# Module spectral

#### Spectral

A library for analysing timeseries data, specifically for neural event identification, detection and classification.

Consists of three submodules, which can be used either independently or together:

- 1. Contrast: This module enables the contrasting between two categories of timeseries data (with multiple trials per category). This enables the identification of the frequency bins that have the most difference between the categories.
- 2. Cluster: This module enables the clustering of the timeseries data based on the similarity of it's spectral decomposition using STFT. It finds the optimal number of clusters and returns a vector with labels of which class each STFT segment belongs to.
- 3. Classify: This module provides code to train classifiers for classifying timeseries data based on the clusters identified using the spectral.cluster module.

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Version 0.1

#### **Sub-modules**

- spectral.classify
- spectral.cluster
- spectral.contrast
- spectral.test\_contrast

# Module spectral.classify

Module for classifying timeseries data based on their spectral properties.

This module enables the classification of STFT transformed data by training an SVM classifier.

This facilitates the identification of various events (defined as transient spatio-temporal patterns of activity) present within the timeseries data, that have been identified.

### **Functions**

### Function classifySVM

```
def classifySVM(
    self,
    X,
    y
)
```

Trains an SVM-classifier on the data.

Parameters

```
X: array Training data with shape: nobs x features y: array vector with training labels
```

Returns

```
scores : array a vector of scores with length equal to number of CV-folds
clf : python object the trained classifier as a python object
```

#### Function generate\_features

```
def generate_features(
   data,
   labels,
   **kwargs
)
```

Generate a feature vector for training a classifier

Parameters

```
data : array array with structure
labels : array vector with class labels
```

Returns

X: array an array with the features in the 1st axis and trials on the 0-th axis y: array a vector with the same number of rows as X containing class labels

# Module spectral.cluster

Module for clustering timeseries data based on their spectral properties.

This module enables the clustering of STFT transformed data by mapping the STFT arrays to a low dimensional manifold and then clustering them using DBSCAN.

This facilitates the identification of various events (defined as transient spatio-temporal patterns of activity) present within the timeseries data.

### **Functions**

#### Function cluster

```
def cluster(
    data,
    **kwargs
)
```

Clusters the array using OPTICS and dbscan. Finds the best number of clusters.

Parameters

 ${\tt data\_array: array \ STFT \ array \ or \ low-dimensional \ embedding \ from \ embed() \ [nchan \ x \ nobs \ x \ ntrials]}$ 

Returns

```
res: array results with res[0] having the nclust: int number of clusters identified
```

#### Function embed

```
def embed(
    data_stft_norm,
    **kwargs
)
```

Returns a low-dimensional embedding of an STFT array.

Parameters

data\_norm: array normalized stft array [nchan x nfreqs x nobs x ntrials]

#### Returns

embedding: array low dimensional embedding of the STFT array

#### Function stft\_norm

```
def stft_norm(
    data,
    **kwargs
)
```

Returns the frequency-normalized STFT for timeseries data.

Parameters

```
data_array : array timeseries data [nchan x nobs x ntrials]
*fs : int sampling frequency in Hz
*nperseg : int number of timepoints for stft window
*noverlap : int number of timepoints for window overlap
Returns
stft_norm : array STFT of the input array [nchan x nfreqs x nobs x trials]
f : array an array of the frequencies of the STFT transform
```

## Module spectral.contrast

Module for contrasting timeseries data based on their spectral properties.

The set of methods are aimed at finding the frequency bands that enable the maximal seperability betwen two sets of timeseries data.

#### **Functions**

#### Function contrast

```
def contrast(
    data,
    y,
    **kwargs
)
```

This method returns the SNR given a data array and vector of labels.

Ideally, this should be the only method that you need to call when contrasting timeseries' spectra.

Parameters

snr: array [nfreqs x nfreqs] a matrix with the SNR for each combination of frequency bandsf: array [nfreqs] a vector that represents the frequencies for interpreting snr.

```
Function filter
```

```
def filter(
    data,
    low_pass,
    high_pass,
    fs,
    order=4
)
```

Generates an n-th order butterworth filter and performs forward-backward pass on the signal.

Parameters

```
data : array same as data structure [nchans x nobs x ntrials]
low_pass : param low pass frequency
high_pass : param high pass frequency
fs : param sampling frequency
order : param filter order
```

Returns

filt\_data: array array with same shape as data but bandpass filtered

### Function generate\_ts

```
def generate_ts(
    nsamples,
    fs,
    **kwargs
)
```

Generates an LFP-like timeseries sampled at fs obeying the power law.

### Function get\_bands

```
def get_bands(
    target_stft_norm,
    baseline_stft_norm,
    f
)
```

Calculates the mean power across all possible combinations of frequencies for each channel.

Parameters

```
\label{target_stft_norm: array stft decomposed target array [nchan x nfreqs x nobs x ntrials] \\ \textbf{baseline_stft_norm: array stft decomposed baseline array [nchan x nfreqs x nobs x ntrials]} \\ \textbf{f: array } vector of frequencies obtained from STFT transform (see get_stft()).} \\
```

Returns

 $\label{target_bands: array array of mean power across all possible band permutations [nchan x nfreqs x nobs x ntrials]$ 

 $\verb|baseline_bands|: array | same | as | target\_bands| [nchan | x | nfreqs| x | nfreqs| x | nobs| x | ntrials]$ 

#### Function get\_norm\_array

```
def get_norm_array(
    data,
    **kwargs
)
```

Returns the normalization array for timeseries data.

```
Parameters
```

```
data: array, timeseries data [nchan x nobs x ntrials]
**fs: int, sampling frequency in Hz
**nperseg: int, number of timepoints for stft window
**noverlap: int, number of timepoints for window overlap
Returns
norm_array: normalized array with mean power per frequency [nchan x freqs]
Function get_snr
     def get_snr(
         target,
         baseline
Returns the SNR given two vectors: target and baseline.
Parameters
target: array [nchan x nfreqs x nobs x ntrials] an array obtained by using get_bands()
baseline: array (same as target)
Returns
snr: array [nfreqs x nfreqs] a lower triangular matix representing the contrast between bands
Function get stft
     def get_stft(
         data_array,
         norm array=[],
         normalize=True,
         **kwargs
     )
Returns the STFT for timeseries data.
Parameters
data_array : array timeseries data [nchan x nobs x ntrials]
norm_array : array for spectral normalization (see get_norm_array())
*fs: int sampling frequency in Hz
*nperseg: int number of timepoints for stft window
*noverlap: int number of timepoints for window overlap
Returns
stft_array: array STFT of the input array [nchan x nfreqs x nobs x trials]
f: array an array of the frequencies of the STFT transform
Function simulate_recording
     def simulate_recording(
         **kwargs
```

Simulates an LFP recording with bursts in power of certain bands.

)

#### Function test

def test()

Simple test method to ensure that the pipeline and dependencies work.

Returns True if everything works.

# ${\bf Module\ spectral.test\_contrast}$

### **Functions**

 $Function\ {\tt test\_contrast}$ 

def test\_contrast()

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