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# **pyLCSIM Documentation**

***Release 0.1***

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## DOCUMENTATION FOR PYLCSIM

### 1.1 Introduction

pyLCSIM is a python package to simulate X-ray lightcurves from coherent signals and power spectrum models.

Coherent signals can be specified as a sum of one or more sinusoids, each with its frequency, pulsed fraction and phase shift; or as a series of harmonics of a fundamental frequency (each with its pulsed fraction and phase shift).

Power spectra can be simulated from a model of the power spectrum density (PSD), using as a template one or more of the built-in library functions. The user can also define his/her custom models. Models are additive.

A PDF version of these notes is available [here](#).

**Warning:** the current release (0.x.y) is HIGHLY EXPERIMENTAL! Use at your own risk...

### 1.2 Prerequisites

pyLCSIM requires [Numpy](#) (at least v1.8) and [Astropy](#) (at least v0.3).

[Matplotlib](#) is highly recommended if you want to plot your simulations.

### 1.3 Installation

The package can be downloaded [here](#).

The installation follows the usual steps:

```
$ tar xzvf pyLCSIM-0.2.0.tar.gz
```

```
$ cd pyLCSIM-0.2.0
```

```
$ python setup.py install
```

The last step may require administrator privileges.

### 1.4 Mailing list

A dedicated mailing list is available, to host announcements of new releases, to report issues and to get support. If interested, subscribe [here](#).

## 1.5 Example 1

Let's begin with a PSD model simulation.

We import the usual packages:

```
import matplotlib.pyplot as plt
import numpy as np
import pyLCSIM
```

We assume that our source has a rate of 30000 counts/s. Moreover, we have a 5000 counts/s background rate, and we have made a 50 s exposure. Our observation has a time resolution of 10 ms. We want to simulate a QPO at a frequency of 10 Hz, superimposed to a continuum modelled as a smoothly-varying broken power law (spectral indices 1 and 2, with a steepness change at 1 Hz). The required fractional RMS variation of the signal is 1%:

```
rate_src    = 30000.0
rate_bkg    = 5000.0
t_exp       = 50.0
dt          = 0.01
frms        = 0.01
```

The total bins of the lightcurve are therefore:

```
nbins = long(t_exp/dt)
```

The simulation follows as:

```
# Instantiate a simulation object
sim = pyLCSIM.Simulation()

# Add two PSD models: a smooth broken power law and a Lorentzian representing a QPO.
# See the documentation for details.
sim.addModel('smoothbknp', [1., 1., 2., 1.])
sim.addModel('lorentzian', [10., 1., 100., 10])

# Run the simulation
sim.run(dt, nbins, rate_src, rms=frms)

# Add Poisson noise to the light curve
sim.poissonRandomize(dt, rate_bkg)

# Get lightcurve and power spectrum as 1-D arrays
time, rate = sim.getLightCurve()
f, psd = sim.getPowerSpectrum()
```

Done! We can save the results as FITS files:

```
# Save FITS files with lightcurve and spectrum
pyLCSIM.saveFITS LC("myLC.fits", time, rate)
pyLCSIM.saveFITSPSD("myPSD.fits", f, psd)
```

