1. Getting started

1.1 Example filterbank files

Use one of the following filterbank files as an example:

- $\bullet \ \ \text{``https://git.dev.ti-more.net/uploads/-/system/personal_snippet/2/e6015ec024ad1f53d4c2f39511620db1/personal_snippet/2/e60160db1/personal_snippet/2/e60$

1.2 Import

from filterbank.filterbank import *

2. Filterbank Tutorial

2.1 Create a filterbank object

filterbank = Filterbank(<PATH TO FILTERBANK FILE>)

This is an example without parameters, see 2.3 for an example with the parameters. ## 2.2 Read the header from filterbank data

filterbank.header

Header data contains the following:

Variable	Description
source_name	name of filterbank file
P	period of pulsar in ms
DM	dispersion measure of pulsar
$machine_id$	id of machine used to receive signal data
$telescope_id$	id of telescope used to receive signal data
data_type	type of file filterbank, time series
fch1	center frequency of first filterbank channel (MHz)
foff	filterbank channel bandwidth (MHz)
nchans	number of filterbank channels
nbits	number of bits per time sample 8, 16 or 32
tstart	timestamp of first sample (MJD)
nifs	number of seperate intermediate-frequency channels

```
freq\_range = fch1 + (foff * nchans + fch1)

time\_range = tstart + (tsamp/24/60/60)
```

freq_range is a tuple with a frequency start and a frequency stop. The same applies to time_range

The header data, including the center frequency, can be retrieved by calling the header attribute from the Filterbank object.

2.3 Read filterbank file

The attributes time_range and freq_range can be passed as parameters to select a specific portion of the filterbank file. To make the Filterbank object read the filterbank file at once, set the read all parameter to True.

filterbank = Filterbank(<PATH TO FILTERBANK FILE>, freq_range, time_range, read_all)

2.4 Select a range of data from the filterbank file

The select_data method can be used to retrieve data from the Filterbank object. The user has the option to give a time and/or frequency range to select a subset from the entire dataset. The time-range can either be an index of a sample or a float that is represents a moment in time (in seconds).

```
filterbank.select_data(freq_range, time_range)
```

The select_data method returns an array of all different channels/frequencies and a large matrix with all the received radio signals.

The matrix contains for each time sample an array which has the intensity per channel/frequency.

2.5 Read filterbank as stream

When reading the filterbank file as a stream, the user should let the read_all-parameter stay False when initializing the filterbank object.

Each time the user calls the next_row method, it will retrieve an array with intensitiy per frequency for a new time sample from the filterbank file. When the last iteration of the filterbank is reached, the new_row method will return False.

The same goes for the next_n_rows method, where the user is able to define the amount of rows that should be returned.

```
filterbank.next_row()
```

filterbank.next_n_rows(n_rows=10)

3. Fourier Transformations

3.1 DFT

The fourier.dft_slow function is a plain implementation of discrete Fourier transformation.

The fourier.fft_vectorized function depends on this function.

3.1.1 Parameters

Parameter	Description
input_data	Array containing the values to be transformed.

Returns an array containing the transformed values.

3.1.2 Example usage

```
>>> from Asteria import fourier
>>> fourier.dft_slow([1,2,3,4])
array([10.+0.00000000e+00j, -2.+2.00000000e+00j, -2.-9.79717439e-16j, -2.-2.00000000e+00j])
```

3.2 FFT

The $fourier.fft_vectorized$ function is a vectorized, non-recursive version of the Cooley-Tukey FFT

Gives the same result as fourier.dft_slow but is many times faster.

3.2.1 Parameters

Parameter	Description	
input_data	Array containing the values to be transformed.	

Returns an array containing the transformed values.

3.2.2 Example usage

```
>>> from Asteria import fourier
>>> fourier.fft_vectorized([1,2,3,4])
array([10.+0.00000000e+00j, -2.+2.00000000e+00j, -2.-9.79717439e-16j, -2.-2.00000000e+00j])
```

4. Plotting diagrams

4.1 PSD

The plot.psd function can be used to generate the data for a Power Spectral Density plot.

