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```
.. module:: BiometricWizard
:platform: Unix, Windows
:synopsis: This module provides various functions to...
.. moduleauthor:: Carlos Carreiras
```

Modules

[copy](#)
[numpy](#)

[os](#)
[matplotlib.pyplot](#)

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Functions

Classify(expConfig, run, clf_name, subjects)

Cluster(tasks, parameters, case)

DimReduction(expConfig, testData, trainData, subjects)

FeatureExtraction(tasks, parameters, case)

FeatureSelection(tasks, parameters, case)

Filter(tasks, parameters, case)

Load(expConfig, store)

Load2Clf(subjects, case, run)

Main(expConfig, expID)

MongoReport(expConfig, nbRuns)

Outlier(tasks, parameters, case)

PopulationStatistics(nbRuns, subjects, key=None)

Prepare(tasks, expConfig, case, run)

Results(expConfig, run, clf_name)

Segment(tasks, parameters, case)

TemplateSelection(tasks, parameters, case)

Visualize(expConfig, run, clf_name)

blockPrepare(data, normalization=None, subpattern=None, number_mean_waves=None, number_median_waves=None, quantization=None, patterns2strings=None, nbTemplates=None, manList=None, subject=None)

concatenate(data, axis, vec)

features(data, normalization=None, subpattern=None, number_mean_waves=None, number_median_waves=None, quantization=None, patterns2strings=None)

leadLoader(expConfig, subjects, lbl2sub, subjectDict, xtrainRecs, xtestRecs, commonRecs)

```
leadTaskLoader(dbParams, lock, tasker, path, subject, trainSR, testSR, commonSR, data_name, data_type,  
run_setup, concatenate_records=False, concatenate_signals=True, axis=0)

loocvLoader(expConfig, subjects, lbl2sub, subjectDict, xtrainRecs, xtestRecs, commonRecs)

loocvTaskLoader(dbParams, lock, tasker, path, subject, trainSR, testSR, commonSR, data_name, data_type,  
run_setup, concatenate_records=False, concatenate_signals=True, axis=0)

popStatsHelper(data, key=None)

prepareTemplates(tasks, src=None, subject=None, manList=None, normalization=None, subpattern=None,  
number_mean_waves=None, number_median_waves=None, quantization=None, patterns2strings=None,  
nbTemplates=None)

selector(step)

simpleLeadLoader(expConfig, subjects, lbl2sub, subjectDict, xtrainRecs, xtestRecs, commonRecs)

simpleLeadTaskLoader(dbParams, lock, tasker, path, subject, trainSR, testSR, commonSR, data_name, data_type,  
run_setup, concatenate_records=False, concatenate_signals=True, axis=0)

simpleLoader(expConfig, subjects, lbl2sub, subjectDict, xtrainRecs, xtestRecs, commonRecs)

simpleTaskLoader(dbParams, lock, tasker, path, subject, trainSR, testSR, commonSR, data_name, data_type,  
run_setup, concatenate_records=False, concatenate_signals=True, axis=0)

visualizeSubject(thresholds, rates, filePath)

windowLoader(expConfig, subjects, lbl2sub, subjectDict, xtrainRecs, xtestRecs, commonRecs)
```

Data

```
CLF = ['SVM', 'KNN', 'Agrafioti', 'Dissimilarity', 'DissimilaritySimple', 'DissimilarityMore']
```

```
.. module:: classifiers
:platform: Unix, Windows
:synopsis: This module provides various functions to...

.. moduleauthor:: Filipe Canento, Carlos Carreiras
```

Modules

[scipy.spatial.distance](#)

[numpy](#)

[os](#)

[shutil](#)

[sklearn.svm](#)

[warnings](#)

Classes

[__builtin__.object](#)

[BaseClassifier](#)

[KNN](#)

[Agrafioti](#)
[Dissimilarity](#)

[DissimilaritySimple](#)

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[Odinaka](#)
[SVM](#)

[ClusterSVM](#)
[LeanSVM](#)

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[ClassifierDict](#)
[fisher](#)

[bidict.bidict\(__builtin__.object\)](#)

[SubjectDict](#)

[exceptions.Exception\(exceptions.BaseException\)](#)

[EmptyCombination](#)
[UnknownSubjectError](#)
[UntrainedError](#)

class **Agrafioti(KNN)**

Class to train and classify AC coefficients from ECG Signals

Performs LDA if coefficient number is bigger than subject number

Method resolution order:

[Agrafioti](#)
[KNN](#)
[BaseClassifier](#)
[__builtin__.object](#)

Methods defined here:

[__init__\(self, io=None, k=3, metric='euclidean', h=4.0, M=100.0, w=None, **kwargs\)](#)

[autoRejectionThresholds\(self\)](#)

[re_train\(self, data\)](#)

[train\(self, data=None, updateThresholds=True\)](#)

Data and other attributes defined here:

EER_IDX = 0

EXT = '.hdf5'

NAME = 'agrafiotiClassifier'

Methods inherited from [KNN](#):

authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)

Methods inherited from [BaseClassifier](#):

authThreshold(self, subject, ready=False)

checkSubject(self, subject)

dirIterator(self)

evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:

For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.

rejection_thresholds (array): Thresholds used to compute the ROCs.

dstPath (string): Path for multiprocessing.

log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

classification (dict): Results of the classification.

assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:

... [1]

fileIterator(self)

idThreshold(self, subject, ready=False)

identify(self, data, threshold=None, ready=False, **kwargs)

io_load(self, label)

io_save(self, label, data)

listSubjects(self)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:

For each test subject and for each threshold, test authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.

rejection_thresholds (array): Thresholds used to compute the ROCs.

dstPath (string): Path for multiprocessing.

```
    log2file (bool): Flag to control the use of logging in multiprocessing.

    Kwargs:
        classification (dict): Results of the classification.

        assessment (dict): Biometric statistics.

    See Also:

    Notes:

    Example:

    References:
        ... [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)
```

Class methods inherited from [BaseClassifier](#):

load(cls, srcPath, dstPath=None) from [builtin.type](#)

Data descriptors inherited from [BaseClassifier](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class **BaseClassifier**([builtin.object](#))

Base Classifier.

Methods defined here:

```
__init__(self, io=None)

authThreshold(self, subject, ready=False)

authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)

autoRejectionThresholds(self)

checkSubject(self, subject)

dirIterator(self)

evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)
    Assess the performance of the classifier in both biometric scenarios: authentication and identification.
```

Workflow:
For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:
data (dict): Dictionary holding the testing samples for each subject.

rejection_thresholds (array): Thresholds used to compute the ROCs.

dstPath (string): Path for multiprocessing.

log2file (bool): Flag to control the use of logging in multiprocessing.

Kwargs:
classification (dict): Results of the classification.

assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
.. [1]

fileIterator(self)

idThreshold(self, subject, ready=False)

identify(self, data, threshold=None, ready=False, **kwargs)

io_load(self, label)

io_save(self, label, data)

listSubjects(self)

re_train(self, data)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:

For each test subject and for each threshold, test authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.
rejection_thresholds (array): Thresholds used to compute the ROCs.
dstPath (string): Path for multiprocessing.
log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

classification (dict): Results of the classification.
assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
.. [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

train(self, data=None, updateThresholds=True)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)

Class methods defined here:

load(cls, srcPath, dstPath=None) from [_builtin_.type](#)

Data descriptors defined here:

_dict
dictionary for instance variables (if defined)

_weakref
list of weak references to the object (if defined)

Data and other attributes defined here:

EER_IDX = 0

EXT = '.hdf5'

NAME = 'BaseClassifier'

class **ClassifierDict**([__builtin__.object](#))

Class to handle hierarchical subject - label - classifier disambiguation.

Methods defined here:

__init__(self)

assign(self, n, label)

getClassifiers(self, label)

getLabel(self, clf)

getLabelWeight(self, label)

getPairWeight(self, pair)

Data descriptors defined here:

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class **ClusterSVM**([SVM](#))

Perform clustering on the training data.

Method resolution order:

[ClusterSVM](#)
[SVM](#)
[BaseClassifier](#)
[__builtin__.object](#)

Methods defined here:

__init__(self, io=None, nensemble=100, partMethod='average', **kwargs)

authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)

re_train(self, data)

train(self, data=None, updateThresholds=True)

Data and other attributes defined here:

EER_IDX = -1

EXT = '.hdf5'

NAME = 'clusterSvmClassifier'

Methods inherited from [SVM](#):

autoRejectionThresholds(self)

Methods inherited from [BaseClassifier](#):

authThreshold(self, subject, ready=False)

checkSubject(self, subject)

dirIterator(self)

evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:
For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:
data (dict): Dictionary holding the testing samples for each subject.
rejection_thresholds (array): Thresholds used to compute the ROCs.
dstPath (string): Path for multiprocessing.
log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:
classification (dict): Results of the classification.
assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
.. [1]

fileIterator(self)

idThreshold(self, subject, ready=False)

identify(self, data, threshold=None, ready=False, **kwargs)

io_load(self, label)

io_save(self, label, data)

listSubjects(self)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:
For each test subject and for each threshold, test authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:
data (dict): Dictionary holding the testing samples for each subject.
rejection_thresholds (array): Thresholds used to compute the ROCs.
dstPath (string): Path for multiprocessing.
log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:
classification (dict): Results of the classification.
assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
.. [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)

Class methods inherited from [BaseClassifier](#):

load(cls, srcPath, dstPath=None) from [__builtin__.type](#)

Data descriptors inherited from [BaseClassifier](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class Dissimilarity([KNN](#))

Class to extract multilead distances subjectwise

Performs kNN for classification based on those distances

Method resolution order:

[Dissimilarity](#)
[KNN](#)
[BaseClassifier](#)
[__builtin__.object](#)

Methods defined here:

__init__(self, io=None, k=3, metric='euclidean', featmetric='euclidean', leads=(I, II, III), reflead='I', median=False, **kwargs)

autoRejectionThresholds(self)

dict2list(self, dictionary)

Gets a dictionary as input and returns a list

Needed so as to join all average ECGs leadwise and feed them to `scipy.spatial.distance.cdist()`

extract_feats(self, data, label=None)

data - dictionary whose entries are discriminated by leads; entries have a matrix of ECG segments

re_train(self, data)

train(self, data=None, updateThresholds=True)

Data and other attributes defined here:

EER_IDX = 0

EXT = '.hdf5'

NAME = 'dissimilarityClassifier'

Methods inherited from [KNN](#):

authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)

Methods inherited from [BaseClassifier](#):

authThreshold(self, subject, ready=False)

checkSubject(self, subject)

dirIterator(self)

evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:

For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by `evaluation.assessClassification`.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.

rejection_thresholds (array): Thresholds used to compute the ROCs.

dstPath (string): Path for multiprocessing.

```

log2file (bool): Flag to control the use of logging in multiprocessing.

Kwargs:
    classification (dict): Results of the classification.

    assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
    ... [1]

fileIterator(self)

idThreshold(self, subject, ready=False)

identify(self, data, threshold=None, ready=False, **kwargs)

io_load(self, label)

io_save(self, label, data)

listSubjects(self)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)
    Assess the performance of the classifier in both biometric scenarios: authentication and identification.

    Workflow:
        For each test subject and for each threshold, test authentication and identification;
        Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
        Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
        Subject and global statistics are then computed by evaluation.assessClassification.

    Kwargs:
        data (dict): Dictionary holding the testing samples for each subject.

        rejection_thresholds (array): Thresholds used to compute the ROCs.

        dstPath (string): Path for multiprocessing.

        log2file (bool): Flag to control the use of logging in multiprocessing.

    Kwargs:
        classification (dict): Results of the classification.

        assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
    ... [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)



---



Class methods inherited from BaseClassifier:



load(cls, srcPath, dstPath=None) from \_builtin\_.type



---



Data descriptors inherited from BaseClassifier:



__dict__  
dictionary for instance variables (if defined)


```

weakref

list of weak references to the object (if defined)

class DissimilarityMore([KNN](#))

Class to extract multilead distances subjectwise

Performs kNN for classification based on those distances

Method resolution order:

[DissimilarityMore](#)
[KNN](#)
[BaseClassifier](#)
[builtin__object](#)

Methods defined here:

`__init__(self, io=None, k=3, metric='euclidean', featmetric='euclidean', leads=(I, II, III), reflead=I, median=False, meansubs=None, percent=0.15, tn=5, **kwargs)`

`autoRejectionThresholds(self)`

`dict2list(self, dictionary)`

Gets a dictionary as input and returns a list

Needed so as to join all average ECGs leadwise and feed them to `scipy.spatial.distance.cdist()`

`extract_feats(self, data, label=None)`

data - dictionary whose entries are discriminated by leads; entries have a matrix of ECG segments

`re_train(self, data)`

`train(self, data=None, updateThresholds=True)`

Data and other attributes defined here:

EER_IDX = 0

EXT = '.hdf5'

NAME = 'dissimilarityClassifier'

Methods inherited from [KNN](#):

`authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)`

Methods inherited from [BaseClassifier](#):

`authThreshold(self, subject, ready=False)`

`checkSubject(self, subject)`

`dirIterator(self)`

`evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)`

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

Workflow:

For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by `evaluation.assessClassification`.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.

rejection_thresholds (array): Thresholds used to compute the ROCs.

dstPath (string): Path for multiprocessing.

log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

classification (dict): Results of the classification.

assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
... [1]

fileIterator(self)

idThreshold(self, subject, ready=False)

identify(self, data, threshold=None, ready=False, **kwargs)

io_load(self, label)

io_save(self, label, data)

listSubjects(self)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:

For each test subject and for each threshold, test authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.

rejection_thresholds (array): Thresholds used to compute the ROCs.

dstPath (string): Path for multiprocessing.

log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

classification (dict): Results of the classification.

assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
... [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)

Class methods inherited from [BaseClassifier](#):

load(cls, srcPath, dstPath=None) from [builtins.type](#)

Data descriptors inherited from [BaseClassifier](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class [DissimilaritySimple\(Dissimilarity\)](#)

Class to extract multilead distances subjectwise

Performs kNN for classification based on those distances

Method resolution order:

[DissimilaritySimple](#)
[Dissimilarity](#)
[KNN](#)
[BaseClassifier](#)
[_builtin__object](#)

Methods defined here:

`__init__(self, io=None, k=3, metric='euclidean', featmetric='euclidean', leads=(I', 'II', 'III'), reflead='I', median=False, tn=5, **kwargs)`
`extract_feats(self, data, label=None)`
 data - dictionary whose entries are discriminated by leads; entries have a matrix of ECG segments
`re_train(self, data)`
`train(self, data=None, updateThresholds=True)`

Data and other attributes defined here:

EER_IDX = 0
EXT = '.hdf5'
NAME = 'dissimilaritySimpleClassifier'

Methods inherited from [Dissimilarity](#):

`autoRejectionThresholds(self)`
`dict2list(self, dictionary)`
 Gets a dictionary as input and returns a list
 Needed so as to join all average ECGs leadwise and feed them to `scipy.spatial.distance.cdist()`

Methods inherited from [KNN](#):

`authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)`

Methods inherited from [BaseClassifier](#):

`authThreshold(self, subject, ready=False)`
`checkSubject(self, subject)`
`dirIterator(self)`
`evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)`
 Assess the performance of the classifier in both biometric scenarios: authentication and identification.

Workflow:
For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by `evaluation.assessClassification`.

Kwargs:
 data (dict): Dictionary holding the testing samples for each subject.
 rejection_thresholds (array): Thresholds used to compute the ROCs.
 dstPath (string): Path for multiprocessing.
 log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:
 classification (dict): Results of the classification.
 assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
.. [1]

fileIterator(self)

idThreshold(self, subject, ready=False)

identify(self, data, threshold=None, ready=False, **kwargs)

io_load(self, label)

io_save(self, label, data)

listSubjects(self)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)
Assess the performance of the classifier in both biometric scenarios: authentication and identification.

Workflow:
For each test subject and for each threshold, test authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:
data (dict): Dictionary holding the testing samples for each subject.
rejection_thresholds (array): Thresholds used to compute the ROCs.
dstPath (string): Path for multiprocessing.
log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:
classification (dict): Results of the classification.
assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
.. [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)

Class methods inherited from [BaseClassifier](#):

load(cls, srcPath, dstPath=None) from [_builtin_.type](#)

Data descriptors inherited from [BaseClassifier](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class DissimilaritySimple_old([Dissimilarity_old](#))
Class to extract multilead distances subjectwise
Performs kNN for classification based on those distances

Method resolution order:

[DissimilaritySimple_old](#)
[Dissimilarity_old](#)
[KNN](#)

[BaseClassifier](#)
[builtin_object](#)

Methods defined here:

```
_init_(self, io=None, k=3, metric='euclidean', featmetric='euclidean', leads=(I, II, III), reflead=I, **kwargs)  
extract_feats(self, data, label=None)  
    data - dictionary whose entries are discriminated by leads; entries have a matrix of ECG segments  
re_train(self, data)  
train(self, data=None, updateThresholds=True)
```

Data and other attributes defined here:

```
EER_IDX = 0  
EXT = '.hdf5'  
NAME = 'dissimilaritySimpleClassifier'
```

Methods inherited from [Dissimilarity_old](#):

```
autoRejectionThresholds(self)  
dict2list(self, dictionary)  
    Gets a dictionary as input and returns a list  
    Needed so as to join all average ECGs leadwise and feed them to scipy.spatial.distance.cdist()
```

Methods inherited from [KNN](#):

```
authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)
```

Methods inherited from [BaseClassifier](#):

```
authThreshold(self, subject, ready=False)  
checkSubject(self, subject)  
dirIterator(self)
```

```
evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)  
    Assess the performance of the classifier in both biometric scenarios: authentication and identification.
```

Workflow:
For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:
data (dict): Dictionary holding the testing samples for each subject.
rejection_thresholds (array): Thresholds used to compute the ROCs.
dstPath (string): Path for multiprocessing.
log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:
classification (dict): Results of the classification.
assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
... [1]

fileIterator(self)

```
idThreshold(self, subject, ready=False)
```

```

identify(self, data, threshold=None, ready=False, **kwargs)

io_load(self, label)

io_save(self, label, data)

listSubjects(self)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)
    Assess the performance of the classifier in both biometric scenarios: authentication and identification.

    Workflow:
        For each test subject and for each threshold, test authentication and identification;
        Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
        Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
        Subject and global statistics are then computed by evaluation.assessClassification.

    Kwargs:
        data (dict): Dictionary holding the testing samples for each subject.

        rejection_thresholds (array): Thresholds used to compute the ROCs.

        dstPath (string): Path for multiprocessing.

        log2file (bool): Flag to control the use of logging in multiprocessing.

    Kwargs:
        classification (dict): Results of the classification.

        assessment (dict): Biometric statistics.

