

# Normalization

January 14, 2016

```
In [1]: # This is a cell to hide code snippets from displaying
        # This must be at first cell!
```

```
from IPython.display import HTML, Image
```

```
hide_me = ''
HTML(''<script>
code_show=true;
function code_toggle() {
  if (code_show) {
    $('div.input').each(function(id) {
      el = $(this).find('.cm-variable:first');
      if (id == 0 || el.text() == 'hide_me') {
        $(this).hide();
      }
    });
    $('div.output_prompt').css('opacity', 0);
  } else {
    $('div.input').each(function(id) {
      $(this).show();
    });
    $('div.output_prompt').css('opacity', 1);
  }
  code_show = !code_show
}
$( document ).ready(code_toggle);
</script>
<form action="javascript:code_toggle()"><input style="opacity:0" type="submit" value="Click here
```

```
Out[1]: <IPython.core.display.HTML at 0x7f279812dc10>
```

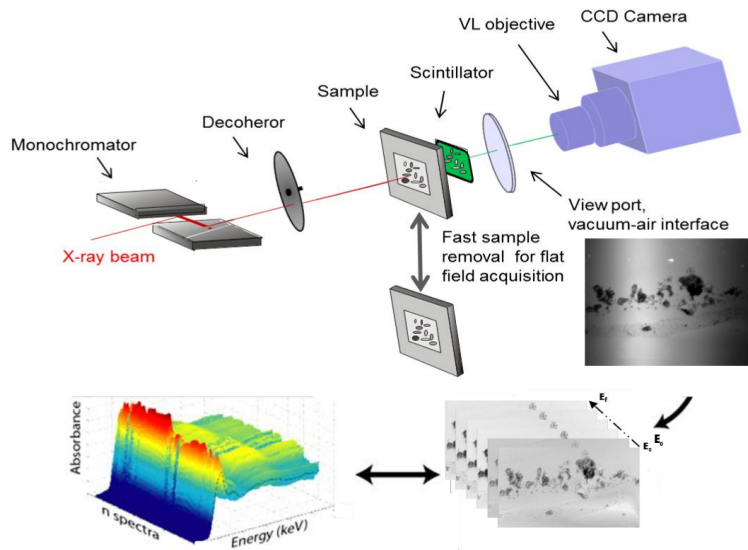
## 1 Image normalization

This example is part of the processing performed during a “Fullfield XANES” experiment at ESRF-ID21 and described in this article: [http://iopscience.iop.org/1742-6596/425/19/192001/pdf/1742-6596\\_425\\_19\\_192001.pdf](http://iopscience.iop.org/1742-6596/425/19/192001/pdf/1742-6596_425_19_192001.pdf)

During this experiment, radiographs of a sample are taken at various energies. As the full beam is used, one should divide the signal by the flat-field image, both corrected for the dark current of the CCD.

```
In [2]: hide_me
        Image(filename='files/setup.png', width=800)
```

```
Out[2]:
```



## Full Field Setup

Energy range: 2 - 9 keV

Spatial resolution: 0.3 to 1.3  $\mu\text{m}$

Field of view:  $\sim 600 \mu\text{m}$  to 2 mm

Acquisition time: 15 to 60 min per full-field stack ( $5 \times 10^6$  spectra)

For an optimal use of the dynamic range, the data frame has a longer exposure time than the flat-field frame. Exposure time is recorded in the header of the file. Each frame needs to be corrected for the correct dark (with the same exposure time).

There are three dark-current files: dark-2.edf, dark-3 and dark-6.edf, two flat-field taken before and after the experiment and a raw data frame in the directory. Correct the image the image for the dark and flat-field effect. Check the correctness of your results often with “imshow” ...

Exercise:

- Read all 6 EDF files and find their exposure time
- Correct the data and flat-field images for their respective dark-current images
- Average the two flat-field images
- Correct the data from the flat-field effect

## 2 Solution

### 2.1 1. Initialization of the scientific environment

```
In [3]: %pylab inline
```

Populating the interactive namespace from numpy and matplotlib

Since iPython 3 or Jupyter 4, one can also use “%pylab nbagg” to provide interactivity on the displayed images.

```
In [4]: import numpy, fabio
```