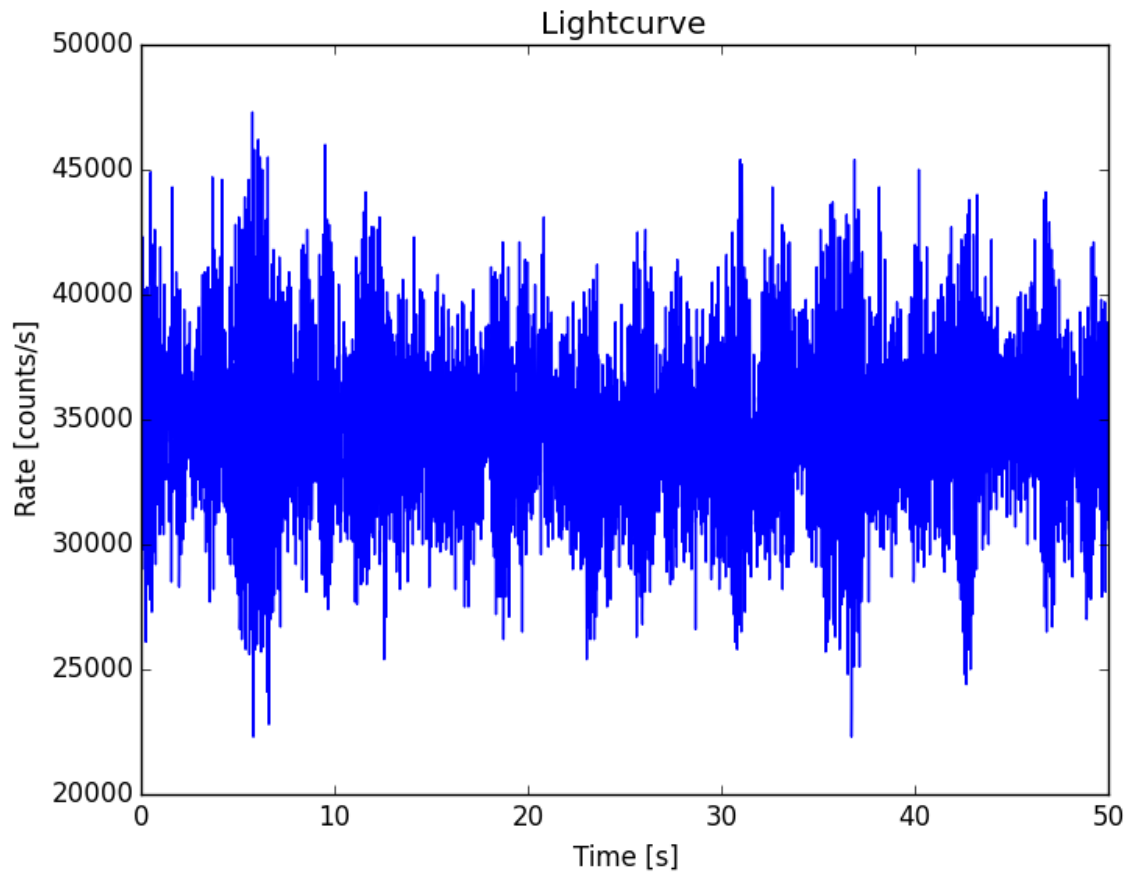
and view the results:

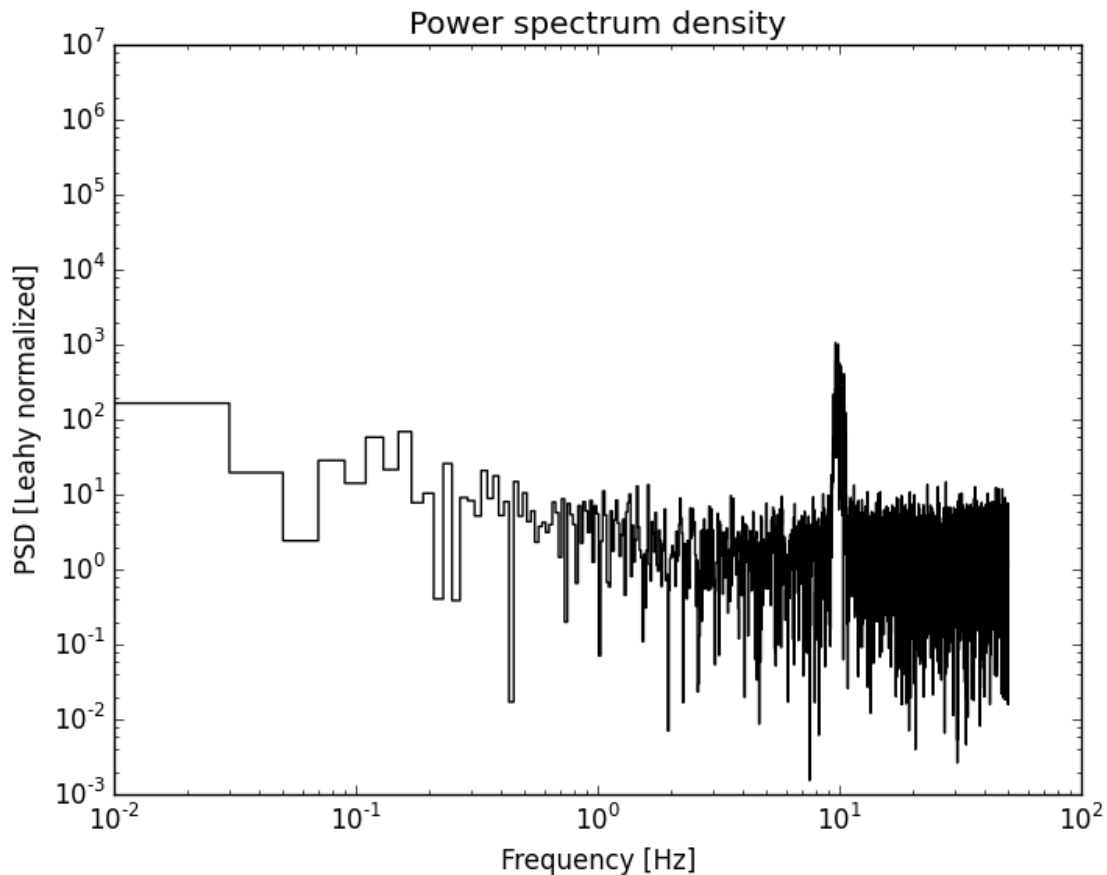
```
# Plot the lightcurve and power spectrum
fig0 = plt.figure()
plt.plot(time, rate)
plt.xlabel("Time [s]")
plt.ylabel("Rate [counts/s]")
plt.title("Lightcurve")
```



```
fig1 = plt.figure()
plt.loglog(f, psd, drawstyle='steps-mid', color='black')
plt.xlabel("Frequency [Hz]")
plt.ylabel("PSD [Leahy normalized]")
plt.title("Power spectrum density")

plt.show()
```



