API Reference

This part of the documentation details the complete **BioSPPy** API.

Packages

• biosppy.signals

Modules

- biosppy.biometrics
- biosppy.clustering
- biosppy.metrics
- biosppy.plotting
- biosppy.storage
- biosppy.timing
- biosppy.utils

biosppy.biometrics

This module provides classifier interfaces for identity recognition (biometrics) applications. The core API methods are: * enroll: add a new subject; * dismiss: remove an existing subject; * identify: determine the identity of collected biometric dataset; * authenticate: verify the identity of collected biometric dataset.

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class biosppy.biometrics.BaseClassifier

Bases: object

Base biometric classifier class.

This class is a skeleton for actual classifier classes. The following methods must be overridden or adapted to build a new classifier:

- __init__
- _authenticate
- _get_thresholds
- identify

- train
- _update

EER IDX

int - Reference index for the Equal Error Rate.

EER_IDX=0

authenticate(data, subject, threshold=None)

Authenticate a set of feature vectors, allegedly belonging to the given subject.

Parameters:

- data (array) Input test data.
- **subject** (hashable) Subject identity.
- threshold (int, float, optional) Authentication threshold.

Returns: decision (array) – Authentication decision for each input sample.

batch_train(data=None)

Train the classifier in batch mode.

Parameters: data (dict) – Dictionary holding training data for each subject; if the object for a

subject is None, performs a dismiss.

check_subject(subject)

Check if a subject is enrolled.

Parameters: subject (hashable) - Subject identity.

Returns: check (bool) – If True, the subject is enrolled.

classmethod cross_validation(data, labels, cv, thresholds=None, **kwargs)

Perform Cross Validation (CV) on a data set.

Parameters:

- data (array) An m by n array of m data samples in an n-dimensional space.
- labels (list, array) A list of m class labels.
- cv (CV iterator) A sklearn.model_selection iterator.
- thresholds (array, optional) Classifier thresholds to use.
- **kwargs (dict, optional) Classifier parameters.

Returns:

- runs (list) Evaluation results for each CV run.
- assessment (dict) Final CV biometric statistics.

Remove a subject.

Parameters:

- subject (hashable) Subject identity.
- deferred (bool, optional) If True, computations are delayed until flush is called.

Raises: SubjectError - If the subject to remove is not enrolled.

Notes

• When using deferred calls, a dismiss overrides a previous enroll for the same subject.

enroll(data=None, subject=None, deferred=False)

Enroll new data for a subject.

If the subject is already enrolled, new data is combined with existing data.

Parameters:

- data (array) Data to enroll.
- subject (hashable) Subject identity.
- **deferred** (*bool*, *optional*) If True, computations are delayed until *flush* is called.

Notes

• When using deferred calls, an enroll overrides a previous dismiss for the same subject.

evaluate(data, thresholds=None, show=False)

Assess the performance of the classifier in both authentication and identification scenarios.

Parameters:

- data (dict) Dictionary holding test data for each subject.
- thresholds (array, optional) Classifier thresholds to use.
- **show** (bool, optional) If True, show a summary plot.

Returns:

- classification (dict) Classification results.
- assessment (dict) Biometric statistics.

flush()

Flush deferred computations.

get_auth_thr(subject, ready=False)

Get the authentication threshold of a subject.

• ready (bool, optional) - If True, subject is the internal classifier label.

Returns: threshold (int, float) – Threshold value.

get_id_thr(subject, ready=False)

Get the identification threshold of a subject.

Parameters: • subject (hashable) – Subject identity.

• ready (bool, optional) – If True, subject is the internal classifier label.

Returns: threshold (int, float) – Threshold value.

get_thresholds(force=False)

Get an array of reasonable thresholds.

Parameters: force (bool, optional) – If True, forces generation of thresholds.

Returns: ths (array) – Generated thresholds.

identify(data, threshold=None)

Identify a set of feature vectors.

Parameters: • data (array) – Input test data.

• threshold (int, float, optional) – Identification threshold.

Returns: subjects (*list*) – Identity of each input sample.

io_del(label)

Delete subject data.

Parameters: label (str) – Internal classifier subject label.

io_load(label)

Load enrolled subject data.

Parameters: label (str) – Internal classifier subject label.

Returns: data (array) – Subject data.

4 of 29 io_save(label, data) 05/04/20, 11:39 am

Parameters:

- label (str) Internal classifier subject label.
- data (array) Subject data.

list_subjects()

List all the enrolled subjects.

Returns: subjects (list) - Enrolled subjects.

classmethod load(path)

Load classifier instance from a file.

Parameters: path (str) – Source file path.

Returns: clf (*object*) – Loaded classifier instance.

save(path)

Save classifier instance to a file.

Parameters: path (str) – Destination file path.

set_auth_thr(subject, threshold, ready=False)

Set the authentication threshold of a subject.

Parameters:

- subject (hashable) Subject identity.
- threshold (int, float) Threshold value.
- ready (bool, optional) If True, subject is the internal classifier label.

set_id_thr(subject, threshold, ready=False)

Set the identification threshold of a subject.

Parameters:

- subject (hashable) Subject identity.
- threshold (int, float) Threshold value.
- ready (bool, optional) If True, subject is the internal classifier label.

update_thresholds(fraction=1.0)

Update subject-specific thresholds based on the enrolled data.

Parameters: fraction (float, optional) – Fraction of samples to select from training data.