### 2.2 2. Load all images

*Nota:* Did you know you can call a shell command from iPython ?

```
In [5]: ls *.edf
```

```
dark-2.edf  dark-3.edf  dark-6.edf  flat-1.edf  flat-2.edf  raw.edf
```

```
In [6]: dark2 = fabio.open("dark-2.edf")
        dark3 = fabio.open("dark-3.edf")
        dark6 = fabio.open("dark-6.edf")
        flat1 = fabio.open("flat-1.edf")
        flat2 = fabio.open("flat-2.edf")
        raw = fabio.open("raw.edf")
        #print the content of the header of the "raw.edf" file
        for key,value in raw.header.items():
            print("%15s: %s"%(key, value))
```

```
HeaderID: EH:000001:000000:000000
ByteOrder: LowByteFirst
DataType: SignedInteger
Size: 16777216
Dim_1: 2048
Dim_2: 2048
Image: 0
acq_frame_nb: 0
time: Thu Dec 16 18:36:31 2010
time_of_day: 1292520991.838942
time_of_frame: 4.146026
energy: 4.99
exposure_time: 6
```

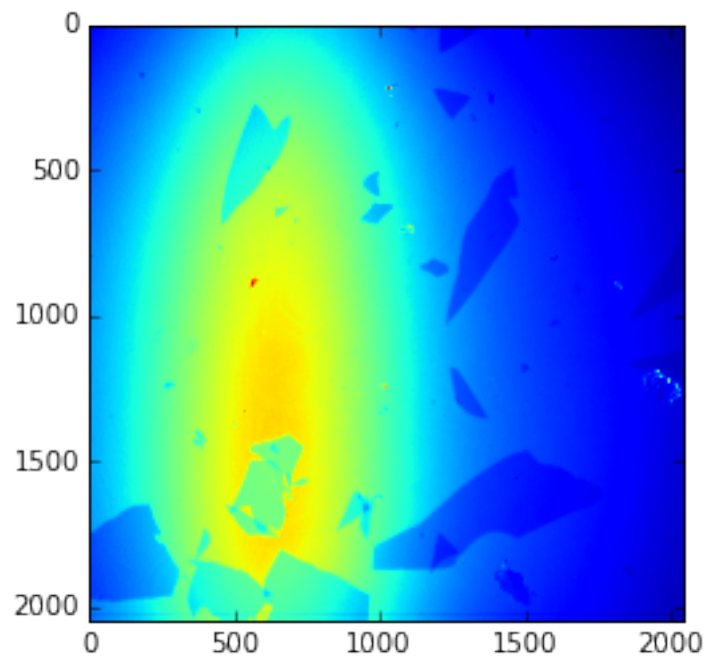
```
In [7]: #print exposure time for all images, from the metadata stored in the header
        for edf in (dark2, dark3, dark6, flat1, raw):
            print("%s : %s"%(edf.filename,edf.header["exposure_time"]))
```

```
dark-2.edf : 2
dark-3.edf : 3
dark-6.edf : 6
flat-1.edf : 3
raw.edf : 6
```

Display the image and its data-type:

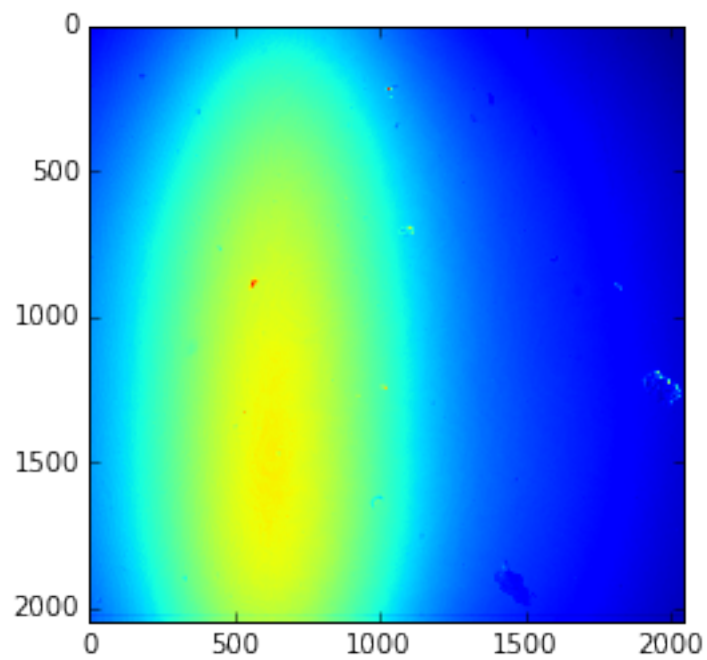
```
In [8]: imshow(raw.data)
        print(raw.data.dtype)
```

```
int32
```



```
In [9]: imshow(flat1.data)
        print(flat1.data.dtype)
```

int32



*nota:* check the data-type of your data as well: intergers are very bad for calculation due to over/under-flow and integer division in Python2. All scientific calculation should be performed using floating point numbers.

