Package ‘ORION’

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Description Functions to handle ordinal relations reflected within the feature space. Those functions allow to search for ordinal relations in multi-class datasets. One can check whether proposed relations are reflected in a specific feature representation. Furthermore, it provides functions to filter, organize and further analyze those ordinal relations.
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as.edgedataframe

Description

Converts from a Subcascades object to a weighted edge list.

Usage

as.edgedataframe(subcascades)
Arguments

subcascades A Subcascades object as it is returned by subcascades-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type '1>2>3' and the entries the sensitivity for each position of the cascade.

Details

Converts a Subcascades object to a data.frame that can be used to generate a graph.

Value

A data.frame that can be used to generate a graph. The first and second column correspond to all pairwise relations (from - to) of the cascades within the Subcascades object. The 'CASC_ID' column contains the same ID for all edges belonging to the same cascade. The 'SIZE' column gives the size of the cascade to which the respective column belongs. The method returns NULL if the object subcascades is empty.

See Also

groupwise, as.subcascades

Examples

library(TunePareto)
library(igraph)
data(esl)
data = esl$data
labels = esl$labels
predMap = predictionMap(data, labels,
classifier = tunePareto.svm(), kernel='linear')
# generate a dataframe
subcascades = subcascades(predMap, thresh=0.65, size=4)
edges = as.edgedataframe(subcascades)
g = graph_from_data_frame(edges[,c(1,2)], directed = TRUE)
E(g)$weight = edges[,3]
plot(g, edge.color=edges[,3], edge.arrow.size=0.5,
     edge.curved = seq(-0.5, 1, length = ecount(g)))
Usage

as.subcascades(groupwise = NULL)

Arguments

groupwise A Groupwise object, which comprises a two-leveled list. The first level collects cascades of the same size. The second level contains a list of unique class combinations, labelled as a character string with '-' separating the different classes. For each unique set of class combinations the corresponding orders and their performance are given as a matrix, where each row contains a cascade, given as a character string of type '1>2>3', and the columns the sensitivity for the class at the corresponding position. Each matrix is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing).

Details

Converts a Groupwise object to a Subcascades object.

Value

A Subcascades object comprising the evaluated cascades and their performances. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type '1>2>3' and the entries the sensitivity for each position of the cascade.

See Also

groupwise, as.edgedataframe

Examples

library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
   ntimes = 2,
   nfold = 2,
   leaveOneOut = FALSE,
   stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
   classifier = tunePareto.svm(), kernel='linear')

# generate a Groupwise object
subc = subcascades(predMap,thresh=0.7)
groupwise = groupwise(subc)

#convert it to a Subcascades object
converted.subcascades = as.subcascades(groupwise)
**Description**

Sensitivities of all pairwise binary classifiers for all classes.

**Usage**

```r
conf(predictionMap = NULL)
```

**Arguments**

- `predictionMap` A PredictionMap object as it is returned by `predictionMap`-function. It is made up of a list of two matrices (pred and meta). Both matrices provide information on individual samples column-wise. The rownames of the pred-matrix (e.g. [0vs1]) show the classes of the binary base classifier. The elements are the prediction result of a specific training. The rows that correspond to base classifiers that would separate the same class consists of -1. Those rows are not used within the analysis. The meta information connects the values in the pred-matrix to a specific fold, run and contains the original label.

**Details**

The Conf object contains all class sensitivities for each binary trained classifiers. The $fC$-part is a matrix with one column and contains the sensitivities for the first class of each pairwise classifier. The rows stand for the pairwise classifiers, whereby 0vs1 means that this classifier was trained for class 0 against class 1, with class 0 being the first class. The number '-1' is used as placeholder. The $sC$-part is a matrix and contains the performance measures for all second classes, which are meant here as all classes except the first class. The rows correspond to the binary classifiers and the columns to the classes.

**Value**

Object of class Conf. Consists of a list of two numeric matrices fC and sC.

**See Also**

- `summary.Conf`, `print.Conf`, `plot.Conf`

**Examples**

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels, ntimes = 2,
```
confusionTable

```r
default = 2,
leaveOneOut = FALSE,
stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
classifier = tunePareto.svm(), kernel='linear')
conf = conf(predMap)
```

---

**confusionTable**

*Construction of a Confusion Table*

**Description**

Confusion table and class assignments of one cascade.

**Usage**

```r
confusionTable(
  predictionMap = NULL,
  cascade = NULL,
  other.classes = NULL,
  sort = TRUE
)
```

**Arguments**

- `predictionMap` A PredictionMap object as it is returned by `predictionMap` function. It is made up of a list of two matrices (pred and meta). Both matrices provide information on individual samples column-wise. The rownames of the pred-matrix (e.g. [0vs1]) show the classes of the binary base classifier. The elements are the prediction result of a specific training. The rows that correspond to base classifiers that would separate the same class consists of -1. Those rows are not used within the analysis. The meta information connects the values in the pred-matrix to a specific fold, run and contains the original label.

- `cascade` A numeric vector of classes or a character string of type '1>2>3' of at least two class labels reflected in `predictionMap`.

- `other.classes` This parameter can be either NULL, 'all' or a numeric vector of classes that are not part of the cascade parameter. If other.classes is: - NULL, only the cascade classes are evaluated. - 'all', all remaining classes are evaluated. - a vector of classes, those classes are evaluated.

- `sort` If TRUE (default) the classes that are not part of cascade are sorted based on their confusion.

**Value**

An object of type ConfusionTable including the sensitivities, with the label of the predicted classes in the rows and the labels of the original class in the columns.
See Also

summary.ConfusionTable, print.ConfusionTable, plot.ConfusionTable

Examples

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
                       ntimes = 2,
nfold = 2,
leaveOneOut = FALSE,
stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
classifier = tunePareto.svm(), kernel='linear')

# Calculation of the confusion matrix for '0>2>3>4'.
confTable = confusionTable(predMap, cascade = '0>2>3>4')
# Calculation of the confusion matrix for '0>2>3>4'
# and the assignment of all samples of the other classes.
confTable = confusionTable(predMap, cascade = '0>2>3>4',
other.classes='all', sort = TRUE)
```

dropSets  

Filter out Subcascades

Description

dropSets filters out specific cascade sets.

Usage

```r
dropSets(
  subcascades = NULL,
  sets = NULL,
  direction = "sub",
  ordered = FALSE,
  neighborhood = "direct",
  type = "any"
)
```

Arguments

subcascades A Subcascades object as it is returned by subcascades-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The
dropSets

rownames show the class order by a character string of type ’1>2>3’ and the entries the sensitivity for each position of the cascade.

sets Contains the set used for filtering. It is either a list of numeric vectors, a numeric vector, or a vector of characters representing a cascade of the following format ’1>2>4’. Empty vectors are not allowed.
direction Either ’sub’, ’super’ or ’exact’, indicating whether the subset, the superset or the exact set is filtered. The ‘exact’ case filters exactly the cascades defined in sets. Please refer to the details section for a detailed description.
ordered Either TRUE or FALSE indicating whether the order of the classes in sets is considered or not.
neighborhood Either ‘direct’ or ‘indirect’ defines whether the given classes have to be direct neighbors (’direct’) or there are other classes allowed inbetween (’indirect’).
type If more than one cascade is defined in sets, this parameter defines whether the filtered cascade has to fit to at least one of the sets (’any’) or to all of the given sets (’all’).