4.1.1 Parameters

Parameter	Description
samples	1-D array or sequence. Array or sequence containing the data
	to be plotted.
nfft	Integer, optional. The number of bins to be used. Defaults to
	256.
$sample_rate$	Integer, optional. The sample rate of the input data in samples
	Defaults to 2.
window	Callable, optional. The window function to be used. Defaults to
	plot.window_hanning.
sides	{'default', 'onesided', 'twosided'}. Specifies which sides of the
	spectrum to return. Default gives the default behavior, which
	returns one-sided for real data and both for complex data.
	'onesided' forces the return of a one-sided spectrum, while
	'twosided' forces two-sided.

4.1.2 Returns

Variable	Description
Pxx	1-D array. The values for the power spectrum before scaling (real valued).
freqs	1-D array. The frequencies corresponding to the elements in Pxx.

4.1.3 Example Usage

```
from filterbank.filterbank import Filterbank
import matplotlib.pyplot as plt
import numpy as np
from plot import psd
from filterbank.header import read_header
# Instatiate the filterbank reader and point to the filterbank file
fb = Filterbank(filename='examples/pspm32.fil')
# read the data in the filterbank file
_, samples = fb.select_data()
# Convert 2D Array to 1D Array with complex numbers
samples = samples[0] + (samples[1] * 1j)
# Read the header of the filterbank file
header = read_header('examples/pspm32.fil')
# Calculate the center frequency with the data in the header
center_freq = header[b'fch1'] + float(header[b'nchans']) * header[b'foff'] / 2.0
print(center_freq)
# Get the powerlevels and the frequencies
PXX, freqs, _ = psd(samples, nfft=1024, sample_rate=80, sides='twosided')
# Calculate the powerlevel dB's
power_levels = 10 * np.log10(PXX/(80))
# Add the center frequency to the frequencies so they match the actual frequencies
freqs = freqs + center_freq
# Plot the PSD
plt.grid(True)
plt.xlabel('Frequency (MHz)')
plt.ylabel('Intensity (dB)')
plt.plot(freqs, power_levels)
plt.show()
```

4.2 Waterfall

The plot.waterfall.Waterfall class can be used to generate waterfall plots.

4.2.1 Construction

Parameter	Description
filter_bank center_freq sample_freq fig mode	A filterbank object. The center frequency of the signal in the filterbank object The sample frequency of the signal in the filterbank object An imaging object, like pyplot.figure() String {discrete, stream}. The mode to operate on. Use discrete for discrete datasets, and stream for stream data. Defaults to stream.

4.2.2 Methods

Method	Description
init_plot(self)	Initialize the plot
update_plot_labes(self	Generate the plot labels
$get_next(self)$	Returns the next row of data in the filterbank object
$get_image(self)$	Returns the image data of the full dataset, if using a
	discrete dataset.
update(self, i)	Updates the image with the next row of data, when
	using a continuous datastream.
$animated_plotter(self)$	Returns the figure and update function for matplotlib
	animation
$get_center_freq(self)$	Returns the center frequency stored in the filterbank
	header

4.2.3 Example Usage

4.2.3.1 With discrete data

```
import matplotlib.animation as animation
from filterbank.header import read_header
from filterbank.filterbank import Filterbank
from plot import waterfall
import pylab as pyl
from plot.plot import next_power_of_2

fb = Filterbank(filename='./pspm32.fil', read_all=True)

wf = waterfall.Waterfall(filter_bank=fb, fig=pyl.figure(), mode="discrete")
img = wf.get_image()
```

```
pyl.show(img)
```

4.2.3.2 With stream data

```
import matplotlib.animation as animation
from filterbank.header import read_header
from filterbank.filterbank import Filterbank
from plot import waterfall
import pylab as pyl
from plot.plot import next_power_of_2

fb = Filterbank(filename='./pspm32.fil')

wf = waterfall.Waterfall(fb=fb, fig=pyl.figure(), mode="stream")

fig, update, frames, repeat = wf.animated_plotter()

ani = animation.FuncAnimation(fig, update, frames=frames,repeat=repeat)
pyl.show()
```

5. Timeseries module

Learn how to use the timeseries object. The timeseries object is used for all timeseries related operations both filterbank and non-filterbank (general array) types.

5.1 Initialize the timeseries object

The Timeseries module can be used with both the filterbank file and a general numpy array.

