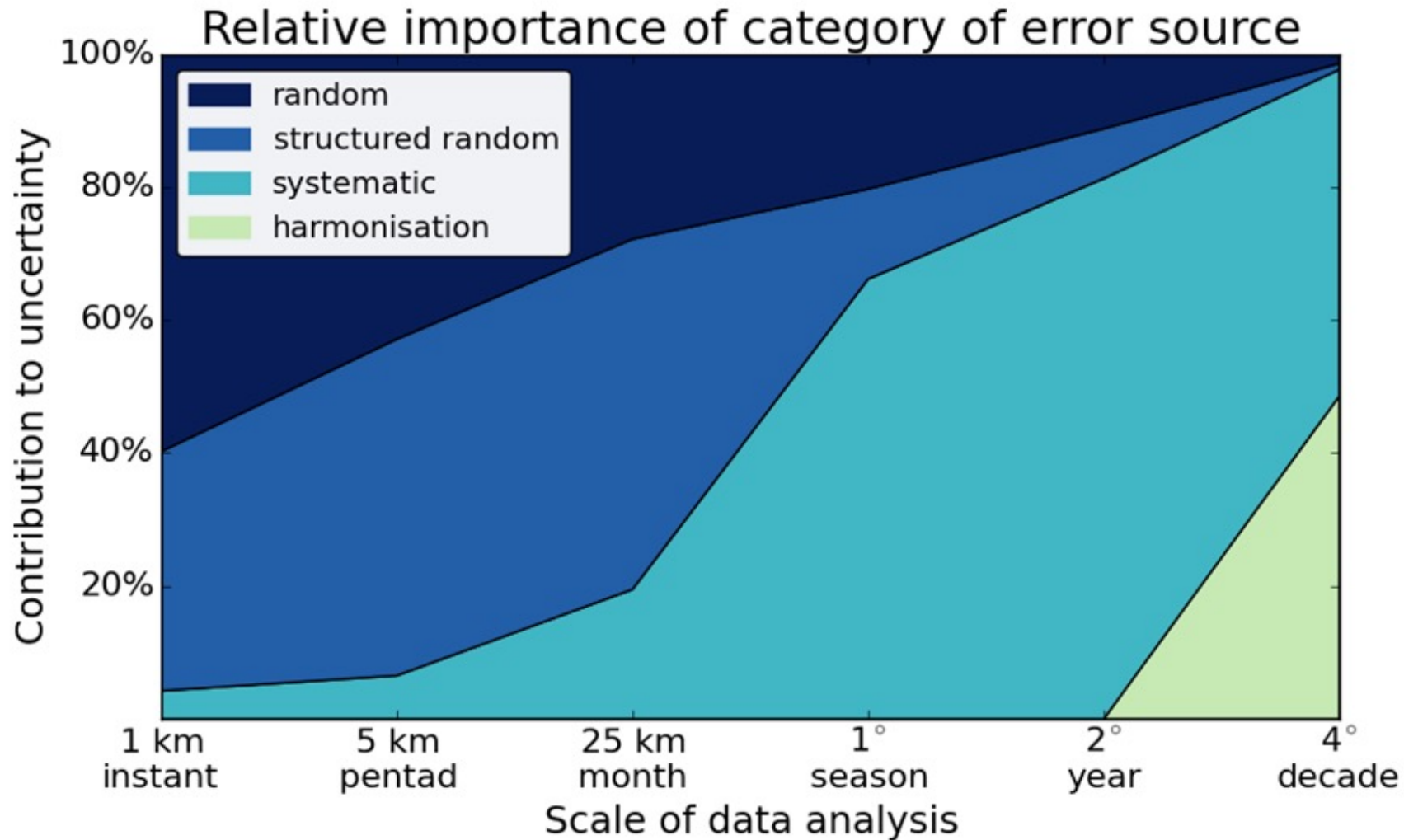


Community Tools for Metrology

QA4EO Cal/Val Workshop #3

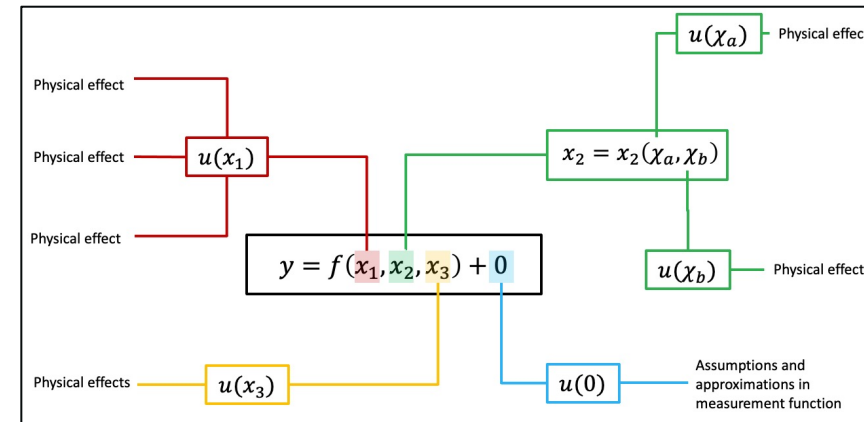
Sam Hunt & Pieter De Vis

Why do we care about error-correlation?



Uncertainty Analysis

- In full error-covariance matrices are impractical to evaluate and store for EO data
- “FIDUCEO-style” approach to uncertainty analysis offers a solution by parameterising error-covariance structure
- How to take this to the next step?
How do I store and make use of this information in data?



Uncertainty Tree Diagram

	Comments	
Name of effect	A unique name	
Affected term in measurement function	Name and standard symbol	
Instruments in the series affected	List names	
Correlation type and form	Pixel-to-pixel [pixels] from scanline to scanline [scanlines] between images [images] Between orbits [orbit] Over time [time]	From a set of defined correlation forms
Correlation scale	Pixel-to-pixel [pixels] from scanline to scanline [scanlines] between images [images] Between orbits [orbit] Over time [time]	As needed to define type