Bases: exceptions.Exception

Exception raised when the combination method fails.

class biosppy.biometrics.KNN(k=3, metric='euclidean', metric_args=None)

Bases: biosppy.biometrics.BaseClassifier

K Nearest Neighbors (k-NN) biometric classifier.

Parameters:

- k (int, optional) Number of neighbors.
- metric (str, optional) Distance metric.
- metric_args (dict, optional) Additional keyword arguments are passed to the distance function.

EER_IDX

int - Reference index for the Equal Error Rate.

EER IDX=0

class biosppy.biometrics.SVM(C=1.0, kernel='linear', degree=3, gamma='auto', coef0=0.0, shrinking=True, tol=0.001, cache_size=200, max_iter=-1, random_state=None)

Bases: biosppy.biometrics.BaseClassifier

Support Vector Machines (SVM) biometric classifier.

Wraps the 'OneClassSVM' and 'SVC' classes from 'scikit-learn'.

Parameters:

- **C** (*float*, *optional*) Penalty parameter C of the error term.
- **kernel** (*str*, *optional*) Specifies the kernel type to be used in the algorithm. It must be one of 'linear', 'poly', 'rbf', 'sigmoid', 'precomputed' or a callable. If none is given, 'rbf' will be used. If a callable is given it is used to precompute the kernel matrix.
- **degree** (*int*, *optional*) Degree of the polynomial kernel function ('poly'). Ignored by all other kernels.
- gamma (*float*, *optional*) Kernel coefficient for 'rbf', 'poly' and 'sigmoid'. If gamma is 'auto' then 1/n_features will be used instead.
- **coef0** (*float*, *optional*) Independent term in kernel function. It is only significant in 'poly' and 'sigmoid'.
- **shrinking** (*bool*, *optional*) Whether to use the shrinking heuristic.
- tol (float, optional) Tolerance for stopping criterion.
- cache_size (float, optional) Specify the size of the kernel cache (in MB).
- max_iter (int, optional) Hard limit on iterations within solver, or -1 for no limit.
- random_state (int, RandomState, optional) The seed of the pseudo random number generator to use when shuffling the data for probability estimation.

int - Reference index for the Equal Error Rate.

EER_IDX=-1

exception biosppy.biometrics.SubjectError(subject=None)

Bases: exceptions.Exception

Exception raised when the subject is unknown.

exception biosppy.biometrics.UntrainedError

Bases: exceptions.Exception

Exception raised when classifier is not trained.

biosppy.biometrics.assess_classification(results=None, thresholds=None)

Assess the performance of a biometric classification test.

Parameters: • results (dict) – Classification results.

• thresholds (array) – Classifier thresholds.

Returns: assessment (dict) – Classification assessment.

biosppy.biometrics.assess_runs(results=None, subjects=None)

Assess the performance of multiple biometric classification runs.

• subjects (list) - Common target subject classes.

• results (list) - Classification assessment for each run.

Returns: assessment (*dict*) – Global classification assessment.

biosppy.biometrics.combination(results=None, weights=None)

Combine results from multiple classifiers.

Parameters: • results (dict) – Results for each classifier.

• weights (dict, optional) – Weight for each classifier.

Returns:

Parameters:

• decision (object) - Consensus decision.

• **confidence** (*float*) – Confidence estimate of the decision.

• counts (array) - Weight for each possible decision outcome.

• classes (array) – List of possible decision outcomes.

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random state=None)

Return a Cross Validation (CV) iterator.

Wraps the StratifiedShuffleSplit iterator from sklearn.model_selection. This iterator returns stratified randomized folds, which preserve the percentage of samples for each class.

Parameters:

- labels (list, array) List of class labels for each data sample.
- n_iter (int, optional) Number of splitting iterations.
- **test_size** (*float*, *int*, *optional*) If float, represents the proportion of the dataset to include in the test split; if int, represents the absolute number of test samples.
- train_size (float, int, optional) If float, represents the proportion of the dataset to include in the train split; if int, represents the absolute number of train samples.
- random_state (int, RandomState, optional) The seed of the pseudo random number generator to use when shuffling the data.

Returns:

cv (CV iterator) - Cross Validation iterator.

biosppy.biometrics.get_auth_rates(TP=None, FP=None, TN=None, FN=None, thresholds=None)

Compute authentication rates from the confusion matrix.

Parameters:

- TP (array) True Positive counts for each classifier threshold.
- FP (array) False Positive counts for each classifier threshold.
- TN (array) True Negative counts for each classifier threshold.
- FN (array) False Negative counts for each classifier threshold.
- thresholds (array) Classifier thresholds.

Returns:

- Acc (array) Accuracy at each classifier threshold.
- TAR (array) True Accept Rate at each classifier threshold.
- FAR (array) False Accept Rate at each classifier threshold.
- FRR (array) False Reject Rate at each classifier threshold.
- TRR (array) True Reject Rate at each classifier threshold.
- **EER** (array) Equal Error Rate points, with format (threshold, rate).
- Err (array) Error rate at each classifier threshold.
- **PPV** (array) Positive Predictive Value at each classifier threshold.
- FDR (array) False Discovery Rate at each classifier threshold.
- NPV (array) Negative Predictive Value at each classifier threshold.
- FOR (array) False Omission Rate at each classifier threshold.
- MCC (array) Matthrews Correlation Coefficient at each classifier threshold.

biosppy.biometrics.get_id_rates(H=None, M=None, R=None, N=None, thresholds=None)

Compute identification rates from the confusion matrix.