5.1.1 Initialize using a filterbank file

- Create an filterbank object using the standard filterbank module.
- Read the filterbank file in memory (do not use streams)
- Initialize the timeseries object using the filterbank object.

```
import filterbank.filterbank as Filterbank
import timeseries.timeseries as Timeseries
```

Read the filterbank file from a file.

```
filterbank_obj = Filterbank('./pspm32.fil')

# Read the filterbank as a whole instead of as a stream.
filterbank_obj = filterbank_obj.read_filterbank()

# Initialize the timeseries object.
ts = Timeseries().from_filterbank(filterbank_obj)

5.1.2 Initialize using a numpy array
import numpy as np
import timeseries.timeseries as Timeseries
input_array = np.array([1, 2, 3, 4, 5, 7, 9, 10, 11])
ts = Timeseries(input_array)
```

5.2 Retrieve the timeseries array from timeseries object

5.2.1 Retrieve timeseries object

```
# Assumed that your timeseries object has been initialized.
timeseries_array = timeseries.get()
```

5.3 Downsample the timeseries

The first implemented feature for the timeseries object is the downsample/decimate function. This enables you to downsample your timeseries by **q** scale. This will make your input array smaller and basically 'cuts' the other parts off.

5.3.1 Downsample/Decimate

Called downsample in the current release because no anti-alliasing is used, might be renamed to decimate once more advanced operations (such as antialiasing) are used.

```
# Downsampled array shall be 3 times smaller than the current timeseries (as initialized)
scale = 3
# Returns an array with the downsampled timeseries, can also be retreived lated user timeser
downsampled_array = timeseries.downsample(scale)
```

6. Clipping

6.1 Clipping

The clipping clipping function can be used to remove noise from the data.

It combines all the individual methods of the clipping module to do this. The individual methods are described below in order.

clipping(<FREQUENCY_CHANNELS>, <TIME_SAMPLES>)

6.2 Filter samples

The clipping.filter_samples function can be used to remove entire time samples, that have noise in them.

The noise for a time sample is calculated by summing all the individual frequencies for one sample, and then comparing it with the average (mean) of all samples. If the intensity of a sample is lower than average intensity per sample * factor, the sample will be added back to the array. Otherwise, it's removed.

It is recommended to give only the first 2000 samples to this method, because adding more will only hurt the performance.

filter_samples(<TIME_SAMPLES>)

6.3 Filter channels

The clipping.filter_channels function can be used to remove entire channels/frequencies with noise from the data.

The noise for a frequency channel is identified by calculating the mean and standard deviation for each column in the data. After that, samples are removed if the mean or standard deviation of the intensity is higher than the average intensity per channel * factor. The channels with noise are removed from both the list with channels, as well as the entire dataset.

filter_channels(<FREQUENCY_CHANNELS>, <TIME_SAMPLES>)

6.4 Filter individual channels

The clipping.filter_indv_channels function can be used to replace all the remaining samples with noise.

The noise for each individual sample is identified by calculating the mean intensity per channel. After that, samples are replaced with the median of a channel if their intensity is higher than the average intensity per channel * factor.

filter_indv_channels(<TIME_SAMPLES>)

7. Dedispersion

7.1 Introduction

- Pulsars produce a narrow beam of electromagnetic radiation which rotates like a lighthouse beam, so a pulse is seen as it sweeps over a radiotelescope. The signal is spread over a wide frequency range.
- If space was an empty vacuum, all the signals would travel at the same speed, but due to free electrons different frequencies travel at slightly different speeds (dispersion).

7.2 Dedisperse

This method performs dedispersion on the filterbank data.

7.2.1 Parameters

Parameters	Description
Samples DM	Array or sequence containing the data to be plotted. Dispersion measure (cm-3 pc)

7.2.2 Returns

Variable	Description
Samples	Dedispersed samples

7.3 find_dm

This method finds the dispersion measure.

7.3.1 Parameters

Parameters	Description
Samples	1-D array or sequence. Array or sequence containing the data to be plotted.

7.3.2 Returns

Variable	Description
$\overline{\mathrm{dm}}$	Dispersion measure

$7.4 \ \mathrm{find_line}$

This method finds a line starting from the sample index given in the parameters. This will stop if there isn't a intensity within the max_delay higher than the average_intensity.

7.4.1 Parameters

Parameters	Description
Samples	1-D array or sequence. Array or sequence containing the data to be plotted.
start_sample_index max_delay	The start sample index The max delay of the recieved signal
average_intensity	The average intensity

7.4.2 Returns

Variable	Description
start_sample_index previous_sample_index	The first frequency The higher frequency

7.5 find_estimation_intensity

This method finds the average intensity for top x intensities. The average_intensity is considered a requirement for intensities to be

considered a pulsar.