    See Also:

    Notes:

    Example:

    References:
        .. [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)



---


Class methods inherited from BaseClassifier:

```

load(cls, srcPath, dstPath=None) from [__builtin__.type](#)

Data descriptors inherited from [BaseClassifier](#):

— dict	dictionary for instance variables (if defined)
— weakref	list of weak references to the object (if defined)

class Dissimilarity_old([KNN](#))

Class to extract multilead distances subjectwise

Performs kNN for classification based on those distances

Method resolution order:

[Dissimilarity_old](#)
[KNN](#)
[BaseClassifier](#)
[__builtin__.object](#)

Methods defined here:

__init__(self, io=None, k=3, metric='euclidean', featmetric='euclidean', leads=(I, II, III), reflead=I, **kwargs)

```
autoRejectionThresholds(self)

dict2list(self, dictionary)
    Gets a dictionary as input and returns a list
        Needed so as to join all average ECGs leadwise and feed them to scipy.spatial.distance.cdist()

extract_feats(self, data, label=None)
    data - dictionary whose entries are discriminated by leads; entries have a matrix of ECG segments

re_train(self, data)

train(self, data=None, updateThresholds=True)
```

Data and other attributes defined here:

```
EER_IDX = 0

EXT = '.hdf5'

NAME = 'dissimilarityClassifier'
```

Methods inherited from [KNN](#):

```
authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)
```

Methods inherited from [BaseClassifier](#):

```
authThreshold(self, subject, ready=False)

checkSubject(self, subject)

dirIterator(self)
```

```
evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)
    Assess the performance of the classifier in both biometric scenarios: authentication and identification.
```

Workflow:
For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by `evaluation.assessClassification`.

Kwargs:
data (dict): Dictionary holding the testing samples for each subject.
rejection_thresholds (array): Thresholds used to compute the ROCs.
dstPath (string): Path for multiprocessing.
log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:
classification (dict): Results of the classification.
assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
... [1]

```
fileIterator(self)
```

```
idThreshold(self, subject, ready=False)
```

```
identify(self, data, threshold=None, ready=False, **kwargs)
```

```
io_load(self, label)
```

```
io_save(self, label, data)
```

```
listSubjects(self)
```

```
save(self, dstPath)
```

```
seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)
    Assess the performance of the classifier in both biometric scenarios: authentication and identification.

    Workflow:
        For each test subject and for each threshold, test authentication and identification;
        Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
        Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
        Subject and global statistics are then computed by evaluation.assessClassification.

    Kwargs:
        data (dict): Dictionary holding the testing samples for each subject.

        rejection_thresholds (array): Thresholds used to compute the ROCs.

        dstPath (string): Path for multiprocessing.

        log2file (bool): Flag to control the use of logging in multiprocessing.

    Kwargs:
        classification (dict): Results of the classification.

        assessment (dict): Biometric statistics.

    See Also:

    Notes:

    Example:

    References:
        ... [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)

Class methods inherited from BaseClassifier:
load(cls, srcPath, dstPath=None) from \_\_builtin\_\_.type
```

Data descriptors inherited from [BaseClassifier](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class [EmptyCombination](#)([exceptions.Exception](#))

Method resolution order:

[EmptyCombination](#)
[exceptions.Exception](#)
[exceptions.BaseException](#)
[__builtin__.object](#)

Methods defined here:

__str__(self)

Data descriptors defined here:

__weakref__
list of weak references to the object (if defined)

Methods inherited from [exceptions.Exception](#):

__init__(...)
x.[__init__](#)(...) initializes x; see help(type(x)) for signature

Data and other attributes inherited from [exceptions.Exception](#):

__new__ = <built-in method `__new__` of type object>
T. `__new__(S, ...)` -> a new `object` with type S, a subtype of T

Methods inherited from [exceptions.BaseException](#):

__delattr__(...)
x. `__delattr__('name')` <==> del x.name

__getattribute__(...)
x. `__getattribute__('name')` <==> x.name

__getitem__(...)
x. `__getitem__(y)` <==> x[y]

__getslice__(...)
x. `__getslice__(i, j)` <==> x[i:j]
Use of negative indices is not supported.

__reduce__(...)

__repr__(...)
x. `__repr__()` <==> repr(x)

__setattr__(...)
x. `__setattr__('name', value)` <==> x.name = value

__setstate__(...)

__unicode__(...)

Data descriptors inherited from [exceptions.BaseException](#):

__dict__
args
message

class KNN([BaseClassifier](#))

k-NN classifier.

Method resolution order:

[KNN](#)
[BaseClassifier](#)
[builtin.object](#)

Methods defined here:

`__init__(self, io=None, k=3, metric='euclidean', **kwargs)`
`authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)`
`autoRejectionThresholds(self)`
`re_train(self, data)`
`train(self, data=None, updateThresholds=True)`

Data and other attributes defined here:

EER_IDX = 0
EXT = '.hdf5'
NAME = 'KNN'

Methods inherited from [BaseClassifier](#):

`authThreshold(self, subject, ready=False)`
`checkSubject(self, subject)`
`dirIterator(self)`
`evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)`

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

Workflow:

For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification; Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples); Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples); Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.
rejection_thresholds (array): Thresholds used to compute the ROCs.
dstPath (string): Path for multiprocessing.
log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

classification (dict): Results of the classification.
assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
.. [1]

fileIterator(self)

idThreshold(self, subject, ready=False)

identify(self, data, threshold=None, ready=False, **kwargs)

io_load(self, label)

io_save(self, label, data)

listSubjects(self)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:

For each test subject and for each threshold, test authentication and identification; Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples); Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples); Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.
rejection_thresholds (array): Thresholds used to compute the ROCs.
dstPath (string): Path for multiprocessing.
log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

classification (dict): Results of the classification.
assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
.. [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

```
updateThresholds_old(self, overwrite=False, N=1)
```

Class methods inherited from [BaseClassifier](#):

```
load(cls, srcPath, dstPath=None) from \_\_builtin\_\_.type
```

Data descriptors inherited from [BaseClassifier](#):

__dict__
 dictionary for instance variables (if defined)

__weakref__
 list of weak references to the object (if defined)

class [LeanSVM\(SVM\)](#)

Lean [SVM](#) classifier: the individual [SVM](#) classifiers are saved as .dict files.

Method resolution order:

```
LeanSVM
SVM
BaseClassifier
\_\_builtin\_\_.object
```

Methods defined here:

```
__init__(self, io=None, **kwargs)
```

```
dirIterator(self)
```

```
fileIterator(self)
```

Data and other attributes defined here:

```
EXT = '.hdf5'
```

```
NAME = 'leanSvmClassifier'
```

Methods inherited from [SVM](#):

```
authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)
```

```
autoRejectionThresholds(self)
```

```
re_train(self, data)
```

```
train(self, data=None, updateThresholds=True)
```

Data and other attributes inherited from [SVM](#):

```
EER_IDX = -1
```

Methods inherited from [BaseClassifier](#):

```
authThreshold(self, subject, ready=False)
```

```
checkSubject(self, subject)
```

```
evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)
```

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

Workflow:

For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification; Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples); Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples); Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.

rejection_thresholds (array): Thresholds used to compute the ROCs.

dstPath (string): Path for multiprocessing.

log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

```
classification (dict): Results of the classification.  
assessment (dict): Biometric statistics.
```

See Also:

Notes:

Example:

References:
.. [1]

idThreshold(self, subject, ready=False)

identify(self, data, threshold=None, ready=False, **kwargs)

io_load(self, label)

io_save(self, label, data)

listSubjects(self)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:

For each test subject and for each threshold, test authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.
rejection_thresholds (array): Thresholds used to compute the ROCs.
dstPath (string): Path for multiprocessing.
log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

classification (dict): Results of the classification.
assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
.. [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)

Class methods inherited from [BaseClassifier](#):

load(cls, srcPath, dstPath=None) from [builtin.type](#)

Data descriptors inherited from [BaseClassifier](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

```
class Odinaka(BaseClassifier)
    Odinaka classifier.
```

Method resolution order:

```
Odinaka
BaseClassifier
__builtin__.object
```

Methods defined here:

```
__init__(self, io=None, n=None, overlap=0, nfft=None, size=None, window='hamming', k=0.0)
authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)
autoRejectionThresholds(self)

identify(self, data, ready=False, **kwargs)
re_train(self, data)
train(self, data=None)
```

Data and other attributes defined here:

```
EXT = '.hdf5'
```

```
NAME = 'odinakaClassifier'
```

Methods inherited from [BaseClassifier](#):

```
authThreshold(self, subject, ready=False)
```

```
checkSubject(self, subject)
```

```
dirIterator(self)
```

```
evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)
```

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

Workflow:

For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification; Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples); Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples); Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.

rejection_thresholds (array): Thresholds used to compute the ROCs.

dstPath (string): Path for multiprocessing.

log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

classification (dict): Results of the classification.

assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:

... [1]

```
fileIterator(self)
```

```
idThreshold(self, subject, ready=False)
```

```
io_load(self, label)
```

```
io_save(self, label, data)
```

```
listSubjects(self)
```

```
save(self, dstPath)
```

```
seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)
    Assess the performance of the classifier in both biometric scenarios: authentication and identification.

    Workflow:
        For each test subject and for each threshold, test authentication and identification;
        Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
        Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
        Subject and global statistics are then computed by evaluation.assessClassification.
```

Kwargs:

- data (dict): Dictionary holding the testing samples for each subject.
- rejection_thresholds (array): Thresholds used to compute the ROCs.
- dstPath (string): Path for multiprocessing.
- log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

- classification (dict): Results of the classification.
- assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:

- .. [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)

Class methods inherited from [BaseClassifier](#):

load(cls, srcPath, dstPath=None) from [builtin.type](#)

Data descriptors inherited from [BaseClassifier](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

Data and other attributes inherited from [BaseClassifier](#):

EER_IDX = 0

```
class SVM(BaseClassifier)
    SVM classifier.
```

Method resolution order:

[SVM](#)
[BaseClassifier](#)
[builtin.object](#)

Methods defined here:

```
__init__(self, io=None, **kwargs)
    Wrapper for the scikit-learn SVM classifier, using OneClassSVM, SVC or LinearSVC,
    depending on usage case.

    sklearn.svm.SVC kwargs:
        C (float): Penalty parameter C of the error term. Default=1.0
        kernel (string): The kernel to be used ('linear', 'poly', 'rbf', 'sigmoid'). Default='rbf'
        degree (int): Degree of kernel function; only significant for 'poly' and 'sigmoid'. Default=3
```

```
gamma (float): Kernel coefficient for 'rbf' and 'poly'. Default=0.0
coef0 (float): Independent term in kernel function; it is only significant in 'poly' and 'sigmoid'. Default=0.0
probability (bool): Whether to enable probability estimates. Default=False
shrinking (bool): Whether to use the shrinking heuristic. Default=True
tol (float): Tolerance for stopping criterion. Default=1e-3
cache_size (float): Specify the size of the kernel cache (in MB).
class_weight (dict, 'auto'): Weights for each class; the 'auto' mode automatically adjusts the
                           weights inversely proportional to class frequencies.

verbose (bool): Enable verbose output. Default=False
max_iter (int): Hard limit on iterations within solver, or -1 for no limit. Default=-1

authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)

autoRejectionThresholds(self)

re_train(self, data)

train(self, data=None, updateThresholds=True)
```

Data and other attributes defined here:

EER_IDX = -1

EXT = '.hdf5'

NAME = 'svmClassifier'

Methods inherited from [BaseClassifier](#):

authThreshold(self, subject, ready=False)

checkSubject(self, subject)

dirIterator(self)

evaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:

For each test subject and for each threshold, create a multiprocessing task that tests authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.

rejection_thresholds (array): Thresholds used to compute the ROCs.

dstPath (string): Path for multiprocessing.

log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

classification (dict): Results of the classification.

assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:

.. [1]

fileIterator(self)

idThreshold(self, subject, ready=False)

identify(self, data, threshold=None, ready=False, **kwargs)

io_load(self, label)

```

io_save(self, label, data)

listSubjects(self)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)
    Assess the performance of the classifier in both biometric scenarios: authentication and identification.

    Workflow:
        For each test subject and for each threshold, test authentication and identification;
        Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
        Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
        Subject and global statistics are then computed by evaluation.assessClassification.

    Kwargs:
        data (dict): Dictionary holding the testing samples for each subject.

        rejection_thresholds (array): Thresholds used to compute the ROCs.

        dstPath (string): Path for multiprocessing.

        log2file (bool): Flag to control the use of logging in multiprocessing.

    Kwvals:
        classification (dict): Results of the classification.

        assessment (dict): Biometric statistics.

    See Also:

    Notes:

    Example:

    References:
        ... [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)



---



Class methods inherited from BaseClassifier:



load(cls, srcPath, dstPath=None) from builtin.type



---



Data descriptors inherited from BaseClassifier:



__dict__  
dictionary for instance variables (if defined)



__weakref__  
list of weak references to the object (if defined)


```

```

class SifterSVM(BaseClassifier)
    SVM classifier.

```

Method resolution order:

```

    SifterSVM
    BaseClassifier
    builtin.object

```

Methods defined here:

```

__init__(self, io=None, edges=None, weights=None, **kwargs)

authenticate(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)

autoRejectionThresholds(self)

combination(self, results)

dirIterator(self)

```

```
evaluate(self, data, rejection_thresholds='auto', dstPath=None, log2file=False)  
fileIterator(self)  
identify(self, data, ready=False, **kwargs)  
re_train(self, data)  
train(self, data=None)
```

Data and other attributes defined here:

EXT = '.hdf5'

NAME = 'SifterSVM'

Methods inherited from [BaseClassifier](#):

authThreshold(self, subject, ready=False)

checkSubject(self, subject)

idThreshold(self, subject, ready=False)

io_load(self, label)

io_save(self, label, data)

listSubjects(self)

save(self, dstPath)

seqEvaluate(self, data, rejection_thresholds=None, dstPath=None, log2file=False)

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

Workflow:

For each test subject and for each threshold, test authentication and identification;
Authentication results stored in a 3 dimensional array of booleans, shape = (N thresholds, M subjects, K samples);
Identification results stored in a 2 dimensional array, shape = (N thresholds, K samples);
Subject and global statistics are then computed by evaluation.assessClassification.

Kwargs:

data (dict): Dictionary holding the testing samples for each subject.
 rejection_thresholds (array): Thresholds used to compute the ROCs.
 dstPath (string): Path for multiprocessing.
 log2file (bool): Flag to control the use of logging in multiprocessing.

Kwrvals:

classification (dict): Results of the classification.
 assessment (dict): Biometric statistics.

See Also:

Notes:

Example:

References:
 ... [1]

setAuthThreshold(self, subject, threshold, ready=False)

setIdThreshold(self, subject, threshold, ready=False)

updateThresholds(self, overwrite=False, fraction=1.0)

updateThresholds_old(self, overwrite=False, N=1)

Class methods inherited from [BaseClassifier](#):

load(cls, srcPath, dstPath=None) from [__builtin__.type](#)

Data descriptors inherited from [BaseClassifier](#):

__dict__

dictionary for instance variables (if defined)
 __weakref__
 list of weak references to the object (if defined)

Data and other attributes inherited from [BaseClassifier](#):

EER_IDX = 0

class SubjectDict([bidict.bidict](#))

Method resolution order:

[SubjectDict](#)
[bidict.bidict](#)
[_builtin__object](#)

Methods defined here:

[__getitem__\(self, keyorslice\)](#)

Methods inherited from [bidict.bidict](#):

[__contains__\(self, *args, **kwds\)](#)
 D.[.contains_\(k\)](#) -> True if D has a key k, else False

[__delitem__\(self, keyorslice\)](#)

[__eq__\(self, other\)](#)

[__init__\(self, *args, **kwds\)](#)

[__invert__\(self\)](#)
 Called when unary ~ operator is applied.

[__inverted__\(self\)](#)

[__iter__\(self, *args, **kwds\)](#)
 x.[.iter_\(\)](#) <=> iter(x)

[__len__\(self, *args, **kwds\)](#)
 x.[.len_\(\)](#) <=> len(x)

[__neq__\(self, other\)](#)

[__repr__\(self\)](#)

[__setitem__\(self, keyorslice, keyorval\)](#)

[__str__ = __repr__\(self\)](#)

[clear\(self\)](#)

[copy\(self\)](#)

[get\(self, *args, **kwds\)](#)
 D.[.get_\(k\[,d\]\)](#) -> D[k] if k in D, else d. d defaults to None.

[invert\(self\)](#)

[items\(self, *args, **kwds\)](#)
 D.[.items_\(\)](#) -> list of D's (key, value) pairs, as 2-tuples

[keys\(self, *args, **kwds\)](#)
 D.[.keys_\(\)](#) -> list of D's keys

[pop\(self, key, *args\)](#)

[popitem\(self\)](#)

[setdefault\(self, key, default=None\)](#)

[update\(self, *args, **kwds\)](#)

[values\(self, *args, **kwds\)](#)
 D.[.values_\(\)](#) -> list of D's values

Data descriptors inherited from [bidict.bidict](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

inv
Called when unary ~ operator is applied.

Data and other attributes inherited from [bidict.bidict](#):

methodname = 'values'

class UnknownSubjectError([exceptions.Exception](#))

Method resolution order:

[UnknownSubjectError](#)
[exceptions.Exception](#)
[exceptions.BaseException](#)
[_builtin_.object](#)

Methods defined here:

__init__(self, subject)
__str__(self)

Data descriptors defined here:

__weakref__
list of weak references to the object (if defined)

Data and other attributes inherited from [exceptions.Exception](#):

__new__ = <built-in method __new__ of type object>
T.[__new__](#)(S, ...) -> a new [object](#) with type S, a subtype of T

Methods inherited from [exceptions.BaseException](#):

__delattr__(...)
x.[__delattr__](#)('name') <==> del x.name

__getattribute__(...)
x.[__getattribute__](#)('name') <==> x.name

__getitem__(...)
x.[__getitem__](#)(y) <==> x[y]

__getslice__(...)
x.[__getslice__](#)(i, j) <==> x[i:j]

Use of negative indices is not supported.

__reduce__(...)

__repr__(...)
x.[__repr__](#)() <==> repr(x)

__setattr__(...)
x.[__setattr__](#)('name', value) <==> x.name = value

__setstate__(...)

__unicode__(...)

Data descriptors inherited from [exceptions.BaseException](#):

__dict__

args

message

class UntrainedError([exceptions.Exception](#))

Method resolution order:

[UntrainedError](#)
[exceptions.Exception](#)
[exceptions.BaseException](#)
[__builtin__.object](#)

Methods defined here:

[__str__\(self\)](#)

Data descriptors defined here:

[__weakref__](#)

list of weak references to the object (if defined)

Methods inherited from [exceptions.Exception](#):

[__init__\(...\)](#)

x.[__init__\(...\)](#) initializes x; see help(type(x)) for signature

Data and other attributes inherited from [exceptions.Exception](#):

[__new__](#) = <built-in method __new__ of type object>

T.[__new__\(S, ...\)](#) -> a new [object](#) with type S, a subtype of T

Methods inherited from [exceptions.BaseException](#):

[__delattr__\(...\)](#)

x.[__delattr__\('name'\)](#) <==> del x.name

[__getattribute__\(...\)](#)

x.[__getattribute__\('name'\)](#) <==> x.name

[__getitem__\(...\)](#)

x.[__getitem__\(y\)](#) <==> x[y]

[__getslice__\(...\)](#)

x.[__getslice__\(i, j\)](#) <==> x[i:j]

Use of negative indices is not supported.

[__reduce__\(...\)](#)

[__repr__\(...\)](#)

x.[__repr__\(\)](#) <==> repr(x)

[__setattr__\(...\)](#)

x.[__setattr__\('name', value\)](#) <==> x.name = value

[__setstate__\(...\)](#)

[__unicode__\(...\)](#)

Data descriptors inherited from [exceptions.BaseException](#):

[__dict__](#)

[args](#)

[message](#)

```
class fisher(\_\_builtin\_\_.object)
```

```
# OLD STUFF
```

Methods defined here:

[__init__\(self, cStep=0.1, **kwargs\)](#)
Init

[authenticate\(self, data, subject, threshold=1.0, labels=False, **kwargs\)](#)

[evaluate\(self, data, rejection_thresholds='auto', dstPath=None\)](#)

[identify\(self, data, **kwargs\)](#)

train(self, data=None)

Data descriptors defined here:

__dict__
 dictionary for instance variables (if defined)
__weakref__
 list of weak references to the object (if defined)

Functions

authenticate_Comb(self, data, subject, threshold=None, ready=False, labels=False, **kwargs)

combinationDecorator(cls)

eerIndex(method)

selector(method)

 Selector for the classifier functions and methods.

Input:
 method (str): The desired function or method.

Output:
 fcn (function): The function pointer.

 Configurable fields:

 See Also:

 Example:

 References:

 .. [1]

train_Comb(self, data=None, updateThresholds=True, **kwargs)

updateThresholds_Comb(self, overwrite=False, N=1, **kwargs)

bpkclone.classifiers.rules

d:\work\productioncode\clones\biometricspykit\bpkclone\classifiers\rules.py

```
.. module:: rules
:platform: Unix, Windows
:synopsis: This module provides various functions to...
.. moduleauthor:: Carlos Carreiras
```

Modules

[numpy](#)

Classes

[exceptions.Exception\(exceptions.BaseException\)](#)

[EmptyCombination](#)

class EmptyCombination([exceptions.Exception](#))

Method resolution order:

[EmptyCombination](#)
[exceptions.Exception](#)
[exceptions.BaseException](#)
[builtin__object](#)

Methods defined here:

[__str__\(self\)](#)

Data descriptors defined here:

[__weakref__](#)
list of weak references to the object (if defined)

Methods inherited from [exceptions.Exception](#):

[__init__\(...\)](#)
x.[__init__\(...\)](#) initializes x; see help(type(x)) for signature

Data and other attributes inherited from [exceptions.Exception](#):

[__new__](#) = <built-in method __new__ of type object>
T.[__new__\(S, ...\)](#) -> a new object with type S, a subtype of T

Methods inherited from [exceptions.BaseException](#):

[__delattr__\(...\)](#)
x.[__delattr__\('name'\)](#) <==> del x.name

[__getattribute__\(...\)](#)
x.[__getattribute__\('name'\)](#) <==> x.name

[__getitem__\(...\)](#)
x.[__getitem__\(y\)](#) <==> x[y]

```
__getslice__(...)
    x.__getslice__(i, j) <==> x[i:j]

    Use of negative indices is not supported.