- M (array) Miss counts for each classifier threshold.
- R (array) Reject counts for each classifier threshold.
- N (int) Number of test samples.
- thresholds (array) Classifier thresholds.

Returns:

- Acc (array) Accuracy at each classifier threshold.
- Err (array) Error rate at each classifier threshold.
- MR (array) Miss Rate at each classifier threshold.
- RR (array) Reject Rate at each classifier threshold.
- **EID** (array) Error of Identification points, with format (threshold, rate).
- **EER** (array) Equal Error Rate points, with format (threshold, rate).

biosppy.biometrics.get_subject_results(results=None, subject=None, thresholds=None, subjects=None, subject_dict=None, subject_idx=None)

Compute authentication and identification performance metrics for a given subject.

Parameters:

- results (dict) Classification results.
- subject (hashable) True subject label.
- thresholds (array) Classifier thresholds.
- subjects (list) Target subject classes.
- subject_dict (bidict) Subject-label conversion dictionary.
- subject_idx (list) Subject index.

Returns: assessment (dict) – Authentication and identification results.

biosppy.biometrics.majority_rule(labels=None, random=True)

Determine the most frequent class label.

Parameters:

- labels (array, list) List of clas labels.
- random (bool, optional) If True, will choose randomly in case of tied classes, otherwise the first element is chosen.

Returns:

- **decision** (*object*) Consensus decision.
- **count** (*int*) Number of elements of the consensus decision.

biosppy.clustering

This module provides various unsupervised machine learning (clustering) algorithms.

biosppy.clustering.centroid_templates(data=None, clusters=None, ntemplates=1)

Template selection based on cluster centroids.

Parameters:

- data (array) An m by n array of m data samples in an n-dimensional space.
- clusters (dict) Dictionary with the sample indices (rows from 'data') for each cluster.
- **ntemplates** (int, optional) Number of templates to extract; if more than 1, k-means is used to obtain more templates.

Returns:

templates (array) - Selected templates from the input data.

biosppy.clustering.coassoc_partition(coassoc=None, k=0, linkage='average')

Extract the consensus partition from a co-association matrix using hierarchical agglomerative methods.

Parameters:

- coassoc (array) Co-association matrix.
- k (int, optional) Number of clusters to extract; if 0 uses the life-time criterion.
- linkage (str, optional) Linkage criterion for final partition extraction; one of 'average', 'complete', 'single', or 'weighted'.

Returns:

clusters (dict) - Dictionary with the sample indices (rows from 'data') for each found cluster; outliers have key -1; clusters are assigned integer keys starting at 0.

biosppy.clustering.consensus(data=None, k=0, linkage='average', fcn=None, grid=None)

Perform clustering based in an ensemble of partitions.

Parameters:

- data (array) An m by n array of m data samples in an n-dimensional space.
- k (int, optional) Number of clusters to extract; if 0 uses the life-time criterion.
- linkage (str, optional) Linkage criterion for final partition extraction; one of 'average', 'centroid', 'complete', 'median', 'single', 'ward', or 'weighted'.
- fcn (function) A clustering function.
- grid (dict, list, optional) A (list of) dictionary with parameters for each run of the clustering method (see sklearn.model_selection.ParameterGrid).

Returns:

clusters (dict) - Dictionary with the sample indices (rows from 'data') for each found cluster; outliers have key -1; clusters are assigned integer keys starting at 0.

Parameters:

- data (array) An m by n array of m data samples in an n-dimensional space.
- k (int, optional) Number of clusters to extract; if 0 uses the life-time criterion.
- **linkage** (*str*, *optional*) Linkage criterion for final partition extraction; one of 'average', 'centroid', 'complete', 'median', 'single', 'ward', or 'weighted'.
- nensemble (int, optional) Number of partitions in the ensemble.
- **kmin** (int, optional) Minimum k for the k-means partitions; defaults to $\sqrt{m}/2$.
- kmax (int, optional) Maximum k for the k-means partitions; defaults to \sqrt{m} .

Returns:

clusters (*dict*) – Dictionary with the sample indices (rows from 'data') for each found cluster; outliers have key -1; clusters are assigned integer keys starting at 0.

biosppy.clustering.create_coassoc(ensemble=None, N=None)

Create the co-association matrix from a clustering ensemble.

Parameters:

- ensemble (list) Clustering ensemble partitions.
- N (int) Number of data samples.

Returns: coassoc (array) – Co-association matrix.

biosppy.clustering.create_ensemble(data=None, fcn=None, grid=None)

Create an ensemble of partitions of the data using the given clustering method.

Parameters:

- data (array) An m by n array of m data samples in an n-dimensional space.
- fcn (function) A clustering function.
- **grid** (*dict*, *list*, *optional*) A (list of) dictionary with parameters for each run of the clustering method (see sklearn.model_selection.ParameterGrid).

Returns:

ensemble (list) - Obtained ensemble partitions.

biosppy.clustering.dbscan(data=None, min_samples=5, eps=0.5, metric='euclidean', metric args=None)

Perform clustering using the DBSCAN algorithm [EKSX96].

The algorithm works by grouping data points that are closely packed together (with many nearby neighbors), marking as outliers points that lie in low-density regions.

- min_samples (int, optional) Minimum number of samples in a cluster.
- eps (float, optional) Maximum distance between two samples in the same cluster.
- metric (str, optional) Distance metric (see scipy.spatial.distance).
- metric_args (dict, optional) Additional keyword arguments to pass to the

distance function.