Channels/bands	List of channels / bands affected Error correlation coefficient matrix	Channel names A matrix
Uncertainty	PDF shape units magnitude	Functional form Units
Sensitivity coefficient	Value, equation or parameterisation of sensitivity of measurand to term	

Effect Table



Encoded Observations

Geospatial data is encoded with complex metadata, though users typically never have to interact with it.



Encoded Observations

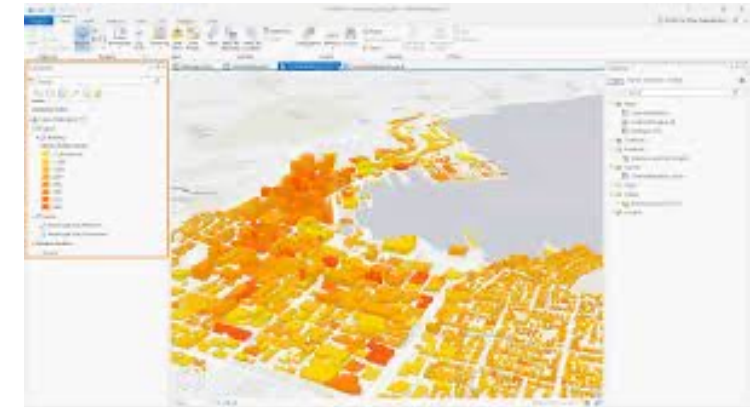
Geospatial data is encoded with complex metadata, though users typically never have to interact with it.

Example: Geocoding

1. Data is accompanied with standardised metadata
2. Tools provide means to

A. Interface with this information

B. Interpret and make use of this information



Encoded Observations

Geospatial data is encoded with complex metadata, though users typically never have to interact with it.

Why not take the same approach for error-covariance information for observations?



Encoded Observations

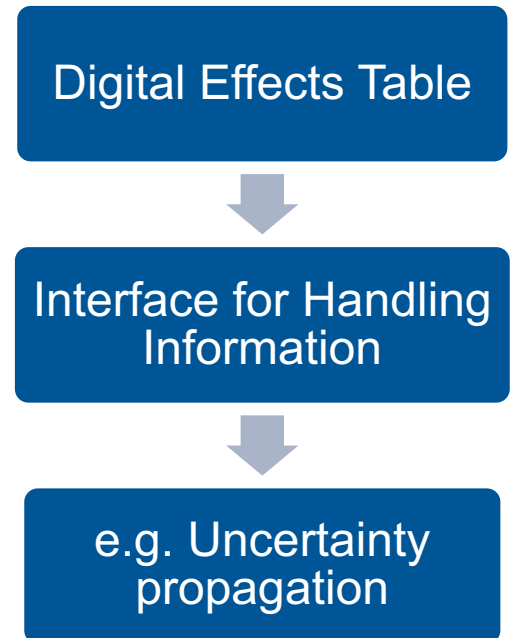
Geospatial data is encoded with complex metadata, though users typically never have to interact with it.