__reduce__(...)

__repr__(...)
    x.__repr__() <==> repr(x)

__setattr__(...)
    x.__setattr__('name', value) <==> x.name = value

__setstate__(...)

__unicode__(...)
```

Data descriptors inherited from [exceptions.BaseException](#):

```
__dict__
args
message
```

Functions

```
combination(labels)
majorityRule(labels)
pluralityRule(labels)
```

```
.. module:: client
:platform: Unix, Windows
:synopsis: This module implements the BITCloud client.

.. moduleauthor:: Carlos Carreiras
```

Modules

[copy](#)

[os](#)

[pymongo](#)

Classes

[connection](#)

class connection

Class to connect to a BITCloud server.

Kwargs:

Kwrvals:

See Also:

Notes:

Example:

References:

.. [1]

Methods defined here:

__init__(self, host='localhost', user=None, passwd=None, queue='BiometricsQ', dbName='BiometricsExperiments')

Establish the [connection](#) to the server.

Kwargs:

Kwrvals:

See Also:

Notes:

Example:

References:
.. [1]

addTask(self, task=None)
cancelTask(self, expID)
close(self)
collectResults(self, expIDList, mapper, basePath, headerOrder=None)
getTaskInfo(self, expID)
listTasks(self, status='all', seen=None)
markSeen(self, expID)
requeue(self, expID)
search(self, spec=None, fields=None)

Data and other attributes defined here:

MONGO_HOST = '193.136.222.234'
MONGO_PORT = 27017
RABBIT_PORT = 5672
TRANSPORT = 'amqp'
VHOST = '/'

```
.. module:: server
:platform: Unix, Windows
:synopsis: This module implements the BITCloud parallelization services.

.. moduleauthor:: Carlos Carreiras
```

Modules

[errno](#) [multiprocessing](#) [os](#)
[multiprocessing.managers](#) [numpy](#) [time](#) [traceback](#)

Classes

[_builtin_.object](#)

[DoWork](#)

[DoWork_Clf](#)
[DoWork_Distances](#)
[DoWork_File](#)

class **DoWork**([_builtin_.object](#))

Methods defined here:

[__init__](#)(self, workQueue, store, timeout, loggerDict, lock=None)

[go](#)(self)

Data descriptors defined here:

[__dict__](#)
dictionary for instance variables (if defined)

[__weakref__](#)
list of weak references to the object (if defined)

class **DoWork_Clf**([DoWork](#))

Method resolution order:

[DoWork_Clf](#)
[DoWork](#)
[_builtin_.object](#)

Methods defined here:

[go](#)(self)

Methods inherited from [DoWork](#):

[__init__](#)(self, workQueue, store, timeout, loggerDict, lock=None)

Data descriptors inherited from [DoWork](#):

dict
 dictionary for instance variables (if defined)

weakref
 list of weak references to the object (if defined)

class DoWork_Distances([DoWork](#))

Method resolution order:

[DoWork_Distances](#)
[DoWork](#)
[builtin.object](#)

Methods defined here:

go(self)

Methods inherited from [DoWork](#):

init(self, workQueue, store, timeout, loggerDict, lock=None)

Data descriptors inherited from [DoWork](#):

dict
 dictionary for instance variables (if defined)

weakref
 list of weak references to the object (if defined)

class DoWork_File([DoWork](#))

Method resolution order:

[DoWork_File](#)
[DoWork](#)
[builtin.object](#)

Methods defined here:

go(self)

Methods inherited from [DoWork](#):

init(self, workQueue, store, timeout, loggerDict, lock=None)

Data descriptors inherited from [DoWork](#):

dict
 dictionary for instance variables (if defined)

weakref
 list of weak references to the object (if defined)

Functions

cleanStore(store)
copy(data)
getCondition()
getDictManager()
getEventManager()
getIntValue(value=None)
getListManager()
getLock()
getManager()
getPipe()
getPool(processes)
getQueue()
loadStore(store, taskid)
randomTester(data)
runMultiprocess(workQueue, store, mode='data', numberProcesses=None, lock=None, timeout=None, log2file=False)
workerProcess(workClass, workQueue, store, timeout, loggerDict, randomSeed, lock=None)

```
.. module:: WebServer
:platform: Windows, Linux
:synopsis: WebSockets server.

.. moduleauthor:: Ana Priscila Alves, Carlos Carreiras
```

Modules

Queue	twisted.internet.protocol	time
twisted.internet.error	multiprocessing.queues	traceback
multiprocessing	threading	uuid

Classes

[multiprocessing.queues.Queue\(`__builtin__.object`\)](#)
[LockQueue](#)
[DummyComQueue](#)
[threading.Thread\(threading._Verbose\)](#)
[SPApp](#)
[Echo](#)
[SPCom](#)
[twisted.internet.protocol.Protocol\(twisted.internet.protocol.BaseProtocol\)](#)
[SP](#)

class **DummyComQueue**([LockQueue](#))

Method resolution order:

[DummyComQueue](#)
[LockQueue](#)
[multiprocessing.queues.Queue](#)
[__builtin__.object](#)

Methods defined here:

get(self, *args, **kwargs)
put(self, item)

Methods inherited from [LockQueue](#):

__getstate__(self)
__init__(self, cleanup=True, *args, **kwargs)
__setstate__(self, state)
lock(self)
unlock(self)

Methods inherited from [multiprocessing.queues.Queue](#):

cancel_join_thread(self)
close(self)

empty(self)
full(self)
get_nowait(self)
join_thread(self)
put_nowait(self, obj)
qsize(self)

Data descriptors inherited from [multiprocessing.queues.Queue](#):

_dict
 dictionary for instance variables (if defined)
_weakref
 list of weak references to the object (if defined)

class Echo([SPApp](#))

Method resolution order:

[Echo](#)
[SPApp](#)
[threading.Thread](#)
[threading.Verbose](#)
[__builtin__.object](#)

Methods defined here:

on_close(self)
on_message(self, message)
on_open(self)

Methods inherited from [SPApp](#):

__init__(self, queue=None)
restartReactor(self)
run(self)
send(self, message)
startCom(self)
stopCom(self)
stopReactor(self)

Class methods inherited from [SPApp](#):

instantiate(cls, *args, **kwargs) from [__builtin__.type](#)

Methods inherited from [threading.Thread](#):

__repr__(self)
getName(self)
isAlive(self)
isDaemon(self)
is_alive = isAlive(self)

```
join(self, timeout=None)  
setDaemon(self, daemonic)  
setName(self, name)  
start(self)
```

Data descriptors inherited from [threading.Thread](#):

daemon

ident

name

Data descriptors inherited from [threading.Verbose](#):

_dict
 dictionary for instance variables (if defined)

_weakref
 list of weak references to the object (if defined)

class LockQueue([multiprocessing.queues.Queue](#))

Method resolution order:

[LockQueue](#)
[multiprocessing.queues.Queue](#)
[builtin.object](#)

Methods defined here:

```
__getstate__(self)  
__init__(self, cleanup=True, *args, **kwargs)  
__setstate__(self, state)  
get(self, *args, **kwargs)  
lock(self)  
put(self, *args, **kwargs)  
unlock(self)
```

Methods inherited from [multiprocessing.queues.Queue](#):

```
cancel_join_thread(self)  
close(self)  
empty(self)  
full(self)  
get_nowait(self)  
join_thread(self)  
put_nowait(self, obj)  
qsize(self)
```

Data descriptors inherited from [multiprocessing.queues.Queue](#):

_dict
 dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class **SP**([twisted.internet.protocol.Protocol](#))

Method resolution order:

SP
[twisted.internet.protocol.Protocol](#)
[twisted.internet.protocol.BaseProtocol](#)

Methods defined here:

__init__(self, app, durable, *args, **kwargs)
connectionLost(self, reason)
connectionMade(self)
dataReceived(self, data)
send(self, data)

Methods inherited from [twisted.internet.protocol.Protocol](#):

__provides__
Special descriptor for class **__provides__**

The descriptor caches the implementedBy info, so that
we can get declarations for objects without instance-specific
interfaces a bit quicker.

logPrefix(self)
Return a prefix matching the class name, to identify log messages
related to this protocol instance.

Data and other attributes inherited from [twisted.internet.protocol.Protocol](#):

__implemented__ = <implementedBy [twisted.internet.protocol.Protocol](#)>

Methods inherited from [twisted.internet.protocol.BaseProtocol](#):
__providedBy__ = <zope.interface.declarations.Declaration object>

makeConnection(self, transport)
Make a connection to a transport and a server.

This sets the 'transport' attribute of this [Protocol](#), and calls the
[connectionMade\(\)](#) callback.

Data and other attributes inherited from [twisted.internet.protocol.BaseProtocol](#):

connected = 0

transport = None

class **SPApp**([threading.Thread](#))

Method resolution order:

SPApp
[threading.Thread](#)
[threading.Verbose](#)
[_builtin_.object](#)

Methods defined here:

__init__(self, queue=None)
on_close(self)

```
on_message(self, message)  
on_open(self)  
restartReactor(self)  
run(self)  
send(self, message)  
startCom(self)  
stopCom(self)  
stopReactor(self)
```

Class methods defined here:

instantiate(cls, *args, **kwargs) from [_builtin_.type](#)

Methods inherited from [threading.Thread](#):

```
_repr_(self)  
getName(self)  
isAlive(self)  
isDaemon(self)  
is_alive = isAlive(self)  
join(self, timeout=None)  
setDaemon(self, daemonic)  
setName(self, name)  
start(self)
```

Data descriptors inherited from [threading.Thread](#):

daemon
ident
name

Data descriptors inherited from [threading.Verbose](#):

```
_dict  
    dictionary for instance variables (if defined)  
_weakref  
    list of weak references to the object (if defined)
```

class [SPCom\(threading.Thread\)](#)

Method resolution order:

[SPCom](#)
[threading.Thread](#)
[threading.Verbose](#)
[_builtin_.object](#)

Methods defined here:

```
_init_(self, go, queue)  
run(self)
```

Methods inherited from [threading.Thread](#):

__repr__(self)
getName(self)
isAlive(self)
isDaemon(self)
is_alive = isAlive(self)
join(self, timeout=None)
setDaemon(self, daemonic)
setName(self, name)
start(self)

Data descriptors inherited from [threading.Thread](#):

daemon
ident
name

Data descriptors inherited from [threading.Verbose](#):

__dict__
 dictionary for instance variables (if defined)
__weakref__
 list of weak references to the object (if defined)

Functions

serverPort(cls, port='COM30', baud=9600, setupFcn=None, cleanupFcn=None, durable=True, *args, **kwargs)

Data

default_timeout = None
reactor = <twisted.internet.selectreactor.SelectReactor object>

bpkclone.Cloud.server

d:\work\productioncode\clones\biometricspykit\bpkclone\cloud\server.py

```
.. module:: server
:platform: Unix, Windows
:synopsis: This module implements the BITCloud server.

.. moduleauthor:: Carlos Carreiras
```

Modules

collections	matplotlib	signal	time
json	os	smtplib	
logging	pymongo	sys	

Classes

[__builtin__.object](#)
[MQWorker](#)
[exceptions.Exception\(exceptions.BaseException\)](#)
[KillSignal](#)

class **KillSignal**([exceptions.Exception](#))

Method resolution order:

[KillSignal](#)
[exceptions.Exception](#)
[exceptions.BaseException](#)
[__builtin__.object](#)

Methods defined here:

[__init__\(self, sig\)](#)
[__str__\(self\)](#)

Data descriptors defined here:

[**__weakref__**](#)
list of weak references to the object (if defined)

Data and other attributes inherited from [exceptions.Exception](#):

[**__new__**](#) = <built-in method [__new__](#) of type object>
T.[__new__\(S, ...\)](#) -> a new [object](#) with type S, a subtype of T

Methods inherited from [exceptions.BaseException](#):

[**__delattr__\(...\)**](#)
x.[__delattr__\('name'\)](#) <=> del x.name

```
__getattribute__(...)
    x.getattribute('name') <==> x.name

__getitem__(...)
    x.getitem(y) <==> x[y]

__getslice__(...)
    x.getslice(i, j) <==> x[i:j]

        Use of negative indices is not supported.

__reduce__(...)

__repr__(...)
    x.repr() <==> repr(x)

__setattr__(...)
    x.setattr('name', value) <==> x.name = value

__setstate__(...)

__unicode__(...)
```

Data descriptors inherited from [exceptions.BaseException](#):

__dict__

args

message

class MQWorker([_builtin_.object](#))

Consumer class

Methods defined here:

```
__init__(self, rabbitHost='localhost', mongoHost='localhost', workerName=None,
        dbName='BiometricsExperiments', queue='BiometricsQ', dstPath='~', mode='production',
        parameters=None)
```

acknowledge_message(self, message, taskID)

declare_queue(self)

go(self)

notify(self, username, taskID, message)

on_message(self, message)

stop(self)

Data descriptors defined here:

```
__dict__
    dictionary for instance variables (if defined)
```

__weakref__

list of weak references to the object (if defined)

Data and other attributes defined here:

EMAIL_SENDER = 'carlos.carreiras@lx.it.pt'

EMAIL SUBJECT = '[BiometricsPyKit] Status of Task %d'

FAILURE_MSG = '\n Dear %s,\n \n Unfortunalety, Task %d ha...le on the %s machine.\n \n Kind regards.\n '

MONGO_PORT = 27017

PASSWD = 'worker'

RABBIT_PORT = 5672

REPEAT_MSG = '\n Dear %s,\n \n Task %d has already been ...ht by the %s machine.\n \n Kind regards.\n '

SUCCESS_MSG = '\n Dear %s,\n \n Task %d has been success...le on the %s machine.\n \n Kind regards.\n '

TIMEOUT = 30

TRANSPORT = 'amqp'

VHOST = '/'

Functions

convertUniDict(data)

onKillSignal(sig, func=None)

processTask(conn, task, taskID, parameters)

processTask_test(conn, task, taskID, parameters)

sendEmail(To, From, CC='', ReplyTo='', Subject='', Text='', FilePath=None, Host='cascais.lx.it.pt', Port=25)

setKillHandler(func)

Data

Encoders = <email.LazyImporter object>

bpkclone.Cloud.webServer

[index](#)

```
.. module:: WebServer
:platform: Windows, Linux
:synopsis: WebSockets server.

.. moduleauthor:: Ana Priscila Alves, Carlos Carreiras
```

Modules

[Queue](#)
[ujson](#)
[logging](#)

[multiprocessing](#)
[twisted.internet.protocol](#)
[multiprocessing.queues](#)

[sys](#)
[threading](#)
[time](#)

[traceback](#)
[uuid](#)
[websocket](#)

Classes

[__builtin__.object](#)

[WSClient](#)

[TestClient](#)

[multiprocessing.queues.Queue](#)([__builtin__.object](#))

[LockQueue](#)

[DummyComQueue](#)

[threading.Thread](#)([threading.Verbose](#))

[WSApp](#)

[Echo](#)

[WSCom](#)

[twisted.internet.protocol.Factory](#)

[WSFactory](#)

[twisted.internet.protocol.Protocol](#)([twisted.internet.protocol.BaseProtocol](#))

[WS](#)

class DummyComQueue([LockQueue](#))

Method resolution order:

[DummyComQueue](#)
[LockQueue](#)
[multiprocessing.queues.Queue](#)
[__builtin__.object](#)

Methods defined here:

get(self, *args, **kwargs)

put(self, item)

Methods inherited from [LockQueue](#):

__getstate__(self)
__init__(self, cleanup=True, *args, **kwargs)
__setstate__(self, state)
lock(self)
unlock(self)

Methods inherited from [multiprocessing.queues.Queue](#):

cancel_join_thread(self)
close(self)
empty(self)
full(self)
get_nowait(self)
join_thread(self)
put_nowait(self, obj)
qsize(self)

Data descriptors inherited from [multiprocessing.queues.Queue](#):

__dict__
 dictionary for instance variables (if defined)
__weakref__
 list of weak references to the object (if defined)

class Echo(WSApp)

Method resolution order:

[Echo](#)
[WSApp](#)
[threading.Thread](#)
[threading.Verbose](#)
[builtin.object](#)

Methods defined here:

on_close(self)
on_message(self, message)
on_open(self)

Methods inherited from [WSApp](#):

__init__(self, queue=None)
run(self)
send(self, message)
startCom(self)

stopCom(self)

Class methods inherited from [WSApp](#):

instantiate(cls, *args, **kwargs) from [__builtin__.type](#)

Methods inherited from [threading.Thread](#):

__repr__(self)

getName(self)

isAlive(self)

isDaemon(self)

is_alive = isAlive(self)

join(self, timeout=None)

setDaemon(self, daemonic)

setName(self, name)

start(self)

Data descriptors inherited from [threading.Thread](#):

daemon

ident

name

Data descriptors inherited from [threading.Verbose](#):

— dict
 dictionary for instance variables (if defined)

— weakref
 list of weak references to the object (if defined)

class LockQueue([multiprocessing.queues.Queue](#))

Method resolution order:

[LockQueue](#)
[multiprocessing.queues.Queue](#)
[__builtin__.object](#)

Methods defined here:

__getstate__(self)

__init__(self, cleanup=True, *args, **kwargs)

__setstate__(self, state)

get(self, *args, **kwargs)

lock(self)

put(self, *args, **kwargs)

unlock(self)

Methods inherited from [multiprocessing.queues.Queue](#):

cancel_join_thread(self)

close(self)

empty(self)

full(self)

get_nowait(self)

join_thread(self)

put_nowait(self, obj)

qsize(self)

Data descriptors inherited from [multiprocessing.queues.Queue](#):

__dict__

dictionary for instance variables (if defined)

__weakref__

list of weak references to the object (if defined)

class **TestClient**([WSClient](#))

Method resolution order:

[TestClient](#)
[WSClient](#)
[builtin_object](#)

Methods defined here:

__init__(self, *args, **kwargs)

on_close(self)

on_message(self, message)

on_open(self)

Methods inherited from [WSClient](#):

close(self)

close websocket connection.

on_error(self, reason)

run_forever(self, sockopt=None, sslopt=None, ping_interval=0)

run event loop for WebSocket framework.

This loop is infinite loop and is alive during websocket is available.

sockopt: values for socket.setsockopt.

sockopt must be tuple and each element is argument of sock.setsockopt.

sslopt: ssl socket optional dict.

ping_interval: automatically send "ping" command every specified period(second)
if set to 0, not send automatically.