Returns:

clusters (*dict*) – Dictionary with the sample indices (rows from 'data') for each found cluster; outliers have key -1; clusters are assigned integer keys starting at 0.

References

[EKSX96] M. Ester, H. P. Kriegel, J. Sander, and X. Xu, "A Density-Based Algorithm for Discovering Clusters in Large Spatial Databases with Noise", Proceedings of the 2nd International Conf. on Knowledge Discovery and Data Mining, pp. 226-231, 1996.

 $\begin{tabular}{ll} biosppy.clustering.hierarchical(data=None, k=0, linkage='average', metric='euclidean', metric_args=None) \end{tabular}$

Perform clustering using hierarchical agglomerative algorithms.

Parameters:

- data (array) An m by n array of m data samples in an n-dimensional space.
- k (int, optional) Number of clusters to extract; if 0 uses the life-time criterion.
- **linkage** (*str*, *optional*) Linkage criterion; one of 'average', 'centroid', 'complete', 'median', 'single', 'ward', or 'weighted'.
- metric (str, optional) Distance metric (see 'biosppy.metrics').
- metric_args (dict, optional) Additional keyword arguments to pass to the distance function.

Returns:

clusters (*dict*) – Dictionary with the sample indices (rows from 'data') for each found cluster; outliers have key -1; clusters are assigned integer keys starting at 0.

Raises:

- TypeError If 'metric' is not a string.
- ValueError When the 'linkage' is unknown.
- ValueError When 'metric' is not 'euclidean' when using 'centroid', 'median', or 'ward' linkage.
- **ValueError** When 'k' is larger than the number of data samples.

biosppy.clustering.kmeans(data=None, k=None, init='random', max_iter=300, n_init=10, tol=0.0001)

API Reference BioSPPy 0.6.1 documentation by n array of m data samples in an indimensional space; table/biosppy...

- k (int) Number of clusters to extract.
- init (str, array, optional) If string, one of 'random' or 'k-means++'; if array, it should be of shape (n_clusters, n_features), specifying the initial centers.
- max_iter (int, optional) Maximum number of iterations.
- **n_init** (int, optional) Number of initializations.
- tol (float, optional) Relative tolerance to declare convergence.

Returns:

clusters (*dict*) – Dictionary with the sample indices (rows from 'data') for each found cluster; outliers have key -1; clusters are assigned integer keys starting at 0.

biosppy.clustering.mdist_templates(data=None, clusters=None, ntemplates=1, metric='euclidean',
metric_args=None)

Template selection based on the MDIST method [UIRJ04].

Extends the original method with the option of also providing a data clustering, in which case the MDIST criterion is applied for each cluster [LCSF14].

Parameters:

- data (array) An m by n array of m data samples in an n-dimensional space.
- **clusters** (*dict*, *optional*) Dictionary with the sample indices (rows from *data*) for each cluster.
- **ntemplates** (int, optional) Number of templates to extract.
- metric (str, optional) Distance metric (see scipy.spatial.distance).
- metric_args (dict, optional) Additional keyword arguments to pass to the distance function.

Returns:

templates (array) – Selected templates from the input data.

References

[UIRJ04] U. Uludag, A. Ross, A. Jain, "Biometric template selection and update: a case study in fingerprints", Pattern Recognition 37, 2004

[LCSF14] A. Lourenco, C. Carreiras, H. Silva, A. Fred, "ECG biometrics: A template selection approach", 2014 IEEE International Symposium on Medical Measurements and Applications (MeMeA), 2014

 $\begin{tabular}{ll} biosppy.clustering.outliers_dbscan({\it data=None, min_samples=5, eps=0.5, metric='euclidean', metric_args=None}) \end{tabular}$

Perform outlier removal using the DBSCAN algorithm.

- min_samples (int, optional) Minimum number of samples in a cluster.
- eps (float, optional) Maximum distance between two samples in the same cluster.
- metric (str, optional) Distance metric (see scipy.spatial.distance).
- metric_args (dict, optional) Additional keyword arguments to pass to the distance function.

Returns:

- **clusters** (*dict*) Dictionary with the sample indices (rows from 'data') for the outliers (key -1) and the normal (key 0) groups.
- **templates** (*dict*) Elements from 'data' for the outliers (key -1) and the normal (key 0) groups.

biosppy.clustering.outliers_dmean(data=None, alpha=0.5, beta=1.5, metric='euclidean', metric args=None, max idx=None)

Perform outlier removal using the DMEAN algorithm [LCSF13].

A sample is considered valid if it cumulatively verifies:

- distance to average template smaller than a (data derived) threshold 'T';
- sample minimum greater than a (data derived) threshold 'M';
- sample maximum smaller than a (data derived) threshold 'N';
- position of the sample maximum is the same as the given index [optional].

For a set of $\{X_1,...,X_n\}n$ samples:

$$\widetilde{X} = \frac{1}{n} \sum_{i=1}^{n} X_i$$

$$d_i = dist(X_i, \widetilde{X})$$

$$D_m = \frac{1}{n} \sum_{i=1}^{n} d_i$$

$$D_s = \sqrt{\frac{1}{n-1} \sum_{i=1}^{n} (d_i - D_m)^2}$$

$$T = D_m + \alpha * D_s$$

$$M = \beta * median(\{\max X_i, i = 1, ..., n\})$$

$$N = \beta * median(\{\min X_i, i = 1, ..., n\})$$

Parameters:

- data (array) An m by n array of m data samples in an n-dimensional space.
- alpha (float, optional) Parameter for the distance threshold.
- beta (float, optional) Parameter for the maximum and minimum thresholds.
- metric (str, optional) Distance metric (see scipy.spatial.distance).
- metric_args (dict, optional) Additional keyword arguments to pass to the distance function.
- max_idx (int, optional) Index of the expected maximum.