Parallel: Error-covariance encoding

1. Data is accompanied with standardised metadata
2. Tools provide means to
 - A. Interface with this information
 - B. Interpret and make use of this information

F|duceo

obsarray
punpy



Standardised Error-Covariance Metadata: Digital Effects Tables

	Comments	
Name of effect	A unique name	
Affected term in measurement function	Name and standard symbol	
Instruments in the series affected	List names	
Correlation type and form	Pixel-to-pixel [pixels]	From a set of defined correlation forms
	from scanline to scanline [scanlines]	
	between images [images]	
	Between orbits [orbit]	
	Over time [time]	
Correlation scale	Pixel-to-pixel [pixels]	As needed to define type
	from scanline to scanline [scanlines]	
	between images [images]	
	Between orbits [orbit]	
	Over time [time]	
Channels/bands	List of channels / bands affected	Channel names
	Error correlation coefficient matrix	A matrix
Uncertainty	PDF shape	Functional form
	units	Units
	magnitude	
Sensitivity coefficient	Value, equation or parameterisation of sensitivity of <u>measurand</u> to term	



```
double u_str_temperature(x=2, y=2, time=3);
  :_FillValue = 9.969209968386869E36; // double
  :err_corr_1_dim = "x";
  :err_corr_1_form = "custom";
  :err_corr_1_units = ; // double
  :err_corr_1_params = "err_corr_str_temperature_x";
  :err_corr_2_dim = "y";
  :err_corr_2_form = "systematic";
  :err_corr_2_units = ; // double
  :err_corr_2_params = ; // double
  :err_corr_3_dim = "time";
  :err_corr_3_form = "systematic";
  :err_corr_3_units = ; // double
  :err_corr_3_params = ; // double
  :pdf_shape = "gaussian";
```

Print out of uncertainty variable attributes for netCDF file

F|duceo Effects Table

Digital Effects Table



Interface to Error-Covariance Metadata: `obsarray`

The ***obsarray*** python module provides an extension to the widely used ***xarray*** package to interface with measurement error-covariance information encoded in datasets

```
print(ds.temperature)
```

```
<xarray.DataArray 'temperature' (x: 2, y: 2, time: 3)>  
array([[[16.969685,  4.8038  , 27.2702  ],  
        [23.128404, 18.645507, 20.901654]],
```

```
        [[19.035036,  8.91408  ,  4.718095],  
         [ 6.079486, 18.107981, 11.597675]])])
```

```
Dimensions without coordinates: x, y, time
```

```
Attributes:
```

```
    unc_comps: ['u_ran_temperature', 'u_str_temperature', 'u_sys_temperature']
```



Interface to Error-Covariance Metadata: **obsarray**

The *obsarray* python module provides an extension to the widely used *xarray* package to interface with measurement error-covariance information encoded in datasets

```
# Inspect uncertainty variables for a particular variable
```

```
print(ds.unc["temperature"])
```

```
<VariableUncertainty>
```

```
Variable Uncertainties: 'temperature'
```

```
Data variables:
```

```
u_ran_temperature (x, y, time) float64 0.8485 0.2402 ... 0.9054 0.5799  
u_str_temperature (x, y, time) float64 0.5091 0.1441 ... 0.5432 0.3479  
u_sys_temperature (x, y, time) float64 0.5091 0.1441 ... 0.5432 0.3479
```



Interface to Error-Covariance Metadata: **obsarray**

The ***obsarray*** python module provides an extension to the widely used ***xarray*** package to interface with measurement error-covariance information encoded in datasets

```
# Get total variable uncertainty
```

```
ds.unc["temperature"].total
```

```
xarray.DataArray (x: 2, y: 2, time: 3)
```

```
array([[[[1.11277669, 0.31500626, 1.78822662],  
         [1.51663088, 1.22266768, 1.37061312]],  
       [[1.24821081, 0.58453532, 0.3093862 ],  
         [0.39865853, 1.18741975, 0.76051039]]]])
```