### 2.3 3. Dark-current correction

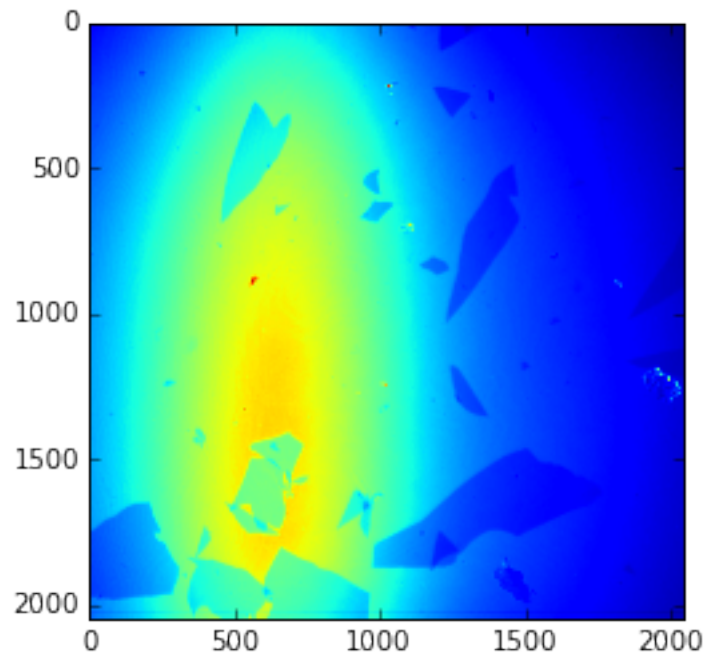
Convert the data-type for the calculation in *float*.

*nota:* Only one member of the equation needs to be *float* as the other will be casted automatically to *float*

```
In [10]: raw_dc = raw.data.astype(float) - dark6.data
         flat1_dc = flat1.data.astype(float) - dark3.data
         flat2_dc = flat2.data.astype(float) - dark3.data
         print(raw_dc.dtype)
         imshow(raw_dc)
```

float64

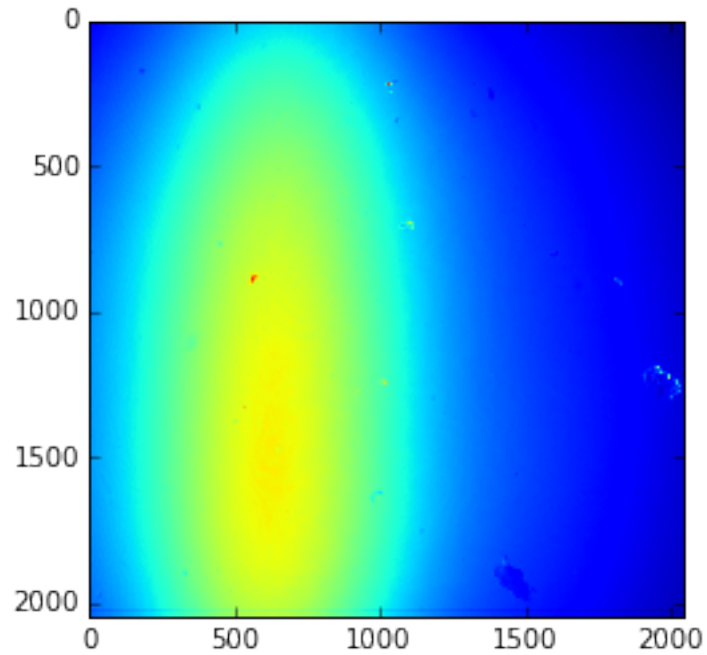
```
Out[10]: <matplotlib.image.AxesImage at 0x7f276c8ad8d0>
```



