

# Copernicus FICE 2025

Training on  
In situ Ocean Colour Above-Water Radiometry towards Satellite Validation

## Uncertainty analysis in HyperCP

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fiducial reference  
measurements for  
satellite ocean colour



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6-20 July 2025  
Venice, Italy



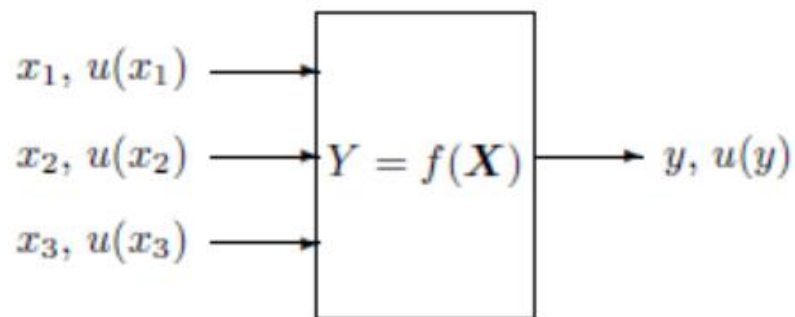


## UNCERTAINTY EVALUATION

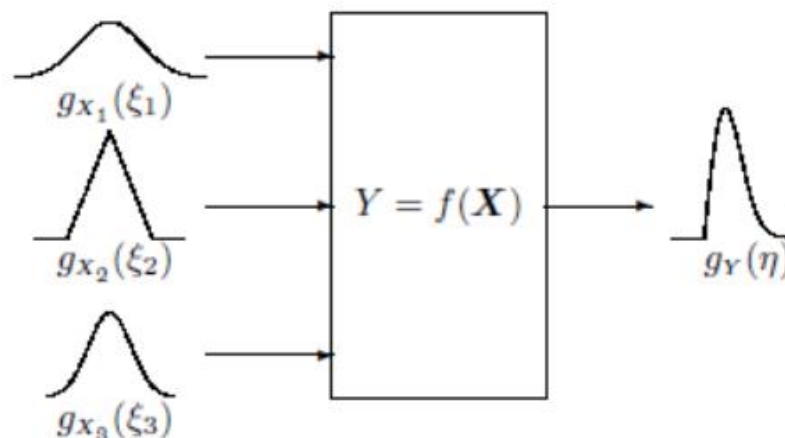
- > GUM – Law of propagation of uncertainty
- > GUM supplement 1 – Monte Carlo Methods



### GUM



### MCM



### GUM ASSUMPTIONS AND RESTRICTIONS

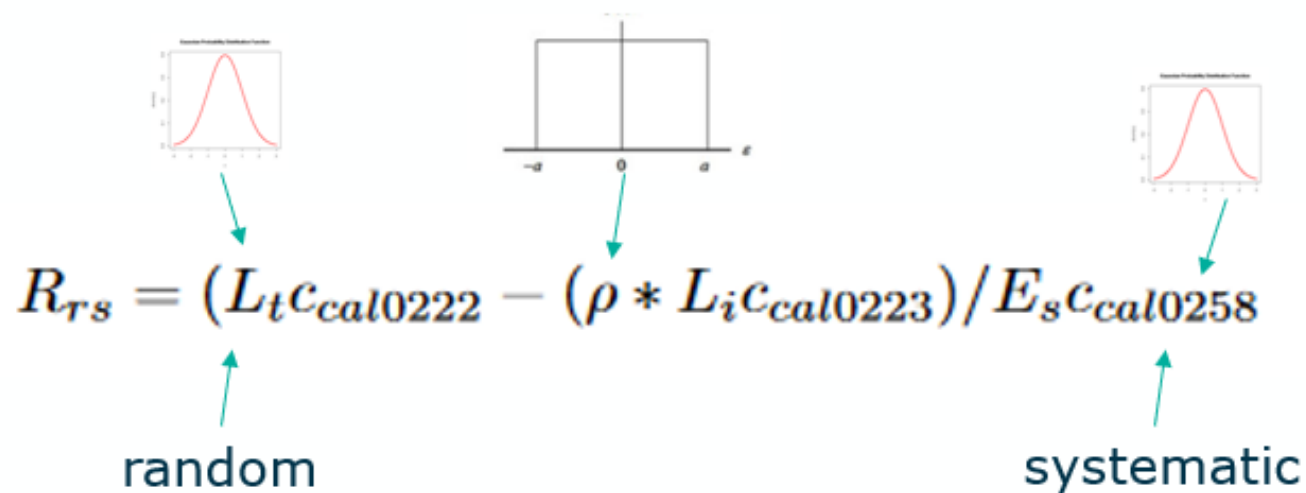
- > Output value has Normal distribution
- > First order approximation applies to linear models
- > Symmetric distribution of inputs

<http://www.bipm.org/en/publications/guides/>

**Bureau**  
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– the intergovernmental organization through which Member States act together on matters related to measurement science and measurement standards.