```
send(self, data, opcode=None)
    send message.
    data: message to send. If you set opcode to OPCODE_TEXT, data must be utf-8 string or unicode.
    opcode: operation code of data. default is OPCODE_TEXT.
```

Class methods inherited from [WSClient](#):

instantiate(cls, url, *args, **kwargs) from [__builtin__.type](#)

Data descriptors inherited from [WSClient](#):

__dict__
 dictionary for instance variables (if defined)

__weakref__
 list of weak references to the object (if defined)

class **WS**([twisted.internet.protocol.Protocol](#))

Method resolution order:

[WS](#)
[twisted.internet.protocol.Protocol](#)
[twisted.internet.protocol.BaseProtocol](#)

Methods defined here:

__init__(self, app, durable, *args, **kwargs)

connectionLost(self, reason)

connectionMade(self)

dataReceived(self, data)

send(self, data)

Methods inherited from [twisted.internet.protocol.Protocol](#):

__provides__

Special descriptor for class **__provides__**

The descriptor caches the implementedBy info, so that we can get declarations for objects without instance-specific interfaces a bit quicker.

logPrefix(self)

Return a prefix matching the class name, to identify log messages related to this protocol instance.

Data and other attributes inherited from [twisted.internet.protocol.Protocol](#):

__implemented__ = <implementedBy twisted.internet.protocol.Protocol>

Methods inherited from [twisted.internet.protocol.BaseProtocol](#):

__providedBy__ = <zope.interface.declarations.Declaration object>

makeConnection(self, transport)

Make a connection to a transport and a server.

This sets the 'transport' attribute of this [Protocol](#), and calls the [connectionMade\(\)](#) callback.

Data and other attributes inherited from [twisted.internet.protocol.BaseProtocol](#):

connected = 0

transport = None

class **WSApp**([threading.Thread](#))

Method resolution order:

[WSApp](#)
[threading.Thread](#)
[threading.Verbose](#)
[builtin.object](#)

Methods defined here:

__init__(self, queue=None)

on_close(self)

on_message(self, message)

on_open(self)

run(self)

send(self, message)

startCom(self)

stopCom(self)

Class methods defined here:

instantiate(cls, *args, **kwargs) from [builtin.type](#)

Methods inherited from [threading.Thread](#):

__repr__(self)

getName(self)

isAlive(self)

isDaemon(self)

is_alive = isAlive(self)

join(self, timeout=None)

setDaemon(self, daemonic)

setName(self, name)

start(self)

Data descriptors inherited from [threading.Thread](#):

daemon

ident

name

Data descriptors inherited from [threading.Verbose](#):

__dict__

dictionary for instance variables (if defined)

__weakref__

list of weak references to the object (if defined)

class WSCClient([builtin.object](#))

Cloned from `websocket.WebSocketApp`

The interface is like JavaScript `WebSocket` [object](#).

Methods defined here:

__init__(self, url, header=[], keep_running=True, get_mask_key=None)

url: websocket url.

header: custom header for websocket handshake.

keep_running: a boolean flag indicating whether the app's main loop should keep running, defaults to True

get_mask_key: a callable to produce new mask keys, see the `WebSocket.set_mask_key`'s docstring for more information

close(self)

close websocket connection.

on_close(self)

on_error(self, reason)

on_message(self, message)

on_open(self)

run_forever(self, sockopt=None, sslopt=None, ping_interval=0)

run event loop for WebSocket framework.

This loop is infinite loop and is alive during websocket is available.

sockopt: values for socket.setsockopt.

sockopt must be tuple and each element is argument of `sock.setsockopt`.

sslopt: ssl socket optional dict.

ping_interval: automatically send "ping" command every specified period(second) if set to 0, not send automatically.

send(self, data, opcode=None)

send message.

data: message to send. If you set opcode to OPCODE_TEXT, data must be utf-8 string or unicode.

opcode: operation code of data. default is OPCODE_TEXT.

Class methods defined here:

instantiate(cls, url, *args, **kwargs) from [builtin.type](#)

Data descriptors defined here:

__dict__

dictionary for instance variables (if defined)

__weakref__

list of weak references to the object (if defined)

class WSCom([threading.Thread](#))

Method resolution order:

[WSCom](#)
[threading.Thread](#)
[threading.Verbose](#)
[builtin.object](#)

Methods defined here:

`__init__(self, go, queue)`
`run(self)`

Methods inherited from [threading.Thread](#):

`__repr__(self)`
`getName(self)`
`isAlive(self)`
`isDaemon(self)`
`is_alive = isAlive(self)`
`join(self, timeout=None)`
`setDaemon(self, daemonic)`
`setName(self, name)`
`start(self)`

Data descriptors inherited from [threading.Thread](#):

`daemon`
`ident`
`name`

Data descriptors inherited from [threading.Verbose](#):

`__dict__`
 dictionary for instance variables (if defined)
`__weakref__`
 list of weak references to the object (if defined)

class **WSFactory**([twisted.internet.protocol.Factory](#))

Methods defined here:

`__init__(self, app, durable, *args, **kwargs)`
`buildProtocol(self, addr)`

Methods inherited from [twisted.internet.protocol.Factory](#):

`__providedBy__` = <zope.interface.declarations.Declaration object>
`__provides__`
 Special descriptor for class `__provides__`

The descriptor caches the implementedBy info, so that we can get declarations for objects without instance-specific interfaces a bit quicker.

doStart(self)

Make sure startFactory is called.

Users should not call this function themselves!

doStop(self)

Make sure stopFactory is called.

Users should not call this function themselves!

logPrefix(self)

Describe this factory for log messages.

startFactory(self)

This will be called before I begin listening on a Port or Connector.

It will only be called once, even if the factory is connected to multiple ports.

This can be used to perform 'unserialization' tasks that are best put off until things are actually running, such as connecting to a database, opening files, etcetera.

stopFactory(self)

This will be called before I stop listening on all Ports/Connectors.

This can be overridden to perform 'shutdown' tasks such as disconnecting database connections, closing files, etc.

It will be called, for example, before an application shuts down, if it was connected to a port. User code should not call this function directly.

Class methods inherited from [twisted.internet.protocol.Factory](#):

forProtocol(cls, protocol, *args, **kwargs) from __builtin__.classobj

Create a factory for the given protocol.

It sets the C{protocol} attribute and returns the constructed factory instance.

@param protocol: A L{Protocol} subclass

@param args: Positional arguments for the factory.

@param kwargs: Keyword arguments for the factory.

@return: A L{Factory} instance wired up to C{protocol}.

Data and other attributes inherited from [twisted.internet.protocol.Factory](#):

__implemented__ = <implementedBy twisted.internet.protocol.Factory>

noisy = True

numPorts = 0

protocol = None

```
connectWS(cls, host=None, port=9999, SSL=False, *args, **kwargs)  
serveWS(cls, port=9999, setupFcn=None, cleanupFcn=None, durable=True, SSL=False, *args, **kwargs)
```

Data

```
SSL_OK = False  
default_timeout = None  
logger = <logging.Logger object>  
reactor = <twisted.internet.selectreactor.SelectReactor object>  
ssl = None
```

```
.. module:: cluster
:platform: Unix, Windows
:synopsis: This module provides various functions to cluster data

.. moduleauthor:: Filipe Canento, Andre Lourenco, Carlos Carreiras

.. conventions:
    output of type dict {'-1': indexes, '0': indexes, '1': indexes, ...}
```

Modules[scipy.spatial.distance](#)
[glob](#)[scipy.cluster.hierarchy](#)
[numpy](#)[os](#)
[pylab](#)[scipy](#)**Functions****clustersFromFile**(filePath, method, criterion, nsegs=None)**convertMatFiles**(srcPath, dstPath, method, criterion, nsegs=None)**dbSCAN**(data, min_samples=10, eps=0.95, metric='euclidean')Perform clustering using the DBSCAN method. Uses: `sklearn.cluster.DBSCAN`.**Input:**

data (array): input data array of shape (number samples x number features).

min_samples (int): minimum number of samples in a cluster.

eps (float): maximum distance between two samples in the same cluster.

metric (string): distance metric

Output:

res (dict): output dict with indexes for each cluster determined.

Example: res = {'clusters': {'-1': noise indexes list,
 '0': cluster 0 indexes list,
 '1': cluster 1 indexes list}}

Configurable fields:

{ "name": "cluster.dbSCAN", "config": { "min_samples": "10", "eps": "0.95", "metric": "euclidean"}, "inputs": ["data"], "outputs": ["core_samp"] }

See Also:[sklearn.cluster.DBSCAN](#)
[scipy.spatial.distance](#)**Notes:**

conversion to similarity uses the same metric distance as dbSCAN

Example:**References:**.. [1] DBSCAN: Ester, M., Kriegel, H.-P., Sander, J. and Xu, X. (1996), "A density-based algorithm for discovering clusters in large spatial databases with noise". Proc of 2nd Int Conf on Knowledge Discovery and Data Mining
.. [2] dbSCAN: <http://scikit-learn.org/dev/modules/generated/sklearn.cluster.DBSCAN.html>
.. [3] pdist: <http://docs.scipy.org/doc/scipy/reference/generated/scipy.spatial.distance.pdist.html#scipy.spatial.distance.pdist>**floor**(...)[floor](#)(x)

Return the floor of x as a float.

This is the largest integral value <= x.

hierarchical(data, k=0, metric='euclidean', method='average', showDendrogram=True)

Clustering using hierarchical agglomerative methods

Input:

data : (array): input data array of shape (number samples x number features).
k: (int) : number of clusters (k=0 using life-time criteria)
method: (str) 'single' --- nearest distance (default)
 'complete' --- furthest distance
 'average' --- average distance
 'centroid' --- center of mass distance
 'ward' --- inner squared distance

Output:

res (dict): output dict with indexes for each cluster determined.

Example: res = {"labels": list(labels),
 "k":k,
 "nsamples_in_cluster":list(nsamples_in_cluster),
 "algorithm":"hierarchical-"+method}

Configurable fields:

{ "name": "cluster.dbSCAN", "config": { "min_samples": "10", "eps": "0.95", "metric": "euclidean"}, "inputs": ["data"], "outputs": ["core_samp"] }

See Also:

Example:

References:
.. [1]

leadHierarchical(data, **kwargs)

Perform hierarchical clustering for each lead independently.

run(data, tasks, parameters, outPath)

Run the clustering methods for each item in data, according to the given parameters, then perform the prototype selection.

Input:

 data

 tasks

 parameters

 outPath

Output:

 output

Configurable fields:{ "name": "???.???", "config": {}, "inputs": [], "outputs": [] }

See Also:

Notes:

Example:

selector(method)

Selector for the clustering functions and methods.

Input:

 method (str): The desired function or method.

Output:

 fcn (function): The function pointer.

Configurable fields:

See Also:

Example:

References:
.. [1]

templateSelection(data, clusters=None, method='MDIST', ntemplatesPerCluster=1, metric='euclidean')

Template Selection

This function implements several template selection methods. It may use the information of cluster labels [passed as argument] in conjunction with the criteria that are implemented. The number of templates per cluster can be controlled by ntemplatesPerCluster argument. The number returned templates is variable and depends on the number of clusters and on the number of templates per clusters [If clusters are singletons the number of templates per cluster may be violated]. It also works with different distance Measures [distMeasure argument]

Currently the implemented criteria are:

1) Centroids: requires that clusters are provided; if number of templates per cluster is 1, then templates are just the centroids; if the number of clusters is > 1 then k-means is used to determine more templates on each cluster.

2) MDIST: criterion described in U. Uludag, A. Ross, A. Jain, Pattern Recognition 37 (2004). Originally it does not require any clustering labels, computing the templates based on distances to the mean wave. The number of templates supplied by the user, in this cases we assume ntemplatesPerCluster argument, as the number of templates. We extended the idea, including cluster label information. We compute for each cluster ntemplatesPerCluster using the distance to mean wave criterion.

3) DEND: criterion described in U. Uludag, A. Ross, A. Jain, Pattern Recognition 37 (2004). The implemented version is an extension that works with any clustering algorithm (uses the clusters labels passed by argument). The original version used Complete-linkage. It computes medoids on each cluster, based on the average pairwise distance to other points in the cluster. Templates are the points whose distance is minimum; we allow to chose more than 1 template, selecting ntemplatesPerCluster using the same criterion

Input:

 data : (array): input data array of shape (number samples x number features).
 clusters: (dict): dictionary with labels (can be None)
 method: (string): method to be used ('Centroids', 'MDIST', 'DEND')/
 ntemplatesPerCluster: (int): number of templates on each cluster
 distMeasure: (string): distance measure to be used ('Euclidean' or 'Cosine')

Output:

 res (dict): output dict with indexes of prototypes for each cluster .
 Example: res = {'templates': list(),
 'algorithm': method}

Configurable fields:

See Also:

Example:

References: [1] U. Uludag, A. Ross, A. Jain, Pattern Recognition 37 (2004)

```
templateSelectionFixingK(data, clusters=None, method='MDIST', ntemplates=10, distMeasure=None)
    Template Selection Fixing the number of templates (argument ntemplates)

time(...)
    time() -> floating point number

    Return the current time in seconds since the Epoch.
    Fractions of a second may be present if the system clock provides them.

transform2dissimilaritySpace(data, templates, distMeasure=None)
    Transform to Dissimilarity Representation

    This function computes a new representation for the data, based on the distance to the templates.
    This representation has as number of dimensions the number of templates (columns)
```

```
#####
# Evidence Accumulation Clustering (EAC)
# Script used for building the clustering ensembles its combination using EAC
#
# Andre Lourenco, arlourenco@lx.it.pt
# Ana Fred, afred@lx.it.pt
# Instituto de Telecomunicacoes, Instituto Superior Tecnico,
# Technical University of Lisbon, Portugal
# IST - Torre Norte, Av. Rovisco Pais, 1049-001, Lisbon, Portugal
#####
```

Modules

[scipy.cluster.hierarchy](#)
[numpy](#)

[os](#)
[pylab](#)

[scipy](#)
[scipy.sparse](#)

Functions

coassociationCreation(ensemble, ns, debug=False)
 Construction of the co-association matrix

Input:
 ensemble (dic):

ns (int): number of samples

Output:
 assocs : co-association matrix

Configurable fields:
 {"name": "cluster.dbscan", "config": {"min_samples": "10", "eps": "0.95", "metric": "euclidean"}, "inputs": ["data"], "outputs": ["core_samples"]}

See Also:

Example:

References:
 .. [1]

consensusExtraction(coassocs, k=0, method='single', showDendrogram=False)

Extraction of the consensus partition using hierarchical agglomerative methods

Input:

assocs : co-association matrix
 method='single' --- nearest distance (default)
 'complete' --- furthest distance
 'average' --- average distance
 'centroid' --- center of mass distance
 'ward' --- inner squared distance
 k : number of clusters (k=0 using life-time criteria)

Output:

res (dict): output dict with indexes for each cluster determined.
 Example: res = {"labels": list(labels),
 "k":k,
 "nsamples_in_cluster":list(nsamples_in_cluster),
 "algorithm":"EAC-"+method}

Configurable fields:

{"name": "cluster.dbscan", "config": {"min_samples": "10", "eps": "0.95", "metric": "euclidean"}, "inputs": ["data"], "outputs": ["core_samples"]}

See Also:

Example:

References:
 .. [1]

ensembleCreation(data, nensemble=100, showHist=False, fun='KMeans', kmin=None, kmax=None)

creation of the ensemble using kmeans

Input:

data (array): input data array of shape (number samples x number features).

N_ensemble (int): number of partitions on the ensemble (if empty using 100)

eps (float): maximum distance between two samples in the same cluster.

fun (string): algorithm for producing clusters (default KMeans)

Output:

res (dict): output dict with partitions

Configurable fields:

See Also:

Example:

References:
.. [1]

ensembleCreationParallel(data, nensemble=100, fun='KMeans', kmin=None, kmax=None, dstPath=None)

ensembleInstance(data, fun='KMeans', k=1, init='random', max_iter=100, batch_size=100)

life_time(Zpy, ns)

life-time criterion for automatic selection of the number of clusters
[porting from life-time implementation on matlab]

Input:

Zpy (array): input data array of shape (number samples x number features).

ns (int): number of samples.

Output:

res (dict): output dict with indexes for each cluster determined.

Example: res = {'-1': noise indexes list,
'0': cluster 0 indexes list,
'1': cluster 1 indexes list}

Configurable fields:

{"name": "cluster.dbscan", "config": {"min_samples": "10", "eps": "0.95", "metric": "euclidean"}, "inputs": ["data"], "outputs": ["core_samples"]}

See Also:

Example:

References:

.. [1]

main()

#####
#

plotClustering(X, labels, figNumber=2)

plot clustering

Input:

Output:

relativeSimilarities(sim, clusters, n)

sim - similarity matrix
clusters - clusters
n - number of samples

runEnsembleCreation(data, tasks, parameters)

time(...)

`time()` -> floating point number

Return the current time in seconds since the Epoch.

Fractions of a second may be present if the system clock provides them.

bpkclone.cluster.templateSelection [index](#) [d:\work\productioncode\clones\biometricspykit\bpkclone\cluster\templateselection.py](#)

```
.. module:: templateSelection
:platform: Unix, Windows
:synopsis: This module provides various functions to cluster data

.. moduleauthor:: Andre Lourenco, Carlos Carreiras

.. conventions:
   output of type dict {'-1': indexes, '0': indexes, '1': indexes, ...}
```

Modules

[scipy.spatial.distance](#) [numpy](#)

Functions

avgDist(data=None, fdist='Euclidean', norm=None)

centroids(data, clusters=None, ntemplates=1)
Template Selection: Centroids

Requires that clusters are provided;
if number of templates per cluster is 1, then templates
are just the centroids; if the number of clusters is >1
then k-means is used to determine more templates on each cluster.

highestAveragesAllocator(cardinals, k, divisor='dHondt', check=False)

leadCentroids(data, **kwargs)
Perform centroids template selection for each lead independently.

WARNING: Has a dirty hack to be compatible with Francis' dissimilarity code.

mdist(data, clusters=None, ntemplates=1, distMeasure='Euclidean')

Template Selection using MDIST criterion
Proposed in U. Uludag, A. Ross, A. Jain, Pattern Recognition 37 (2004).
Originally it does not require any clustering labels, computing the templates
based on mean distances. The number of templates supplied by the user,
in this cases we assume ntemplates argument, as the number of templates.
We extended the idea, including cluster label information. We compute,
for each cluster, the mean distance criterion.
We select, at most, ntemplates. With cluster labels, templates are attributed
to clusters using an allocation method (e.g. d'Hondt)

remainderAllocator(cardinals, k, reverse=True, check=False)

selector(method)
Selector for the template selection functions and methods.

Input:
method (str): The desired function or method.

Output:
fcn (function): The function pointer.

Configurable fields:

See Also:

Example:

References:
.. [1]

```
.. module:: config
:platform: Unix, Windows
:synopsis: This module provides various configuration parameters to be shared across all modules of BiometricsPyKit.