- **clusters** (*dict*) Dictionary with the sample indices (rows from 'data') for the outliers (key -1) and the normal (key 0) groups.
- **templates** (*dict*) Elements from 'data' for the outliers (key -1) and the normal (key 0) groups.

References

[LCSF13] A. Lourenco, H. Silva, C. Carreiras, A. Fred, "Outlier Detection in Non-intrusive ECG Biometric System", Image Analysis and Recognition, vol. 7950, pp. 43-52, 2013

biosppy.metrics

This module provides pairwise distance computation methods.

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biosppy.metrics.cdist(XA, XB, metric='euclidean', p=2, V=None, VI=None, w=None)

Computes distance between each pair of the two collections of inputs.

Wraps scipy.spatial.distance.cdist.

Parameters:

- XA (array) An m_A by n array of m_A original observations in an n-dimensional space.
- XB (array) An m_B by n array of m_B original observations in an n-dimensional space.
- metric (str, function, optional) The distance metric to use; the distance can be 'braycurtis', 'canberra', 'chebyshev', 'cityblock', 'correlation', 'cosine', 'dice', 'euclidean', 'hamming', 'jaccard', 'kulsinski', 'mahalanobis', 'matching', 'minkowski', 'pcosine', 'rogerstanimoto', 'russellrao', 'seuclidean', 'sokalmichener', 'sokalsneath', 'sqeuclidean', 'yule'.
- p (*float*, *optional*) The p-norm to apply (for Minkowski, weighted and unweighted).
- w (array, optional) The weight vector (for weighted Minkowski).
- **V** (array, optional) The variance vector (for standardized Euclidean).
- VI (array, optional) The inverse of the covariance matrix (for Mahalanobis).

Returns:

Y (array) – An m_A by m_B distance matrix is returned. For each i and j, the metric dist(u=XA[i], v=XB[j]) is computed and stored in the ij th entry.

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$$d(u,v) = 1 - abs\left(\frac{u \cdot v}{||u||_2||v||_2}\right)$$

where $u \cdot v$ is the dot product of u and v.

Parameters:

- u (array) Input array.
- v (array) Input array.

Returns:

cosine (float) - Cosine distance between u and v.

biosppy.metrics.pdist(X, metric='euclidean', p=2, w=None, V=None, VI=None)

Pairwise distances between observations in n-dimensional space.

Wraps scipy.spatial.distance.pdist.

Parameters:

- X (array) An m by n array of m original observations in an n-dimensional space.
- metric (str, function, optional) The distance metric to use; the distance can be
 'braycurtis', 'canberra', 'chebyshev', 'cityblock', 'correlation', 'cosine', 'dice',
 'euclidean', 'hamming', 'jaccard', 'kulsinski', 'mahalanobis', 'matching', 'minkowski',
 'pcosine', 'rogerstanimoto', 'russellrao', 'seuclidean', 'sokalmichener', 'sokalsneath',
 'sqeuclidean', 'yule'.
- p (*float*, *optional*) The p-norm to apply (for Minkowski, weighted and unweighted).
- w (array, optional) The weight vector (for weighted Minkowski).
- V (array, optional) The variance vector (for standardized Euclidean).
- VI (array, optional) The inverse of the covariance matrix (for Mahalanobis).

Returns:

Y (array) – Returns a condensed distance matrix Y. For each i and j (where i < j < n), the metric <code>dist(u=X[i], v=X[j])</code> is computed and stored in entry <code>ij</code>.

biosppy.metrics.squareform(X, force='no', checks=True)

Converts a vector-form distance vector to a square-form distance matrix, and vice-versa.

Wraps scipy.spatial.distance.squareform.

Parameters:

16 of 29

- X (array) Either a condensed or redundant distance matrix.
- **force** (*str*, *optional*) As with MATLAB(TM), if force is equal to 'tovector' or 'tomatrix', the input will be treated as a distance matrix or distance vector respectively.
- **checks** (*bool*, *optional*) If *checks* is set to False, no checks will be made for matrix symmetry nor zero diagonals. This is useful if it is known that x x.T1 is small and diag(x) is close to zero. These values are ignored any way so they do not disrupt the squareform transformation.

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biosppy.plotting

This module provides utilities to plot data.

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biosppy.plotting.plot_biometrics(assessment=None, eer_idx=None, path=None, show=False)

Create a summary plot of a biometrics test run.

Parameters:

- assessment (dict) Classification assessment results.
- eer_idx (int, optional) Classifier reference index for the Equal Error Rate.
- path (str. optional) If provided, the plot will be saved to the specified file.
- **show** (bool, optional) If True, show the plot immediately.

biosppy.plotting.plot_bvp(ts=None, raw=None, filtered=None, onsets=None, heart_rate_ts=None, heart_rate_ts=None, show=False)

Create a summary plot from the output of signals.bvp.bvp.