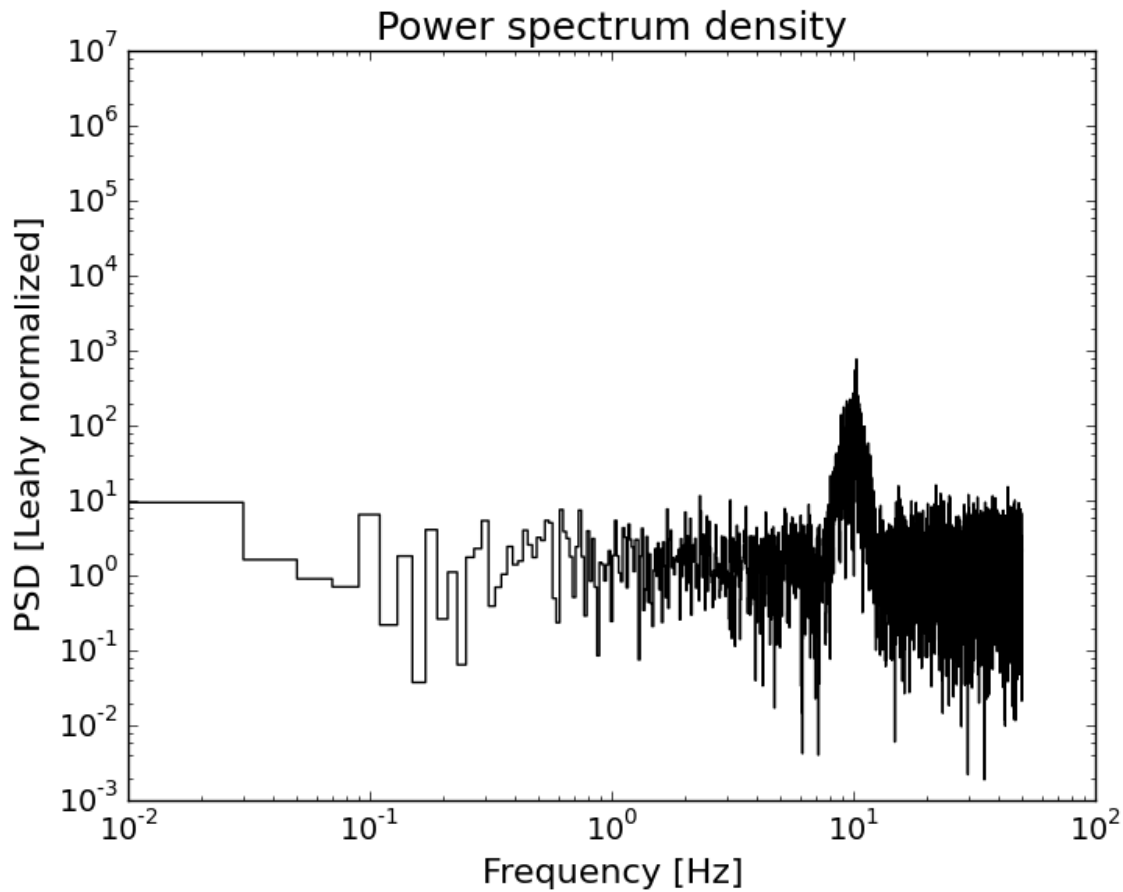


## 1.6 Example 2

The user can also define his/her own PSD models, simply as python functions. The only caveat is that the function should be positive-valued, and it is suggested to avoid too small output values.

The following example shows a simulation using an user-defined function (a Gaussian centered at 10 Hz, in this case):

```
def myFunc(f, p):  
    """  
    Example of user-defined function: a Gaussian.  
    User-defined PSD models should be positive-valued!  
    Moreover, in this example the output is clipped at 1e-32 to avoid too small values.  
    """  
    f = p[0]*np.exp(-(f-p[1])**2/p[2]**2)  
    return np.clip(f, 1e-32, np.max(f))  
  
sim = pyLCSIM.Simulation()  
sim.addModel('smoothbknp', [1., 1, 2, 1])  
sim.addModel(myFunc, [1000., 10, 1.])  
  
# Run the simulation  
sim.run(dt, nbins, rate_src, rms=frms)
```



## 1.7 Example 3

The following example shows the simulation of a coherent signal using a sum of sinusoids:

```
import matplotlib.pyplot as plt
import numpy as np
import pyLCSIM
```

```
rate_src    = 30000.0
rate_bkg    = 5000.0
t_exp       = 1.0
dt          = 0.0001
nbins = long(t_exp/dt)
```