Details

This function allows for filtering the Subcascades object.

If direction is set to ’exact’ the parameter neighborhood is ignored and the parameter type is used as its default and cannot be changed. There has to be an exact match between the classes of the sets parameter cascades and the cascade of the Subcascades object, so not more but also not less classes are allowed. If the ordered parameter is set to TRUE also the order of the classes within the sets parameter and within the Subcascades object has to be exactly the same.

If direction is set to ’sub’ the Subcascades object is filtered for subsets of the given cascades. If the type parameter is set to ’any’ each cascade is considered individually otherwise all cascades are used as reference for filtering. This means that either for each cascade of the sets parameter individually or for taking all together the Subcascades object is filtered for cascades that are made up of the same classes, a subset of classes (ordered = FALSE) or a cascade, part of a cascade (ordered = TRUE), resulting in a set of cascades that might contain less classes. The parameter neighbourhood is not considered.

If direction is set to ’super’ the Subcascades object is filtered for a superset of the given cascades. If the type parameter is set to ’any’ each cascade is considered individually otherwise all cascades are used as reference for filtering. A superset are cascades that are made up of (parts) of the given cascades (ordered = TRUE) or classes (ordered = FALSE) but can contain also more classes. Depending on the parameter neighborhood only classes defined in the sets parameter are allowed as neighbors (’direct’). If neighborhood is set to ’indirect’ the filtering is less strict and the direct neighborhood defined in sets is not considered.

Value

A Subcascades object comprising the evaluated cascades and their performances. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type ’1>2>3’ and the entries the sensitivity for each position of the cascade.
See Also

dropSize, keepSize, keepSets, dropThreshold, keepThreshold

Examples

library(TunePareto)

data(esl)

data = esl$data

t = esl$labels

foldList = generateCVRuns(labels = t,
                         ntimes = 2,
                         nfold = 2,
                         leaveOneOut = FALSE,
                         stratified = TRUE)

predMap = predictionMap(data, labels, foldList = foldList,
                         classifier = tunePareto.svm(), kernel='linear')

# generate Subcascades object

tsc = subcascades(predMap, thresh=0.7)

# define sets

sP = list(c(1,2,3), c(2,3,4))

sQ = c('1>2>3', '2>3>4')

# filter for the subset cascades that contain either the classes # {1,2,3} or {2,3,4} independent of the order, but neighbored
dropSets(tsc, sets = sP, direction = 'sub',
         ordered = FALSE, neighborhood = 'direct')
dropSets(tsc, sets = sQ, direction = 'sub',
         ordered = FALSE, neighborhood = 'direct')

# filter for the superset cascades that contain either the classes # {1,2,3} or {2,3,4} independent of the order, but neighbored
dropSets(tsc, sets = sP, direction = 'super',
         ordered = FALSE, neighborhood = 'direct')
dropSets(tsc, sets = sQ, direction = 'super',
         ordered = FALSE, neighborhood = 'direct')

# filter for the superset cascades that contain both the classes # {1,2,3} and {2,3,4} in exactly the given order, but allowing # for other classes inbetween

dropSets(tsc, sets = sP, direction = 'super',
         ordered = TRUE, neighborhood = 'indirect', type = 'all')
dropSets(tsc, sets = sQ, direction = 'super',
         ordered = TRUE, neighborhood = 'indirect', type = 'all')

# filter for the exact cascades

# sets can be a numeric list

dropSets(tsc, list(c(0,1,2),c(2,3,4,1)),
         direction = 'exact', ordered=TRUE)

unlist(t(lapply(result, rownames)))

# or sets can be a character vector

result <- dropSets(tsc, c('0>1>2','2>3>4>1'),
         direction = 'exact', ordered=TRUE)

unlist(t(lapply(result, rownames)))
dropSize

Description

Filters out the Subcascades object for the given sizes.

Usage

dropSize(subcascades = NULL, size = NA)

Arguments

subcascades  A Subcascades object as it is returned by subcascades-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type ‘1>2>3’ and the entries the sensitivity for each position of the cascade.

size  A numeric value that defines the size of the cascades that should be returned. The smallest size is 2 and the largest the maximal number of classes of the current dataset.

Value

A Subcascades object as in subcascades, that does not include cascades of the specific lengths that have been filtered.

See Also

keepSize, dropSets, keepSets, dropThreshold, keepThreshold

Examples

library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
    ntimes = 2,
    nfold = 2,
    leaveOneOut = FALSE,
    stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
    classifier = tunePareto.svm(), kernel='linear')
# generate Subcascades object

direction = 'exact', ordered=TRUE)
unlist(t(lapply(result, rownames)))
dropThreshold

```r
subc = subcascades(predMap, thresh=0.7)

# filters out cascades that have a length of 3
dropSize(subc, size=3)
# filters out cascades that have a length of 3 or 4
dropSize(subc, size=c(3,4))
```

### dropThreshold

**Exclude Cascades Based on Threshold**

**Description**

Filters out all cascades that match the comparison with a minimal classwise sensitivity threshold.

**Usage**

```r
dropThreshold(subcascades = NULL, comparison = ">=", thresh = 0)
```

**Arguments**

- `subcascades` A Subcascades object as it is returned by `subcascades`-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type '1>2>3' and the entries the sensitivity for each position of the cascade.
- `comparison` Defines the comparison type (<,>,<=,>=) for the threshold.
- `thresh` A numeric value between 0 and 1. The minimal sensitivity threshold used to filter the returned cascades. Only cascades that pass this threshold are returned. If 0 is used the returned cascades are filtered for >0 and otherwise >= thresh. For low thresholds the calculation lasts longer.

**Value**

A Subcascades object comprising the evaluated cascades and their performances. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type ‘1>2>3’ and the entries the sensitivity for each position of the cascade.

**See Also**

`dropSize, keepSize, dropSets, keepSets, keepThreshold`
Examples

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
                          ntimes = 2,
nfold = 2,
leaveOneOut = FALSE,
stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
classifier = tunePareto.svm(), kernel='linear')
# generate Subcascades object
subc = subcascades(predMap, thresh=0.5)
# filters for cascades that
# 1. have a minimal classwise sensitivity >= 0.6
dropThreshold(subc, thresh=0.6)
# 2. have a minimal classwise sensitivity <= 0.6
dropThreshold(subc, comparison = '<=', thresh=0.6)
```

---

### esl

**ESL Dataset**

**Description**

The original ESL data set contains 488 profiles of applicants for certain industrial jobs. Expert psychologists of a recruiting company, based upon psychometric test results and interviews with the candidates, determined the values of the input attributes. The output is the an overall score corresponding to the degree of fitness of the candidate to this type of job. Here classes 1-3 are merged to one group and all classes >6 are merged to one group. Furthermore, the data is relabeled to start at 0 and only 20 samples per class are included.

**Usage**