# GUM Methodology applied in CoMET tool

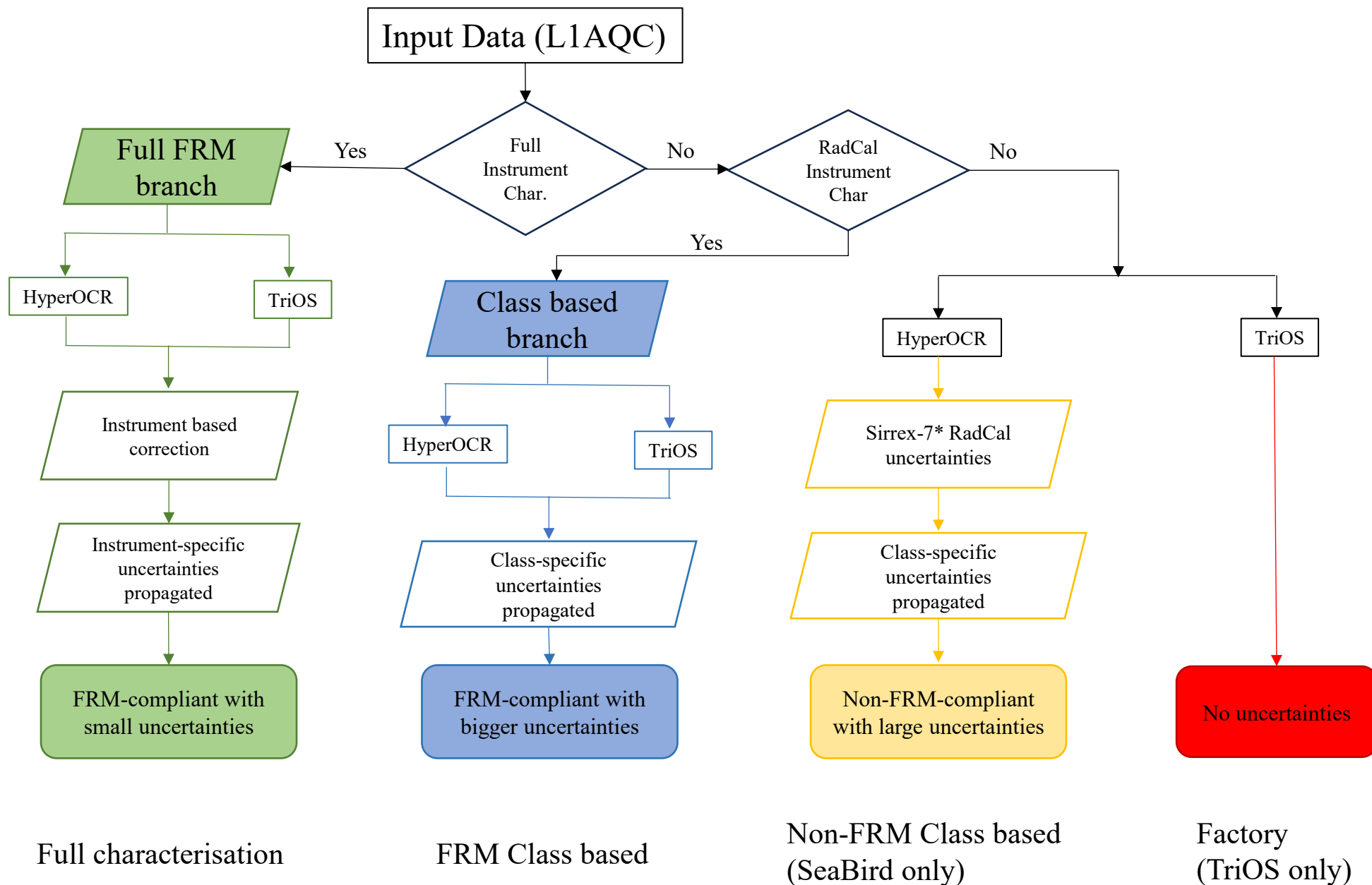


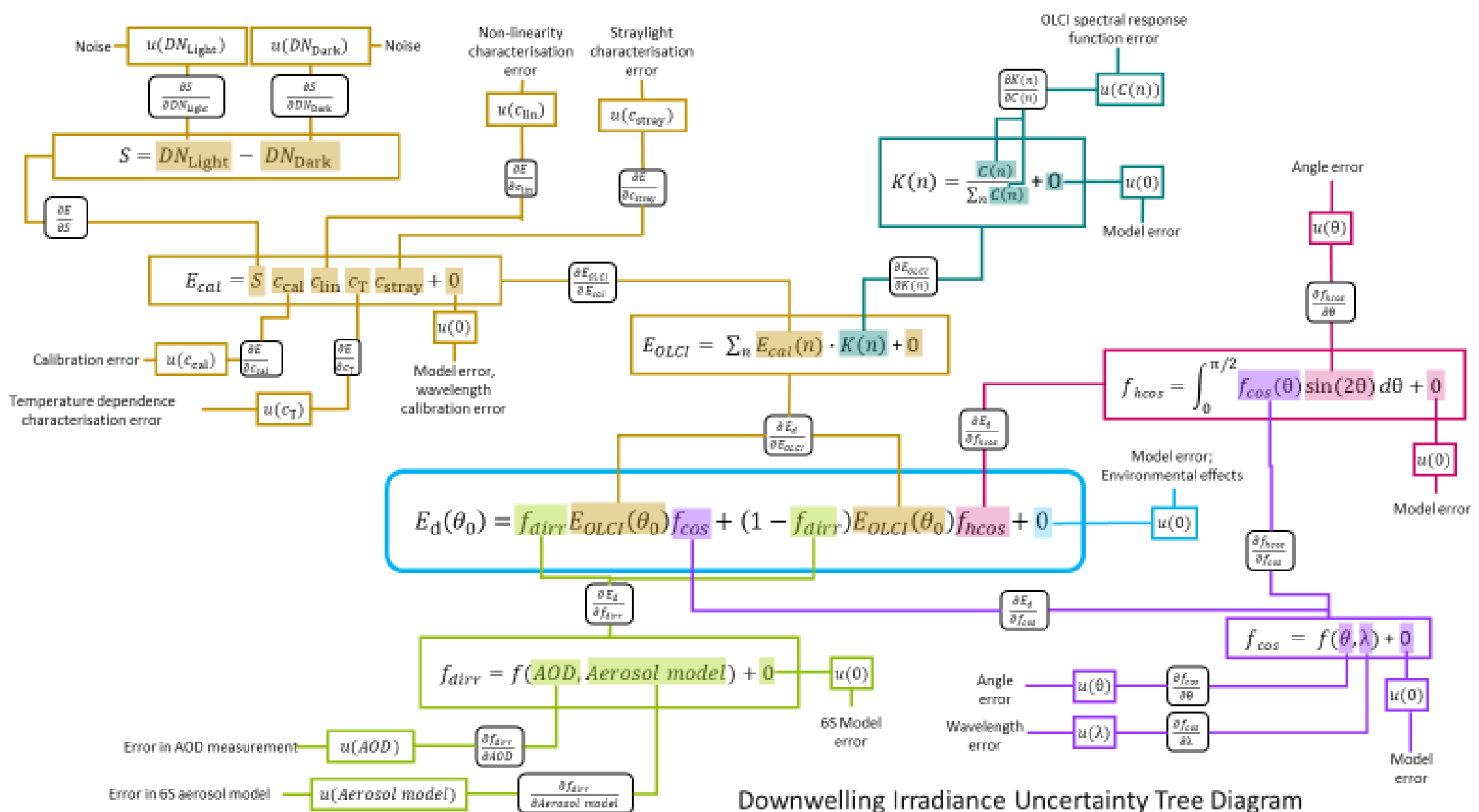
| $L_t$ | $c_{c2}$ | $\rho$ | $L_i$ | $c_{c3}$ | $E_s$ | $c_{c8}$ |
|-------|----------|--------|-------|----------|-------|----------|
| 1     | 0        | 0      | 0     | 0        | 0     | 0        |
| 0     | 1        | 0      | 0     | 1        | 0     | 1        |
| 0     | 0        | 1      | 0     | 0        | 0     | 0        |
| 0     | 0        | 0      | 1     | 0        | 0     | 0        |
| 0     | 1        | 0      | 0     | 1        | 0     | 1        |
| 0     | 0        | 0      | 0     | 0        | 1     | 0        |
| 0     | 1        | 0      | 0     | 1        | 0     | 1        |

$$u_c^2(y) = \sum_{i=1}^N c_i^2 u^2(x_i) + 2 \sum_{i=1}^{N-1} \sum_{j=1+1}^N c_i c_j u(x_i) u(x_j) r(x_i, x_j),$$

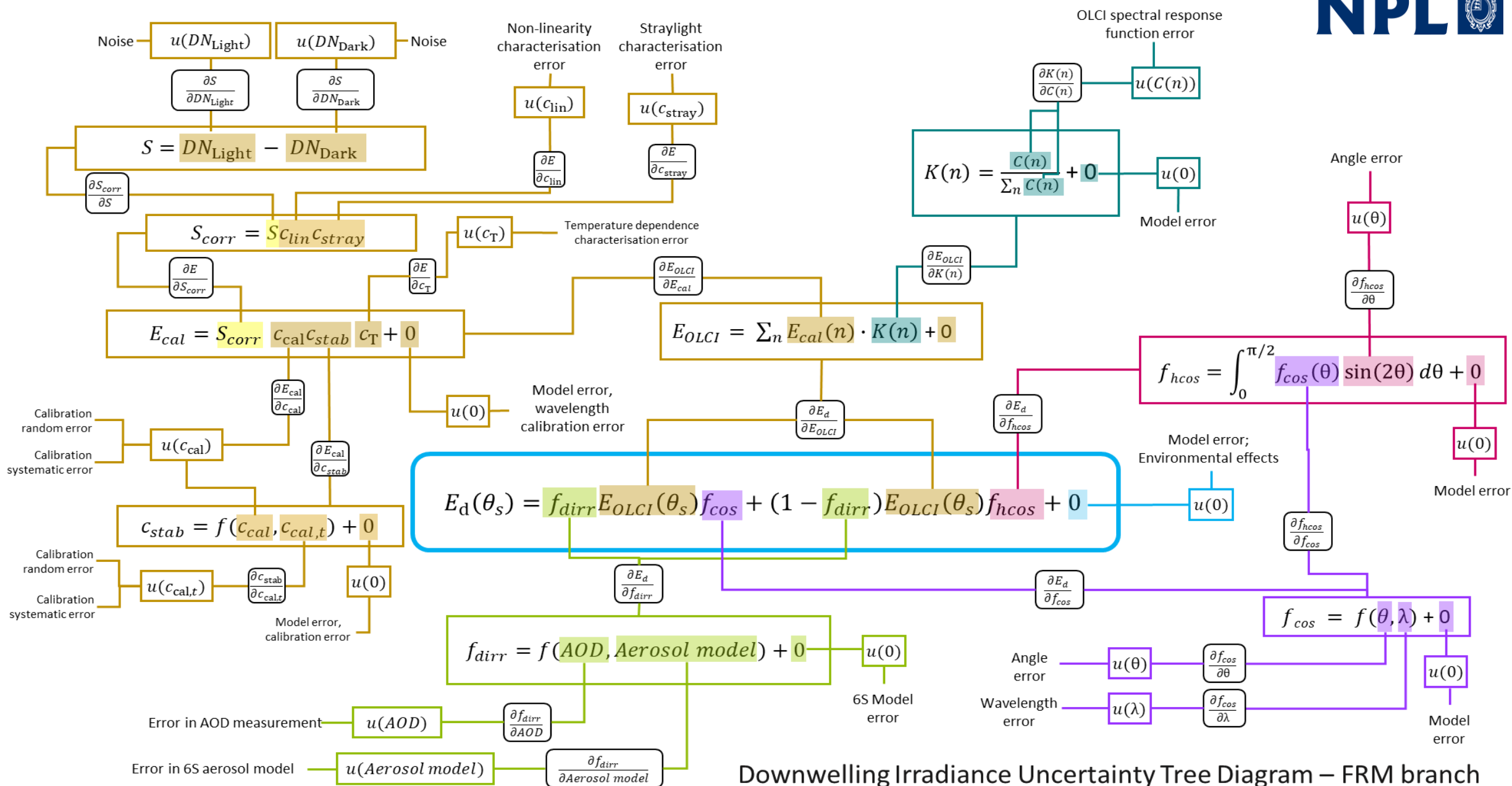
### Instrument Characterization:

- Dark current noise
- Linearity of response
- Calibration/stability
- Straylight response
- Angularity of response
- Thermal response
- Polarization response





### Downwelling Irradiance Uncertainty Tree Diagram



Downwelling Irradiance Uncertainty Tree Diagram – FRM branch

Radiance,  $L_i, L_t$

$$S_{rad} = DN_{Light} - DN_{Dark}$$

$$c_{stab} = f(c_{cal}, c_{cal,t}) + 0$$

$$S_{corr} = S_{rad} c_{lin} c_{stray}$$

$$L_{cal} = S_{corr} c_{cal} c_{stab} c_T c_{pol} + 0$$

**Instrument characterization** - the uncertainty contribution of non-linearity, temperature, polarisation, stray-light, & cosine response

Irradiance,  
 $E_d$

$$c_{stab} = f(c_{cal}, c_{cal,t}) + 0$$

$$S = DN_{Light} - DN_{Dark}$$

$$S_{corr} = S c_{lin} c_{stray}$$

$$E_{cal} = S_{corr} c_{cal} c_{stab} c_T + 0$$

Sea-surface reflectivity constant,  $\rho$

$$\rho_{sun} = f(\lambda, \theta, w_s, AOD, L_{sun}, L_{sky}) + 0$$

$$\rho_{sky} = f(\lambda, \theta, w_s, AOD, L_{sky}) + 0$$

$$\rho(\lambda) = \rho_{sky}(\lambda) + \rho_{sun}(\lambda) + 0$$



CoMet is a python toolkit, developed at NPL, which is used to propagate uncertainty using Monte Carlo Propagation.

**Sea surface reflectance ( $\rho$ )** – the uncertainty of  $\rho$  comes from model error and uncertainty from the ancillary data it relies upon.

$$L_w(\theta, \Delta\phi, \theta_s) = L_t(\theta, \Delta\phi, \theta_s) - \rho(\theta, \Delta\phi, \theta_0, w_s) L_i(\theta', \Delta\phi, \theta_s)$$

$$R_{rs}(\theta, \Delta\phi, \theta_s) = \frac{L_w(\theta, \Delta\phi, \theta_s, \theta_0, W)}{E_d(\theta_s)} + 0$$

**Model error** – Other sources of uncertainty can be included as a model error for the processor.

$$E_d(\theta_s) = f_{dirr} E_{cal}(\theta_s) f_{cos} + (1 - f_{dirr}) E_{cal}(\theta_s) f_{hcos} + 0$$

$$f_{cos} = f(\theta, \lambda) + 0$$

$$f_{dirr} = f(AOD, \text{Aerosol model}) + 0$$

$$f_{hcos} = \int_0^{\pi/2} f_{cos}(\theta) \sin(2\theta) d\theta + 0$$

# Default branch measurements equations **NPL**

- Irradiance

$$E_d(\lambda) = \overline{E_d(\lambda)} \cdot c_{cal}(\lambda) c_{stab}(\lambda) c_{lin}(\lambda) c_{stray}(\lambda) c_T(\lambda) f_{cos}$$

- Radiance

$$L_t(\lambda) = \overline{L_t(\lambda)} \cdot c_{cal}(\lambda) c_{stab}(\lambda) c_{lin}(\lambda) c_{stray}(\lambda) c_T(\lambda) c_{pol}(\lambda)$$



