
FFT-tools Documentation

Release 0.5

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January 28, 2015

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Module documentation:

`FFT_tools` Wrap Numpy's `fft` module to reduce clutter.

FFT_tools

Wrap Numpy's `fft` module to reduce clutter.

Provides a unitary discrete FFT and a windowed version based on `numpy.fft.rfft()`.

Main entry functions:

- `unitary_rfft()`
- `power_spectrum()`
- `unitary_power_spectrum()`
- `avg_power_spectrum()`
- `unitary_avg_power_spectrum()`

Functions

<code>avg_power_spectrum(data[, freq, chunk_size, ...])</code>	Compute the average power spectrum of <i>data</i> .
<code>ceil_pow_of_two(num)</code>	Round <i>num</i> up to the closest exact a power of two.
<code>floor_pow_of_two(num)</code>	Round <i>num</i> down to the closest exact a power of two.
<code>power_spectrum(data[, freq])</code>	Compute the power spectrum of the time series <i>data</i> .
<code>round_pow_of_two(num)</code>	Round <i>num</i> to the closest exact a power of two on a log scale.
<code>unitary_avg_power_spectrum(data[, freq, ...])</code>	Compute the unitary average power spectrum of <i>data</i> .
<code>unitary_power_spectrum(data[, freq])</code>	Compute the unitary power spectrum of the time series <i>data</i> .
<code>unitary_rfft(data[, freq])</code>	Compute the unitary Fourier transform of real data.
<code>window_hann(length)</code>	Returns a Hann window array with <i>length</i> entries

Classes

<code>TestRFFT([methodName])</code>	Ensure Numpy's FFT algorithm acts as expected.
<code>TestUnitaryAvgPowerSpectrum([methodName])</code>	
<code>TestUnitaryPowerSpectrum([methodName])</code>	
<code>TestUnitaryRFFT([methodName])</code>	Verify <code>unitary_rfft()</code> .

```
class FFT_tools.TestRFFT (methodName='runTest')
    Ensure Numpy's FFT algorithm acts as expected.
```

The expected return values are ¹:

$$X_k = \sum_{m=0}^{n-1} x_m \exp^{-2\pi i m k / n}$$

test_rfft ()

Test NumPy's builtin `numpy.fft.rfft` ()

class `FFT_tools.TestUnitaryRFFT` (*methodName='runTest'*)

Verify `unitary_rfft` ().

gaussian (*a, t*)

Gaussian function.

$$\text{gaussian}(a, t) = \exp^{-at^2}$$

rect (*t*)

Rectangle function.

$$\text{rect}(t) = \begin{cases} 1 & \text{if } |t| < 0.5, \\ 0 & \text{if } |t| \geq 0.5. \end{cases}$$

run_gaussian (*a=1.0, time_shift=5.0, samp_freq=25.6, samples=256*)

Test `unitary_rttf` () on known function `gaussian` ().

Analytic result:

$$\text{rfft}(\text{gaussian}(a, t)) = \sqrt{\pi/a} \cdot \text{gaussian}(1/a, \pi f)$$

run_parsevals (*xs, freq, freqs, Xs*)

Check the discretized integral form of Parseval's theorem

Which is:

$$\sum_{m=0}^{n-1} |x_m|^2 dt = \sum_{k=0}^{n-1} |X_k|^2 df$$

run_rect (*a=1.0, time_shift=5.0, samp_freq=25.6, samples=256*)

Test `unitary_rfft` () on known function `rect` ().

Analytic result:

$$\text{rfft}(\text{rect}(at)) = 1/|a| \cdot \text{sinc}(f/a)$$

test_gaussian ()

Test unitary FFTs on variously shaped gaussian functions.

test_parsevals ()

Test unitary rfft on Parseval's theorem

test_rect ()

Test unitary FFTs on variously shaped rectangular functions.

`FFT_tools.avg_power_spectrum` (*data, freq=1.0, chunk_size=2048, overlap=True, window=<function window_hann at 0x7f9cf37df410>*)

Compute the average power spectrum of *data*.

¹ See the *Background information* section of `numpy.fft`.

data [iterable] Real (not complex) data taken with a sampling frequency *freq*.

freq [real] Sampling frequency.

chunk_size [int] Number of samples per chunk. Use a power of two.

overlap: {True,False} If *True*, each chunk overlaps the previous chunk by half its length. Otherwise, the chunks are end-to-end, and not overlapping.

window: iterable Weights used to “smooth” the chunks, there is a whole science behind windowing, but if you’re not trying to squeeze every drop of information out of your data, you’ll be OK with the default Hann window.

freq_axis,power [numpy.ndarray] Arrays ready for plotting.

The average power spectrum is computed by breaking *data* into chunks of length *chunk_size*. These chunks are transformed individually into frequency space and then averaged together.

See Numerical Recipes 2 section 13.4 for a good introduction to the theory.

If the number of samples in *data* is not a multiple of *chunk_size*, we ignore the extra points.

`FFT_tools.ceil_pow_of_two(num)`

Round *num* up to the closest exact a power of two.

```
>>> ceil_pow_of_two(3)
4
>>> ceil_pow_of_two(11)
16
>>> ceil_pow_of_two(15)
16
```

`FFT_tools.floor_pow_of_two(num)`

Round *num* down to the closest exact a power of two.

```
>>> floor_pow_of_two(3)
2
>>> floor_pow_of_two(11)
8
>>> floor_pow_of_two(15)
8
```

`FFT_tools.power_spectrum(data, freq=1.0)`

Compute the power spectrum of the time series *data*.

data [iterable] Real (not complex) data taken with a sampling frequency *freq*.

freq [real] Sampling frequency.

freq_axis,power [numpy.ndarray] Arrays ready for plotting.

If the number of samples in *data* is not an integer power of two, the FFT ignores some of the later points.

unitary_power_spectrum,avg_power_spectrum

`FFT_tools.round_pow_of_two(num)`

Round *num* to the closest exact a power of two on a log scale.

```
>>> round_pow_of_two(2.9) # Note rounding on *log scale*
4
>>> round_pow_of_two(11)
8
```

```
>>> round_pow_of_two(15)
16
```

`FFT_tools.unitary_avg_power_spectrum` (*data*, *freq*=1.0, *chunk_size*=2048, *overlap*=True, *window*=<function `window_hann` at 0x7f9cf37df410>)

Compute the unitary average power spectrum of *data*.

avg_power_spectrum, *unitary_power_spectrum*

`FFT_tools.unitary_power_spectrum` (*data*, *freq*=1.0)

Compute the unitary power spectrum of the time series *data*.

power_spectrum, *unitary_avg_power_spectrum*

`FFT_tools.unitary_rfft` (*data*, *freq*=1.0)

Compute the unitary Fourier transform of real data.

Unitary = preserves power [Parseval's theorem].

data [iterable] Real (not complex) data taken with a sampling frequency *freq*.

freq [real] Sampling frequency.

freq_axis,trans [numpy.ndarray] Arrays ready for plotting.

If the units on your data are Volts, and your sampling frequency is in Hz, then *freq_axis* will be in Hz, and *trans* will be in Volts.

`FFT_tools.window_hann` (*length*)

Returns a Hann window array with *length* entries

The Hann window with length *L* is defined as

$$w_i = \frac{1}{2}(1 - \cos(2\pi i/L))$$

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