```r
data(esl)
```

**Format**

An object of class list of length 2.

**Source**

[Weka dataset](datasets-arie_ben_david.tar.gz, 11,348 Bytes)
References

ordinal, real-world datasets donated by Dr. Arie Ben David (Holon Inst. of Technology/Israel)
Donor: Arie Ben David MIS, Dept. of Technology Management Holon Academic Inst. of Technology 52 Golomb St. Holon 58102 Israel abendav@hait.ac.il Owner: Yoav Ganzah Business Administration School Tel Aviv Univerity Ramat Aviv 69978 Israel

Examples

data(esl)

---

esl_org  ESL Dataset

Description

The ESL data set contains 488 profiles of applicants for certain industrial jobs. Expert psychologists of a recruiting company, based upon psychometric test results and interviews with the candidates, determined the values of the input attributes. The output is the an overall score corresponding to the degree of fitness of the candidate to this type of job.

Usage

data(esl_org)

Format

An object of class data.frame with 488 rows and 5 columns.

Source

Weka dataset (datasets-arie_ben_david.tar.gz, 11,348 Bytes)

References

ordinal, real-world datasets donated by Dr. Arie Ben David (Holon Inst. of Technology/Israel)
Donor: Arie Ben David MIS, Dept. of Technology Management Holon Academic Inst. of Technology 52 Golomb St. Holon 58102 Israel abendav@hait.ac.il Owner: Yoav Ganzah Business Administration School Tel Aviv Univerity Ramat Aviv 69978 Israel

Examples

data(esl_org)
**groupwise**  
*Coerce to a Groupwise object*

### Description

Converts from a Subcascades object to a Groupwise object.

### Usage

```r
groupwise(subcascades = NULL, maxCl = 50)
```

### Arguments

- `subcascades`: A Subcascades object as it is returned by `subcascades`-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type '1>2>3' and the entries the sensitivity for each position of the cascade.

- `maxCl`: An integer defining the lower bound for the maximal number of classes. Has only to be set if the analyzed dataset has more than 50 classes.

### Details

This function re-sorts the Subcascades object in a way that the cascades made up of the same classes are grouped.

### Value

A Groupwise object, which comprises a two-leveled list. The first level collects cascades of the same size. The second level contains a list of unique class combinations, labelled as a character string with '-' separating the different classes. For each unique set of class combinations the corresponding orders and their performance are given as a matrix, where each row contains a cascade, given as a character string of type '1>2>3', and the columns the sensitivity for the class at the corresponding position. Each matrix is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing).

### See Also

`as.subcascades`, `summary.Groupwise`, `print.Groupwise`, `plot.Groupwise`, `as.edgedataframe`

### Examples

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
```
foldList = generateCVRuns(labels = labels,
    ntimes = 2,
    nfold = 2,
    leaveOneOut = FALSE,
    stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
    classifier = tunePareto.svm(), kernel='linear')

# generate Subcascades object
subcascades = subcascades(predMap, thresh=0.7)

# create a Groupwise object
groupwise = groupwise(subcascades)

---

keepSets | **Filter Subcascades**

**Description**
keepSets filters specific cascade sets.

**Usage**
keepSets(
    subcascades = NULL,
    sets = NULL,
    direction = "sub",
    ordered = FALSE,
    neighborhood = "direct",
    type = "any"
)

**Arguments**

- **subcascades** A Subcascades object as it is returned by `subcascades`-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type '1>2>3' and the entries the sensitivity for each position of the cascade.

- **sets** Contains the set used for filtering. It is either a list of numeric vectors, a numeric vector, or a vector of characters representing a cascade of the following format '1>2>3'. Empty vectors are not allowed.

- **direction** Either 'sub','super' or 'exact', indicating whether the subset, the superset or the exact set is filtered. The 'exact' case filters exactly the cascades defined in `sets`. Please refer to the details section for a detailed description.

- **ordered** Either TRUE or FALSE indicating whether the order of the classes in `sets` is considered or not.
keepSets neighborhood  Either 'direct' or 'indirect' defines whether the given classes have to be direct neighbors ('direct') or there are other classes allowed inbetween ('indirect').

type  If more than one cascade is defined in sets, this parameter defines whether the filtered cascade has to fit to at least one of the sets ('any') or to all of the given sets ('all').

Details
This function allows for filtering the Subcascades object.

If direction is set to 'exact' the parameter neighborhood is ignored and the parameter type is used as its default and cannot be changed. There has to be an exact match between the classes of the sets parameter cascades and the cascade of the Subcascades object, so not more but also not less classes are allowed. If the ordered parameter is set to TRUE also the order of the classes within the sets parameter and within the Subcascades object has to be exactly the same.

If direction is set to 'sub' the Subcascades object is filtered for subsets of the given cascades. If the type parameter is set to 'any' each cascade is considered individually otherwise all cascades are used as reference for filtering. This means that either for each cascade of the sets parameter individually or for taking all together the Subcascades object is filtered for cascades that are made up of the same classes, a subset of classes (ordered = FALSE) or a cascade, part of a cascade (ordered = TRUE), resulting in a set of cascades that might contain less classes. The parameter neighborhood is not considered.

If direction is set to 'super' the Subcascades object is filtered for a superset of the given cascades. If the type parameter is set to 'any' each cascade is considered individually otherwise all cascades are used as reference for filtering. A superset are cascades that are made up of (parts) of the given cascades (ordered = TRUE) or classes (ordered=FALSE) but can contain also more classes. Depending on the parameter neighborhood only classes defined in the sets parameter are allowed as neighbors ('direct'). If neighborhood is set to 'indirect' the filtering is less strict and the direct neighborhood defined in sets is not considered.

Value
A Subcascades object comprising the evaluated cascades and their performances. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type '1>2>3' and the entries the sensitivity for each position of the cascade.

See Also
dropSize, keepSize, dropSets, dropThreshold, keepThreshold

Examples

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels, ntimes = 2,
```
keepSize

Filter for Size

Description

Filters the Subcascades object for cascades of the given sizes.
Usage

keepSize(subcascades = NULL, size = NA)

Arguments

subcascades  A Subcascades object as it is returned by subcascades-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type ‘1>2>3’ and the entries the sensitivity for each position of the cascade.

size  A numeric value that defines the size of the cascades that should be returned. The smallest size is 2 and the largest the maximal number of classes of the current dataset.

Value

A Subcascades object as in subcascades, that only includes cascades of specific lengths.

See Also

dropSize, dropSets, keepSets, dropThreshold, keepThreshold

Examples

library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
ntimes = 2,
nfold = 2,
leaveOneOut = FALSE,
stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
classifier = tunePareto.svm(), kernel='linear')

# generate Subcascades object
subc = subcascades(predMap, thresh=0.7)

# filters for cascades that have a length of 3
keepSize(subc, size=3)

# filters for cascades that have a length of 3 or 4
keepSize(subc, size=c(3,4))
**Description**

Filters for all cascades that match the comparison with a minimal classwise sensitivity threshold.

**Usage**

```r
keepThreshold(subcascades = NULL, thresh = 0, comparison = "\geq")
```

**Arguments**

- **subcascades**: A Subcascades object as it is returned by `subcascades`-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type ‘1>2>3’ and the entries the sensitivity for each position of the cascade.

- **thresh**: A numeric value between 0 and 1. The minimal sensitivity threshold used to filter the returned cascades. Only cascades that pass this threshold are returned. If 0 is used the returned cascades are filtered for >0 and otherwise >= thresh. For low thresholds the calculation lasts longer.

- **comparison**: Defines the comparison type (<,>,<=,>=) for the threshold.

**Value**

A Subcascades object comprising the evaluated cascades and their performances. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type ‘1>2>3’ and the entries the sensitivity for each position of the cascade.

**See Also**

- dropSize,
- keepSize,
- dropSets,
- keepSets,
- dropThreshold

**Examples**

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
        ntimes = 2,
nfold = 2,
leaveOneOut = FALSE,
stratified = TRUE)
```
predMap = predictionMap(data, labels, foldList = foldList,
        classifier = tunePareto.svm(), kernel='linear')

# generate Subcascades object
subc = subcascades(predMap, thresh=0.5)

# filters for cascades that
# 1. have a minimal classwise sensitivity >= 0.6
keepThreshold(subc, thresh=0.6)
# 2. have a minimal classwise sensitivity <= 0.6
keepThreshold(subc, comparison = '<=', thresh=0.6)

mergeSubcascades  Merge Subcascades

Description
mergeSubcascades adds the cascades from subcascades2 to the subcascades1, that have not been part of subcascades1.