.. moduleauthor:: Carlos Carreiras
```

Modules

[multiprocessing](#)[numpy](#)[os](#)[pylab](#)

Functions

deploy(folders=None)**getFilePath**(sequence)**getPath**(sequence, verify=True)**getRandomSeed**()**setRandomSeed**(seed=None)

Data

```
baseFolder = r'C:\Users\Carlos\BiometricsPyKitRun'
folder = r'C:\Users\Carlos\BiometricsPyKitRun'
font = {'family': 'Bitstream Vera Sans', 'size': 16, 'weight': 'normal'}
manager = None
nP = 12
numberProcesses = 6
queueTimeOut = 2
random = <mtrand.RandomState object>
randomSeed = None
```

```
.. module:: datamanager
:platform: Unix, Windows
:synopsis: This module provides various functions to load and store data.

.. moduleauthor:: Carlos Carreiras
```

Modules

[cPickle](#)
[csv](#)
[errno](#)

[glob](#)
[gzip](#)
[h5py](#)

[sklearn.externals.joblib](#)
[numpy](#)
[os](#)

[shutil](#)
[warnings](#)
[zipfile](#)

Classes

[DBUnit](#)
[H5Unit](#)
[MPUnit](#)

class DBUnit
StorageBIT (online) database manager class.

Input:
None

Output:
None

Configurable fields:

See Also:

Notes:

Example:

Methods defined here:

__init__(self, dbConn, mapper, experiments)
Instantiate the class.

Input:
dbConn (dict): The DB connection parameters.
mapper (dict): To map user variables to the database representation.
experiments (string or list of strings): The experiment(s) to load.

Output:
None

Configurable fields:

See Also:

Notes:

Example:
dbConn = {'host': '193.136.222.234',
 'dstPath': '/path/to/files',
 'dbName': 'database'}
mapper = {'raw': 'signals/ECG/hand/raw',
 'filtered': 'signals/ECG/hand/filtered'}
experiments = ['T1', 'T2', 'test']
store = [DBUnit](#)(dbConn, mapper, experiments)

addExperiment(self, experiment=None)
Add a new experiment to the database.

Input:
experiment (dict): The new experiment to add. Has to have the key 'name'.

Output:
experimentName (str): The name of the new experiment.

Configurable fields:

See Also:

Notes:

Example:

```
experiment = {'name': 'Test', 'goals': 'To test the addition of experiments.', 'description': 'This is a sample experiment.'}
experimentName = store.addExperiment(experiment)
```

addRecord(self, header)

Add a new record to the database.

Input:

header (dict): The new record to add. Has to have 'subject' and 'experiment' keys.

Output:

recordId (int): The ID of the new record.

Configurable fields:

See Also:

Notes:

Example:

```
header = {'experiment': 'test', 'subject': 1, 'date': 'yyyy-mm-ddThh:mm:ss'}
store.addRecord(header)
```

addSubject(self, subject=None)

Add a new subject to the database.

Input:

experiment (dict): The new subject to add. Has to have the key 'name'.

Output:

subjectId (int): The ID of the new subject.

Configurable fields:

See Also:

Notes:

Example:

```
subject = {'name': 'John Smith', 'birthdate': 'yyyy-mm-ddThh:mm:ss', 'email': 'john.smith@domain.com'}
subjectId = store.addExperiment(subject)
```

close(self)

Close the manager.

Input:

None

Output:

None

Configurable fields:

See Also:

Notes:

Example:

```
store.close()
```

data2db(self, data, data_type, updateDuration=False)

Insert data of specified types in the database.

Input:

data (dict): Input data where keys correspond to record IDs.
There is also one key 'info' with specific information about the data.

data_type (list of strings): Input data types to be added.

updateDuration (bool): Flag to update the duration of the record. Default: False.

Output:

None

Configurable fields:

See Also:

Notes:

Example:
data = {'info': {'smappleRate': 1000.}, 0: {'raw': [0, 1, 2], 'filtered': [2, 3, 4]}, 1: {'raw': [0, 1, 2], 'filtered': [2, 3, 4]}}
store.[data2db](#)(data, ['raw', 'filtered'])

db2data(self, data_type, refine=None, tasks=False, mdataSplit=False, concatenate=False, axis=0)
Retrieves data of the specified type from database. Refine rules may be applied.

Input:
data_type (string): The data type to retrieve.
refine (dict): Refine retrieved information.
tasks (bool): If True, also return task indices for each data element.
mdataSplit (bool): Flag to separate the metadata from the signals or events.
concatenate (bool): If True, concatenate the existing datasets into a single array, along axis.
axis (int): Axis along which to concatenate the datasets.

Output:
output (dict): Output data of data_type where keys correspond to record id numbers.

Configurable fields:

See Also:

Notes:

Example:
output = store.[db2data](#)('raw')
output is {0: [{'signal': [0, 1, 2], 'mdata': {'sampleRate': 1000.}}, ...], 1: ...}
output = store.[db2data](#)('raw', refine={'_id': [0]})
output is {0: [{'signal': [0, 1, 2], 'mdata': {'sampleRate': 1000.}}, ...]}
output = store.[db2data](#)('R', refine={'_id': [0]})
output is {0: [{'timeStamps': [0, 1, 2], 'mdata': {'sampleRate': 1000.}}, ...]}

dbmetadata(self, refine=None)
Get metadata of records from database. Refine rules may be applied.

Input:
refine (dict): Refine rules
Output:
output (dict): Output metadata where keys correspond to record id numbers.

Configurable fields:

See Also:

Notes:

Example:
output = store.[dbmetada](#)()
output is {0: record 0 metadata, 1: record 1 metadata, ...}
output = store.[dbmetada](#)(refine={'_id': [0]})
output is {0: record 0 metadata}

getDataSet(self, tags=None)
Returns the IDs of the records that have the given tags (AND operator).

Input:
tags (list): List of tags.
Output:
output (list): Output list with the record IDs.

Configurable fields:

See Also:

Notes:

Example:
output = store.[getDataSet](#)(['Tag 1', 'Tag 2', ...])
output is e.g. [0, 1, 2, ...]

getSubjectInfo(self, refine=None)
Get metadata of subjects from database. Refine rules may be applied.

Input:
refine (dict): Refine rules
Output:
output (dict): Output metadata where keys correspond to subject id numbers.

Configurable fields:

See Also:

Notes:

Example:

```
output = store.getSubjectInfo()
output is {0: subject 0 metadata, 1: subject 1 metadata, ...}
output = store.getSubjectInfo(refine={'_id': [0]})
output is {0: subject 0 metadata}
```

getTTSets(self, data, tasks, trainTags, testTags)

Split the data and tasks dictionaries into train and test sets.

Input:

data (dict): Data dictionary.

tasks (dict): Tasks dictionary.

trainTags (list): List of tags for the train set.

testTags (list): List of tags for the test set.

Output:

```
output (dict): Output dictionary with keys:
    trainData (dict): Train data.
    trainTasks (dict): Train tasks.
    testData (dict): Test data.
    testTasks (dict): Test tasks.
```

Configurable fields:

See Also:

Notes:

Example:

subjectTTSets(self, info)

Returns subjects information according to the parameters in info.

Input:

```
info (dict): Input dict with the following keys:
    'train_set' (list): train tags
    'test_set' (list): test tags
    'train_time' (tuple of ints): train record time in seconds
    'test_time' (tuple of ints): test record time in seconds
```

Output:

output (list): Output list where each entry is tuple (subject id, train record id(s), test record id(s)).

Configurable fields:

See Also:

Notes:

Example:

```
info = {'train_set': ['T1'],
        'test_set': ['T2'],
        'train_time': (0, 60),
        'test_time': (0, 60)}
output = store.subjectTTSets(info)
output is [(0, [0, 1], [2]), (1, [3], [4, 5]), (2, [6], [7]), ...]
```

class H5Unit

StorageBIT (offline) database manager class.

Input:

None

Output:

None

Configurable fields:

See Also:

Notes:

Example:

Methods defined here:

__init__(self, path, mapper, experiments)

Instantiate the class.

Input:

path (string): The path to the HDF5 files.

mapper (dict): To map user variables to the database representation.

experiments (string or list of strings): The experiment(s) to load.

Output:

None

Configurable fields:

See Also:

Notes:

Example:

```
path = '/path/to/files'
mapper = {'raw': 'signals/ECG/hand/raw',
          'filtered': 'signals/ECG/hand/filtered'}
experiments = ['T1', 'T2', 'test']
store = HSUnit(dbConn, mapper, experiments)
```

addExperiment(self, experiment=None)

Add a new experiment to the database.

Input:

experiment (dict): The new experiment to add. Has to have the key 'name'.

Output:

experimentName (str): The name of the new experiment.

Configurable fields:

See Also:

Notes:

Example:

```
experiment = {'name': 'Test', 'goals': 'To test the addition of experiments.', 'description': 'This is a sample experiment.'}
experimentName = store.addExperiment(experiment)
```

addRecord(self, header)

Add a new record to the database.

Input:

header (dict): The new record to add. Has to have 'subject' and 'experiment' keys.

Output:

recordId (int): The ID of the new record.

Configurable fields:

See Also:

Notes:

Example:

```
header = {'experiment': 'test', 'subject': 1, 'date': 'yyyy-mm-ddThh:mm:ss'}
store.addRecord(header)
```

addSubject(self, subject=None)

Add a new subject to the database.

Input:

experiment (dict): The new subject to add. Has to have the key 'name'.

Output:

subjectId (int): The ID of the new subject.

Configurable fields:

See Also:

Notes:

Example:
subject = {'name': 'John Smith', 'birthdate': 'yyyy-mm-ddThh:mm:ss', 'email': 'john.smith@domain.com'}
subjectId = store.[addExperiment](#)(subject)

close(self)

Close the manager.

Input:
None

Output:
None

Configurable fields:

See Also:

Notes:

Example:
store.[close\(\)](#)

data2db(self, data, data_type, updateDuration=False)

Insert data of specified types in the database.

Input:
data (dict): Input data where keys correspond to record IDs.
There is also one key 'info' with specific information about the data.
data_type (list of strings): Input data types to be added.
updateDuration (bool): Flag to update the duration of the record. Default: False.

Output:
None

Configurable fields:

See Also:

Notes:

Example:
data = {'info': {'sampleRate': 1000.}, 0: {'raw': [0, 1, 2], 'filtered': [2, 3, 4]}, 1: {'raw': [0, 1, 2], 'filtered': [2, 3, 4]}}
store.[data2db](#)(data, ['raw', 'filtered'])

db2data(self, data_type, refine=None, tasks=False, mdataSplit=False, concatenate=False, axis=0)

Retrieves data of the specified type from database. Refine rules may be applied.

Input:
data_type (string): The data type to retrieve.
refine (dict): Refine retrieved information.
tasks (bool): If True, also return task indices for each data element.
mdataSplit (bool): Flag to separate the metadata from the signals or events.
concatenate (bool): If True, concatenate the existing datasets into a single array, along axis.
axis (int): Axis along which to concatenate the datasets.

Output:
output (dict): Output data of data_type where keys correspond to record id numbers.

Configurable fields:

See Also:

Notes:

Example:
output = store.[db2data](#)('raw')
output is {0: [{'signal': [0, 1, 2], 'mdata': {'sampleRate': 1000.}}, ...], 1: ...}
output = store.[db2data](#)('raw', refine={'_id': [0]})
output is {0: [{'signal': [0, 1, 2], 'mdata': {'sampleRate': 1000.}}, ...]}
output = store.[db2data](#)('R', refine={'_id': [0]})
output is {0: [{'timeStamps': [0, 1, 2], 'mdata': {'sampleRate': 1000.}}, ...]}

dbmetadata(self, refine=None)

Get metadata of records from database. Refine rules may be applied.

Input:
refine (dict): Refine rules

Output:
output (dict): Output metadata where keys correspond to record id numbers.

Configurable fields:

See Also:

Notes:

Example:

```
output = store.dbmeta()
output is {0: record 0 metadata, 1: record 1 metadata, ...}
output = store.dbmeta(refine={'_id': [0]})
output is {0: record 0 metadata}
```

getDataSet(self, tags=None)

Returns the IDs of the records that have the given tags (AND operator).

Input:

tags (list): List of tags.

Output:

output (list): Output list with the record IDs.

Configurable fields:

See Also:

Notes:

Example:

```
output = store.getDataSet(['Tag 1', 'Tag 2', ...])
output is e.g. [0, 1, 2, ...]
```

getSubjectInfo(self, refine=None)

Get metadata of subjects from database. Refine rules may be applied.

Input:

refine (dict): Refine rules

Output:

output (dict): Output metadata where keys correspond to subject id numbers.

Configurable fields:

See Also:

Notes:

Example:

```
output = store.getSubjectInfo()
output is {0: subject 0 metadata, 1: subject 1 metadata, ...}
output = store.getSubjectInfo(refine={'_id': [0]})
output is {0: subject 0 metadata}
```

getTTSets(self, data, tasks, trainTags, testTags)

Split the data and tasks dictionaries into train and test sets.

Input:

data (dict): Data dictionary.

tasks (dict): Tasks dictionary.

trainTags (list): List of tags for the train set.

testTags (list): List of tags for the test set.

Output:

```
output (dict): Output dictionary with keys:
    trainData (dict): Train data.
    trainTasks (dict): Train tasks.
    testData (dict): Test data.
    testTasks (dict): Test tasks.
```

Configurable fields:

See Also:

Notes:

Example:

subjectTTSets(self, info)

Returns subjects information according to the parameters in info.

Input:

```

info (dict): Input dict with the following keys:
    'train_set' (list): train tags
    'test_set' (list): test tags
    'train_time' (tuple of ints): train record time in seconds
    'test_time' (tuple of ints): test record time in seconds

Output:
    output (list): Output list where each entry is tuple (subject id, train record id(s), test record id(s)).

Configurable fields:

See Also:

Notes:

Example:
    info = {'train_set': ['T1'],
            'test_set': ['T2'],
            'train_time': (0, 60),
            'test_time': (0, 60)}
    output = store.subjectTTSets(info)
    output is [(0, [0, 1], [2]), (1, [3], [4, 5]), (2, [6], [7]), ...]

```

class MPUnit

Methods defined here:

`__init__(self, path, mapper)`

Functions

`Store(config)`

Method to obtain the appropriate database manager: StorageBIT (online or offline), MonitorPlux, GZIP.

Input:

`config (dict): Configuration parameters.`

Output:

`unit (DBUnit, H5Unit, MPUnit or GZUnit): The database manager instance.`

Configurable fields:

See Also:

Notes:

Example:

`allocH5(path)`

`copyExpResults(basePath, expID, dstPath, images=False)`

`cpLoad(path)`

`cpStore(path, data)`

`csvLoad(path)`

`csvStore(path, data)`

`deleteFile(filePath)`

`export2csv(fname, data)`

`fileBenchmark(path, it)`

`gzLoad(path)`

`gzStore(path, data)`

`h5Load(path, label)`

`h5Store(path, label, data)`

`listFiles(path, spec='*')`

`skLoad(path)`

`skStore(path, data, compress=3)`

```
taskify(data)
zipArchiveLoad(zipPath, dstPath)
zipArchiveStore(srcPath, zipPath)
```

```
.. module:: parted
:platform: Unix, Windows
:synopsis: This module provides various functions to partition data into train and test sets.

.. moduleauthor:: Carlos Carreiras
```

Modules

[copy](#)[numpy](#)

Functions

deterministicFraction(idx, run, fraction=None, order='start')

Deterministically select a fraction of the elements for the training set.

Input:

idx (list, array): Indices to partition.

run (int): Number of the run.

fraction (int, float): The fraction to select.

order (str): Determines whether the training data is selected from the start ('start', the default), or from the end ('end') of the set.

Output:

use (list, array): The selected indices.

unuse (list, array): The excluded indices.

Configurable fields:**See Also:****Example:****References:**

.. [1]

equalFraction(idx, run)

Equal sets (no separation).

Input:

idx (list, array): Indices to partition.

run (int): Number of the run.

Output:

use (list, array): The selected indices.

unuse (list, array): The excluded indices.

Configurable fields:**See Also:****Example:****References:**

.. [1]

leaveKOut(idx, run, k=1, random=True)

Leave k items out of the training set.

Input:

idx (list, array): Indices to partition.

run (int): Number of the run.

k (int): Number of elements to leave out. Default=1

random (bool): If True, select randomly the items. Default=True

Output:

use (list, array): The selected indices.

unuse (list, array): The excluded indices.

Configurable fields:**See Also:****Example:**

References:
.. [1]

randomFraction(indx, run, fraction=None)
Randomly select a fraction of the elements for the training set.

Input:
indx (list, array): Indices to partition.
run (int): Number of the run.
fraction (int, float): The fraction to select.

Output:
use (list, array): The selected indices.
unuse (list, array): The excluded indices.

Configurable fields:

See Also:

Example:

References:
.. [1]

randomSelection(indx, run, k=None)
Randomly select a k elements for the training set.

Input:
indx (list, array): Indices to partition.
run (int): Number of the run.
k (int): The number of elements to select.

Output:
use (list, array): The selected indices.
unuse (list, array): The excluded indices.

Configurable fields:

See Also:

Example:

References:
.. [1]

selector(method)
Selector for the parted functions and methods.

Input:
method (str): The desired function or method.
Output:
fcn (function): The function pointer.

Configurable fields:

See Also:

Example:

References:
.. [1]

bpkclone.devices.bitalinoTools

[d:\work\productioncode\clones\biometricspykit\bpkclone\devices\bitalinotools.py](#)

```
.. module:: bitalinoProcess
:platform: Windows
:synopsis: BITalino acquisition process.

