# hvsrpy

Release 0.1.0

Joseph P. Vantassel

# **CONTENTS:**

1	Summary	1
2	License Information	2
3	Sensor3c Class	3
4	Hysr Class	5

## **CHAPTER**

# **ONE**

# **SUMMARY**

hvsrpy is a Python module for horizontal-to-vertical spectral ratio processing. hvsrpy was developed by Joseph P. Vantassel with contributions from Dana M. Brannon under the supervision of Professor Brady R. Cox at the University of Texas at Austin. The fully-automated frequency-domain rejection algorithm implemented in *hvsrpy* was developed by Tianjian Cheng under the supervision of Professor Brady R. Cox at the University of Texas at Austin and detailed in Cox et al. (in review).

The module includes two main class definitons *Sensor3c* and *Hvsr*. These classes include various methods for creating and manipulating 3-component sensor and horizontal-to-vertical spectral ratio objects, respectively.

**CHAPTER** 

**TWO** 

# LICENSE INFORMATION

Copyright (C) 2019 Joseph P. Vantassel (jvantassel@utexas.edu)

This program is free software: you can redistribute it and/or modify it under the terms of the GNU General Public License as published by the Free Software Foundation, either version 3 of the License, or (at your option) any later version.

This program is distributed in the hope that it will be useful, but WITHOUT ANY WARRANTY; without even the implied warranty of MERCHANTABILITY or FITNESS FOR A PARTICULAR PURPOSE. See the GNU General Public License for more details.

You should have received a copy of the GNU General Public License along with this program. If not, see <a href="https://www.gnu.org/licenses/">https://www.gnu.org/licenses/</a>>.

# THREE

# SENSOR3C CLASS

#### class Sensor3c(ns, ew, vt)

Class for creating and manipulating 3-component sensor objects.

#### **Attributes:**

ns, ew, vt [Timeseries] TimeSeries object for each component.

ns\_f, ew\_f, vt\_f [FourierTransform] FourierTransform object for each component.

**normalization\_factor** [float] Maximum value of *ns*, *ew*, and *vt* amplitude used for normalization when plotting.

```
___init___(ns, ew, vt)
```

Initalize a 3-component sensor (Sensor3c) object.

#### Args

ns, ew, vt [timeseries] TimeSeries object for each component.

**Returns:** Initialized 3-component sensor (Sensor3c) object.

#### bandpassfilter (flow, fhigh, order)

Bandpassfilter component TimeSeries.

Refer to SigProPy documentation for details.

# combine\_horizontals (method='squared-average')

Combine two horizontal components (ns and ew).

#### Args:

**method** [{'squared-averge', 'geometric-mean'}, optional] Defines how the two horizontal components are combined to represent a single horizontal component, the default is 'squared-average'.

**Returns:** A FourierTransform object representing the combined horizontal components.

#### cosine\_taper(width)

Cosine taper component TimeSeries.

Refer to SigProPy documentation for details.

#### detrend()

Detrend component TimeSeries.

Refer to SigProPy documentation for details.

# ${\tt classmethod\ from\_mseed}\,(\mathit{fname}\,)$

Initialize a 3-component sensor (Sensor3c) object from a .miniseed file.

#### Args:

**fname** [str] Name of miniseed file, full path may be used if desired. The file should contain three traces with the appropriate channel names. Refer to the *SEED* Manual here. for specifics.

**Returns:** Initialized 3-component sensor (*Sensor3c*) object.

**hv** (windowlength, bp\_filter, taper\_width, bandwidth, resampling, method)

Prepare time series and Fourier transforms then compute H/V.

#### Args:

windowlength [float] Length of time windows in seconds.

**bp\_filter** [dict] Bandpass filter settings, of the form {'flag':bool, 'flow':float, 'fhigh':float, 'order':int}.

taper\_width [float] Width of cosine taper.

bandwidth [float] Bandwidth of the Konno and Ohmachi smoothing window.