Usage
mergeSubcascades(subcascades1 = NULL, subcascades2 = NULL)

Arguments
subcascades1  A Subcascades object as it is returned by subcascades-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length. The rownames show the class order and the entries the sensitivity for each position of the cascade.

subcascades2  A Subcascades object like subcascades1

Value
A Subcascades object comprising the evaluated cascades and their performances. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type '1>2>3' and the entries the sensitivity for each position of the cascade.

Examples
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
        ntimes = 2,
        nfold = 2,
        leaveOneOut = FALSE,
nonOrdinalData

stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
classifier = tunePareto.svm(), kernel='linear')

# make two Subcascades objects
subc1 = subcascades(predMap, size = c(3,4), thresh = 0.6)
subc2 = subcascades(predMap, size = c(4), thresh = 0.5)
# add the cascades of subcascades2 to subcascades1
mergeSubcascades(subc1, subc2)

---

nonOrdinalData  non-ordinal data

Description

It comprises a list of two elements: A data matrix ("data") and a label vector ("labels"). The samples are given in the rows and the features in the columns of the matrix. The sampled dataset comprises 5 classes with 40 samples, each. It is embedded in a two dimensional feature space and contains 5 classes with no ordinal sequence.

Usage

data(nonOrdinalData)

Format

An object of class list of length 2.

Examples

data(nonOrdinalData)

---

ordinalData  ordinal data

Description

It comprises a list of two elements: A data matrix ("data") and a label vector ("labels"). The samples are given in the rows and the features in the columns of the matrix. The sampled dataset comprises 5 classes with 40 samples, each. It is embedded in a two dimensional feature space and contains 1 ordinal sequence with 5 classes.

Usage

data(ordinalData)
An object of class list of length 2.

Examples

```
data(ordinalData)
```

Description

Plots a heatmap that shows base classifier performance.

Usage

```r
## S3 method for class 'Conf'
plot(
  x = NULL,
  classNames = NULL,
  onlySens = FALSE,
  symmetric = FALSE,
  main = "summary of base classifiers",
  xlab = "classes",
  ylab = "base classifiers",
  digits = 3,
  ignore = 0,
  first.colors = c("#f5f5f5", "#b2182b"),
  second.colors = c("#f5f5f5", "#2166ac"),
  other.colors = c("#f5f5f5", "#1b7837"),
  na.color = c("yellow"),
  las = 1,
  srt = 30,
  color.key = FALSE,
  cex = 1,
  cex.lab = 1,
  ...
)
```

Arguments

- `x`: A Conf object as it is returned by `conf`-function.
- `classNames`: Vector of the original class names. If not given the class number is used instead.
- `onlySens`: A logical indicating if the plot should should be restricted to the class-wise sensitivities of the base classifiers.
symmetric Logical indicating whether a symmetric base classifier (TRUE) is used. Only used when onlySens is TRUE.
main See plot.
xlab A title for the x axis (see plot).
ylab A title for the y axis (see plot).
digits Integer indicating the number of decimal places to be used (see round).
ignore A numeric value between 0 and 1. All confusion and purity values below this number are not written as string into the corresponding element.
first.colors A 2-element vector of the color for the minimal and maximal class-wise sensitiv-
ty of the first class. The color palette is calculated by an interpolation between the 2 given colors.
second.colors A 2-element vector of the color for the minimal and maximal class-wise sen-
sitivity of the second class. The color palette is calculated by an interpolation between the 2 given colors.
other.colors A 2-element vector of the color for the minimal and maximal class-wise sen-
sitivity of the other class. The color palette is calculated by an interpolation between the 2 given colors.
na.color Color, which is used for indicating the empty elements (if the given class is not part of the cascade).
las See par.
srt Angle used to rotate the strings of the x-axis and y-axis labels (see par).
color.key Specifies whether a color key is drawn (TRUE) or not (FALSE).
cex See par.
cex.lab See par.
... Further arguments passed from other methods.

Details
This function plots a heatmap of the base classifier performance.

Value
No return value, called to generate the heatmap plot.

See Also
conf, plot.Subcascades, plot.ConfusionTable

Examples
library(TunePareto)
data(esl)
data <- esl$data
labels <- esl$labels
foldList <- generateCVRuns(labels = labels,
predMap <- predictionMap(data, labels, foldList = foldList, classifier = tunePareto.svm(), kernel='linear')

# generate Subcascades object
conf <- conf(predMap)

plot(conf, onlySens=TRUE, symmetric=TRUE)

---

**plot.ConfusionTable**  
Extended Confusion Table Plot

### Description
Plots an extended confusion table.

### Usage
```r
## S3 method for class 'ConfusionTable'
plot(
x = NULL,  
classNames = NULL,  
main = "extended confusion table",  
xlab = "original class labels",  
ylab = "predicted class labels",  
cascLab = "inner classes",  
otherLab = "outer classes",  
digits = 3,  
ignore = 0,  
casc.colors = c("#f5f5f5", "#8c510a"),  
other.colors = c("#f5f5f5", "#01665e"),  
colSep = "#b2182b",  
las = 1,  
color.key = TRUE,  
cex = 1,  
cex.lab = 1,  
...  
)
```

### Arguments
- **x**: A ConfusionTable object as it is returned by `confusionTable`-function.
- **classNames**: Vector of the original class names. If not given the class number is used instead.
- **main**: See `plot`.
- **xlab**: A title for the x axis (see `plot`).
plot.ConfusionTable

ylab A title for the y axis (see plot).
cascLab Character string used as header for the cascade part of the extended confusion table.
otherLab Character string used as header for the other class part of the extended confusion table.
digits Integer indicating the number of decimal places to be used (see round).
ignore A numeric value between 0 and 1. All confusion and purity values below this number are not written as string into the corresponding element.
casc.colors A 2-element vector of the color for the minimal and maximal class-wise sensitivity of the first class. The color palette is calculated by an interpolation between the 2 given colors.
other.colors A 2-element vector of the color for the minimal and maximal class-wise sensitivity of the second class. The color palette is calculated by an interpolation between the 2 given colors.
colSep Color, which is used for the vertical line separating the cascade classes and the other classes.
las See par.
color.key Specifies whether a color key is drawn (TRUE) or not (FALSE).
cex See par.
cex.lab See par.
...
Further arguments passed from other methods.

Value
No return value, called to generate the confusion table plot.