Parameters:

- ts (array) Signal time axis reference (seconds).
- raw (array) Raw BVP signal.
- filtered (array) Filtered BVP signal.
- onsets (array) Indices of BVP pulse onsets.
- heart_rate_ts (array) Heart rate time axis reference (seconds).
- heart_rate (array) Instantaneous heart rate (bpm).
- path (str. optional) If provided, the plot will be saved to the specified file.
- **show** (bool, optional) If True, show the plot immediately.

biosppy.plotting.plot_clustering(data=None, clusters=None, path=None, show=False)

Create a summary plot of a data clustering.

Parameters:

- data (array) An m by n array of m data samples in an n-dimensional space.
- **clusters** (*dict*) Dictionary with the sample indices (rows from *data*) for each cluster.
- path (str, optional) If provided, the plot will be saved to the specified file.
- **show** (bool, optional) If True, show the plot immediately.

API Reference process of the Reference process

Create a summary plot from the output of signals.ecg.ecg.

Parameters:

- ts (array) Signal time axis reference (seconds).
- raw (array) Raw ECG signal.
- filtered (array) Filtered ECG signal.
- rpeaks (array) R-peak location indices.
- templates_ts (array) Templates time axis reference (seconds).
- templates (array) Extracted heartbeat templates.
- heart_rate_ts (array) Heart rate time axis reference (seconds).
- heart_rate (array) Instantaneous heart rate (bpm).
- path (str. optional) If provided, the plot will be saved to the specified file.
- **show** (bool, optional) If True, show the plot immediately.

biosppy.plotting.plot_eda(ts=None, raw=None, filtered=None, onsets=None, peaks=None, amplitudes=None, path=None, show=False)

Create a summary plot from the output of signals.eda.eda.

Parameters:

- ts (array) Signal time axis reference (seconds).
- raw (array) Raw EDA signal.
- filtered (array) Filtered EDA signal.
- onsets (array) Indices of SCR pulse onsets.
- peaks (array) Indices of the SCR peaks.
- amplitudes (array) SCR pulse amplitudes.
- path (str, optional) If provided, the plot will be saved to the specified file.
- show (bool, optional) If True, show the plot immediately.

 $\begin{tabular}{l} \textbf{biosppy.plotting.plot}_\textbf{eeg} (ts=None, raw=None, filtered=None, labels=None, features_ts=None, theta=None, alpha_low=None, alpha_high=None, beta=None, gamma=None, plf_pairs=None, plf=None, path=None, show=False) \\ \end{tabular}$

Create a summary plot from the output of signals.eeg.eeg.

- raw (array) Raw EEG signal.
- filtered (array) Filtered EEG signal.
- labels (list) Channel labels.
- features_ts (array) Features time axis reference (seconds).
- theta (array) Average power in the 4 to 8 Hz frequency band; each column is one EEG channel.
- alpha_low (array) Average power in the 8 to 10 Hz frequency band; each column is one EEG channel.
- alpha_high (array) Average power in the 10 to 13 Hz frequency band; each column is one EEG channel.
- **beta** (*array*) Average power in the 13 to 25 Hz frequency band; each column is one EEG channel.
- gamma (array) Average power in the 25 to 40 Hz frequency band; each column is one EEG channel.
- plf_pairs (list) PLF pair indices.
- plf (array) PLF matrix; each column is a channel pair.
- path (str, optional) If provided, the plot will be saved to the specified file.
- **show** (bool, optional) If True, show the plot immediately.

biosppy.plotting.plot_emg(ts=None, sampling_rate=None, raw=None, filtered=None, onsets=None, processed=None, path=None, show=False)

Create a summary plot from the output of signals.emg.emg.

Parameters:

- ts (array) Signal time axis reference (seconds).
- sampling_rate (int, float) Sampling frequency (Hz).
- raw (array) Raw EMG signal.
- filtered (array) Filtered EMG signal.
- onsets (array) Indices of EMG pulse onsets.
- **processed** (*array*, *optional*) Processed EMG signal according to the chosen onset detector.
- path (str, optional) If provided, the plot will be saved to the specified file.
- **show** (bool, optional) If True, show the plot immediately.

biosppy.plotting.plot_filter(ftype='FIR', band='lowpass', order=None, frequency=None, sampling_rate=1000.0, path=None, show=True, **kwargs)

Plot the frequency response of the filter specified with the given parameters.

Filter type:

- Finite Impulse Response filter ('FIR');
- Butterworth filter ('butter');
- Chebyshev filters ('cheby1', 'cheby2');
- Elliptic filter ('ellip');
- Bessel filter ('bessel').
- band (str) -

Band type:

- Low-pass filter ('lowpass');
- o High-pass filter ('highpass');
- Band-pass filter ('bandpass');
- o Band-stop filter ('bandstop').
- order (int) Order of the filter.
- **frequency** (int, float, list, array) –

Cutoff frequencies; format depends on type of band:

- o 'lowpass' or 'bandpass': single frequency;
- o 'bandpass' or 'bandstop': pair of frequencies.
- sampling_rate (int, float, optional) Sampling frequency (Hz).
- path (str, optional) If provided, the plot will be saved to the specified file.
- show (bool, optional) If True, show the plot immediately.
- **kwargs (dict, optional) Additional keyword arguments are passed to the underlying scipy.signal function.

biosppy.plotting.plot_resp(ts=None, raw=None, filtered=None, zeros=None, resp_rate_ts=None, resp_rate=None, path=None, show=False)

Create a summary plot from the output of signals.bvp.bvp.