# Approach

Table 3. Summary information about each uncertainty component values for class-based approach (blue branch, Fig. 5)

| Variable symbol  | Variable name/description  | Exemplary uncertainty magnitude for class-based characterisation  |  | PDF shape   | Correlation ' <u>corr_x</u> ' | Correlation between ' <u>corr_between</u> ' |
|--|--|---|--|-------------|-------------------------------|---|
|  |  | TRIOS   | HyperOCR   |             |                               |   |
| $(DN_{light,LX} - DN_{dark,LX})$<br>$(DN_{light,ES} - DN_{dark,ES})$ | Mean value of DNs measured by a single instrument at a "station" | Standard deviation calculated per measurement from data statistics  |  | Normal      | Random                        | N/A   |
| $c_{cal}$  | Absolute radiometric calibration                                 | Uncertainty values from calibration certificate divided by 2 to convert them back into standard uncertainty, $k=1$                      |  | Normal      | Systematic                    | Between all three instruments               |
| $c_{stab}$   | Absolute calibration stability                                   | 1%  |  | Rectangular | Systematic                    | N/A   |
| $c_{lin}$  | Detector non-linearity   | 2%  |  | Normal      | Systematic                    | Between all three instruments               |
| $c_{stray}$  | Spectral stray light   | Vary spectrally and per instrument due to difference in spectral shape of the signal, should come from the class-based stray light file |  | Normal      | Systematic                    | Between all three instruments               |
| $c_T$  | Temperature sensitivity  | Vary spectrally come from the class-based temperature sensitivity file  |  | Normal      | Systematic                    | Between all three instruments               |
| $c_{pol}$  | Polarisation sensitivity (Radiance only)                         | Vary spectrally and per instrument to use published data from (Talone and Zibordi, 2016)  | Vary spectrally and per instrument triple values for TRIOS, as shown in [AD-1] | Normal      | Systematic                    | Between two radiance instruments            |
| $c_{cos}$  | Cosine response (Irradiance only)                                | Directional 3.5%  | Directional 2%   | Normal      | Systematic                    | N/A   |

# Approach

Table 3 . Summary information about each uncertainty component for sea surface reflectance factor ( $\rho$ ) estimation using Mobley method to estimate the sea-surface reflectance factor ( $\rho$ ).



| Variable symbol | Variable name/description | Exemplary uncertainty magnitude   | PDF shape   | Correlation 'corr_x' | Correlation between 'corr_between' |
|-----------------|---------------------------|---|-------------|----------------------|------------------------------------|
| $\rho$          | Sea surface reflectance   | Calculated for each cast depends on all input components, especially wind speed | Normal      | Random               | N/A                                |
| $w_s$           | Wind speed                | $2 \text{ ms}^{-1}$   | Normal      | Random               | N/A                                |
| $\Delta\phi$    | Relative azimuth          | $3^\circ$   | Normal      | Random <sup>1</sup>  | N/A                                |
| $\theta_s$      | Solar zenith angle        | $0.5^\circ$   | Normal      | Random               | N/A                                |
| +0              | Model error               | Difference between Mobley and Zhang method                                      | Rectangular | Systematic           | N/A                                |



# Default Branch CP Implementation Example

| Source of Uncertainty | Input Uncertainty |
|-----------------------|-------------------|
| $DN_{light}$          | Std ( $k=1$ )     |
| $DN_{dark}$           | Std ( $k=1$ )     |
| $c_{cal}$             | Tartu file        |
| $c_{stab}$            | 1%                |
| $c_{lin}$             | 2%                |
| $c_{stray}$           | FRM4SOC-1         |
| $c_{temp}$            | Tartu file        |
| $c_{cos}$             | 2%                |

```
179
180     @staticmethod
181     def Es(DNLight, DNDark, Ccal, Cstab, Clin, Cstray, Ct, Ccos):
182         """(DNLIGHT-DNDARK).Ccal.Cstab.Clin.Cstray.Ct.Ccos"""
183         return (DNLight - DNDark) * Ccal * Cstab * Clin * Cstray * Ct * Ccos
184
```

- Occurs at L1B during dark correction,  $c_{cal}$  is taken from Tartu file.
- Time average  $DN_{light}$  &  $DN_{dark}$ .
- Remaining coefficients are set to 1.
- Measurement function is defined in python.
- Punpy generates samples from inputs and uncertainties.
- Runs M=10000 Monte Carlo uncertainty propagation according to the GUM.
- Output is divided by signal to generate relative uncertainty.
- Saved in the uncertainty budget group.




# Using Monte Carlo

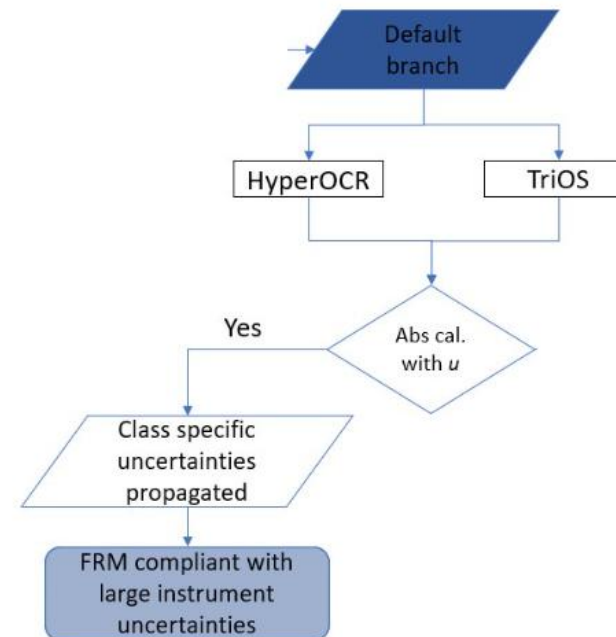
- First, we need to identify our measurement function,  $f$
- We need our inputs to the measurement function with their associated uncertainties
- Then we build samples of  $M$  draws, based on known input correlation
- We run those samples through  $f$
- $u_c(y)$  can be found from the statistics of the output

GUM Methodology applied in CoMET tool



$$R_{rs} = (L_t c_{cal0222} - (\rho * L_i c_{cal0223}) / E_s c_{cal0258}$$

 random             systematic

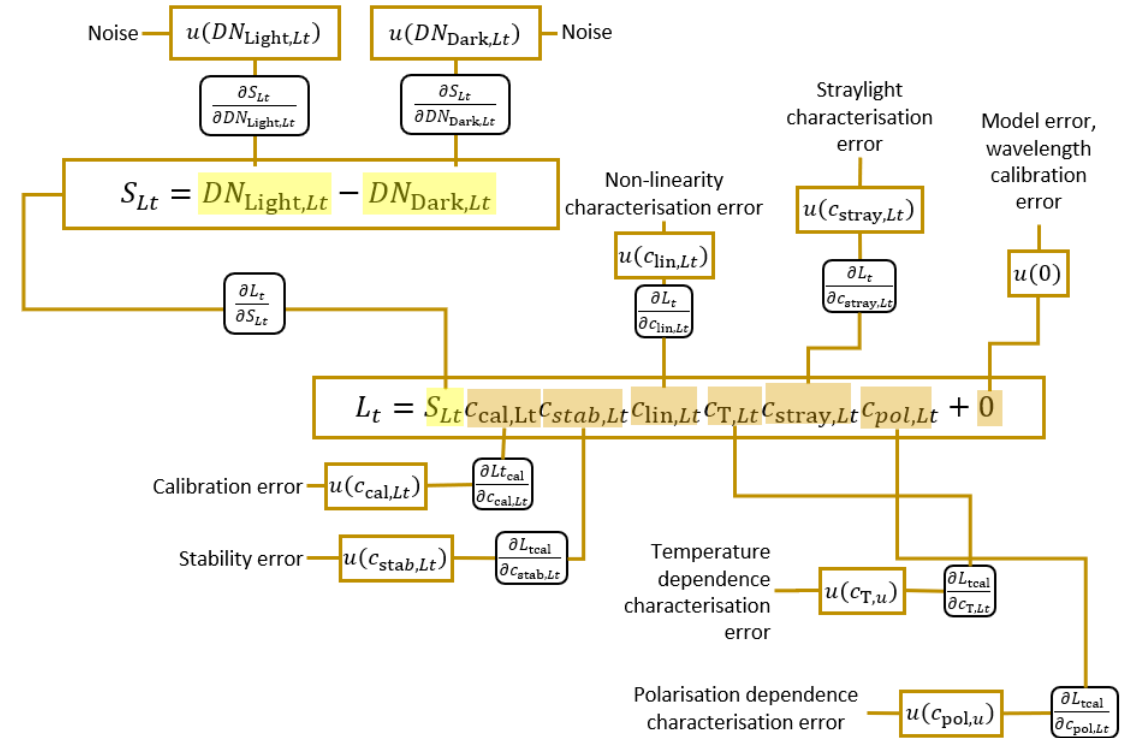


|          | $L_t$ | $c_{c2}$ | $\rho$ | $L_i$ | $c_{c3}$ | $E_s$ | $c_{c8}$ |
|----------|-------|----------|--------|-------|----------|-------|----------|
| $L_t$    | 1     | 0        | 0      | 0     | 0        | 0     | 0        |
| $c_{c2}$ | 0     | 1        | 0      | 0     | 1        | 0     | 1        |
| $\rho$   | 0     | 0        | 1      | 0     | 0        | 0     | 0        |
| $L_i$    | 0     | 0        | 0      | 1     | 0        | 0     | 0        |
| $c_{c3}$ | 0     | 1        | 0      | 0     | 1        | 0     | 1        |
| $E_s$    | 0     | 0        | 0      | 0     | 0        | 1     | 0        |
| $c_{c8}$ | 0     | 1        | 0      | 0     | 1        | 0     | 1        |



# How FRM Uncertainties are Propagated

- We combine and process samples directly
- We calculate uncertainty (in theory) alongside the processor
- Correlations are engendered within the samples
- L2 uncertainties ( $R_{rs}$ ,  $Lw$ ,  $NLw$ ), are calculated using distributions of  $E_s$ ,  $Li$ ,  $Lt$ , &  $Rho$  which contain information of uncertainty and correlation