See Also
confusionTable, plot.Subcascades, plot.Con

Examples

library(TunePareto)
data(esl)
data <- esl$data
labels <- esl$labels
foldList <- generateCVRuns(labels = labels,
ntimes = 2,
nfold = 2,
leaveOneOut = FALSE,
stratified = TRUE)
predMap <- predictionMap(data, labels, foldList = foldList,
classifier = tunePareto.svm(), kernel='linear')
# generate Subcascades object
cconf.table <- confusionTable(predMap,cascade='0>1>3>4',other.classes = 'all')
plot(conf.table)
plot.Groupwise  

Heatmap of a Groupwise Object

Description

Plots a heatmap, that shows the performance of cascades of a Groupwise object.

Usage

```r
## S3 method for class 'Groupwise'
plot(
  x = NULL,
  class.sort = "",
  row.sort = "sens",
  main = "groupwise",
  xlab = "classes",
  ylab = "cascades",
  digits = 3,
  ignore = 0,
  casc.colors = c("#f5f5f5", "#01665e"),
  na.color = c("#f5f5f5"),
  color.key = TRUE,
  las = 1,
  cex = 1,
  cex.lab = 1,
  srt = 30,
  ...
)
```

Arguments

- **x**: A Groupwise object as it is returned by `groupwise`-function.
- **class.sort**: The classes can be sorted either according to the first cascade in the Groupwise object (""), or based on the class frequency ("max").
- **row.sort**: The cascade can be sorted either based on the maximal-minimal class-wise sensitivity ("sens") or according the class frequency.
- **main**: See `plot`.
- **xlab**: A title for the x axis (see `plot`).
- **ylab**: A title for the y axis (see `plot`).
- **digits**: Integer indicating the number of decimal places to be used (see `round`).
- **ignore**: A numeric value between 0 and 1. All confusion and purity values below this number are not written as string into the corresponding element.
- **casc.colors**: A 2-element vector of the color for the minimal and maximal class-wise sensitivity. The color palette is calculated by an interpolation between the 2 given colors.
Plotting a heatmap for Groupwise objects

```r
plot.Groupwise
```

- **na.color**: Color, which is used for indicating the empty elements (if the given class is not part of the cascade).
- **color.key**: Specifies whether a color key is drawn (TRUE) or not (FALSE).
- **las**: See `par`.
- **cex**: See `par`.
- **cex.lab**: See `par`.
- **srt**: Angle used to rotate the strings of the x-axis and y-axis labels (see `par`).
- **...**: Further arguments passed from other methods.

**Details**

This function plots a heatmap with the cascades of the Groupwise object in the rows and all classes present in any of the cascades in the columns. The colors indicate whether a given class is present in the corresponding cascade and with which sensitivity. Internally converts the Groupwise object to a Subcascades object and plots the corresponding heatmap.

**Value**

No return value, called to generate the heatmap plot of the Groupwise Object.

**See Also**

`groupwise`, `subcascades`, `plot.Subcascades`

**Examples**

```r
library(TunePareto)
data(esl)
data <- esl$data
labels <- esl$labels
foldList <- generateCVRuns(labels = labels, ntimes = 2, nfold = 2, leaveOneOut = FALSE, stratified = TRUE)
predMap <- predictionMap(data, labels, foldList = foldList, classifier = tunePareto.svm(), kernel='linear')
# generate Subcascades object
subc <- subcascades(predMap, thresh=0.7, size=c(3,4))
groupwise <- groupwise(subc)
plot(groupwise, row.sort='max')
```
plot.PredictionMap  

Heatmap of a PredictionMap Object

Description

Plots a heatmap, that shows the predictions of a PredictionMap object and the real labels in a cross-validation or reclassification experiment.

Usage

```r
## S3 method for class 'PredictionMap'
plot(
  x = NULL,
  xlab = "samples",
  ylab = "base classifiers",
  main = "Prediction map",
  las = 1,
  srt = 30,
  cex = 1,
  cex.lab = 1,
  label.colors = NULL,
  plot.sampleIDs = FALSE,
  plot.cv.runs = TRUE,
  plot.class.labels = TRUE,
  ...
)
```

Arguments

- `x`: A PredictionMap object as it is returned by `predictionMap`-function.
- `xlab`: A title for the x axis (see `plot`).
- `ylab`: A title for the y axis (see `plot`).
- `main`: See `plot`.
- `las`: See `par`.
- `srt`: Angle used to rotate the strings of the x-axis and y-axis labels (see `par`).
- `cex`: See `par`.
- `cex.lab`: See `par`.
- `label.colors`: A vector of the color for the class labels. If NULL, automatically use rainbow color scheme.
- `plot.sampleIDs`: Specifies if the sample IDs should be plotted along the x axis (TRUE or FALSE).
- `plot.cv.runs`: Specifies if the cross-validation runs should be plotted (TRUE or FALSE). Cross-validation runs are visually separated by straight lines.
- `plot.class.labels`: Specifies if the numerical class labels should additionally plotted (TRUE or FALSE).
- `...`: Further arguments passed from other methods.
Details
This function plots a heatmap with color-decoded predictions made by the specified classifier. Here, the rows indicate the different binary base classifiers and the columns the samples in the specified resampling experiment (reclassification or cross-validation). Labels are visualized in the top row and decoded color-wise.

Value
No return value, called to a heatmap plot of the predictionMap Object.

See Also
predictionMap

Examples
library(TunePareto)
data(esl)
data <- esl$data
labels <- esl$labels
foldList <- generateCVRuns(labels = labels,
                          ntimes = 2,
                          nfold = 2,
                          leaveOneOut = FALSE,
                          stratified = TRUE)
predMap <- predictionMap(data, labels, foldList = foldList,
classifier = tunePareto.svm(), kernel='linear')

plot(predMap)
```r
xlab = "cascades",
digits = 3,
ignore = 0,
casc.colors = c("#f5f5f5", "#01665e"),
na.color = c("#f5f5f5"),
color.key = TRUE,
las = 1,
cex = 1,
cex.lab = 1,
srt = 30,
...)
```

### Arguments

- **x**: A Subcascades object as it is returned by `subcascades`-function.
- **classNames**: Character vector including the names of the classes.
- **class.sort**: The classes can be sorted either by: 
  - "" = unsorted, 
  - "max": sorted according to maximal occurrence of a class or 
  - "sens": sorted according to highest class-wise sensitivity
- **row.sort**: The cascade can be sorted by: 
  - "" = unsorted, 
  - "max": sorted according to maximal occurrence of maximal length or 
  - "sens": sorted according to highest class-wise sensitivity
- **main**: See `plot`.
- **xlab**: A title for the x axis (see `plot`).
- **ylab**: A title for the y axis (see `plot`).
- **digits**: Integer indicating the number of decimal places to be used (see `round`).
- **ignore**: A numeric value between 0 and 1. All confusion and purity values below this number are not written as string into the corresponding element.
- **casc.colors**: A 2-element vector of the color for the minimal and maximal class-wise sensitivity. The color palette is calculated by an interpolation between the 2 given colors.
- **na.color**: Color, which is used for indicating the empty elements (if the given class is not part of the cascade).
- **color.key**: Specifies whether a color key is drawn (TRUE) or not (FALSE).
- **las**: See `par`.
- **cex**: See `par`.
- **cex.lab**: See `par`.
- **srt**: Angle used to rotate the strings of the x-axis and y-axis labels (see `par`).
- **...**: Further arguments passed from other methods.