Note the different exposure time (1 s) and time resolution (100 us):

```
# Instantiate a simulation object, this time as coherent
sim = pyLCSIM.Simulation(kind='coherent')
```

```
# Run the simulation, using:
```

```
# four sinusoidal frequencies: 340, 550, 883, 1032 Hz;
```

```
# with pulsed fractions 10%, 5%, 7% and 15% respectively;
```

```
# the third frequency has a 35 degree phase shift with respect to the others
```

```
sim.run(dt, nbins, rate_src, freq=[340, 550, 883, 1032], amp=[0.1, 0.05, 0.07, 0.15], phi=[0., 0, 35
```

```
# Add Poisson noise to the light curve
sim.poissonRandomize(dt, rate_bkg)

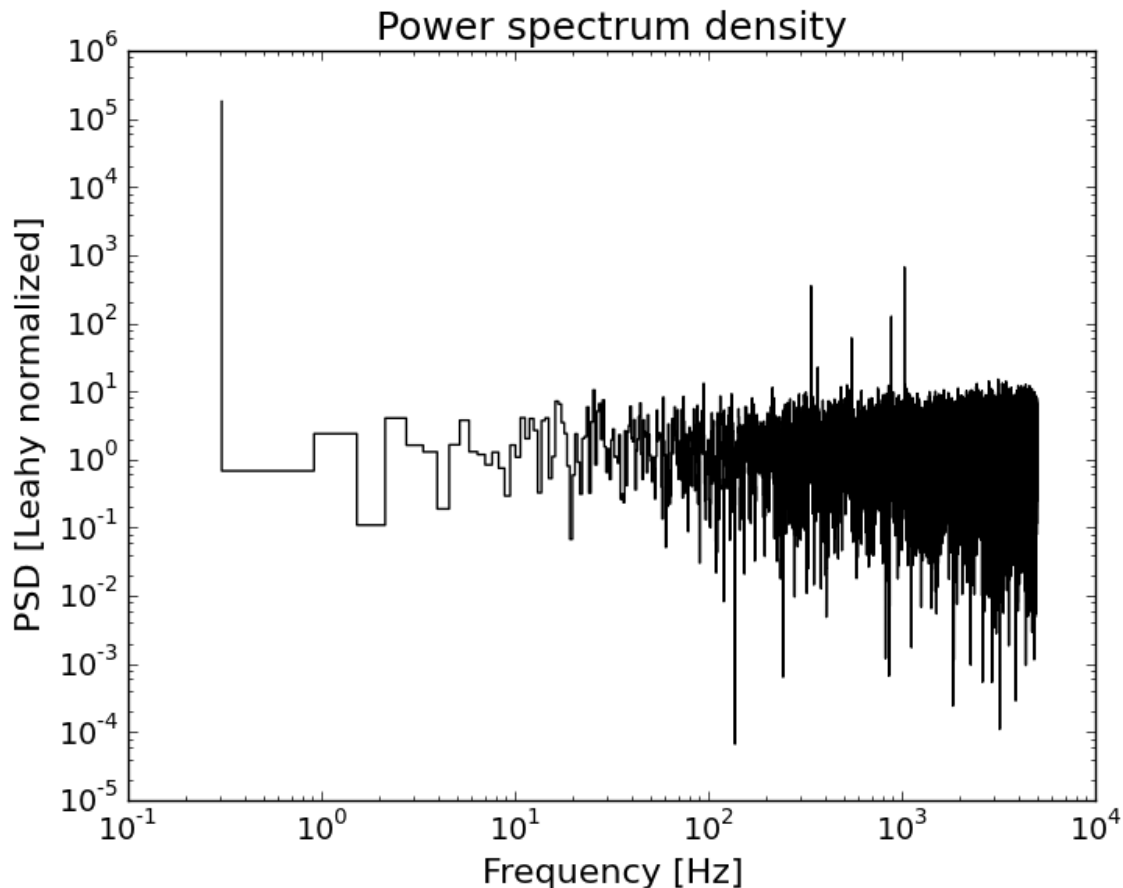
# Get lightcurve and power spectrum as 1-D arrays
time, rate = sim.getLightCurve()
f, psd = sim.getPowerSpectrum()

# Plot the lightcurve and power spectrum
fig0 = plt.figure()
plt.plot(time, rate)
plt.xlabel("Time [s]")
plt.ylabel("Rate [counts/s]")
plt.title("Lightcurve")

fig1 = plt.figure()
plt.loglog(f, psd, drawstyle='steps-mid', color='black')
plt.xlabel("Frequency [Hz]")
plt.ylabel("PSD [Leahy normalized]")
plt.title("Power spectrum density")

# Save FITS files with lightcurve and spectrum
pyLCSIM.saveFITS LC("myLC.fits", time, rate)
pyLCSIM.saveFITSPSD("myPSD.fits", f, psd)

plt.show()
```



## 1.8 Example 4

Finally, an example with a fundamental frequency and two harmonics:

```
import matplotlib.pyplot as plt
import numpy as np
import pyLCSIM

rate_src      = 300000.0
rate_bkg      = 3000.0
t_exp         = 1.0
dt            = 0.0001
phase_shift   = 0.
nbins         = long(t_exp/dt)

print nbins

# Instantiate a simulation object, this time as coherent
sim = pyLCSIM.Simulation(kind='coherent')

# Run the simulation:
# Fundamental at 500 Hz, 3 harmonics (500, 1000, 1500 Hz)
# with pulsed fractions 10%, 5% and 15% respectively
sim.run(dt, nbins, rate_src, freq=500, nha=3, amp=[0.1, 0.05, 0.15])

# Add Poisson noise to the light curve
sim.poissonRandomize(dt, rate_bkg)

# Get lightcurve and power spectrum as 1-D arrays
time, rate = sim.getLightCurve()
f, psd = sim.getPowerSpectrum()

# Plot the lightcurve and power spectrum
fig0 = plt.figure()
plt.plot(time, rate)

fig1 = plt.figure()
plt.loglog(f, psd, drawstyle='steps-mid', color='black')

# Save FITS files with lightcurve and spectrum
pyLCSIM.saveFITSLC("myLC.fits", time, rate)
pyLCSIM.saveFITSPSD("myPSD.fits", f, psd)

plt.show()
```

## 1.9 Main module

A python package to simulate X-ray lightcurves from coherent signals and power spectrum models.