# FRM Uncertainties – an example

$$L_w(\theta, \Delta\phi, \theta_s) = L_t(\theta, \Delta\phi, \theta_s) - \rho(\theta, \Delta\phi, \theta_0, W) L_i(\theta', \Delta\phi, \theta_s) + 0$$

$\frac{\partial L_w}{\partial \rho}$

$u(0)$  Model error, Environmental effects

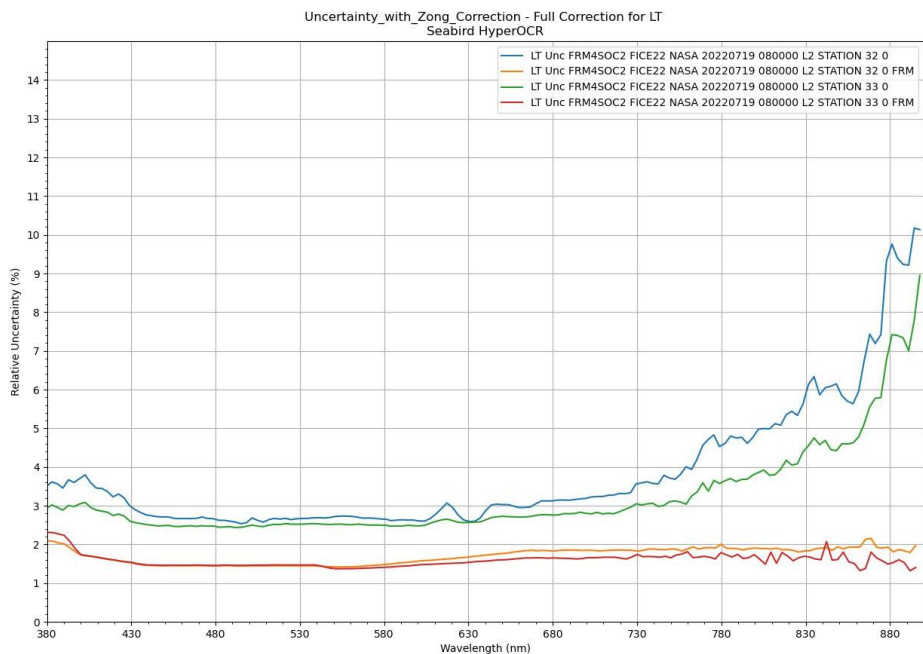
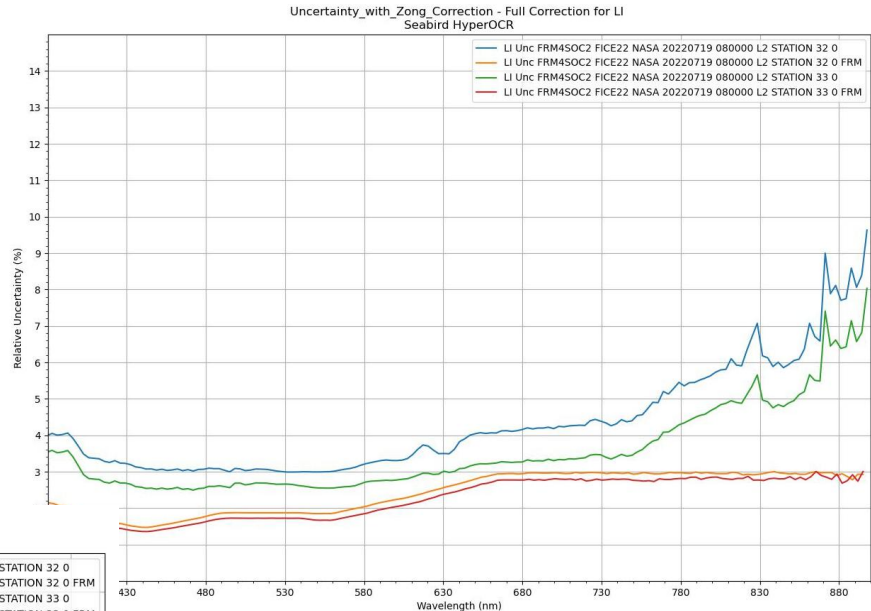
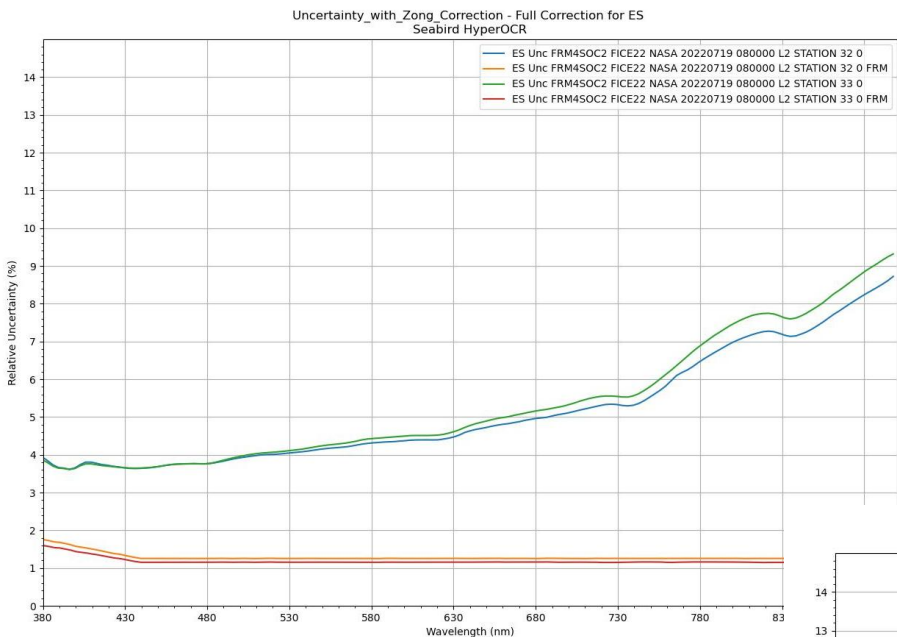
```
404 sample_Lw = Propagate_L2_FRM.run_samples(Propagate.Lw_FRM, MC_x: [ltSample, rhoSample, liSample])
405 sample_Rrs = Propagate_L2_FRM.run_samples(Propagate.Rrs_FRM, MC_x: [ltSample, rhoSample, liSample, esSample])
```

```
344 @staticmethod
345 def Lw_FRM(lt, rho, li):
346     """ Lw FRM branch measurment function """
347     return lt - (rho * li)
348
```

```
349 @staticmethod
350 def Rrs_FRM(lt, rho, li, es):
351     """ Rrs FRM branch measurment function """
352     return (lt - (rho * li)) / es
353
```

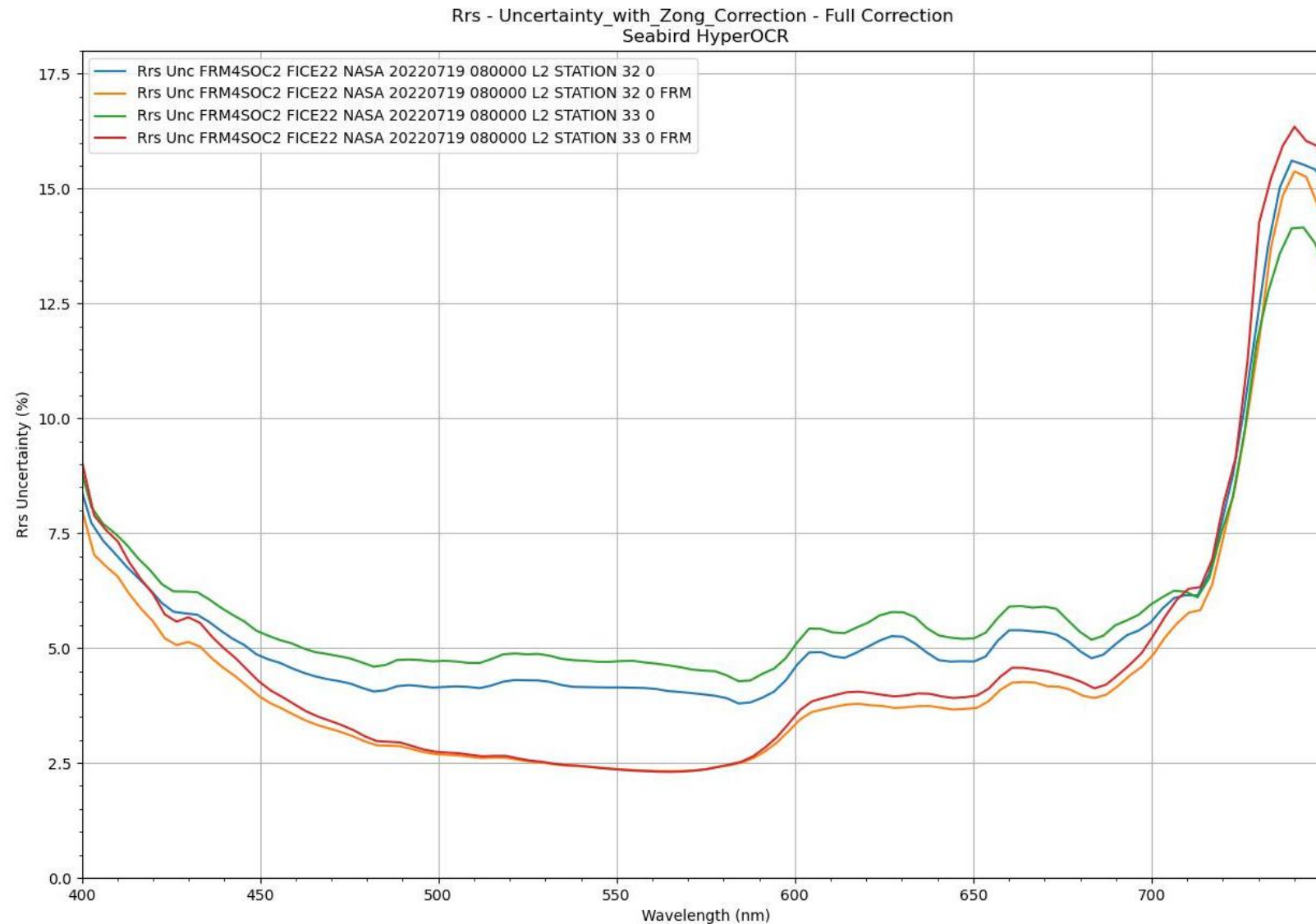
| Variable Symbol  | Variable Name  | Uncertainty Source   | Class Based        | FRM |
|--|--|--|--------------------|-----|
|  |  |  | Correction Applied |     |
| $\left( DN_{\text{light},L_X} - DN_{\text{dark},L_X} \right)$<br>$\left( DN_{\text{light},E_S} - DN_{\text{dark},E_S} \right)$ | Mean value of DNs measured by a single instrument at a “station” | Standard deviation calculated from statistics of filtered measurements | NA                 | NA  |
| $c_{\text{cal}}$   | Absolute Radiometric Calibration                                 | Instrument specific characterisation                                   | NA                 | NA  |
| $c_{\text{stab}}$  | Absolute Calibration Stability                                   | Instrument specific characterisation                                   | NA                 | NA  |
| $c_{\text{lin}}$   | Detector Non-Linearity   | Instrument specific characterisation                                   | No                 | Yes |
| $c_{\text{stray}}$   | Spectral Stray Light   | Zong stray light correction method                                     | No                 | Yes |
| $c_T$  | Temperature Sensitivity  | Instrument specific characterisation                                   | No                 | Yes |
| $c_{\text{pol}}$   | Polarisation Sensitivity (Radiance)                              | Class specific characterisation  | No                 | No  |
| $c_{\text{cos}}$   | Cosine Response (Irradiance)                                     | Instrument specific characterisation                                   | No                 | Yes |