**resampling** [dict] Resampling settings, of the form {'minf':float, 'maxf':float, 'nf':int, 'res\_type':str}.

method [{'squared-averge', 'geometric-mean'}] Refer to combine\_horizontals for details.

Returns: Initialized Hvsr object.

**Notes:** More information for the above arguements can be found in the documenation of SigProPy.

#### property normalization\_factor

Return sensor time history normalization factor.

#### resample (fmin, fmax, fn, res\_type, inplace)

Resample component FourierTransforms.

Refer to SigProPy documentation for details.

#### smooth (bandwidth)

Smooth component FourierTransforms.

Refer to SigProPy documentation for details.

#### split (windowlength)

Split component *TimeSeries*.

Refer to SigProPy documentation for details.

# transform()

Perform Fourier transform on component TimeSeries.

**Returns:** None, redefines attributes ew\_f, ns\_f, and vt\_f as FourierTransform objects for each component.

# **FOUR**

# **HVSR CLASS**

```
class Hvsr(amplitude, frequency, find_peaks=True)
```

Class for creating and manipulating horizontal-to-vertical spectral ratio objects.

#### **Attributes:**

**amp** [ndarray] Array of H/V amplitudes. Each row represents an individual curve and each column a frequency.

frq [ndarray] Vector of frequencies corresponds to each column.

**n\_windows** [int] Number of windows in *Hvsr* object.

valid\_window\_indices [ndarray] Array of indices indicating valid windows.

rejected\_window\_indices [ndarray] Array of indices indicating rejected windows.

**peak\_frq** [ndarray] Frequency peaks, one per valid time window.

**peak\_amp** [ndarray] Amplitude(s) which correspond to *peak\_frq*.

\_\_init\_\_ (amplitude, frequency, find\_peaks=True)

Initialize a *Hvsr* oject from an amplitude and frequency vector.

# Args:

**amplitude** [ndarray] Array of H/V amplitudes. Each row represents an individual curve and each column a frequency.

frequency [ndarray] Vector of frequencies, corresponding to each column.

**Returns:** Initialized *Hvsr* object.

```
static find_peaks(amp, **kwargs)
```

Indices of all peaks in amp.

Wrapper method for scipy.signal.find\_peaks function.

#### Args:

**amp** [ndarray] Vector or array of amplitudes. See *amp* attribute for details.

\*\*kwargs [dict] Refer to scipy.signal.find\_peaks documentation.

# **Returns:**

**peaks** [ndarray or list] *ndarray* or *list* of *ndarrays* (one per window) of peak indices.

**properties** [dict] Refer to scipy.signal.find\_peaks documentation.

```
mc_peak_amp (distribution='log-normal')
```

Amplitude of the peak of the mean H/V curve.

```
Args:
         distribution [{'normal', 'log-normal'}, optional] Refer to mean_curve for details.
     Returns: Ampltiude associated with the peak of the mean H/V curve.
mc_peak_frq(distribution='log-normal')
     Frequency of the peak of the mean H/V curve.
     Args:
         distribution [{'normal', 'log-normal'}, optional] Refer to mean_curve for details.
     Returns: Frequency associated with the peak of the mean H/V curve.
mean curve(distribution='log-normal')
     Return mean H/V curve.
     Args:
         distribution [{'normal', 'log-normal'}, optional] Assumed distribution of mean curve, default is
              'log-normal'.
     Returns: Mean H/V curve as ndarray according to the distribution specified.
     Raises:
         KeyError: If distribution does not match the available options.
mean f0 amp (distribution='log-normal')
     Mean amplitude of f0 of valid time windows.
     Args:
         distribution [{'normal', 'log-normal'}] Assumed distribution of f0, default is 'log-normal'.
     Returns: Mean amplitude of f0 according to the distribution specified.
     Raises:
         KeyError: If distribution does not match the available options.
mean_f0_frq(distribution='log-normal')
     Mean f0 of valid time windows.
     Args:
         distribution [{'normal', 'log-normal'}] Assumed distribution of f0, default is 'log-normal'.
     Returns: Mean value of f0 according to the distribution specified.
     Raises:
         KeyError: If distribution does not match the available options.
nstd curve(n, distribution)
     Return nth standard deviation curve.
     Args:
         n [float] Number of standard deviations away from the mean curve.
         distribution [{'log-normal', 'normal'}, optional] Assumed distribution of mean curve, the default is
              'log-normal'.
     Return: nth standard deviation curve as an ndarray.
nstd_f0_amp (n, distribution)
```

nth sample standard deviation of amplitude of f0 from time windows.

