scientific community is involved to optimize the design of the instrument
- The MAG members made the recommendation on the compatibility of each proposed instrumental configuration for each application needs.
- From these results, CNES is studying the best trade-off to design the future HSI sensor to fulfill the mission objectives.
- Article is under review

From HSI MAG report – HPS-NT-SY-014-CNES – 28/06/2022

Table 48: Synthesis (where k.a. means if the estimation is done with a prior knowing the type of aerosol and not k.a. if it is not known). The color code is as follow: █ indicates the best performances close to the objective, █ indicates a small degradation of the performances but remain acceptable performances close to the objective, █ indicates the performances remain acceptable, █ indicates the performances are of the same order of the acceptable uncertainty, █ indicates the performances cannot be obtained.

Thematic	Optimistic		Realistic	
	Target	Threshold	Target	Threshold
Spectral feature				
Mineralogy	█ 1, █ 2, █ 3, █ 4, █ 5, █ 6	█ 1, █ 2, █ 3, █ 4, █ 5, █ 6	█ 1, █ 2, █ 3, █ 4, █ 5, █ 6	█ 1, █ 2, █ 3, █ 4, █ 5, █ 6
Gas of Industrial Plant: ΔCH_4 (ppm.m)	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	█ 1, 2, 3, 4, 5, 6	1, █ 2, 3, 4, 5, 6
Gas of Industrial Plant: ΔCO_2 (150000 ppm.m)	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	█ 1, 2, 3, 4, 5, 6	1, █ 2, 3, 4, 5, 6
Atmosphere Gas H_2O	█ 1, █ 2, 3, 4, 5, 6	Not tested	Not tested	1, 2, 3, 4, 5, 6
Atmosphere Gas CO_2	█ 1, █ 2, 3, 4, 5, 6	Not tested	Not tested	1, 2, 3, 4, 5, 6
Spectral continuum				
Soil Moisture Content	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6
Tree Species Classification	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6
Forest EBV	1, 2, 3, 4, 5, 6	█ 1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6
Bathymetry	1, 2, 3, 4, 5, 6	█ 1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	█ 1, 2, 3, 4, 5, 6
Bottom Classification of Shallow Water	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	█ 1, 2, 3, 4, 5, 6
Classification of Coastal habitats (without Fusion)	1, 2, 3, 4, 5, 6	█ 1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6
Urban Land Cover	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	█ 1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6
Aerosols of Industrial Plant: aerosol model known PM1	80 $\mu\text{g}/\text{cm}^2$: 1, 2, 3, 4, 5, 6	80 $\mu\text{g}/\text{cm}^2$: 1, 2, 3, 4, 5, 6	150 $\mu\text{g}/\text{cm}^2$: 1, 2, 3, 4, 5, 6	150 $\mu\text{g}/\text{cm}^2$: 1, 2, 3, 4, 5, 6
Aerosols of Industrial Plant: aerosol model not known PM1	150 $\mu\text{g}/\text{cm}^2$: 1, 2, 3, 4, 5, 6	150 $\mu\text{g}/\text{cm}^2$: 1, 2, 3, 4, 5, 6		
Cryosphere Spec. Surf. area	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6
Equivalent Black Carbon content	█ 1, 2, 3, 4, 5, 6	█ 1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	█ 1, 2, 3, 4, 5, 6
Atmosphere Aerosol with revisit or auxiliary	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6	1, 2, 3, 4, 5, 6