### Details

This function plots a heatmap with the cascades of the Subcascades object in the rows and all classes present in any of the cascades in the columns. The colors indicate whether a given class is present in the corresponding cascade and with which sensitivity.
plotBaseClassifier

Base Classifier Performance Heatmap

Description

Plots a heatmap that shows base classifier performance.

Usage

plotBaseClassifier(
  x = NULL,
  classNames = NULL,
  symmetric = FALSE,
  main = "summary of base classifiers",
  xlab = "second class",
  ylab = "first class",
  digits = 3,
  ignore = 0,
  first.colors = c("#f5f5f5", "#8c510a"),
  second.colors = c("#f5f5f5", "#01665e"),
  na.color = c("yellow"),
  las = 1,
)

Value

No return value, called to generate the heatmap plot of the subcascades Object.

See Also

subcascades, plot.Conf, plot.ConfusionTable

Examples

library(TunePareto)
data(esl)
data <- esl$data
labels <- esl$labels
foldList <- generateCVRuns(labels = labels,
  ntimes = 2,
  nfold = 2,
  leaveOneOut = FALSE,
  stratified = TRUE)
predMap <- predictionMap(data, labels, foldList = foldList,
  classifier = tunePareto.svm(), kernel='linear')
# generate Subcascades object
subc <- subcascades(predMap,thresh=0.7,size=c(3,4))
plot(subc,row.sort='max')
color.key = TRUE,
cex = 1,
cex.lab = 1,
...
)

Arguments

x
A Conf object as it is returned by conf-function.
classNames
Vector of the original class names. If not given the class number is used instead.
symmetric
Logical indicating whether a symmetric base classifier (TRUE) is used.
main
See plot.
xlab
A title for the x axis (see plot).
ylab
A title for the y axis (see plot).
digits
Integer indicating the number of decimal places to be used (see round).
ignore
A numeric value between 0 and 1. All confusion and purity values below this number are not written as string into the corresponding element.
first.colors
A 2-element vector of the color for the minimal and maximal class-wise sensitivity of the first class. The color palette is calculated by an interpolation between the 2 given colors.
second.colors
A 2-element vector of the color for the minimal and maximal class-wise sensitivity of the second class. The color palette is calculated by an interpolation between the 2 given colors.
na.color
Color, which is used for indicating the empty elements (if the given class is not part of the cascade).
las
See par.
color.key
Specifies whether a color key is drawn (TRUE) or not (FALSE).
cex
See par.
cex.lab
See par.
...
Further arguments passed from other methods.

Details
This function plots a heatmap of the base classifier performance.

Value
No return value, called to generate the heatmap plot.
plotConf

Description

Plots a heatmap that shows base classifier performance.

Usage

plotConf(
  x = NULL,
  classNames = NULL,
  main = "summary of base classifiers",
  xlab = "classes",
  ylab = "base classifiers",
  digits = 3,
  ignore = 0,
  first.colors = c("#f5f5f5", "#b2182b"),
  second.colors = c("#f5f5f5", "#2166ac"),
  other.colors = c("#f5f5f5", "#1b7837"),
  las = 1,
  srt = 30,
  color.key = FALSE,
  cex = 1,
  cex.lab = 1,
  ...
)

Arguments

x A Conf object as it is returned by conf-function.
classNames Vector of the original class names. If not given the class number is used instead.
main See plot.
xlab A title for the x axis (see plot).
ylab A title for the y axis (see plot).
digits Integer indicating the number of decimal places to be used (see round).
ignore A numeric value between 0 and 1. All confusion and purity values below this number are not written as string into the corresponding element.
first.colors A 2-element vector of the color for the minimal and maximal class-wise sensitivity of the first class. The color palette is calculated by an interpolation between the 2 given colors.
second.colors A 2-element vector of the color for the minimal and maximal class-wise sensitivity of the second class. The color palette is calculated by an interpolation between the 2 given colors.
other.colors
A 2-element vector of the color for the minimal and maximal class-wise sensitivity of the other class. The color palette is calculated by an interpolation between the 2 given colors.

las
See \texttt{par}.

srt
Angle used to rotate the strings of the x-axis and y-axis labels (see \texttt{par}).

color.key
Specifies whether a color key is drawn (TRUE) or not (FALSE).

cex
See \texttt{par}.

cex.lab
See \texttt{par}.

... Further arguments passed from other methods.

Details
This function plots a heatmap of the base classifier performance.

Value
No return value, called to generate the heatmap plot.

See Also
\texttt{conf}, \texttt{plot.Subcascades}, \texttt{plot.ConfusionTable}

\begin{verbatim}
predictionMap(data = NULL, labels = NULL, foldList = NULL, parallel = FALSE, classifier = NULL, ...
)
\end{verbatim}

Description
Makes a PredictionMap object for the given data.

Usage
predictionMap(
  data = NULL,
  labels = NULL,
  foldList = NULL,
  parallel = FALSE,
  classifier = NULL,
  ...
)
predictionMap

Arguments

data
The data either as matrix or data.frame with samples in rows and features in columns.

labels
A vector of labels of data, consecutively labelled starting with 0.

foldList
A set of partitions for a cross-validation experiment. This list comprises as many elements as cross-validation runs. Each run is a list of as many vectors as folds. The entries are the indices of the samples that are left out in the folds. This fold list can be generated by using `TunePareto::generateCVRuns`. If the foldList is set to NULL (default) reclassification is performed.

parallel
Either TRUE or FALSE (default). If TRUE the pairwise training is performed parallelized.

classifier
A TunePareto classifier object.
For detailed information refer to `TunePareto::tuneParetoClassifier`.

...
Further parameters of the classifier object.

Details

Using a reclassification or a cross-validation set up this function performs a pairwise training for all class combinations and evaluates all samples using the trained classifiers. This means that for each run, fold and binary classifier the predicted class for each sample is calculated.

Value

A PredictionMap object. It is made up of a list of two matrices, which are called meta and pred. Both matrices provide information for individual samples column-wise. The meta information in meta connects the values in the pred-matrix to a specific fold, run, sample and contains the original label. The rownames of the pred-matrix (e.g. [0vs1]) show the classes of the binary base classifier. The elements are the prediction result of a specific training. The rows that correspond to base classifiers that would separate the same class consists of -1. Those rows are not used within the analysis.

See Also


Examples

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
                          ntimes = 2,
                          nfold = 2,
                          leaveOneOut = FALSE,
                          stratified = TRUE)

# svm with linear kernel
```
predMap = predictionMap(data, labels, foldList = foldList,
                        classifier = tunePareto.svm(), kernel='linear')

# knn with k = 3
predMap = predictionMap(data, labels, foldList = foldList,
                        classifier = tunePareto.knn(), k = 3)

# randomForest
predMap = predictionMap(data, labels, foldList = foldList,
                        classifier = tunePareto.randomForest())

print.Conf

---

print.Conf  
*Prints all details about a Conf object.*

### Description

Prints all details about a Conf object.

### Usage

```r
## S3 method for class 'Conf'
print(x, printfC = TRUE, printsC = TRUE, ...)
```

### Arguments

- **x**: A Conf object as it is returned by `conf`-function.
- **printfC**: If TRUE, the sensitivities of the first classes are printed (TP = true positives).
- **printsC**: If TRUE, the values of sC are printed (TN = true negatives). The values of sC correspond to the rate of samples of a given class (column), which are assigned to the second class of the base classifier (row).
- **...**: Further parameters

### Details

Prints all details about a given Conf object. A Conf object provides the conditional prediction rates of all base classifiers applied to all classes.

### Examples

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
                           ntimes = 2,
                           nfold = 2,
                           leaveOneOut = FALSE,
                           stratified = TRUE)
```
predMap = predictionMap(data, labels, foldList = foldList,
    classifier = tunePareto.svm(), kernel='linear')

conf = conf(predMap)

print(conf)

print.ConfusionTable  Prints all details about a ConfusionTable object.