7.5.1 Parameters

Parameters	Description
Samples	Array or sequence containing the data to be plotted.

7.5.2 Returns

Variable	Description
average_intensity	The avarage intesity

8. Pipeline

8.1 Introduction

The Pipeline module is used to execute the different modules in a specific order. There are currently three different options for running the pipeline.

These options include: * read multiple rows, read_n_rows * read single rows, read_rows * read all rows, read_static

The constructor of the pipeline module will recognize which method is fit for running which method, by looking at the given arguments to the constructor.

Parameter	Description
filename	The path to the filterbank file.
as_stream	This parameter decides whether
	the filterbank should be read as
	stream.
DM	The dispersion measure (DM) is
	used for performing
	dedispersion.
scale	The scale is used for performing
	downsampling the time series.
n	The n is the rowsize of chunks
	for reading the filterbank as
	stream.

Parameter	Description
size	The size parameter is used for deciding the size of the filterbank.

After deciding which method to run for running the filterbank in a pipeline, it will measure the time it takes to run each method using measure_method. After running all the different methods, the constructor will append the results (a dictionary) to a txt file.

8.2 Read rows

The read_rows method reads the Filterbank data row per row. Because it only reads the filterbank per row, it is unable to execute most methods. The alternative for this method is the read_n_rows method, which is able to run all methods.

pipeline.Pipeline(<filterbank_file>, as_stream=True)

8.3 Read n rows

The read_n_rows method first splits all the filterbank data into chunks of n samples. After splitting the filterbank data in chunks, it will run the different modules of the pipeline for each chunk. The remaining data, that which does not fit into the sample size, is currently ignored.

The **n** or sample size should be a power of 2 multiplied with the given scale for the downsampling.

pipeline.Pipeline(<filterbank_file>, n=<size> , as_stream=True)

8.4 Read static

The read_static method reads the entire filterbank at once, and applies each method to the entire dataset. If the filterbank file is too large for running it in-memory, the alternative is using read_n_rows.

pipeline.Pipeline(<filterbank_file>)

8.5 Measure methods

The measure_methods is ran for each of the above methods, and calculates the time it takes to run each of the different methods. For each method it will create

a key using the name of the method, and save the time it took to run the method as a value. At the end, it will returns a dictionary with all the keys and values.

7.6 Overview of pipeline

Apart from the different modules described in the previous paragraphs, additional modules are required for this library to make detecting pulsar signals possible. However, these additional modules have not been developed yet, and are required to be developed in the future. In this paragraph the additional modules are listed and described.

Modules that are missing in the pipeline are highlighted using a *.

```
    Read Filterbank as stream
    Reduce RFI using clipping
    Dedisperse radio signal
    Transform dedispersed signal to TimeSeries
    Run fast Fourier transformation on TimeSeries
    * Identify and save birdies in file
    * Perform Harmonic Summing
    * Search and identify single and periodic signals
    * Phase-fold remaining signals
    * Do Transient searches
```

9. Generating mock data

9.1 Creating a filterbank file

To generate a filterbank file you may use the following example code:

import filterbank.header as header

```
header = {
    b'source_name': b'P: 80.0000 ms, DM: 200.000',
    b'machine_id': 10,
    b'telescope_id': 4,
    b'data_type': 1,
    b'fch1': 400,
    b'foff': -0.062,
    b'nchans': 128,
    b'tstart': 6000.0,
    b'tsamp': 8e-05,
    b'nifs': 1,
    b'nbits': 8
```

}

header.generate_file("file_path/file_name", header)

The header data is passed in a header dict. For a more detailed explanation on the file header, please consult chapter 2.2.

9.2 Generate signal

The generate_signal method generates a mock signal based on the following parameters:

Variable	Description
noise_level	the max amplitude of the generated noise
period	period of the signal
t_obs	observation time in s
n_pts	intervals between samples

9.3 Generate header

The generate_header method generates a header string based on the header dict provided in the example in chapter 8.1. The dict provides keys encoded in bytes and the method converts each keyword to a string using the keyword_to_string method (see below).

9.4 Keyword to string

The keyword_to_string method converts a keyword from the header dict to a serialized string.

9.5 Write data

Once all the required data is generated to a filterbank file. The data is written to the file as bytes.