# Uncertainty Results – PySAS sample data



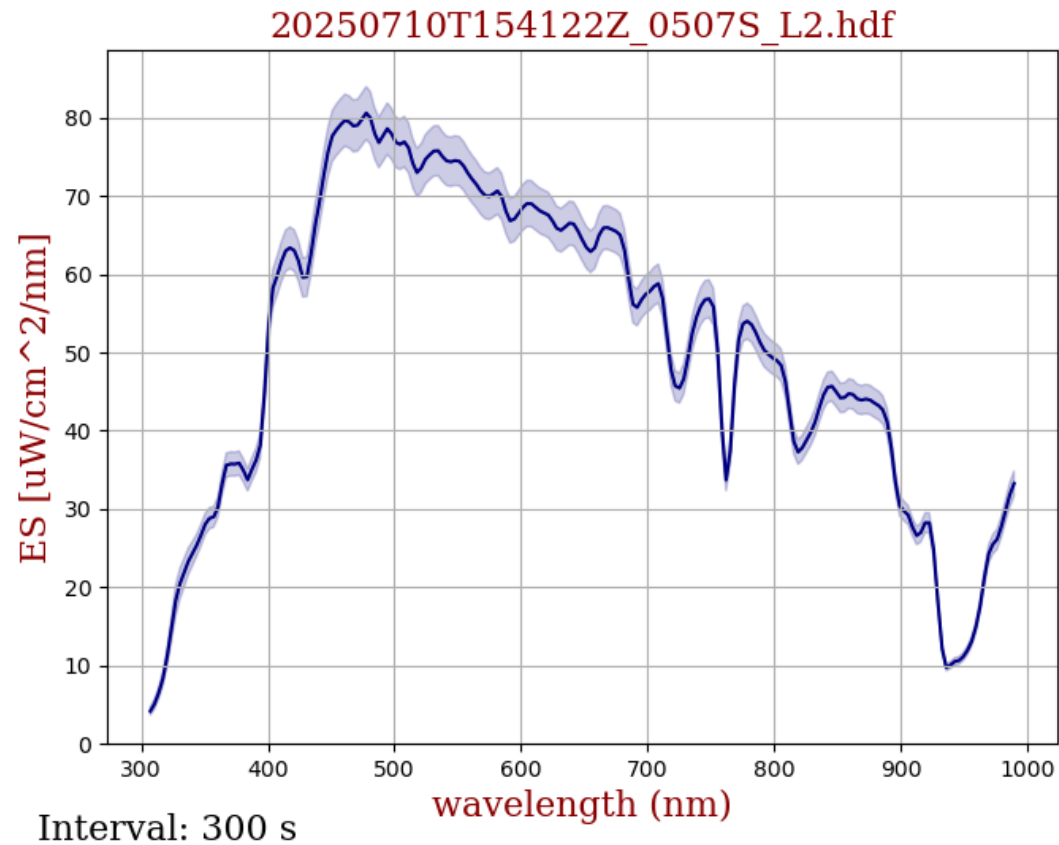


# Uncertainty Results – PySAS sample data

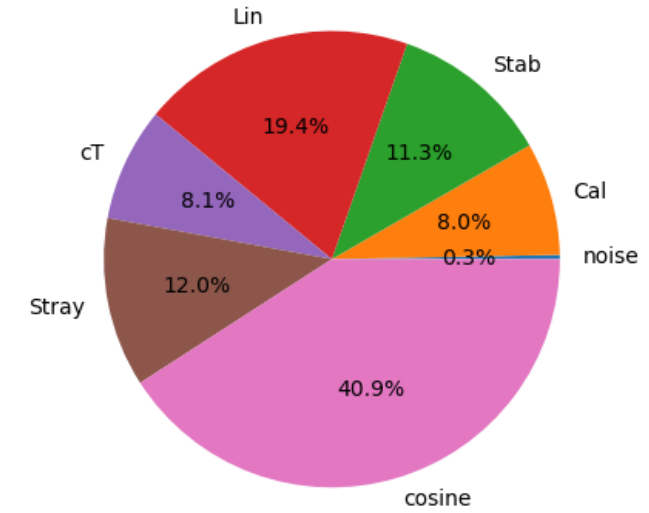


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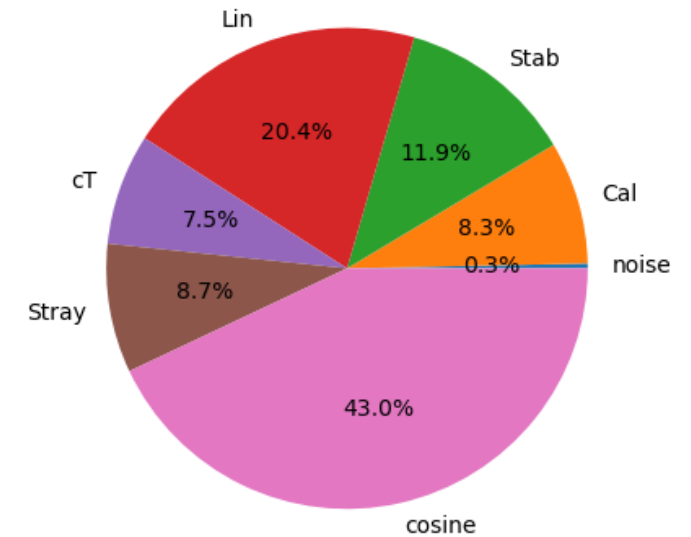
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SZA =59.5



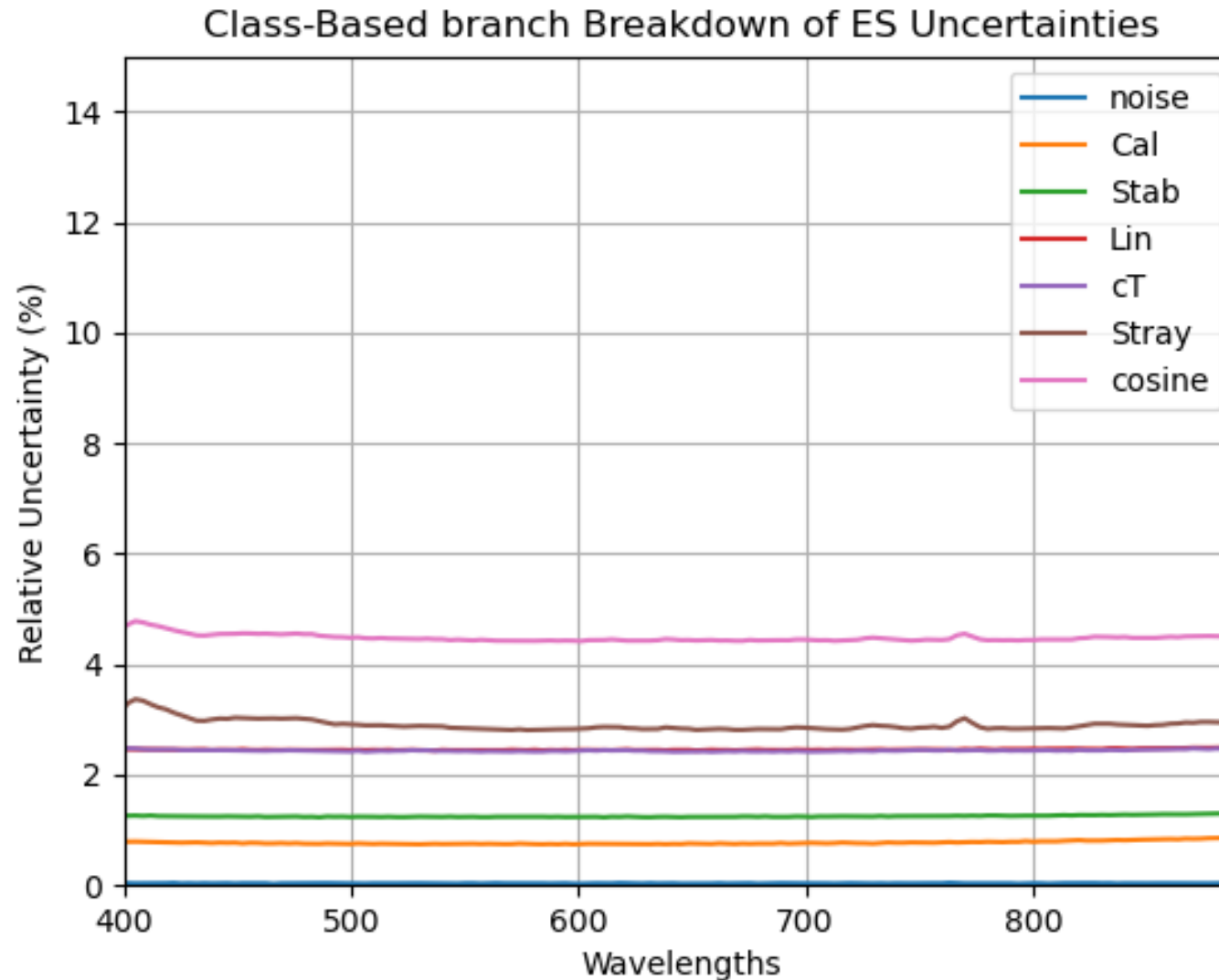
ES Class Based Uncertainty Components at 441.86nm