Tools for Error-Covariance Metadata: punpy

punpy interfaces with **obsarray** to make uncertainty propagation as efficient and easy to use as possible. All flexibility of punpy is included as optional keywords. The `propagate_ds()` function returns an **obsarray** dataset with combined random, systematic and structured uncertainties on measurand.

```
# Define your measurement function inside a subclass of MeasurementFunction
class GasLaw(MeasurementFunction):
    def function(self, pres, temp):
        return pres/(temp*8.134)

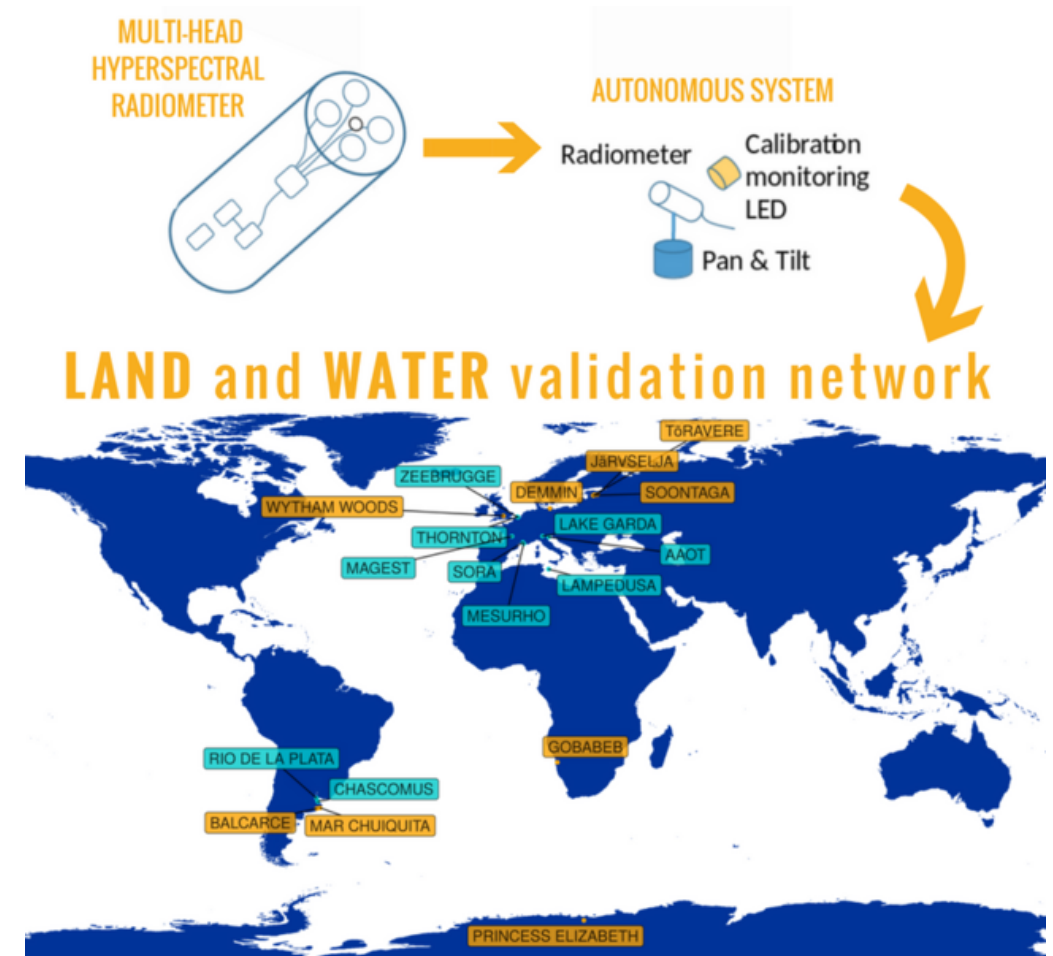
# create class object and pass all optional keywords for punpy
gl = GasLaw(["pressure","temperature"],steps=100000)

# propagate the uncertainties on the input quantities in ds to measurand uncertainties in ds_y
ds_y=gl.propagate_ds("V/n",ds)
```