Description
Prints all details about a ConfusionTable object.

Usage
## S3 method for class 'ConfusionTable'
print(x, ...)

Arguments
x A ConfusionTable object as it is returned by confusionTable-function.
...
Further parameters

Details
Prints all conditional prediction rates for the specified cascade.

Examples
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
    ntimes = 2,
    nfold = 2,
    leaveOneOut = FALSE,
    stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
    classifier = tunePareto.svm(), kernel='linear')

# Calculation of the confusion matrix for '0>2>3>4'.
confTable = confusionTable(predMap, cascade = '0>2>3>4')

print(confTable)
print.Groupwise

Prints all details about a Groupwise object.

Description

Prints all details about a Groupwise object.

Usage

## S3 method for class 'Groupwise'
print(x, printSizes = length(x), ...)

Arguments

x       A Groupwise object as it is returned by `groupwise`-function.
printSizes Integer that specifies how many of the cascade sizes (starting with the largest
cascades) should be printed.
...     Further parameters

Details

Prints all details about a given Groupwise object. A Groupwise object re-sorts a Subcascades object
in a way that the cascades made up of the same classes are grouped. The printSizes parameter can
be used to control the cascade sizes to be printed.

Examples

library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
              ntimes = 2,
nfold = 2,
leaveOneOut = FALSE,
stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
classifier = tunePareto.svm(), kernel='linear')

# generate Subcascades object
subc = subcascades(predMap, thresh=0.7)

# create a Groupwise object
groupwise = groupwise(subc)

print(groupwise, printSizes = 2)
print.PredictionMap

Prints all details about a PredictionMap object.

Description

Prints all details about a PredictionMap object.

Usage

```r
## S3 method for class 'PredictionMap'
print(x, showMeta = TRUE, showPred = TRUE, ...)
```

Arguments

- `x` A PredictionMap object generated by `predictionMap`.
- `showMeta` If TRUE the meta data of the PredictionMap object will be printed.
- `showPred` If TRUE the predictions of the PredictionMap object will be printed.
- `...` Further parameters

Details

Prints all details about a given PredictionMap object and information about the meta information as well as the predictions.

Examples

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
                          ntimes = 2,
nfold = 2,
leaveOneOut = FALSE,
stratified = TRUE)

# svm with linear kernel
predMap = predictionMap(data, labels, foldList = foldList,
                         classifier = tunePareto.svm(), kernel='linear')

print(predMap, showMeta=TRUE, showPred=TRUE)
```
print.Subcascades  

Prints all details about a Subcascades object.

Description
Prints all details about a Subcascades object.

Usage
## S3 method for class 'Subcascades'
print(x, printSizes = length(x), ...)

Arguments
- x: A Subcascades object generated by `subcascades`.
- printSizes: Integer specifying how many elements of a Subcascades object should be printed, sorted decreasingly after cascade length.
- ...: Further parameters

Details
Prints all details about a given Subcascades object.

Examples
```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels, ntimes = 2, nfold = 2, leaveOneOut = FALSE, stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList, classifier = tunePareto.svm(), kernel='linear')

# use default parameter settings
# -> returns cascades of all lengths that show a minimal classwise sensitivity >0.
subc = subcascades(predMap)

# print subcascades for the largest 2 sizes
print(subc, printSize=2)
```
screeningData

**Description**

It comprises a list of two elements: A data matrix ("data") and a label vector ("labels"). The samples are given in the rows and the features in the columns of the matrix. The sampled dataset comprises 6 classes with 40 samples, each. It is embedded in a two dimensional feature space and contains 2 ordinal sequences with 5 classes.

**Usage**

data(screeningData)

**Format**

An object of class `list` of length 2.

**Examples**

data(screeningData)

---

subcascades

**Subcascades Evaluation**

**Description**

Subcascades returns all cascades found within the data or evaluates a set of specific cascades.

**Usage**

subcascades(
  predictionMap = NULL,
  sets = NULL,
  thresh = 0,
  size = NA,
  numSol = 1000
)
Arguments

predictionMap A PredictionMap object as it is returned by predictionMap-function. It is made up of a list of two matrices(pred and meta). Both matrices provide information on individual samples column-wise. The rownames of the pred-matrix (e.g. [0vs1]) show the classes of the binary base classifier. The elements are the prediction result of a specific training. The rows that correspond to base classifiers that would separate the same class consists of -1. Those rows are not used within the analysis. The meta information connects the values in the pred-matrix to a specific fold, run and contains the original label.

sets Contains the set used for filtering. It is either a list of numeric vectors, a numeric vector, or a vector of characters representing a cascade of the following format '1>2>4'. Empty vectors are not allowed.

thresh A numeric value between 0 and 1. The minimal sensitivity threshold used to filter the returned cascades. Only cascades that pass this threshold are returned. If 0 is used the returned cascades are filtered for >0 and otherwise >= thresh. For low thresholds the calculation lasts longer.

size A numeric value that defines the size of the cascades that should be returned. The smallest size is 2 and the largest the maximal number of classes of the current dataset. If size is NA (the default setting), all cascades from 2 to the maximal number of classes are evaluated.

numSol The maximum number of cascades that should pass the first sensitivity bound and are further evaluated.

Details

This function can either be used to evaluate the performance of a specific cascade, a set of cascades or to filter out the set of cascades of a specific size and passing a given threshold. If the sets-variable is given no size can be set.

Value

A Subcascades object comprising the evaluated cascades and their performances. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type '1>2>3' and the entries the sensitivity for each position of the cascade.

See Also

print.Subcascades, plot.Subcascades, summary.Subcascades

Examples

library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
ntimes = 2,
fold = 2,
leaveOneOut = FALSE,
stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
classifier = tunePareto.svm(), kernel='linear')

# use default parameter settings
# -> returns cascades of all lengths that show a minimal classwise sensitivity >0.
subc = subcascades(predMap)
# change the threshold
# -> returns cascades of all lengths that show a minimal classwise sensitivity >=0.6.
subc = subcascades(predMap, thresh=0.6)
# search only for cascades of length 2 and 4
# -> returns cascades of length 2 and 4 that show a minimal classwise sensitivity >=0.6.
subc = subcascades(predMap, thresh=0.6, size=c(2,4))
# evaluates the performance of the cascade '0>1>2>3>4'.
subc = subcascades(predMap, sets = c('0>1>2>3>4'))

summary.Conf

Summary of conf

Description

summary.Conf returns a summarizing overview of a Conf object.

Usage

## S3 method for class 'Conf'
summary(object = NULL, ...)

Arguments

object A Conf object as it is returned by conf-function.
... Further arguments passed from other methods.

Details

This function gives an overview of the Conf object. The overall number of classes, the highest and lowest sensitivities of the first class as well as the highest and lowest performance of the second class are recorded.

See Also

Examples

library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
                          ntimes = 2,
                          nfold = 2,
                          leaveOneOut = FALSE,
                          stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
                         classifier = tunePareto.svm(), kernel='linear')

conf = conf(predMap)
summary(conf)

summary.ConfusionTable

Summary of confusion tables

Description

summary.ConfusionTable returns a summarizing overview.

Usage

## S3 method for class 'ConfusionTable'
summary(object = NULL, ...)

Arguments

object A ConfusionTable object as it is returned by confusionTable-function.
...
Further arguments passed from other methods.

Details

This function gives an overview of the characteristics of the ConfusionTable object. The cascade with the corresponding mean accuracy are given.

See Also

confusionTable, summary.PredictionMap, summary.Subcascades, summary.Groupwise, summary.Conf
Examples

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
                        ntimes = 2,
nfold = 2,
leaveOneOut = FALSE,
stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
                        classifier = tunePareto.svm(), kernel='linear')

# Calculation of the confusion matrix for '0>2>3>4'.
confusionTable = confusionTable(predMap, cascade = '0>2>3>4')

summary(confusionTable)
```