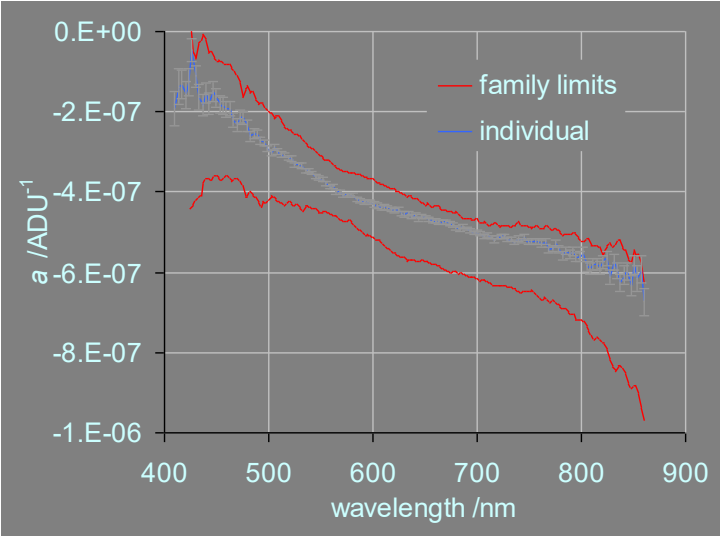
ES Class Based Uncertainty Components at 675.46nm



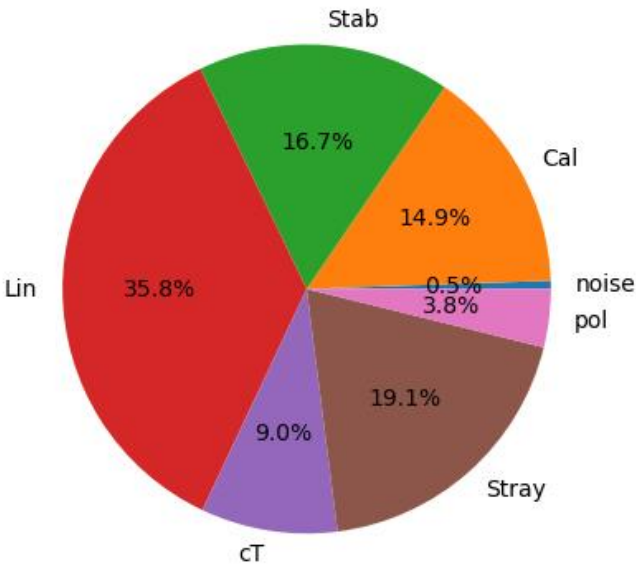
# Cumulative Uncertainty breakdown



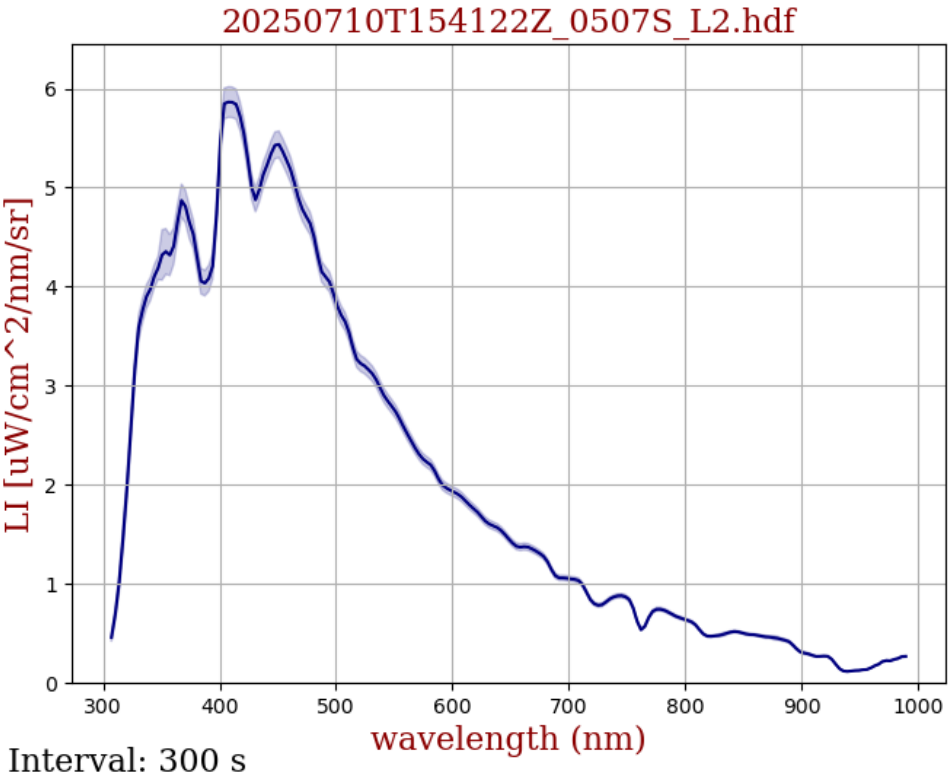
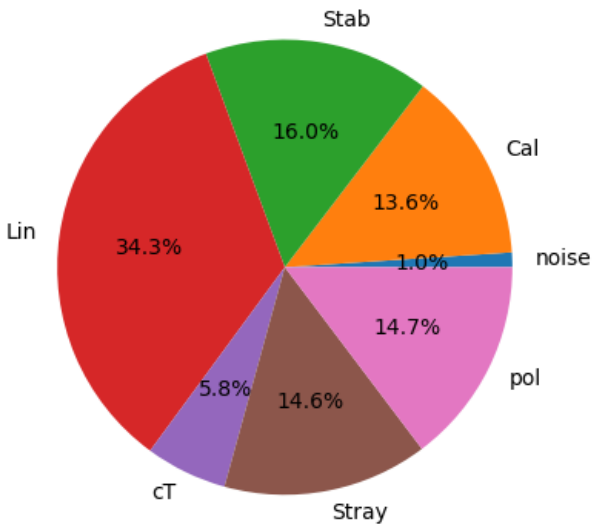
# FICE2025