.. moduleauthor:: Filipe Canento, Carlos Carreiras
```

Modules

[collections](#)

[numpy](#)

[time](#)

Classes

[__builtin__.object](#)

[CheckHands](#)

[CheckHands2](#)

[CheckHands3](#)

[CheckHandsNotch](#)

[OnlineFilter](#)

[OnlineFilterBA](#)

[SSFSegmenter](#)

[SegmentECG](#)

[WSPlotter](#)

[multiprocessing.Process\(__builtin__.object\)](#)

[BITalinoProcess](#)

class BITalinoProcess(multiprocessing.Process)

BITalino acquisition process.

Method resolution order:

[BITalinoProcess](#)

[multiprocessing.Process](#)

[__builtin__.object](#)

Methods defined here:

__init__(self, outQueue, goFlag, acqFlag, mac=None, channels=[0], step=100, SamplingRate=1000, timeout=5, bitCls=None)

connect(self)

exit(self)

on_start(self)

on_stop(self)

processData(self, data)

run(self)

setup(self)

trigger(self, data=[0, 0, 0, 0])

Class methods defined here:

instaStart(cls, *args, **kwargs) from [__builtin__.type](#)

Methods inherited from [multiprocessing.Process](#):

__repr__(self)

is_alive(self)

 Return whether process is alive

join(self, timeout=None)

 Wait until child process terminates

start(self)

 Start child process

terminate(self)

 Terminate process; sends SIGTERM signal or uses TerminateProcess()

Data descriptors inherited from [multiprocessing.Process](#):

__dict__

 dictionary for instance variables (if defined)

__weakref__

 list of weak references to the object (if defined)

authkey

daemon

 Return whether process is a daemon

exitcode

 Return exit code of process or `None` if it has yet to stop

ident

 Return identifier (PID) of process or `None` if it has yet to start

name

pid

 Return identifier (PID) of process or `None` if it has yet to start

class CheckHands([__builtin__.object](#))

Class to check if hands are on the ECG electrodes.

Methods defined here:

__init__(self, step=150, size=300, pad=256, signalFc=None, highFc=None, SamplingRate=1000.0, nbits=10, tol=24, delay=3)

check(self, signal)

reset(self)

Data descriptors defined here:

__dict__

 dictionary for instance variables (if defined)

__weakref__

 list of weak references to the object (if defined)

class CheckHands2([__builtin__.object](#))

Class to check if hands are on the ECG electrodes.

Methods defined here:

__init__(self, step=150, size=300, SamplingRate=1000.0, nbits=10, delay=3)

check(self, signal)

reset(self)

Data descriptors defined here:

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

Data and other attributes defined here:

HF = [45, 65]

PAD = 512

TH_BP = 0.3

TH_BV = 0.0234375

TH_HF = 0.6

TH_TH = 0.02

TH_TL = 2e-05

class **CheckHands3([__builtin__.object](#))**

Class to check if hands are on the ECG electrodes.

Methods defined here:

__init__(self, step=150, size=500, SamplingRate=1000.0, nbits=10, delay=3, bsize=5)

check(self, signal)

reset(self)

Data descriptors defined here:

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

Data and other attributes defined here:

HF = [1.953125, 3.90625]

PAD = 512

THRH = 100.0

THRL = 1.0

TH_BV = 0.0234375

class **CheckHandsNotch([__builtin__.object](#))**

Class to check if hands are on the ECG electrodes.

Methods defined here:

`__init__(self, step=150, size=300, SamplingRate=1000.0, nbits=10, delay=3)`
`check(self, signal)`
`reset(self)`
`test(self, signal)`

Data descriptors defined here:

`__dict__`
 dictionary for instance variables (if defined)
`__weakref__`
 list of weak references to the object (if defined)

Data and other attributes defined here:

`HF = [45, 65]`
`PAD = 512`
`TH_BP = 0.15`
`TH_BV = 0.0234375`
`TH_TH = 0.015529624809234088`
`TH_TL = 0.00015529624809234089`

class **OnlineFilter([__builtin__.object](#))**

Online Filtering class.

Methods defined here:

`__init__(self, truncate=False, *args, **kwargs)`
`filter(self, signal)`
`reset(self)`

Data descriptors defined here:

`__dict__`
 dictionary for instance variables (if defined)
`__weakref__`
 list of weak references to the object (if defined)

class **OnlineFilterBA([OnlineFilter](#))**

[OnlineFilter](#) with explicitly given coefficients.

Method resolution order:

[OnlineFilterBA](#)
[OnlineFilter](#)
[__builtin__.object](#)

Methods defined here:

`__init__(self, b, a, truncate=False)`

Methods inherited from [OnlineFilter](#):

filter(self, signal)

reset(self)

Data descriptors inherited from [OnlineFilter](#):

dict
dictionary for instance variables (if defined)

weakref
list of weak references to the object (if defined)

class **SSFSegmenter**([__builtin__.object](#))

Online ECG segmentation using the SSF algorithm.

Methods defined here:

init(self, SamplingRate=1000.0, size=1.5, threshold=20, before=0.03, after=0.01)

reset(self)

resetBuffer(self)

segment(self, signal)

Static methods defined here:

concatenateSegments(*args)

Data descriptors defined here:

dict
dictionary for instance variables (if defined)

weakref
list of weak references to the object (if defined)

class **SegmentECG**([__builtin__.object](#))

Online ECG segmentation.

Methods defined here:

init(self, size=1500, SamplingRate=1000.0)

reset(self)

resetBuffer(self)

segment(self, signal)

Static methods defined here:

concatenateSegments(*args)

Data descriptors defined here:

dict
dictionary for instance variables (if defined)

weakref
list of weak references to the object (if defined)

class WSPlotter([_builtin__object](#))

Prepare the signal plot for communication to the webpage via websockets.

Methods defined here:

init(self, screenSize=1280, decimation=1, scale=None)

prepare(self, signal)

reset(self)

Data descriptors defined here:

dict
dictionary for instance variables (if defined)

weakref
list of weak references to the object (if defined)

```
.. module:: dimreduction
:platform: Unix, Windows
:synopsis: This module provides various functions to...
.. moduleauthor:: Filipe Canento
```

Modules

[scipy.linalg](#)
[numpy](#)[pylab](#)
[scipy](#)[sklearn.decomposition](#)
[traceback](#)

Classes

[builtin__object](#)[lda](#)
[pca_Ind](#)
[pca_OPop](#)**class lda([builtin__object](#))**

Linear Discriminant Analysis

Methods defined here:

__init__(self)
Init

project(self, data=None)

Transforms input data. Train method must have been performed previously.

Input:
data (array): input data array of shape (number samples x number features).

Output:
transformed_data (matrix): output data array.

Configurable fields:{ "name": "dimreduction.[lda](#).train", "config": { """: ""}, "inputs": ["data"], "outputs": ["transformed_data"]}

See Also:

Notes:

Example:

References:

... [1] ...
... [2] ...
... [3] ...

train(self, data=None, label=None, energy=1)

Perform Linear Discriminant Analysis on input data.

Input:
data (array): input data array of shape (number samples x number features).

label (array): input data label array of shape (number of samples x 1). Must start in 0 (e.g., label = [0,0,0,1,2] for 5 samples).

Output:
success (boolean): indicates whether training was successful (True) or not (False).

Configurable fields:{ "name": "dimreduction.[lda](#).train", "config": { """: ""}, "inputs": ["data", "label"], "outputs": ["success"]}

See Also:

Notes:

Example:

References:

... [1] ...
... [2] ...
... [3] ...

Data descriptors defined here:

__dict__

dictionary for instance variables (if defined)

weakref
list of weak references to the object (if defined)

```
class pca_Ind(\_\_builtin\_\_.object)
    Methods defined here:

        __init__(self)
            # Init

        project(self, data=None)
            Input:
                data : test data, a 2 dimentional list or array [heartbeat, values] with the data of ONE individual
                eigen_vectors : eigenvectors' matrix coming from data training
                train_mean : mean of the training data
            Output:
                Projections : projection coeficients of each observation of the test data projected in each individual of the train set
                            *projections[testID][trainID] = set of projections of the individual testID in the individual trainID

        train(self, data=None, label=None, energy=1)
            Input:
                data : 3 dimentional list or array [individual, heartbeat, values]
                energy : value between 0 and 1 for dimentionality reduction (fraction)
```

Data descriptors defined here:

dict
dictionary for instance variables (if defined)

weakref
list of weak references to the object (if defined)

```
class pca_OPop(\_\_builtin\_\_.object)
    # Principal Component Analysis - Overall Population Eigen-Heartbeat

    Methods defined here:

        __init__(self)
            # Init

        project(self, data=None)
            Input:
                data : test data, a 2 dimentional list or array [heartbeat, values] with the segments of one individue
                eigen_vectors : eigenvectors' matrix coming from data training
                train_mean : mean of the training data
            Output:
                Projections : projection coeficients of the individual's observation in the test data

        train(self, data=None, label=None, energy=1)
            Input:
                data : 3 dimentional list or array [individual, heartbeat, values]
                energy : value between 0 and 1 for dimentionality reduction ##### nao sei se e para fazer isto aqui
```

Data descriptors defined here:

dict
dictionary for instance variables (if defined)

weakref
list of weak references to the object (if defined)

Functions

```
l1da_project(data=None, transform_matrix=None)
    Transforms input data. Train method must have been performed previously.

    Input:
        data (array): input data array of shape (number samples x number features).

    Output:
        transformed_data (matrix): output data array.

Configurable fields:{ "name": "dimreduction.l1da.train", "config": { """: ""}, "inputs": ["data"], "outputs": ["transformed_data"] }

See Also:

Notes:

Example:
```

References:
.. [1] ...
.. [2] ...
.. [3] ...

l1da_train(data=None, label=None)

l1da_train2(data=None, label=None)
Perform Linear Discriminant Analysis on input data.

Input:
data (array): input data array of shape (number samples x number features).

label (array): input data label array of shape (number of samples x 1). Must start in 0 (e.g., label = [0,0,0,1,2] for 5 samples).

Output:
success (boolean): indicates whether training was successful (True) or not (False).

Configurable fields:{ "name": "dimreduction.l1da.train", "config": { """: ""}, "inputs": ["data", "label"], "outputs": ["success"]}

See Also:

Notes:

Example:

References:
.. [1] ...
.. [2] ...
.. [3] ...

selector(method)
Selector for the dimensionality reduction functions and methods.

Input:
method (str): The desired function or method.

Output:
fcn (function): The function pointer.

Configurable fields:

See Also:

Example:

References:
.. [1]

```
.. module:: evaluation
:platform: Unix, Windows
:synopsis: This module provides various functions to...
.. moduleauthor:: Filipe Canento, Carlos Carreiras
```

Modules

[copy](#)
[scipy.interpolate](#)

[numpy](#)
[scipy.optimize](#)

[scipy](#)

Functions

assessClassification(results, rejection_thresholds, ignSubjects=[], dstPath=None, log2file=False)
assessRuns(runResults, subjects)
combineClassifiers(clf, *p)
combineSubjectClass(data, label2subject, subject2label)
computeAuthRates(TP, FP, TN, FN, thresholds)
computeIdRates(H, M, R, N, thresholds)
distanceStats(data, distanceFcn)
findEqual(x1, y1, x2, y2, alpha=1.5, xtol=1e-06, ytol=1e-06)
interDistances(data, data2=None, distanceFcn=None)
intraDistances(data, distanceFcn=None)
partialDistStats(pdata)
pdiff(x, p1, p2)
results2report(title, info, image_path, outfile='report.pdf')
subjectResults(results, subject, rejection_thresholds, subjects, subjectDict, subjectIdx)

Data

A4 = (595.275590551181, 841.8897637795275)
inch = 72.0

bpkclone.featureextraction.featureextraction [d:\work\productioncode\clones\biometricspykit\bpkclone\featureextraction\featureextra](#)

```
.. module:: featureextraction
:platform: Unix, Windows
:synopsis: This module provides various functions to...
.. moduleauthor:: Filipe Canento
```

Modules

[h5py](#)
[math](#)

[numpy](#)
[pylab](#)

[scipy](#)
[traceback](#)

Classes

[builtin__object](#)

[odinaka](#)

class odinaka([builtin__object](#))

Methods defined here:

[__init__\(self, fs, framesz, hop, sinal_length\)](#)

[classify\(self, data=None\)](#)

[train\(self, data=None, data_idx=None, label=None\)](#)

Data descriptors defined here:

[__dict__](#) dictionary for instance variables (if defined)

[__weakref__](#) list of weak references to the object (if defined)

Functions

[all_indices\(value, qlist\)](#)

[getWindow\(name, n\)](#)

[odinakaSTFT\(data, n, overlap, nfft, window, mask=None, scale=False\)](#)

[selector\(method\)](#)

Selector for the feature extraction functions and methods.

Input:
method (str): The desired function or method.

Output:
fcn (function): The function pointer.

Configurable fields:

See Also:

Example:

References:
.. [1]

[shortTimeFT\(signal, n, overlap, nfft, window, scale=False\)](#)

[stft\(x, fs, framesz, hop, nfft, ftype\)](#)

[stft0\(x, fs, framesz, hop\)](#)

[stft1\(x, fs, framesz, hop, nfft\)](#)

[subpattern\(patterns, cols, lines='all'\)](#)

```
.. module:: misc
:platform: Unix, Windows
:synopsis: This module provides miscellaneous tools.

.. moduleauthor:: Filipe Canento
```

Modules

[cPickle](#)
[ctypes](#)
[gzip](#)

[itertools](#)
[scipy.linalg](#)
[logging](#)

[numpy](#)
 [pylab](#)
[scipy](#)

[time](#)

Classes

[builtin__object](#)

[BDIterator](#)
[FiniteCycle](#)
[InfiniteBiterator](#)

[ctypes.Structure\(_ctypes._CData\)](#)

[HardwareInput](#)
[Input](#)
[KeyBdInput](#)
[MouseInput](#)

[ctypes.Union\(_ctypes._CData\)](#)

[Input_I](#)

class BDIterator([builtin__object](#))

Methods defined here:

[__init__\(self, collection, prev=0\)](#)
[__iter__\(self\)](#)
[item\(self\)](#)
[next\(self\)](#)
[prev\(self\)](#)

Data descriptors defined here:

[__dict__](#)
 dictionary for instance variables (if defined)

[__weakref__](#)
 list of weak references to the object (if defined)

class FiniteCycle([builtin__object](#))

Cycles the given finite iterable indefinitely.
Subclasses ```itertools.cycle``` and adds pickle support.

Methods defined here:

__init__(self, finite_iterable)
__reduce__(self)
__setstate__(self, state)
next(self)
peek(self)
 Return the next value in the cycle without moving the iterable forward.

Data descriptors defined here:

_dict
 dictionary for instance variables (if defined)
_weakref
 list of weak references to the object (if defined)
index

class HardwareInput([ctypes.Structure](#))

Method resolution order:

[HardwareInput](#)
[ctypes.Structure](#)
[_ctypes._CData](#)
[_builtin_.object](#)

Data descriptors defined here:

_dict
 dictionary for instance variables (if defined)
_weakref
 list of weak references to the object (if defined)
uMsg
 Structure/Union member
wParamH
 Structure/Union member
wParamL
 Structure/Union member

Methods inherited from [ctypes.Structure](#):

__init__(...)
x.[__init__](#)(...) initializes x; see help(type(x)) for signature

Data and other attributes inherited from [ctypes.Structure](#):

__new__ = <built-in method __new__ of _ctypes.PyCStructType object>
T.[__new__](#)(S, ...) -> a new [object](#) with type S, a subtype of T

Methods inherited from _ctypes._CData:

_ctypes_from_outparam(...)
__hash__(...)

```
x.__hash__() <==> hash(x)  
__reduce__(...)  
__setstate__(...)
```

```
class InfiniteBiterator(\_builtin\_.object)  
    This infinite iterator swings both ways.
```

Methods defined here:

```
__init__(self, collection)  
    The iterator infinitely loops over the given collection, either forwards or backwards.  
  
cur(self)  
    Return the current element.  
  
next(self)  
    Get the next element from the collection.  
  
prev(self)  
    Get the previous element from the collection.
```

Data descriptors defined here:

```
__dict__  
    dictionary for instance variables (if defined)  
  
__weakref__  
    list of weak references to the object (if defined)  
  
index  
    Returns current index.
```

```
class Input(\_ctypes.Structure)
```

Method resolution order:

```
Input  
\_ctypes.Structure  
\_ctypes.\_CData  
\_builtin\_.object
```

Data descriptors defined here:

```
__dict__  
    dictionary for instance variables (if defined)  
  
__weakref__  
    list of weak references to the object (if defined)  
  
ii  
    Structure/Union member  
  
type  
    Structure/Union member
```

Methods inherited from [_ctypes.Structure](#):

```
__init__(...)  
    x.__init__(...) initializes x; see help(type(x)) for signature
```

Data and other attributes inherited from [_ctypes.Structure](#):

__new__ = <built-in method __new__ of _ctypes.PyCStructType object>
T.[__new__](#)(S, ...) -> a new [object](#) with type S, a subtype of T

Methods inherited from _ctypes._CData:

__ctypes_from_outparam__(...)

__hash__(...)
x.[__hash__](#)() <==> hash(x)

__reduce__(...)

__setstate__(...)

class [Input_I](#)([_ctypes.Union](#))

Method resolution order:

[Input_I](#)
[_ctypes.Union](#)
[_ctypes._CData](#)
[_builtin_.object](#)

Data descriptors defined here:

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

hi
Structure/Union member

ki
Structure/Union member

mi
Structure/Union member

Methods inherited from [_ctypes.Union](#):

__init__(...)
x.[__init__](#)(...) initializes x; see help(type(x)) for signature

Data and other attributes inherited from [_ctypes.Union](#):

__new__ = <built-in method __new__ of _ctypes.UnionType object>
T.[__new__](#)(S, ...) -> a new [object](#) with type S, a subtype of T

Methods inherited from _ctypes._CData:

__ctypes_from_outparam__(...)

__hash__(...)
x.[__hash__](#)() <==> hash(x)

__reduce__(...)

__setstate__(...)

class [KeyBdInput](#)([_ctypes.Structure](#))

Method resolution order:

[KeyBdInput](#)
[_ctypes.Structure](#)
[_ctypes._CData](#)
[_builtin_.object](#)

Data descriptors defined here:

_dict
 dictionary for instance variables (if defined)

_weakref
 list of weak references to the object (if defined)

dwExtraInfo
 Structure/Union member

dwFlags
 Structure/Union member

time
 Structure/Union member

wScan
 Structure/Union member

wVk
 Structure/Union member

Methods inherited from [_ctypes.Structure](#):

init(...)
 x.[__init__](#)(...) initializes x; see help(type(x)) for signature

Data and other attributes inherited from [_ctypes.Structure](#):

_new = <built-in method [__new__](#) of [_ctypes.PyCStructType](#) object>
T.[__new__](#)(S, ...) -> a new [object](#) with type S, a subtype of T

Methods inherited from [_ctypes._CData](#):

_ctypes_from_outparam_(...)

hash(...)
 x.[__hash__](#)() <==> hash(x)

reduce(...)

setstate(...)

class **MouseInput**([_ctypes.Structure](#))

Method resolution order:

[MouseInput](#)
[_ctypes.Structure](#)
[_ctypes._CData](#)
[_builtin_.object](#)

Data descriptors defined here:

_dict
 dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

dwExtraInfo
Structure/Union member

dwFlags
Structure/Union member

dx
Structure/Union member

dy
Structure/Union member

mouseData
Structure/Union member

time
Structure/Union member

Methods inherited from [_ctypes.Structure](#):

__init__(...)
x.[__init__](#)(...) initializes x; see help(type(x)) for signature

Data and other attributes inherited from [_ctypes.Structure](#):

__new__ = <built-in method __new__ of _ctypes.PyCStructType object>
T.[__new__](#)(S, ...) -> a new [object](#) with type S, a subtype of T

Methods inherited from _ctypes._CData:

_ctypes_from_outparam__(...)

_hash__(...)
x.[__hash__](#)() <==> hash(x)

_reduce__(...)

_setstate__(...)

Functions

ar(x, M)

centeredFFT(x, fs, oneside=False, ax=None)

checkMouseMov(d)

chunks(l, n)

clickMouse(key)

cosdistance(s1, s2)

cosv(a, b)

dtdistance(s1, s2)

getLogger(name, logPath=None, level='debug')

get_subgroup_by_tags(st, tags)

```

keyboardPress(key, duration=0)

load_information(filename)

mean_waves(segments, nwaves, jump=0)

median_waves(data, nwaves, jump=0)

merge_clusters(data)
    Merge information from different clusters. Ignores cluster -1 (noise).