### 2.4 4. Average out the flat-field images

```
In [11]: flat = (flat1_dc + flat2_dc) / 2.0
         imshow(flat)
```

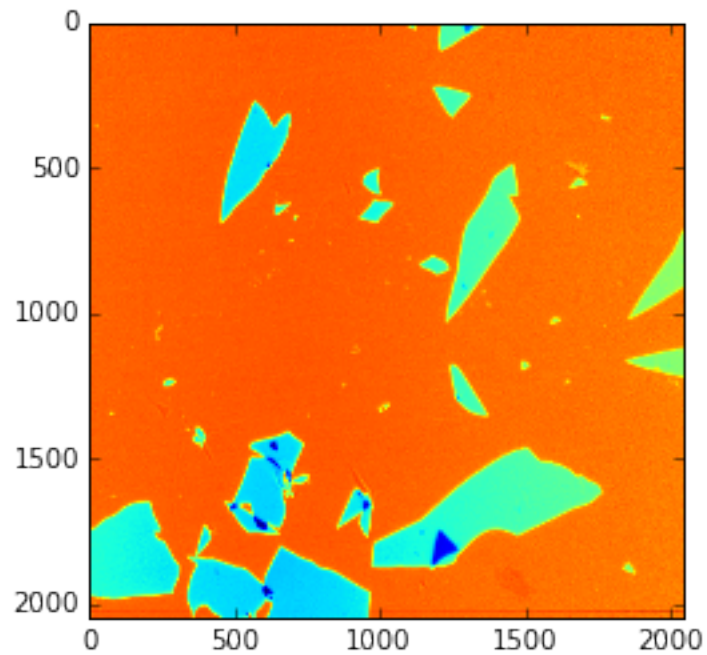
```
Out[11]: <matplotlib.image.AxesImage at 0x7f276c7a2e50>
```



## 2.5 5. Perform the flat-field correction, weighted by exposure time

In [12]: *#Data has been exposed for 6s, while flat was only exposed 3s*  
`corr = (raw_dc / 6.0) / (flat / 3.0)`  
`imshow(corr)`

Out[12]: <matplotlib.image.AxesImage at 0x7f276c744f90>



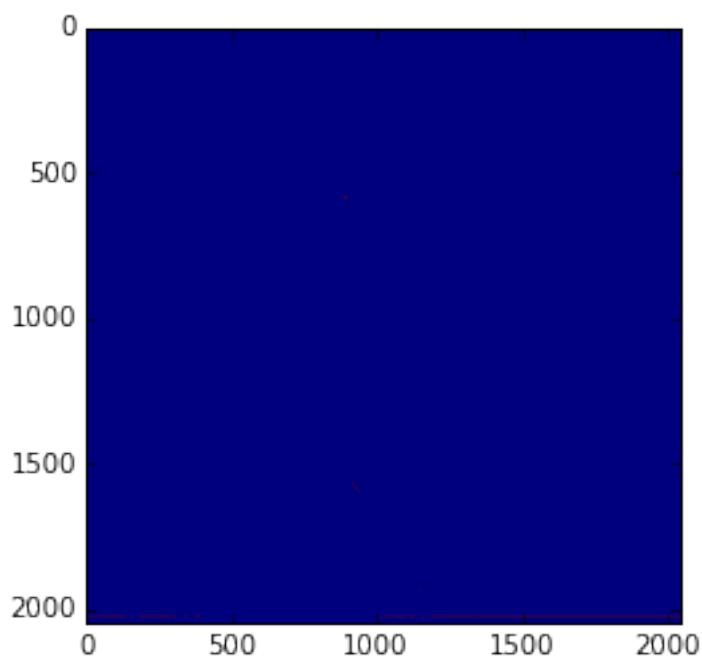
## 2.6 6. Why was it so important to calculate with floats ?

We will perform the calculation using integers and here is the result using Python2

```
In [13]: int_corr = (raw.data - dark6.data) / (flat1.data + flat2.data - 2*dark3.data)
```

```
In [14]: imshow(int_corr)
```

```
Out[14]: <matplotlib.image.AxesImage at 0x7f276c668e10>
```



```
In [15]: print(int_corr)
          print(int_corr.max())
```

```
[[0 0 0 ..., 0 0 0]
 [0 0 0 ..., 0 0 0]
 [0 0 0 ..., 0 0 0]
 ...,
 [0 0 0 ..., 0 0 0]
 [0 0 0 ..., 0 0 0]
 [0 0 0 ..., 0 0 0]]
```

1

What happened ?

The normalisation provides values between 0 and 1 and all calculation performed using integers results in 0 everywhere except for 1 or a few pixels which are 1.

### 3 Conclusion

Most image manipulation can simply be performed using *numpy* together with a library to load images. This library depends on the image type. The *h5py* library is the one with the greatest future as ESRF is moving towards HDF5 and it provides a nice “dictionary-like” interface.

In [] :