**class** pyLCSIM.**Simulation** (*kind='psd'*)  
Main simulation class.

**History:** v1.2: Added reset method. Riccardo Campana, 2014.

v1.1: Bugfix. Riccardo Campana, 2014.

v1: Initial python implementation. Riccardo Campana, 2014.

**addModel** (*modelName, modelParams*)

Append simulation model to model dictionary

**getLightCurve** ()

Get lightcurve as time and rate arrays

**getPowerSpectrum** ()

Get power spectrum as frequency and power arrays

**info** ()

Prints simulation informations

**poissonRandomize** (*dt, bkg*)

Add Poissonian noise to lightcurve (and background, if present)

**reset** ()

Reset the simulation, emptying models array and other members

**run** (*dt, nbins\_old, mean, rms=None, freq=None, nha=None, amp=None, phi=None, verbose=False*)

Run the simulation

`pyLCSIM.lcsinusoid` (*dt=1.0, nbins=65536, mean=0.0, freq=None, nha=1, amp=None, phi=None*)

Generate coherent signals as a sequence of sinusoids (if `len(freq) > 1`) or of the fundamental frequency plus `nha-1` harmonics (if `len(freq) == 1`), each with normalized pulsed fraction `amp[i]`.

**Kwargs:** `dt`: time resolution of the lightcurve to be simulated (default: 1.0).

`nbins`: Number of bins of the simulated lightcurve (default: 65536).

`freq`: if float: frequency of the fundamental harmonic, if array: frequencies of sinusoids.

`nha`: number of harmonics (>1)

`amp`: array with `nha/nfreq` elements; pulsed fraction for each frequency

`phi`: array with `nha/nfreq` elements; phases (in degrees!) for each frequency

**Returns:** `time`: time array

`rate`: array of count rates

**History:** v1: Initial python implementation, from the IDL procedure `lcharmonics.pro` v0.0.3 by I. Donnarumma & R. Campana. Riccardo Campana, 2014.

`pyLCSIM.lcpsd` (*dt=1.0, nbins=65536, mean=0.0, rms=1.0, seed=None, models=None, phase\_shift=None, verbose=False*)

Simulate a light-curve with a general power spectrum shape. For the underlying algorithm see: J. Timmer & M. Koenig, "On generating power law noise", A&A, 300, 707-710 (1995).

**Kwargs:** `dt`: time resolution of the lightcurve to be simulated

**nbins:** Number of bins of the simulated lightcurve (default:65536). Should be power of two for optimum performance (FFT...)

`mean`: Mean count rate of the simulated lightcurve (default: 0.).

`rms`: Total fractional RMS of the simulated lightcurve.

`seed`: Seed for the random number generator.

**models:** List of tuples, each containing:

1. Name of a function returning the desired PSD shape,
2. Parameters of the model (argument to `model`). Total model is the sum of these tuples.

phase\_shift: Constant phase shift (in degrees) to the FFT.

verbose: If True, prints some debugging information.

**Returns:** time: time array.

rate: array of count rates.

**History:** v2: Added the possibility to employ user-defined PSD models. Riccardo Campana, 2014.

v1: Initial python implementation, based on AITLIB IDL procedure timmerlc.pro. Riccardo Campana, 2014.

`pyLCSIM.poisson_randomization` (*rate*, *dt=1.0*, *bkg=0.0*, *seed=None*)

Poisson randomization of the rate array.

**Args:** rate: input array of count rates (in cts/s).

**Kwargs:** dt: time resolution of the lightcurve to be simulated.

bkg: Mean count rate of the simulated lightcurve (default: 0.0).

seed: Seed for the random number generator.