Parameters:

- ts (array) Signal time axis reference (seconds).
- raw (array) Raw Resp signal.
- filtered (array) Filtered Resp signal.
- zeros (array) Indices of Respiration zero crossings.
- resp_rate_ts (array) Respiration rate time axis reference (seconds).
- resp_rate (array) Instantaneous respiration rate (Hz).
- path (str, optional) If provided, the plot will be saved to the specified file.
- **show** (bool, optional) If True, show the plot immediately.

biosppy.plotting.plot_spectrum(signal=None, sampling_rate=1000.0, path=None, show=True)

Plot the power spectrum of a signal (one-sided).

- sampling_rate (int, float, optional) Sampling frequency (Hz).
- path (str, optional) If provided, the plot will be saved to the specified file.
- **show** (bool, optional) If True, show the plot immediately.

biosppy.storage

This module provides several data storage methods.

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class biosppy.storage.HDF(path=None, mode='a')

Bases: object

Wrapper class to operate on BioSPPy HDF5 files.

Parameters:

- path (str) Path to the HDF5 file.
- mode (str, optional) File mode; one of:
 - o 'a': read/write, creates file if it does not exist;
 - o 'r+': read/write, file must exist;
 - o 'r': read only, file must exist;
 - o 'w': create file, truncate if it already exists;
 - o 'w-': create file, fails if it already esists.

 ${\tt add_event} (\textit{ts=None}, \textit{values=None}, \textit{mdata=None}, \textit{group="}, \textit{name=None}, \textit{compress=False})$

Add an event to the file.

Parameters:

- ts (array) Array of time stamps.
- values (array, optional) Array with data for each time stamp.
- mdata (dict, optional) Event metadata.
- **group** (*str*, *optional*) Destination event group.
- name (str, optional) Name of the dataset to create.
- compress (bool, optional) If True, the data will be compressed with gzip.

Returns:

- group (str) Destination group.
- name (str) Name of the created event dataset.

Parameters: header (dict) – Header metadata.

add_signal(signal=None, mdata=None, group=", name=None, compress=False)

Add a signal to the file.

Parameters:

- signal (array) Signal to add.
- mdata (dict, optional) Signal metadata.
- group (str, optional) Destination signal group.
- name (str, optional) Name of the dataset to create.
- compress (bool, optional) If True, the signal will be compressed with gzip.

Returns:

- group (str) Destination group.
- name (str) Name of the created signal dataset.

close()

Close file descriptor.

del_event(group=", name=None)

Delete an event from the file.

Parameters:

- **group** (*str*, *optional*) Event group.
- name (str) Name of the event dataset.

del_event_group(group=")

Delete all events in a file group.

Parameters: str,optional (group) – Event group.

del_signal(group=", name=None)

Delete a signal from the file.

Parameters:

- group (str, optional) Signal group.
- name (str) Name of the dataset.

del_signal_group(group=")

Delete all signals in a file group.

Retrieve an event from the file.

Parameters:

- group (str, optional) Event group.
- name (str) Name of the event dataset.

Returns:

- ts (array) Array of time stamps.
- values (array) Array with data for each time stamp.
- mdata (dict) Event metadata.

Notes

Loads the entire event data into memory.

get_header()

Retrieve header metadata.

Returns: header (dict) – Header metadata.

get_signal(group=", name=None)

Retrieve a signal from the file.

Parameters:

- group (str, optional) Signal group.
- name (str) Name of the signal dataset.

Returns:

- signal (array) Retrieved signal.
- mdata (dict) Signal metadata.

Notes

• Loads the entire signal data into memory.

list_events(group=", recursive=False)

List events in the file.

Parameters:

- group (str, optional) Event group.
- recursive (bool, optional) If True, also lists events in sub-groups.

Returns: events (*list*) – List of (group, name) tuples of the found events.

list_signals(group=", recursive=False)

23 of 29 List signals in the file.

• recursive (bool, optional) – If True, also lists signals in sub-groups.

Returns: signals (list) – List of (group, name) tuples of the found signals.

biosppy.storage.alloc_h5(path)

Prepare an HDF5 file.

Parameters: path (str) – Path to file.

biosppy.storage.deserialize(path)

Deserialize data from a file using sklearn's joblib.

Parameters: path (str) – Source path.

Returns: data (object) – Deserialized object.

biosppy.storage.dumpJSON(data, path)

Save JSON data to a file.

Parameters: • data (dict) – The JSON data to dump.

• path (str) - Destination path.

biosppy.storage.loadJSON(path)

Load JSON data from a file.

Parameters: path (str) – Source path.

Returns: data (dict) – The loaded JSON data.

biosppy.storage.load_h5(path, label)

Load data from an HDF5 file.

Parameters: • path (str) – Path to file.

• label (hashable) - Data label.

Returns: data (array) – Loaded data.

biosppy.storage.load_txt(path)

Returns:

- data (array) Loaded data.
- mdata (dict) Metadata.

biosppy.storage.pack_zip(files, path, recursive=True, forceExt=True)

Pack files into a zip archive.

Parameters:

- files (iterable) List of files or directories to pack.
- path (str) Destination path.
- **recursive** (*bool*, *optional*) If True, sub-directories and sub-folders are also written to the archive.
- forceExt (bool, optional) Append default extension.

Returns: zip_path (*str*) – Full path to created zip archive.

biosppy.storage.serialize(data, path, compress=3)

Serialize data and save to a file using sklearn's joblib.