### summary.Groupwise

#### Overview Class Groups

**Description**

`summary.Groupwise` returns a summarizing overview. For each length the number of permutations consisting of the same set of classes is given.

**Usage**

```r
## S3 method for class 'Groupwise'
summary(object = NULL, ...)
```

**Arguments**

- `object`: A Groupwise object as it is returned by `groupwise`-function.
- `...`: Further arguments passed from other methods.

**Details**

This function gives an overview of the characteristics of the Groupwise object. The number of permutations per size is given. A permutation means that the corresponding cascades contain the same classes but with different order.

**See Also**

Examples

```r
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels,
                           ntimes = 2,
                           nfold = 2, 
                           leaveOneOut = FALSE, 
                           stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList,
                        classifier = tunePareto.svm(), kernel='linear')

# generate Subcascades object
subc = subcascades(predMap, thresh=0.7)
groupwise = groupwise(subc, maxCl=50)

summary(groupwise)
```

summary.PredictionMap  Summary of prediction maps

Description

summary.PredictionMap returns a summarizing overview of a PredictionMap object.

Usage

```r
## S3 method for class 'PredictionMap'
summary(object = NULL, ...)
```

Arguments

- `object`: A PredictionMap object as it is returned by `predictionMap`-function.
- `...`: Further arguments passed from other methods.

Details

This function gives an overview of the PredictionMap object. A short summary about the utilized data and labels is given as well as the number of runs and folds of the cross-validation. The summary also includes if the prediction map was generated in parallel and the name of the utilized TunePareto classifier as well as its specified parameters.

See Also

- `predictionMap`, `summary.Subcascades`, `summary.Groupwise`, `summary.Conf`, `summary.ConfusionTable`
Example

```r
library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
foldList = generateCVRuns(labels = labels, ntimes = 2, nfold = 2, leaveOneOut = FALSE, stratified = TRUE)
predMap = predictionMap(data, labels, foldList = foldList, classifier = tunePareto.svm(), kernel='linear')

summary(predMap)
```

### summary.Subcascades

#### Summary Subcascades Characteristics

**Description**

Generates a general overview of the characteristics of the Subcascades object.

**Usage**

```r
## S3 method for class 'Subcascades'
summary(object = NULL, includeClassSummary = TRUE, digits = 3, ...)
```

**Arguments**

- **object** A Subcascades object as it is returned by `subcascades`-function.
- **includeClassSummary** Boolean indicating if the occurrence of classes by size should be included in the summary.
- **digits** Integer defining the number of decimal places as it is used in the `round`-function.
- **...** Further arguments passed from other methods.

**Details**

This function gives a general overview of characteristics of the Subcascades object, like number of cascades or maximal cascade length.

**See Also**

Examples

```r
library(TunePareto)
data(esl)
data <- esl$data
labels <- esl$labels
foldList <- generateCVRuns(labels = labels,
  ntimes = 2,
nfold = 2,
leaveOneOut = FALSE,
stratified = TRUE)
predMap <- predictionMap(data, labels, foldList = foldList,
  classifier = tunePareto.svm(), kernel='linear')
# generate Subcascades object
subc <- subcascades(predMap, thresh=0.7, numSol=10000)
summary(subc)
```

## summaryClasses

### Occurrence of Classes by Size

#### Description

`summaryClasses` returns the occurrence of classes by size.

#### Usage

```r
summaryClasses(subcascades = NULL)
```

#### Arguments

- `subcascades`: A Subcascades object as it is returned by `subcascades`-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type `"1>2>3"` and the entries the sensitivity for each position of the cascade.

#### Details

This function gives an overview of the classes of the Subcascades object. For each length in the Subcascades object the occurrence of classes is given.

#### Value

A matrix summarizing the overview characteristics of the Subcascades object.
**summarySubcascades**

**Overview Subcascades**

**Description**

`summarySubcascades` returns a summarizing overview. For each length the number of cascades and the minimal class-wise sensitivity is given.

**Usage**

```r
summarySubcascades(subcascades = NULL)
```

**Arguments**

- `subcascades`: A Subcascades object as it is returned by `subcascades`-function. The Subcascades object is made up of a list of matrices. Each matrix comprises the evaluation results of cascades of a specific length and is sorted row-wise according to the achieved minimal classwise sensitivities of the cascades (decreasing). The rownames show the class order by a character string of type ’1>2>3’ and the entries the sensitivity for each position of the cascade.

**Details**

This function gives an overview of the Subcascades object. For each length in the Subcascades object the number of cascades of that length, as well as their minimal classwise sensitivity is given.

**Value**

A matrix summarizing the overview characteristics of the Subcascades object. For each size (rows) the number of cascades within the Subcascades object (number) and the minimal classwise sensitivity (min.class.sens) are given.

---

**tunePareto.occ**

**Ordinal Classifier Cascade Tune Pareto Object**

**Description**

TunePareto wrapper for the ordinal classifier cascade.

**Usage**

```r
tunePareto.occ(base.classifier)
```
Arguments

base.classifier
A predefined TuneParetoClassifier object used as binary classifier for the ordinal classifier cascade. There exist five classifier types that can be used: tunePareto.knn(), tunePareto.svm(), tunePareto.tree(), tunePareto.randomForest(), tunePareto.NaiveBayes(). For more information about these classifier functions please refer to the corresponding help page of like TunePareto::tunePareto.knn.

Details

The "tunePareto.occ" encapsulates the classifier of an ordinal classifier cascade. Additionally, to the parameters of the corresponding base classifier the "class.order" parameter can be provided. It is either a character vector, a numeric vector, or a vector representing a cascade of the following format '1>2>4'.

Value

Returns an object of class TuneParetoClassifier (see: TunePareto::tuneParetoClassifier). This can be passed to the function trainTuneParetoClassifier (see: TunePareto::trainTuneParetoClassifier).

Examples

library(TunePareto)
data(esl)
data = esl$data
labels = esl$labels
# train classifier
model <- trainTuneParetoClassifier(
  classifier = tunePareto.occ( base.classifier = tunePareto.svm()),
  trainData = data,
  trainLabels = labels,
  class.order = as.character(c(4,3,1,0)),
  kernel = "linear",
  cost = 1)

# predict labels
prediction <- predict(object = model, newdata = data)
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