LI Class Based Uncertainty Components at 441.86nm

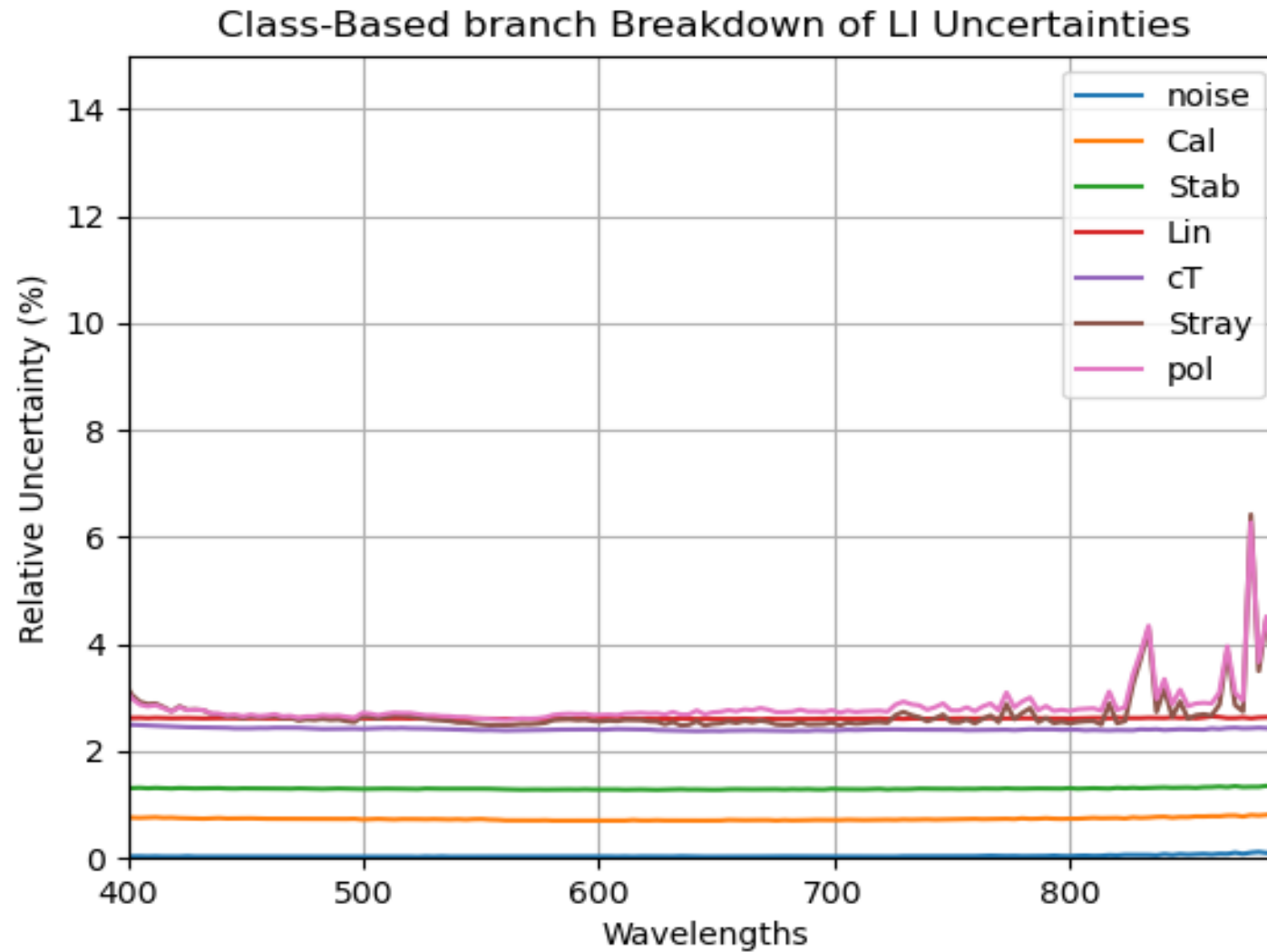


LI Class Based Uncertainty Components at 675.46nm

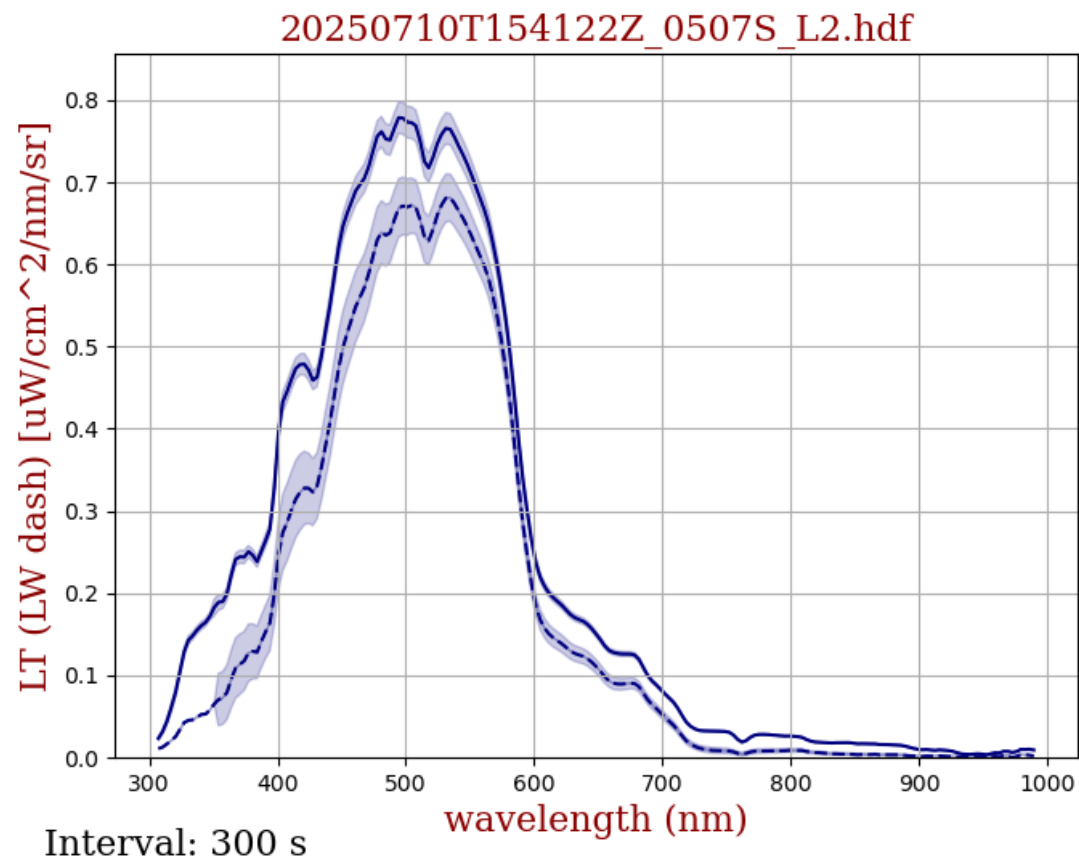




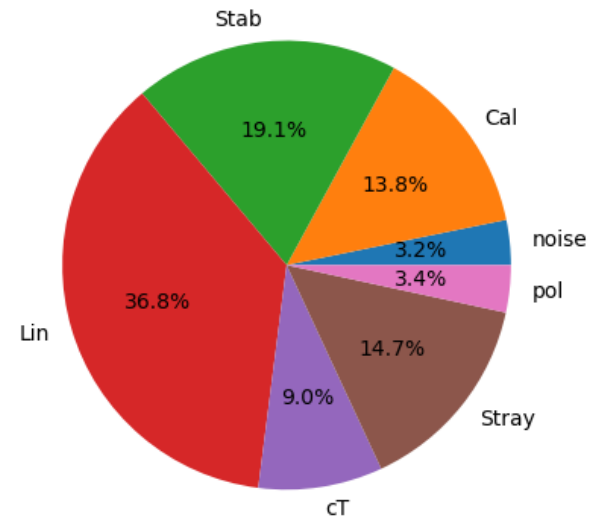
# Cumulative Uncertainty breakdown



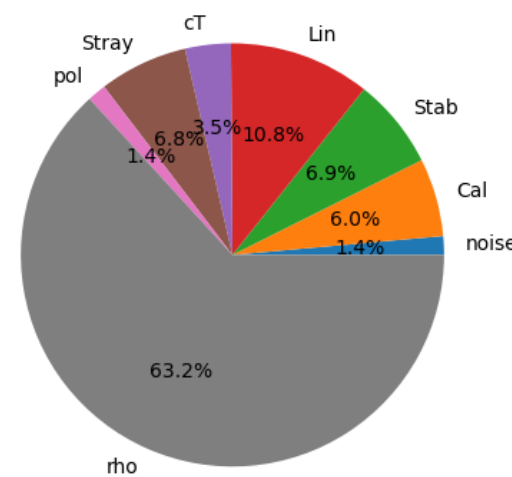
# FICE2025



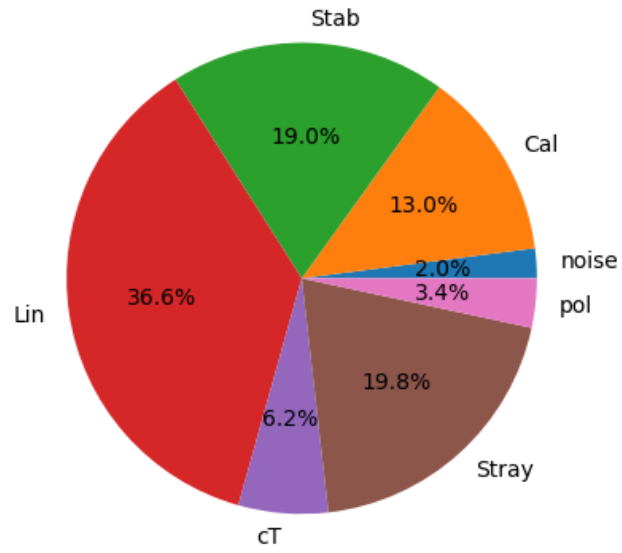
LT Class Based Uncertainty Components at 441.86nm



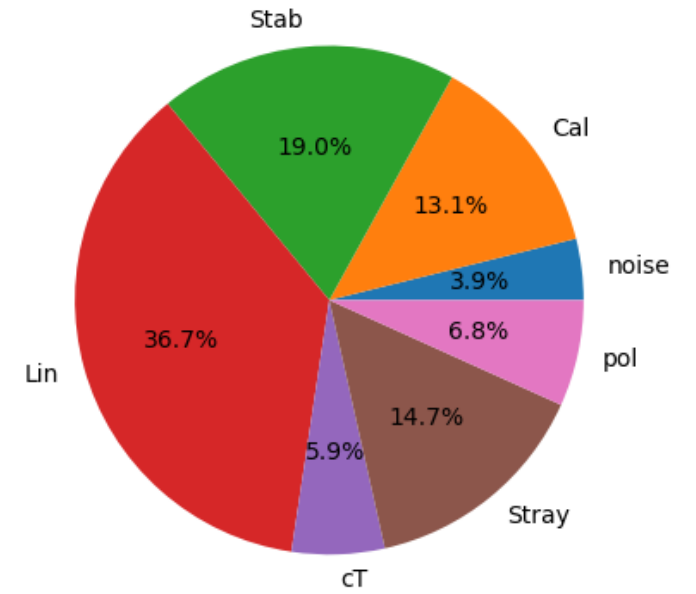
Lw Class Based Uncertainty Components at 441.86nm



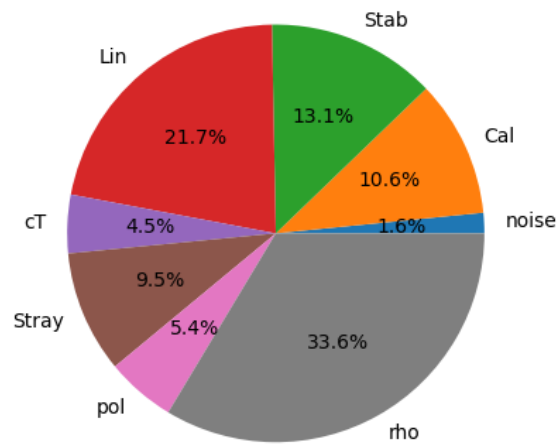
LT Class Based Uncertainty Components at 560.32nm



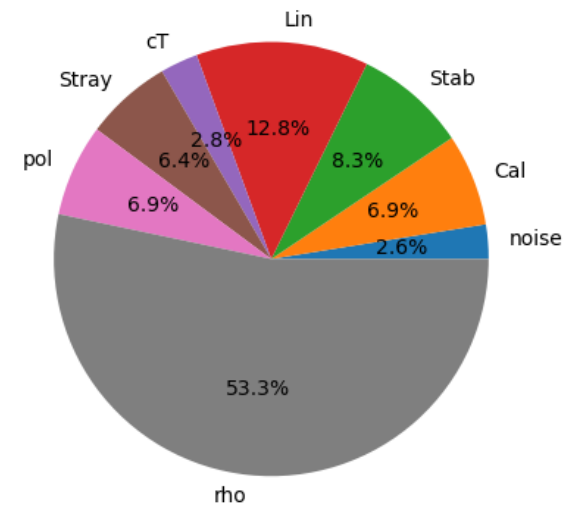
LT Class Based Uncertainty Components at 675.46nm