#### Args:

n [float] Number of standard deviations away from the mean amplitude of f0 from valid time windows.

**distribution** [{'log-normal', 'normal'}, optional] Assumed distribution of f0, the default is 'log-normal'.

**Return:** nth standard deviation of ampltiude of f0 as float.

#### nstd\_f0\_frq(n, distribution)

Return nth standard deviation of f0.

## Args:

**n** [float] Number of standard deviations away from the mean f0 from the valid time windows.

**distribution** [{'log-normal', 'normal'}, optional] Assumed distribution of f0, the default is 'log-normal'.

**Return:** nth standard deviation of f0 as float.

#### property peak\_amp

Valid peak amplitude vector.

## property peak\_frq

Valid peak frequency vector.

#### print\_stats (distribution\_f0)

Print basic statistics of *Hvsr* instance.

reject\_windows (n=2, max\_iterations=50, distribution\_f0='log-normal', distribution\_mc='log-normal')

Perform rejection of H/V windows using the method proposed by Cox et al. (in review).

## Args:

**n** [float, optional] Number of standard deviations from the mean, default value is 2.

max\_iterations [int, optional] Maximum number of rejection iterations, default value is 50.

**distribution\_f0** [{'log-normal', 'normal'}, optional] Assumed distribution of *f0* from time windows, the default is 'log-normal'.

**distribution\_mc** [{'log-normal', 'normal'}, optional] Assumed distribution of mean curve, the default is 'log-normal'.

#### **Returns:**

**c\_iteration** [int] Number of iterations required for convergence.

#### property rejected\_window\_indices

Rejected window indices.

## std\_curve (distribution='log-normal')

Sample standard deviation associated with the mean H/V curve.

#### Args:

**distribution** [{'normal', 'log-normal'}, optional] Assumed distribution of H/V curve, default is 'log-normal'.

Returns: Sample standard deviation of H/V curve as ndarray according to the distribution specified.

#### Raises:

ValueError: If only single time window is defined.

**KeyError:** If *distribution* does not match the available options.

#### std\_f0\_amp (distribution='log-normal')

Sample standard deviation of amplitude of f0 of valid time windows.

#### Args:

**distribution** [{'normal', 'log-normal'}, optional] Assumed distribution of f0, default is 'log-normal'.

**Returns:** Sample standard deviation of the amplitude of f0 according to the distribution specified.

#### Raises:

**KeyError:** If *distribution* does not match the available options.

#### std\_f0\_frq(distribution='log-normal')

Sample standard deviation of f0 of valid time windows.

#### Args:

**distribution** [{'normal', 'log-normal'}, optional] Assumed distribution of f0, default is 'log-normal'.

**Returns:** Sample standard deviation of f0 according to the distribution specified.

#### Raises:

**KeyError:** If *distribution* does not match the available options.

#### to\_file\_like\_geopsy (fname, distribution\_f0, distribution\_mc)

Save H/V data to file following the Geopsy format.

#### Args:

fname [str] Name of file to save the results, may be a full or relative path.

**distribution\_f0** [{'log-normal', 'normal'}, optional] Assumed distribution of f0 from the time windows, the default is 'log-normal'.

**distribution\_mc** [{'log-normal', 'normal'}, optional] Assumed distribution of mean curve, the default is 'log-normal'.

Returns: None, writes file to disk.

#### update\_peaks (\*\*kwargs)

Update *peaks* attribute with the lowest frequency, highest amplitude peak.

# Args:

\*\*kwargs: Refer to find\_peaks documentation.

**Returns:** *None*, update *peaks* attribute.