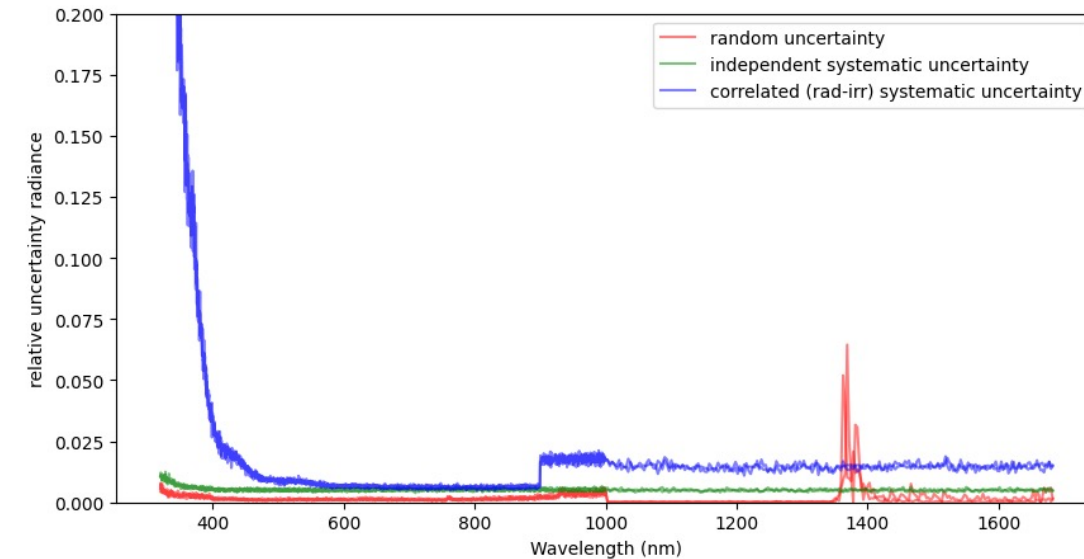
Example: Hypernets Ground Processor

- Hypernets is an underdevelopment network of ground test sites with automated hyperspectral spectrometers for surface reflectance validation



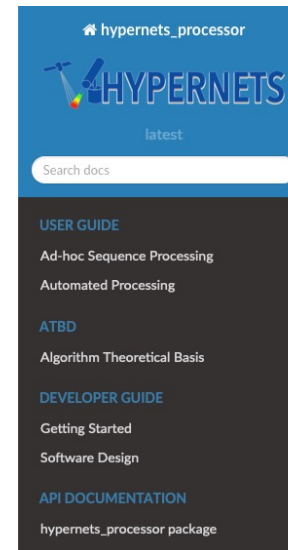
Example: Hypernets Ground Processor

- Hypernets is an underdevelopment network of ground test sites with automated hyperspectral spectrometers for surface reflectance validation
- Uncertainty information is provided with every product, including error-correlation information.



Example: Hypernets Ground Processor

- Hypernets is an underdevelopment network of ground test sites with automated hyperspectral spectrometers for surface reflectance validation
- Uncertainty information is provided with every product, including error-correlation information.
- Implementation in ground processor is powered by CoMet tools



[Docs](#) » [hypernets_processor: Hypernets water and land network data processor](#) [Edit on GitHub](#)

hypernets_processor: Hypernets water and land network data processor

The `hypernets_processor` module is a Python software package to process the [Hypernets](#) land and water network in-situ hyperspectral data to surface reflectance products for distribution to users.

There are two main use cases for the `hypernets_processor` module. The primary function of the software is the automated preparation of data retrieved from network sites for distribution to users. Additionally, the software may also be used for ad-hoc processing of particular field acquisitions, for example for testing instrument operation in the field.

This documentation aimed at both users and developers of the software, find the relevant sections below.

User Guide

- [Ad-hoc Sequence Processing](#)
- [Automated Processing](#)



CoMet: Community tools for Metrology

- An open-source software project to develop Python tools for the handling of error-covariance information in the analysis of measurement data
- Includes **obsarray** and **punpy** as initial offering, to be extended (optimisation next)
- Moving towards initial release on GitHub/PyPI platform



Next Steps

- Developing tools further – including expanding scope to include more functionality, such as optimisation.
- Development of documentation, examples and dissemination approach
- We are looking for beta testers!