Lw Class Based Uncertainty Components at 560.32nm

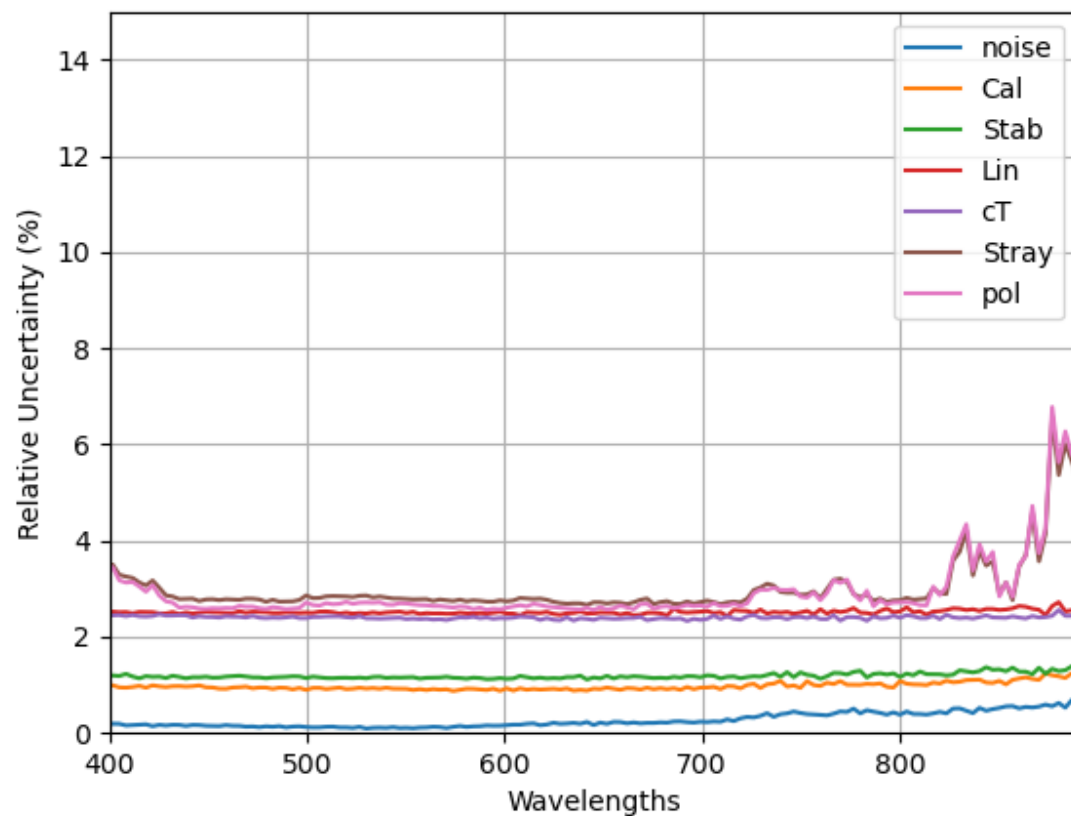


Lw Class Based Uncertainty Components at 675.46nm

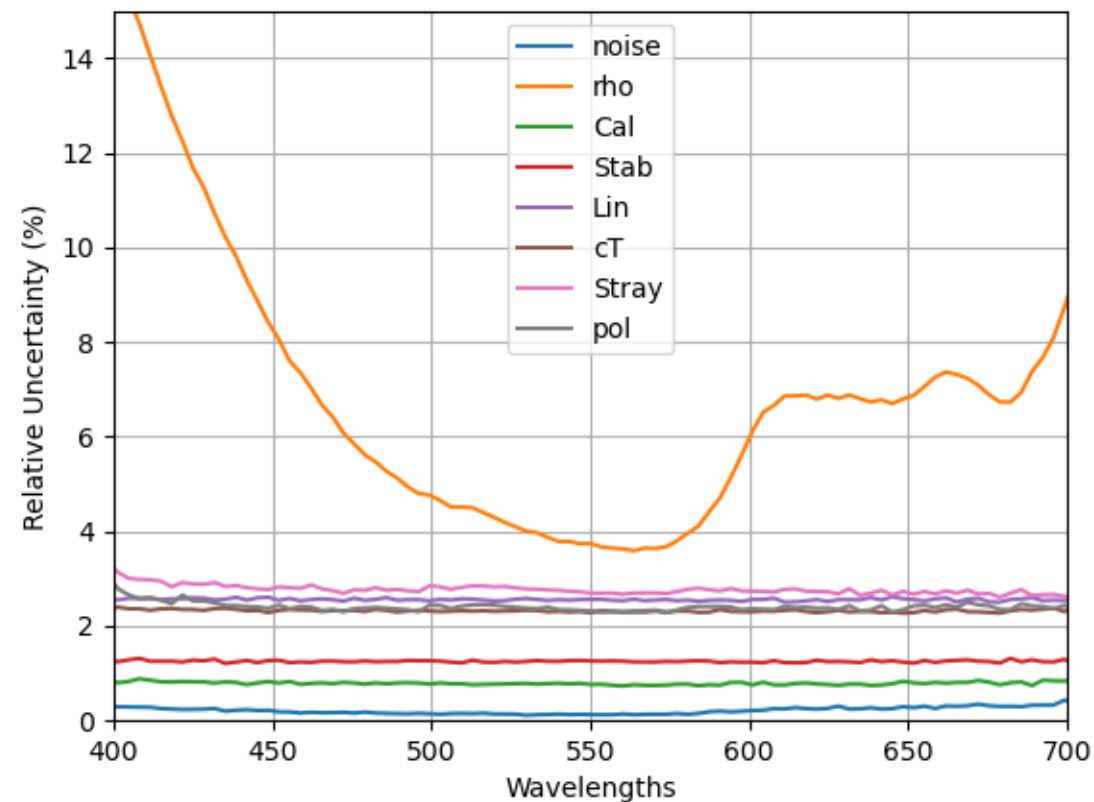


# Cumulative Uncertainty breakdown

Class-Based branch Breakdown of LT Uncertainties

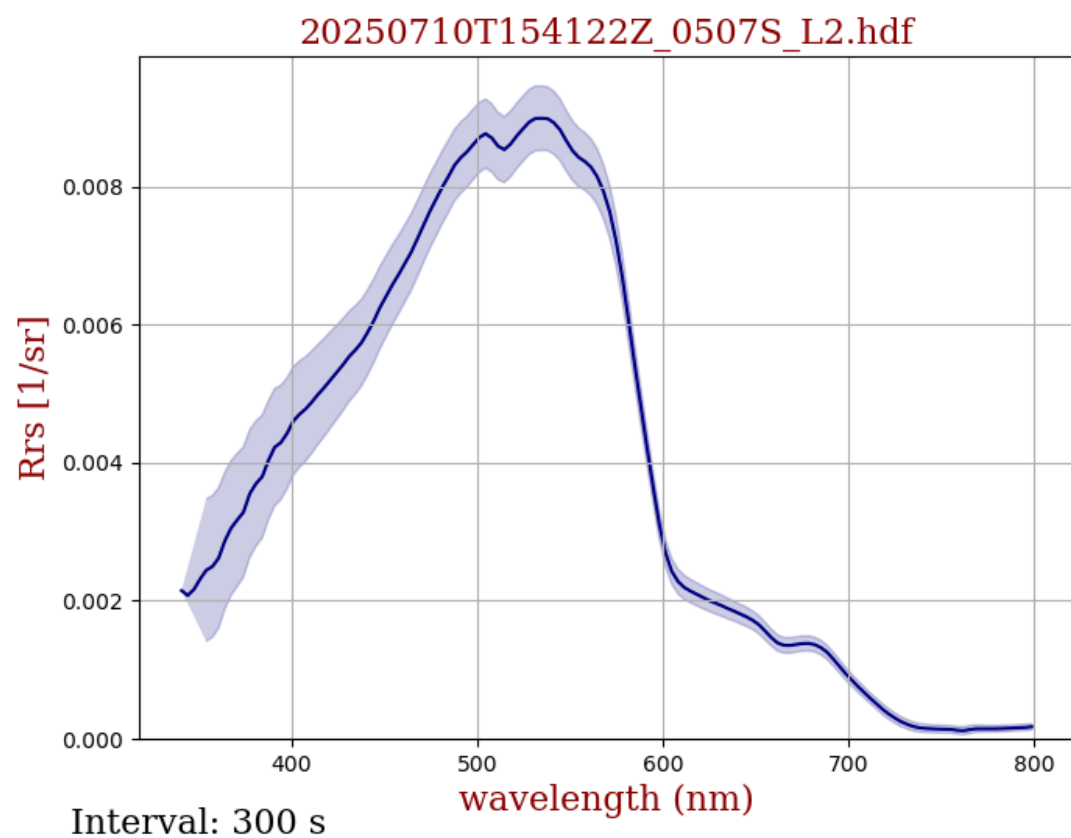


Class-Based branch Breakdown of Lw Uncertainties

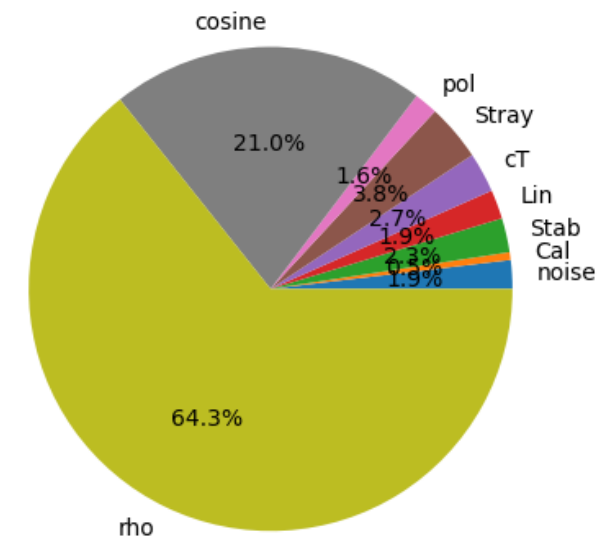




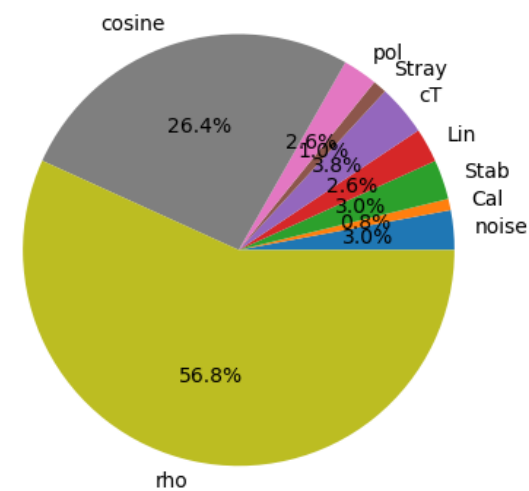
# FICE2025



Rrs Class Based Uncertainty Components at 441.86nm



Rrs Class Based Uncertainty Components at 675.46nm



**Congratulation!**

**I finished and you survived ;-)**