    Input:

        data (dict): input data.
            example: data = {1: {'segments': record 1 ECG segments, 'R': record 1 r peaks},
                            2: {'segments': record 2 ECG segments, 'R': record 2 r peaks},
                            ...}

        data_type (string): data type to be analyzed

        parameters (dict): filter parameters.
            Example: parameters = {'method': 'dbscan', ...}

    Output:

        output (dict): output data where keys correspond to record id numbers.
            example: output = { 1: {-1: record 1 outlier indexes, '0': record 1 cluster 0 indexes},
                                2: {-1: record 2 outlier indexes, '0': record 2 cluster 0 indexes},
                                ...}

    Configurable fields:
    {"name": "???.???", "config": {}, "inputs": ["data", "data_type", "parameters"], "outputs": ["output"]}

    See Also:

    Notes:

    Example:

moveMouse(dx, dy, absolute=True)

msedistance(s1, s2)

plot_data(data, ax, title)

plot_mean_curve(data)

pressKey(key)

quantize(signal, levels=256)

random_idx(low, high, size, exclude=[])

releaseKey(key)

save_information(info, dformat, filename=None)

slasherDict(d, path)

test_real_time()

unclickMouse(key)

wavedistance(testwave, trainwaves, fdistance)

```

Data

SendInput = <[_FuncPtr object](#)>

```
WIN_KEYS = {'*': 106, '+': 187, ',': 188, '-': 189, '!': 190, '/': 111, '0': 48, '1': 49, '2': 50, '3': 51, ...}
```

```
.. module:: outlier detection
:platform: Unix, Windows
:synopsis: This module provides various functions to...

.. moduleauthor:: Filipe Canento, Andre Lourenco
```

Modules

[scipy.spatial.distance](#) [os](#) [scipy](#)
[numpy](#) [pylab](#)

Functions

dmean(data=None, R_Position=200, alpha=0.5, beta=1.5, metric='euclidean', checkR=True)

DMEAN outlier detection heuristic.

Determines which (if any) samples in a block of ECG templates are outliers.

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;
- [optional] position of the sample maximum is the same as the given R position.

```
For a set of {X_1, ..., X_n} n samples,
Y = rac{1}{n} \sum_{i=1}^n {X_i}
d_i = dist(X_i, Y)
D_m = rac{1}{n} \sum_{i=1}^n {d_i}
D_s = \sqrt{rac{1}{n - 1} \sum_{i=1}^n {(d_i - D_m)^2}}
T = D_m + lpha * D_s
M = eta * median({\max(X_i), i=1, ..., n})
N = eta * median({\min(X_i), i=1, ..., n})
```

Input:

data (array): input data (number of samples x number of features)

R_Position (int): Position of the R peak.

alpha (float): Parameter for the distance threshold.

beta (float): Parameter for the maximum and minimum thresholds.

metric (string): Distance metric to use (euclidean or cosine).

checkR (bool): If True checks the R peak position.

Output:

'0' (list): Indices of the normal samples.

'-1' (list): Indices of the outlier samples.

Configurable fields:{}

References:

.. [1]

dmean_old(data=None, R_Position=200, alpha=0.5, metric='euclidean')

dmean_tst(data=None, R_Position=200, alpha=0.5, beta=1.5, metric='euclidean', checkR=True, absolute=False)

eleazar_eskin(data=None, metric='euclidean', threshold=0.99)

A Geometric Framework for Unsupervised Anomaly Detection:
Detecting based on Fixed width-clustering (Alg1)
intrusions in Unlabeled Data Eleazar Eskin et Al....

Input:

data (array): input data (number of observations x number of features).

metric (string): distance metric to be used.

threshold (float): threshold between 0.0 and 1.0.

Output:

res (dict): output dict with keys "normal" and "outlier" and the corresponding data indexes.

Configurable fields:{"name": "outlier.eleazar", "config": {"metric": "euclidean", "threshold": 0.99}, "inputs": ["data"]}

See Also:

Notes:

Example:

References:

.. [1] Eleazar Eskin et Al ...

fiducialMedianFilter(data=None, fil=0.4, NS=20)

Detect outliers using features and medians

Input:

 data (array): input data (number of observations x number of features).

 fil (float): percentage of median

 NS (int): number of points

Output:

 res (dict): output dict with keys "0":normal and "-1":outliers and the corresponding data indexes.

Configurable fields:{ "name": "outlier.fiducialMedianFilter", "config": { "fil": 0.4, "NS": 20}, "inputs": ["data"] }

References:

.. [1] Marta Santos et Al, "EIGEN HEARTBEATS FOR USER IDENTIFICATION"

meanfilter(data=None, metric='euclidean', th=0.8)

Detect outliers using distances over the mean

Input:

 data (array): input data (number of observations x number of features).

 metric : type of metric

 th (float): percentage to be removed

Output:

 res (dict): output dict with keys "0":normal and "-1":outliers and the corresponding data indexes.

Configurable fields:{ "name": "outlier.modefilter", "config": { "fil": 0.4, "NS": 20}, "inputs": ["data"] }

References:

.. [1]

outliers_dbscan(data=None, min_samples=10, eps=0.95, metric='euclidean')

Detect outliers using DBSCAN

Input:

 data (array): input data (number of observations x number of features).

 min_samples (int): minimum number of samples in a cluster.

 eps (float): maximum distance between two samples in the same cluster.

 metric (string): distance metric

Output:

 res (dict): output dict with keys "0":normal and "-1":outliers and the corresponding data indexes.

Configurable fields:

{ "name": "outlier.outliers_dbscan", "config": { "metric": "euclidean", "eps": 0.99, "min_samples": 10}, "inputs": ["data"] }

See Also:

BiometricsPyKit.cluster.dbscan

Notes:

Example:

plotECG(data, res, figProps)

removeOutliers(data=None, outliers=None)

Method to remove the outliers from a given data set.

Input:

 data (array): The initial data set.

```
    outliers (dict): The dictionary with the outlier labels.

Output:
    templates (array): The data set without the outliers.

Configurable fields:

See Also:

Example:

References:
    .. [1]

run(data, tasks, parameters, outPath)
    Run the outlier detection methods for each item in data, according to the given parameters.

Input:
    data

    tasks

    parameters

    outPath

Output:
    output

Configurable fields:{"name": "???.???", "config": {}, "inputs": [], "outputs": []}

See Also:

Notes:

Example:

runMethod(data=None, method=None, **kwargs)
    Runs an outlier detection and removal method.

Input:
    data (array): The initial data set.

    method (str): The desired function or method.

    Additional kwargs are passed to the outlier detection method.

Output:
    templates (array): The data set without the outliers.

    partition (dict): The partition dictionary between normal and outlier templates.

    nbGood (int): Number of normal templates

Configurable fields:

See Also:

Example:

References:
    .. [1]

selector(method)
    Selector for the outlier detection functions and methods.

Input:
    method (str): The desired function or method.

Output:
    fcn (function): The function pointer.

Configurable fields:

See Also:

Example:

References:
```



.. [1]

```
.. module:: preprocessing
:platform: Unix, Windows
:synopsis: This module provides various functions to...
.. moduleauthor:: Filipe Canento, Carlos Carreiras
```

Modules

[numpy](#)[scipy](#)[scipy.signal](#)[scipy.stats](#)

Functions

PLA(signal=None, step=1, epsilon=0.1)

butter(signal=None, low_cutoff=None, high_cutoff=None, zi=None, order=4, sampling_rate=1000.0)
Butterworth filtering.

Input:

 signal (array): input signal.

 'order' (int): filter order.

 'low_cutoff' (float): low cutoff frequency.

 'high_cutoff' (float): high cutoff frequency.

 'sampling rate' (float): sampling rate.

 'zi' (array): initial conditions

Output:

 output_signal (array): output FIR-filtered signal. Output signal length = length(signal) - filter order (n)

Configurable fields:

{ "name": "preprocessing.butter", "config": { "order": 4, "sampling_rate": 1000., "zi": None }, "inputs": ["signal", "low_cutoff", "high_cutoff"] }

See Also:

Notes:

Example:

References:

.. [1]

butterFilt(signal=None, low_cutoff=None, high_cutoff=None, order=1, sampling_rate=1000.0)
Butterworth IIR filtering. Uses filtfilt to avoid phase distortions.

Input:

 signal (array): input signal.

 low_cutoff (float): low cutoff frequency.

 high_cutoff (float): high cutoff frequency.

 order (int): filter order.

 sampling_rate (float): sampling rate.

Output:

 output_signal (array): output filtered signal.

Configurable fields:

{ "name": "preprocessing.butterfilt", "config": { "order": 300, "sampling_rate": 1000. }, "inputs": ["signal", "low_cutoff", "high_cutoff"], "outputs": ["output_signal"] }

See Also:

 scipy.signal.filtfilt

 scipy.signal.butter

Notes:

Example:

References:

.. [1]

butterNotch(signal=None, center=None, width=None, order=1, sampling_rate=1000.0, harmonics=False)
Butterworth IIR notch filtering. Uses filtfilt to avoid phase distortions.

Input:

 signal (array): input signal.

 low_cutoff (float): low cutoff frequency.

```

    high_cutoff (float): high cutoff frequency.
    order (int): filter order.
    sampling_rate (float): sampling rate.
    harmonics (bool): Flag to also remove harmonics; the default is False.

    Output:
        output_signal (array): output FIR-filtered signal.

    Configurable fields:
    {"name": "preprocessing.firfilt", "config": {"order": 300, "sampling_rate": 1000.}, "inputs": ["signal", "low_cutoff", "high_cutoff"], "outputs": ["output"]}

    See Also:
        scipy.signal.filtfilt
        scipy.signal.butter

    Notes:

    Example:

    References:
        .. [1]

do_filter_work(work_queue, output)

filter_data(data, parameters)
    Filters input data according to specified parameters.

    Input:
        data (dict): input data.
            example: data = {1: record 1 signal, 2: record 2 signal, ...}
        parameters (dict): filter parameters.
            Example: parameters = {'filter': 'butter',
                                    'order': 4,
                                    'low_cutoff': 100.,
                                    'sampling_rate': 1000.}

    Output:
        output (dict): output filtered data where keys correspond to record id numbers.
            example: output = {1: record 1 filtered signal, 2: record 2 filtered signal}

    Configurable fields: {"name": "???.???", "config": {}, "inputs": ["data", "parameters"], "outputs": ["output"]}

    See Also:

    Notes:

    Example:

firfilt(signal=None, low_cutoff=None, high_cutoff=None, order=300, sampling_rate=1000.0)
    FIR Filtering. Uses firwin and lfilter from scipy.signal.

    Input:
        signal (array): input signal.
        low_cutoff (float): low cutoff frequency.
        high_cutoff (float): high cutoff frequency.
        order (int): filter order.
        sampling_rate (float): sampling rate.

    Output:
        output_signal (array): output FIR-filtered signal. Output signal length = length(signal) - filter order (n)

    Configurable fields:
    {"name": "preprocessing.firfilt", "config": {"order": 300, "sampling_rate": 1000.}, "inputs": ["signal", "low_cutoff", "high_cutoff"], "outputs": ["output"]}

    See Also:
        scipy.signal.firwin
        scipy.signal.lfilter

    Notes:

    Example:

    References:
        .. [1]

heartRateExtractor(data, sift, selection=None)

heartRateSifter(signal, R, Fs=1000.0, edges=None)

medFIR(signal, sampleRate=1000.0)

```

```

norm1(data)

norm2(data)

norm3(data)

norm4(data, R=200)

norm5(data)

norm6(data)

norm7(data)

norm8(data, R=200)

norm9(data, R=200, sigma=1.5)

normArcLength(data)

normSelector(method)

normalWindow(nb, mu, sigma)

normalization(data, ntype)

notchFilter(signal=None, center=None, width=None, order=300, sampling_rate=1000.0, harmonics=False)
    FIR notch Filtering. Uses firwin and lfilter from scipy.signal.

    Input:
        signal (array): input signal.
        low_cutoff (float): low cutoff frequency.
        high_cutoff (float): high cutoff frequency.
        order (int): filter order.
        sampling_rate (float): sampling rate.
        harmonics (bool): Flag to also remove harmonics; the default is False.

    Output:
        output_signal (array): output FIR-filtered signal. Output signal length = length(signal) - filter order (n)

    Configurable fields:
    {"name": "preprocessing.firfilt", "config": {"order": 300, "sampling_rate": 1000.}, "inputs": ["signal", "low_cutoff", "high_cutoff"], "outputs": ["output_signal"]}

    See Also:
        scipy.signal.firwin
        scipy.signal.lfilter

    Notes:

    Example:

    References:
        .. [1]

```

patterns2strings(patterns)

sparseNorm(data)

Data

NUMBER_PROCESSES = 5

```
.. module:: visualization
:platform: Unix, Windows
:synopsis: This module provides various functions to...
.. moduleauthor:: Carlos Carreiras, Andre Lourenco
```

Modules

[matplotlib.cm](#)
[copy](#)

[csv](#)
[numpy](#)

[matplotlib.pyplot](#)
[pylab](#)

[scipy](#)

Functions

EERCurves(results, idx, figsize=None)

FARFRRCurves(ths, results, figsize=None)

addEERCurve(ax, thresholds, rates, variables, lw=1, colors=None, alpha=1, drawEER=False, labels=False, scale=1.0)

addFARFRR(ax, thresholds, rates, lw=1, colors=None, alpha=1, drawEER=False, labels=False)

buildCSVTable(filePath, resultsCollection)

buildLatexTable(resultsCollection, keyOrder)

clfConfusionMatrix(subjects, rejection_thresholds, clfResults, assessResults)

close(fig=None)

clusterOverTime(clusterlabels, linelabels=None, xlabel=None, ylabel=None, ncols=20, figureNumber=1, figsize=None)

confusionMatrix(confMatrix)

figure(*args, **kwargs)

multiBarPLOT(data, xticklabels, yerr=None, labels=None, xlabel=None, ylabel=None, xlim=None, ylim=None, width=0.15, loc='best', legendAnchor=None, figsize=None, rotation=30, xtickssize=12, vlines=False, btext=False, btextsize=12, xgap=3.0)

multiBarPLOTAxis(ax, data, xticklabels, yerr=None, labels=None, xlabel=None, ylabel=None, xlim=None, ylim=None, width=0.15, loc='best', legendAnchor=None, rotation=30, xtickssize=12, vlines=False, btext=False, btextsize=12, xgap=3.0)

multiBarPLOTAxisText(ax, data, ypos=0, textalpha=4.5)

nestedMultiBarPLOT(data, xticklabels1, xticklabels2, yerr=None, labels=None, xlabel1=None, xlabel2=None, ylabel=None, xlim=None, ylim=None, width=0.15, loc='best', legendAnchor=(0.75, 1.115), figsize=None, rotation=30)

plotClusters(data, clusters, figsize=None, alpha=1)
Plot a clustering partition.

Kwargs:
data (array): The (2D) feature array; each line corresponds to a sample.
clusters (dict): The data partition.
figsize (tuple): The size of the figure to create.
alpha (float): The transparency of the plot lines.

Kwrvals:
fig (matplotlib.figure.Figure): The figure object.

See Also:

Notes:

Example:

References:
.. [1]

plotMeanWaves(data, labels=None, figureNumber=1)

```
plotOutliers(data, partition, alpha=0.7, figsize=None)
plotSegments(data, labels, linewidth=2, legenda=None, figureNumber=1, figsize=None)
plotTTTemplates(x, trainData, testData, alpha=0.7, linewidth=1, xlabel=None, ylabel=None, title=None, loc='best', figsize=None)
rocCurves(results, outfile=None, figNum=None, label=None)
selector(method)
show()
singleFARFRRCurve(thresholds, rates, figsize=None)
superFARFRRCurves(results, figsize=None)
```

```
.. module:: wavelets
:platform: Unix, Windows
:synopsis: This module provides various functions to ...
.. moduleauthor:: Carlos Carreiras
```

Modules

[numpy](#)[pylab](#)[pywt](#)[scipy.signal](#)

Functions

RDWT(signal, wavelet, level, matrix=True, **kwargs)

Perform the Redundant Discrete Wavelet Transform (RDWT). This transform is different from the Discrete Wavelet Transform (DWT) by skipping the decimation step at each decomposition level. One advantage of RDWT over DWT is that it is time-invariant.

This transform is also known as:

- Stationary wavelet transform,
- Algorithme a trous,
- Quasi-continuous wavelet transform,
- Translation invariant wavelet transform,
- Shift invariant wavelet transform,
- Undecimated wavelet transform.

Kwargs:

- signal (array): The input signal.
- wavelet (string, pywt.Wavelet): The mother wavelet.
- level (int): The number of decomposition levels.
- matrix (bool): Flag to return the output in matrix form. Default: True.

Kwrvals:

See Also:

Notes:

The approximation and detail coefficients are arranged by level: [(cAn, cDn), ..., (cA1, cD1)]. In matrix form the output is [cD1, cD2, ..., cDn, cAn].

Example:

```
signal = ones(1000)
wavelet = pywt.Wavelet('rbio3.3')
coefficients, H, G = RDWT(signal, wavelet, 3)
```

References:

.. [1] Fowler, The Redundant Discrete Wavelet Transform and Additive Noise, 2005

RPeaks(matrix, sampleRate)

Determine locations of R peaks in the wavelet domain.

Kwargs:

- matrix (array): Set of coefficients.
- sampleRate (float): The sample rate.

Kwrvals:

See Also:

Notes:

Example:

References:

.. [1] Rooijakkers, Low-complexity R-peak Detection in ECG Signals, 2011

applyThreshold(coefficients, threshold, hard=False, last=False)

Function to apply a threshold to each of the decomposition levels.

Kwargs:

coefficients (array): List of the wavelet decomposition coefficients in matrix form.

threshold (list): List of length L with the threshold to apply to each level.

hard (bool): Flag to perform hard thresholding. Default: False.

last (bool): Flag to also apply the threshold to the last level. Default: False.

Kwrvals:

See Also:

Notes:

Example:

References:

.. [1]

changThr(coefficients)

Compute the Chang Threshold for each level of the RDWT. Noise is estimated from the first level.

Kwargs:

coefficents (array): The coefficients from RDWT in matrix form.

Kwrvals:

See Also:

Notes:

Example:

References:

.. [1] Chang, Adaptive Wavelet Thresholding for Image Denoising and Compression, 2000

coeffs2Matrix(coefficients)

Convert the coefficients list to a matrix format. Note that only the approximation coefficients from the last level are kept.

Kwargs:

coefficents (list): The coefficients from RDWT.

Kwrvals:

See Also:

Notes:

Example:

References:

.. [1]

conv(x, h)

Convolution operator. To avoid boundary effects, the input signal is padded (length of the filter) at both ends.

Kwargs:

x (array): The input signal.

`h` (list): The FIR filter.

Kwrvals:

See Also:

Notes:

Example:

References:
.. [1]

`cutLevel(coefficients, level)`

Set detail coefficients to zero up to (excluding) a given level.

Kwargs:

`coefficients` (array): List of the wavelet decomposition coefficients in matrix form.
`level` (int): The level at which to cut.

Kwrvals:

See Also:

Notes:

Example:

References:
.. [1]

`extractSegments(matrix, R, sampleRate)`

Extract the wavelet segments.

Kwargs:

`matrix` (array): Set of coefficients.
`R` (list): List of indexes locating the R peaks.
`sampleRate` (float): The sample rate.

Kwrvals:

See Also:

Notes:

Example:

References:
.. [1]

`iRDWT(coefficients, wavelet, zeros=False)`

The inverse Redundant Discrete Wavelet Transform.

Kwargs:

`coefficients` (array): The coefficients as produced by the RDWT function (matrix form).
`wavelet` (string, pywt.Wavelet): The mother wavelet.
`zeros` (bool): Flag to replace approximation coefficients with zeros.

Kwrvals:

See Also:

Notes:
The approximation and detail coefficients are arranged by level: [(cAn, cDn), ..., (cA1, cD1)].

Example:

```
signal = ones(1000)
wavelet = pywt.Wavelet('rbio3.3')
coefficients, H, G = RDWT(signal, wavelet, 3)
back = iRDWT(coefficients, H, G)
```

References:

.. [1]

maxLevel(signal, wavelet)

Determine the maximum decomposition level that is possible to apply the RDWT to a given signal, with the specified mother wavelet.

Kwargs:

 signal (int, array): The input signal, or the length of the signal to decompose.
 wavelet (string, pywt.Wavelet): The mother wavelet.

Kwrvals:

See Also:

Notes:

Example:

References:

.. [1] Fowler, The Redundant Discrete Wavelet Transform and Additive Noise, 2005

mchangThr(coefficients)

Compute the modified Chang Threshold for each level of the RDWT. Noise is estimated for each level.

Kwargs:

 coefficents (array): The coefficients from RDWT in matrix form.

Kwrvals:

See Also:

Notes:

Example:

References:

.. [1] Chang, Adaptive Wavelet Thresholding for Image Denoising and Compression, 2000

meanWave(data, axis=2)

medianWave(data, axis=2)

muniversalThr(coefficients)

Compute the modifief Universal Threshold for each level of the RDWT.

Kwargs:

 coefficents (array): The coefficients from RDWT in matrix form.

Kwrvals:

See Also:

Notes:

Example:

References:
.. [1] Donoho, De-Noising by Soft Thresholding, 1995

plotCoeffs(signal, coefficients, start=None, stop=None, step=None)
Plot the decomposition.

Kwargs:
signal (array): The original signal.
coefficients (array): Set of coefficients in matrix form.
start (int): Starting time index to plot. Default: None.
stop (int): Ending time index to plot. Default: None.
step (int): The time step to plot. Default: None.

Kwrvals:

See Also:

Notes:

Example:

References:
.. [1]

plotOutliers(data=None, outliersResult=None, offset=0)
Plot the results of the outlier removal procedure.

Kwargs:
data (array): The segments of wavelet coefficients.
outliersResult (dict): Result of the outlier removal procedure;
has keys '0' (good) and '-1' (bad).
offset (int): Offset to correct the display of the levels. Default: 0

Kwrvals:

See Also:

Notes:

Example:

References:
.. [1]

plotWaveletSpectrum(wavelet, level, sampleRate, npoints=1024, start=None, stop=None)
Plot the transfer function (amplitude) of the wavelet decomposition filters.

Kwargs:
wavelet (string, pywt.Wavelet): The mother wavelet.
level (int): The number of decomposition levels.
sampleRate (float): The sample rate.
npoints (int): Number of points in which to compute the transfer function. Default: 1024.
start (int): Starting frequency to plot. Default: None.
stop (int): Ending frequency to plot. Default: None.

Kwrvals:

See Also:

Notes:

Example:

References:

.. [1]

rcosdmean(data=None, th_min=1.5, th_max=1.5, th_dist=0.5)

Outlier detection based on the mean segment and the medians of maxima and minima (wavelet version).

Kwargs:

 data (array): The segments of wavelet coefficients.

 th_min (float): Parameter for median of minima. Default: 1.5

 th_max (float): Parameter for median of maxima. Default: 1.5

 th_dist (float): Parameter for distances. Default: 0.5

Kwrvals:

See Also:

Notes:

Example:

References:

.. [1]

reverse(x)

Method to reverse a signal.

y[n] = x[-n]

Kwargs:

 x (array): The input signal.

Kwrvals:

See Also:

Notes:

Example:

References:

.. [1]

trous(h)

Add zeros (holes - trous) between the filter coefficients.

Kwargs:

 h (list): The FIR filter.

Kwrvals:

See Also:

Notes:

Example:

References:
.. [1]

universalThr(coefficients)

Compute the Universal Threshold for each level of the RDWT.

Kwargs:
coefficients (array): The coefficients from RDWT in matrix form.

Kwrvals:

See Also:

Notes:

Example:

References:
.. [1] Donoho, De-Noising by Soft Thresholding, 1995

waveDist(coefficients1, coefficients2, thr=400.0, levels=None, indexes=None)

Compute distance between a pair of wavelet coefficients (in matrix form).

Kwargs:
coefficients1 (array): First set of coefficients.
coefficients2 (array): Second set of coefficients.
levels (list, slice): Levels to use in the distance (e.g [0, 1, 2]). Default: all levels.
indexes (list, slice): Time indexes to use in the distance. Default: all.

Kwrvals:

See Also:

Notes:

Example:

References:
.. [1] Chan, Wavelet disntance measure for person identification using ECG, 2008

waveletSegments(matrix, sampleRate, **kwargs)

Determine R peak locations and extract the wavelet segments.

Kwargs:
matrix (array): Set of coefficients.
sampleRate (float): The sample rate.

Kwrvals:
segments (array): The segmented wavelet coefficients (time, levels, segments).
R (list): List of indexes locating the R peaks.

See Also:

Notes:

Example:

References:
.. [1]

```
.. module:: vitalidiFramework
:platform: Unix, Windows
:synopsis: This module interfaces between the Vitalidi demonstrator Python back-end and BiometricsPyKit.

.. moduleauthor:: Ana Priscila Alves, Carlos Carreiras
```

Modules

[copy](#)
[datetime](#)

[glob](#)
[json](#)

[numpy](#)
[os](#)

[shutil](#)
[traceback](#)

Classes

[__builtin__.object](#)

[BaseBrain](#)

[DoorBrain](#)
[KeyBrain](#)
[MixBrain](#)
[MyBrain](#)
[QuickBrain](#)

[multiprocessing.Process\(__builtin__.object\)](#)

[RunEvaluate](#)

```
class BaseBrain(\_\_builtin\_\_.object)
    Base class for Vitalidi brains.
```

Methods defined here:

__init__(self, path, settings, ignorePrevious=False, deployLocations=None)

authenticate(self, data, subject, ready=False)

checkSubject(self, subject)

combination(self, results)

createClassifiers(self)

debugTest(self, data)

deploy(self, deployLocations=None)

evaluate(self, recordTypes=None)

groupAuthenticate(self, data, names=None, ready=False)

identify(self, data, ready=False)

listClassSubjects(self)

listSubjects(self)

loadSegments(self, subject, recordTypes)

```
loadSettings(self, settings)
prepareData(self, data)
removeSubject(self, subject, updateThresholds=False)

saveFile(self, command, name, Rawsignal=None, FilteredSignal=None, Segments=None, currentTime=None,
eventValue=None, good_segments=None, res=None, mode='a')

saveOutliersPlot(self, segments, partition, currentTime=None)

subjectPerformance(self, subject)

train(self, data, subject, updateThresholds=True)

trainFromFiles(self, recordTypes=None, ignoreSubjects=None)

trainFromFolder(self, path=None, ignoreSubjects=None)

updateFile(self, name)

updateThresholds(self)

validateAutoAuth(self, segments)

validation(self, nsegments, segments, case, autoAuth=False)

validationStatus(self, nsegments, case)
```

Class methods defined here:

getCurrentTime(cls) from [__builtin__.type](#)

Static methods defined here:

listFolderSubjects(path)

Data descriptors defined here:

__dict__
 dictionary for instance variables (if defined)

__weakref__
 list of weak references to the object (if defined)

Data and other attributes defined here:

CLF_NAMES = []

CLF_WEIGHTS = {}

```
class DoorBrain(BaseBrain)
    Brain using an SVM classifier with 5 median waves.
    Templates are scaled with a gaussian window.
```

Method resolution order:

[DoorBrain](#)
[BaseBrain](#)
[__builtin__.object](#)

Methods defined here:

prepareData(self, data)

Data and other attributes defined here:

CLF_NAMES = ['5-medianWaves']

CLF_WEIGHTS = {'5-medianWaves': 1.0}

Methods inherited from [BaseBrain](#):

__init__(self, path, settings, ignorePrevious=False, deployLocations=None)

authenticate(self, data, subject, ready=False)

checkSubject(self, subject)

combination(self, results)

createClassifiers(self)

debugTest(self, data)

deploy(self, deployLocations=None)

evaluate(self, recordTypes=None)

groupAuthenticate(self, data, names=None, ready=False)

identify(self, data, ready=False)

listClassSubjects(self)

listSubjects(self)

loadSegments(self, subject, recordTypes)

loadSettings(self, settings)

removeSubject(self, subject, updateThresholds=False)

saveFile(self, command, name, Rawsignal=None, FilteredSignal=None, Segments=None, currentTime=None, eventValue=None, good_segments=None, res=None, mode='a')

saveOutliersPlot(self, segments, partition, currentTime=None)

subjectPerformance(self, subject)

train(self, data, subject, updateThresholds=True)

trainFromFiles(self, recordTypes=None, ignoreSubjects=None)

trainFromFolder(self, path=None, ignoreSubjects=None)

updateFile(self, name)

updateThresholds(self)

validateAutoAuth(self, segments)

validation(self, nsegments, segments, case, autoAuth=False)

validationStatus(self, nsegments, case)

Class methods inherited from [BaseBrain](#):

getCurrentTime(cls) from [__builtin__.type](#)

Static methods inherited from [BaseBrain](#):

listFolderSubjects(path)

Data descriptors inherited from [BaseBrain](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class KeyBrain([BaseBrain](#))

Brain using SVM classifiers, combining all segments, 3 median waves and 5 median waves.
Templates are scaled with a gaussian window.

Method resolution order:

[KeyBrain](#)
[BaseBrain](#)
[__builtin__.object](#)

Methods defined here:

prepareData(self, data)

Data and other attributes defined here:

CLF_NAMES = ['segments', '3-medianWaves', '5-medianWaves']

CLF_WEIGHTS = {'3-medianWaves': 2.0, '5-medianWaves': 3.0, 'segments': 1.0}

Methods inherited from [BaseBrain](#):

__init__(self, path, settings, ignorePrevious=False, deployLocations=None)

authenticate(self, data, subject, ready=False)

checkSubject(self, subject)

combination(self, results)

createClassifiers(self)

debugTest(self, data)

deploy(self, deployLocations=None)

evaluate(self, recordTypes=None)

groupAuthenticate(self, data, names=None, ready=False)

identify(self, data, ready=False)

listClassSubjects(self)

listSubjects(self)

loadSegments(self, subject, recordTypes)

```
loadSettings(self, settings)
removeSubject(self, subject, updateThresholds=False)
saveFile(self, command, name, Rawsignal=None, FilteredSignal=None, Segments=None, currentTime=None,
eventValue=None, good_segments=None, res=None, mode='a')
saveOutliersPlot(self, segments, partition, currentTime=None)
subjectPerformance(self, subject)
train(self, data, subject, updateThresholds=True)
trainFromFiles(self, recordTypes=None, ignoreSubjects=None)
trainFromFolder(self, path=None, ignoreSubjects=None)
updateFile(self, name)
updateThresholds(self)
validateAutoAuth(self, segments)
validation(self, nsegments, segments, case, autoAuth=False)
validationStatus(self, nsegments, case)
```

Class methods inherited from [BaseBrain](#):

getCurrentTime(cls) from [builtin.type](#)

Static methods inherited from [BaseBrain](#):

listFolderSubjects(path)

Data descriptors inherited from [BaseBrain](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class **MixBrain**([BaseBrain](#))

Brain

Method resolution order:

[MixBrain](#)
[BaseBrain](#)
[builtin.object](#)

Methods defined here:

prepareData(self, data)

Data and other attributes defined here:

CLF_NAMES = ['segments', '3-medianWaves-norm', '3-medianWaves', '5-medianWaves']

CLF_WEIGHTS = {'3-medianWaves': 1.0, '3-medianWaves-norm': 1.0, '5-medianWaves': 2.0, 'segments': 1.0}

Methods inherited from [BaseBrain](#):

__init__(self, path, settings, ignorePrevious=False, deployLocations=None)
authenticate(self, data, subject, ready=False)
checkSubject(self, subject)
combination(self, results)
createClassifiers(self)
debugTest(self, data)
deploy(self, deployLocations=None)
evaluate(self, recordTypes=None)
groupAuthenticate(self, data, names=None, ready=False)
identify(self, data, ready=False)
listClassSubjects(self)
listSubjects(self)
loadSegments(self, subject, recordTypes)
loadSettings(self, settings)
removeSubject(self, subject, updateThresholds=False)
saveFile(self, command, name, Rawsignal=None, FilteredSignal=None, Segments=None, currentTime=None, eventValue=None, good_segments=None, res=None, mode='a')
saveOutliersPlot(self, segments, partition, currentTime=None)
subjectPerformance(self, subject)
train(self, data, subject, updateThresholds=True)
trainFromFiles(self, recordTypes=None, ignoreSubjects=None)
trainFromFolder(self, path=None, ignoreSubjects=None)
updateFile(self, name)
updateThresholds(self)
validateAutoAuth(self, segments)
validation(self, nsegments, segments, case, autoAuth=False)
validationStatus(self, nsegments, case)

Class methods inherited from [BaseBrain](#):

getCurrentTime(cls) from [_builtin_.type](#)

Static methods inherited from [BaseBrain](#):

listFolderSubjects(path)

Data descriptors inherited from [BaseBrain](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class **MyBrain**([BaseBrain](#))

Brain using SVM classifiers, combining all segments, 3 mean-waves and 5 mean-waves.

Method resolution order:

[MyBrain](#)
[BaseBrain](#)
[_builtin__object](#)

Methods defined here:

prepareData(self, data)

Data and other attributes defined here:

CLF_NAMES = ['segments', '3-meanWaves', '5-meanWaves']

CLF_WEIGHTS = {'3-meanWaves': 1.0, '5-meanWaves': 1.0, 'segments': 1.0}

Methods inherited from [BaseBrain](#):

__init__(self, path, settings, ignorePrevious=False, deployLocations=None)

authenticate(self, data, subject, ready=False)

checkSubject(self, subject)

combination(self, results)

createClassifiers(self)

debugTest(self, data)

deploy(self, deployLocations=None)

evaluate(self, recordTypes=None)

groupAuthenticate(self, data, names=None, ready=False)

identify(self, data, ready=False)

listClassSubjects(self)

listSubjects(self)

loadSegments(self, subject, recordTypes)

loadSettings(self, settings)

removeSubject(self, subject, updateThresholds=False)

saveFile(self, command, name, Rawsignal=None, FilteredSignal=None, Segments=None, currentTime=None, eventValue=None, good_segments=None, res=None, mode='a')

saveOutliersPlot(self, segments, partition, currentTime=None)

subjectPerformance(self, subject)

```
train(self, data, subject, updateThresholds=True)  
trainFromFiles(self, recordTypes=None, ignoreSubjects=None)  
trainFromFolder(self, path=None, ignoreSubjects=None)  
updateFile(self, name)  
updateThresholds(self)  
validateAutoAuth(self, segments)  
validation(self, nsegments, segments, case, autoAuth=False)  
validationStatus(self, nsegments, case)
```

Class methods inherited from [BaseBrain](#):

getCurrentTime(cls) from [__builtin__.type](#)

Static methods inherited from [BaseBrain](#):

listFolderSubjects(path)

Data descriptors inherited from [BaseBrain](#):

__dict__
 dictionary for instance variables (if defined)
__weakref__
 list of weak references to the object (if defined)

class QuickBrain([BaseBrain](#))

Brain using SVM classifiers, combining all segments, 3 mean-waves and 5 mean-waves.

Method resolution order:

[QuickBrain](#)
[BaseBrain](#)
[__builtin__.object](#)

Methods defined here:

prepareData(self, data)

Data and other attributes defined here:

CLF NAMES = ['5-meanWaves']
CLF WEIGHTS = {'5-meanWaves': 1.0}

Methods inherited from [BaseBrain](#):

__init__(self, path, settings, ignorePrevious=False, deployLocations=None)
authenticate(self, data, subject, ready=False)
checkSubject(self, subject)
combination(self, results)

```
createClassifiers(self)

debugTest(self, data)

deploy(self, deployLocations=None)

evaluate(self, recordTypes=None)

groupAuthenticate(self, data, names=None, ready=False)

identify(self, data, ready=False)

listClassSubjects(self)

listSubjects(self)

loadSegments(self, subject, recordTypes)

loadSettings(self, settings)

removeSubject(self, subject, updateThresholds=False)

saveFile(self, command, name, Rawsignal=None, FilteredSignal=None, Segments=None, currentTime=None,
eventValue=None, good_segments=None, res=None, mode='a')

saveOutliersPlot(self, segments, partition, currentTime=None)

subjectPerformance(self, subject)

train(self, data, subject, updateThresholds=True)

trainFromFiles(self, recordTypes=None, ignoreSubjects=None)

trainFromFolder(self, path=None, ignoreSubjects=None)

updateFile(self, name)

updateThresholds(self)

validateAutoAuth(self, segments)

validation(self, nsegments, segments, case, autoAuth=False)

validationStatus(self, nsegments, case)
```

Class methods inherited from [BaseBrain](#):

getCurrentTime(cls) from [__builtin__.type](#)

Static methods inherited from [BaseBrain](#):

listFolderSubjects(path)

Data descriptors inherited from [BaseBrain](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

class **RunEvaluate**([multiprocessing.process.Process](#))

Run classifier evaluation

Method resolution order:

[RunEvaluate](#)
[multiprocessing_process.Process](#)
[builtin__object](#)

Methods defined here:

__init__(self, brainCls=None, stateDict=None, *args, **kwargs)
run(self)

Methods inherited from [multiprocessing_process.Process](#):

__repr__(self)
is_alive(self)
 Return whether process is alive
join(self, timeout=None)
 Wait until child process terminates
start(self)
 Start child process
terminate(self)
 Terminate process; sends SIGTERM signal or uses TerminateProcess()

Data descriptors inherited from [multiprocessing_process.Process](#):

__dict__
 dictionary for instance variables (if defined)
__weakref__
 list of weak references to the object (if defined)
authkey
daemon
 Return whether process is a daemon
exitcode
 Return exit code of process or `None` if it has yet to stop
ident
 Return identifier (PID) of process or `None` if it has yet to start
name
pid
 Return identifier (PID) of process or `None` if it has yet to start

Data

ERR_MSG = {'AUT-0': 'Authentication: Success', 'AUT-1': 'Authentication: Error: UntrainedError', 'AUT-2': 'Authentication: Error: UnknownSubjectError', 'AUT-3': 'Authentication: Error: Combination confidence below threshold', 'AUT-4': 'Authentication: Error: Empty combination', 'ID-0': 'Identification: Success', 'ID-1': 'Identification: Error: UntrainedError', 'ID-2': 'Identification: Error: Default subject', 'ID-3': 'Identification: Error: Combination confidence below threshold', 'ID-4': 'Identification: Error: Empty combination', ...}
settings = {'MAC': '', 'autoAuth': {'k': 0, 'threshold': 0}, 'classification': {'clf_0': {'method': '', 'parameters': {}}, 'clf_N': {'method': '', 'parameters': {}}, 'confidence_threshold': 0}, 'outlier': {'enroll': {'method': '', 'parameters': {}}, 'id': {'method': ''}}

```
", "parameters": {}}, "validation": {"enroll": {"minimum_templates": 0, "time": 0}, "id": {"minimum_templates": 0, "time": 0}}}
```

Package Contents

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