Parameters:

- data (object) Object to serialize.
- path (str) Destination path.
- **compress** (*int*, *optional*) Compression level; from 0 to 9 (highest compression).

biosppy.storage.store_h5(path, label, data)

Store data to HDF5 file.

Parameters:

- path (str) Path to file.
- label (hashable) Data label.
- data (array) Data to store.

biosppy.storage.store_txt(path, data, sampling_rate=1000.0, resolution=None, date=None, labels=None, precision=6)

Store data to a simple text file.

Parameters:

- path (str) Path to file.
- data (array) Data to store (up to 2 dimensions).
- sampling_rate (int, float, optional) Sampling frequency (Hz).
- resolution (int, optional) Sampling resolution.
- date (*datetime*, *str*, *optional*) Datetime object, or an ISO 8601 formatted datetime string.
- labels (list, optional) Labels for each column of data.
- 05/04/20, 11:39 am

ValueError - If the number of labels is inconsistent with the data.

biosppy.storage.unpack_zip(zip_path, path)

Unpack a zip archive.

Parameters:

- **zip_path** (*str*) Path to zip archive.
- path (str) Destination path (directory).

biosppy.storage.zip_write(fid, files, recursive=True, root=None)

Write files to zip archive.

Parameters:

- fid (file-like object) The zip file to write into.
- files (iterable) List of files or directories to pack.
- **recursive** (*bool*, *optional*) If True, sub-directories and sub-folders are also written to the archive.
- root (str, optional) Relative folder path.

Notes

• Ignores non-existent files and directories.

biosppy.timing

This module provides simple methods to measure computation times.

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biosppy.timing.clear(name=None)

Clear the clock.

Parameters: name (str, optional) – Name of the clock; if None, uses the default name.

biosppy.timing.clear_all()

Clear all clocks.

biosppy.timing.tac(name=None)

Returns: delta (float) – Elapsed time, in seconds.

Raises: KeyError if the name of the clock is unknown.

biosppy.timing.tic(name=None)

Start the clock.

Parameters: name (str, optional) – Name of the clock; if None, uses the default name.

biosppy.utils

This module provides several frequently used functions and hacks.

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class biosppy.utils.ReturnTuple(values, names=None)

Bases: tuple

A named tuple to use as a hybrid tuple-dict return object.

Parameters: • values (iterable) – Return values.

• names (iterable, optional) - Names for return values.

Raises: • valueError – If the number of values differs from the number of names.

• **ValueError** – If any of the items in names: * contain non-alphanumeric characters; * are Python keywords; * start with a number; * are duplicates.

as_dict()

Convert to an ordered dictionary.

Returns: out (OrderedDict) – An OrderedDict representing the return values.

keys()

Return the value names.

Returns: out (*list*) – The keys in the mapping.

Parameters: path (str) – Input file path.

Returns: • dirname (str) – File directory.

fname (str) - File name.ext (str) - File extension.

Notes

• Removes the dot (") from the extension.

biosppy.utils.fullfile(*args)

Join one or more file path components, assuming the last is the extension.

Parameters: *args (list, optional) – Components to concatenate.

Returns: fpath (*str*) – The concatenated file path.

biosppy.utils.highestAveragesAllocator(votes, k, divisor='dHondt', check=False)

Allocate k seats proportionally using the Highest Averages Method.

Parameters: • votes (list) – Number of votes for each class/party/cardinal.

• k (int) - Total number o seats to allocate.

• **divisor** (*str, optional*) – Divisor method; one of 'dHondt', 'Huntington-Hill', 'Sainte-Lague', 'Imperiali', or 'Danish'.

• **check** (*bool*, *optional*) – If True, limits the number of seats to the total number of votes.

Returns: seats (list) – Number of seats for each class/party/cardinal.

biosppy.utils.normpath(path)

Normalize a path.

Parameters: path (str) – The path to normalize.

Returns: npath (*str*) – The normalized path.

biosppy.utils.random_fraction(indx, fraction, sort=True)

Select a random fraction of an input list of elements.

Parameters: • indx (list, array) – Elements to partition.

- fraction (int, float) Fraction to select.
- sort (bool, optional) If True, output lists will be sorted.

- use (list, array) Selected elements.
- unuse (list, array) Remaining elements.

biosppy.utils.remainderAllocator(votes, k, reverse=True, check=False)

Allocate k seats proportionally using the Remainder Method.

Also known as Hare-Niemeyer Method. Uses the Hare quota.

Parameters:

- votes (list) Number of votes for each class/party/cardinal.
- k (int) Total number o seats to allocate.
- reverse (bool, optional) If True, allocates remaining seats largest quota first.
- **check** (*bool*, *optional*) If True, limits the number of seats to the total number of votes.

Returns: seats (list) – Number of seats for each class/party/cardinal.

biosppy.utils.walktree(top=None, spec=None)

Iterator to recursively descend a directory and return all files matching the spec.

Parameters:

- **top** (*str*, *optional*) Starting directory; if None, defaults to the current working directoty.
- spec (str, optional) -

Regular expression to match the desired files; if None, matches all files; typical patterns:

- o r'.txt\$' matches files with '.txt' extension;
- o r'^File_' matches files starting with 'File_'
- o *r'^File_.+.txt\$'* matches files starting with 'File_' and ending with the '.txt' extension.

Yields: fpath (*str*) – Absolute file path.

Notes

• Partial matches are also selected.

See also

- https://docs.python.org/3/library/re.html
- https://regex101.com/