**Returns:** newrate: Array of Poisson randomized count rates of length n.

**History:** v1: Initial python implementation. Riccardo Campana, 2014.

`pyLCSIM.psd` (*time*, *rate*, *norm='leahy'*)

Returns power spectral density from a (real) time series with Leahy normalization.

**Args:** time: array of times (evenly binned).

rate: array of rate in counts/s.

**Kwargs:** norm: Normalization (only Leahy for now).

**Returns:** f: array of frequencies.

p: power spectrum.

**History:** v1: Initial python implementation. Riccardo Campana, 2014.

`pyLCSIM.rebin` (*x*, *y*, *nbins*, *mode='rate'*)

Linearly rebins the (x, y) 1-D arrays using #nbins new bins, taking for each new element the mean of the corresponding x elements, and the sum (if mode=='counts') or the mean (if mode=='rate') of the corresponding y elements.

**Args:** x: x array (same length as y)

y: y array (same length as x)

nbins: Number of new bins. Should be a factor of len(x) == len(y).

**Kwargs:** mode: 'counts' or 'rate'; returns sum or mean of y-array elements

**Returns:** xreb: rebinned x array

yreb: rebinned y array

**History:** v1: Initial python implementation. Riccardo Campana, 2014.

`pyLCSIM.logrebin` (*x*, *y*, *nbins*, *mode='rate'*)

Logarithmically rebins the (x, y) 1-D arrays using #nbins new bins, taking for each new element the logarithmic mean of the corresponding x elements, and the sum (if mode=='counts') or the mean (if mode=='rate') of the corresponding y elements.

**Args:** x: x array (same length as y)

y: y array (same length as x)

nbins: Number of new bins.

**Kwargs:** mode: 'counts' or 'rate'; returns sum or mean of y-array elements

**Returns:** xreb: rebinned x array

yreb: rebinned y array

**History:** v1: Initial python implementation. Riccardo Campana, 2014.

`pyLCSIM.saveFITS LC (outfile name, time, rate, clobber=True)`

Produce an output FITS file containing a lightcurve.

**Args:** outfile name: Name of the output FITS file

time: Array of times

rate: Array of count rates

**Kwargs:** clobber: if True, overwrites existing files with same name

**Returns:** none

**History:** v2: OGIP-compliance (OGIP 93-003). Riccardo Campana, 2014.

v1: Initial python implementation. Riccardo Campana, 2014.

`pyLCSIM.saveFITS PSD (outfile name, freq, psd, clobber=True)`

Produce an output FITS file containing a power spectrum.

**Args:** outfile name: Name of the output FITS file

freq: Array of frequencies

psd: Array of power spectrum

**Kwargs:** clobber: if True, overwrites existing files with same name

**Returns:** none

**History:** v1: Initial python implementation. Riccardo Campana, 2014.

## 1.10 Submodule: psd\_models

Contains the analytic models for the power spectrum density.

`pyLCSIM.psd_models.lorentzian(x, p)`

Generalized Lorentzian function.

(**WARNING:** for  $n \neq 2$  the function is no more normalized!)

**Args:** x: (non-zero) frequencies.

p[0]:  $x_0$  = peak central frequency.

p[1]:  $\gamma$  = FWHM of the peak.

p[2]: value of the peak at  $x = x_0$ .

p[3]:  $n$  = power coefficient.

The quality factor is given by  $Q = x_0/\gamma$ .



**Returns:** f: psd model.

**History:** v1: Initial python implementation. Riccardo Campana, 2014.

`pyLCSIM.psd_models.smoothbknp`(*x*, *p*)

Smooth broken power law.

**Args:** x: (non-zero) frequencies.

p[0]: Normalization.

p[1]: power law index for  $f \rightarrow 0$ .

p[2]: power law index for  $f \rightarrow \infty$ .

p[3]: break frequency.

**Returns:** f: psd model.

**History:** v1: initial python implementation. Riccardo Campana, 2014.

## 1.11 Changelog

v0.2.0: Added the possibility to employ user-defined PSD models.

v0.1.2: Added reset method to Simulation. Lightcurve FITS output is now OGIP-compliant.

v0.1.1: Bugfix. Modified Simulation class.

v0.1.0: Initial release.



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