# Package 'PerfMeas' 

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Description Implementation of different performance measures for classification and ranking tasks in-cluding Area Under the Receiving Characteristic Curve (AUROC) and Area Under the Preci-sion Recall Curve (AUPRC), precision at a given recall, F-score for single and multiple classes.
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## Description

Metrics for ranking and classification tasks: Area Under the Receiving Operating Characteristic Curve (AUROC), Area Under the Precision and Recall Curve (AUPRC), F-scores, and precision at given recall level are implemented.

## Details

| Package: | PerfMeas |
| :--- | :--- |
| Type: | Package |
| Version: | 1.2 .5 |
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| LazyLoad: | yes |

This package implements a set of functions to estimate the AUC, F-score, precision, recall, specificity, accuracy according to the $0 / 1$ loss, and precision at given recall level for ranking and classification problems.

Functions to compute the above measures for single classes or for sets of classes are provided.

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AUC.measures AUC measures

## Description

Set of functions to compute the Area Under the ROC Curve (AUC)

## Usage

```
AUC.single(pred, labels)
AUC.single.over.classes(target, predicted, g, root = "00")
AUC.n.single(pred, labels, n=50)
AUC.n.single.over.classes(target, predicted, g, n=50, root = "00")
compute.mean.AUC.single.over.classes(y)
```


## Arguments

pred numeric vector (scores) of the values of the predicted labels
labels numeric vector of the true labels ( 0 negative, 1 positive examples)
target matrix with the target multilabels: rows correspond to examples and columns to classes. $\operatorname{target}[i, j]=1$ if example i belongs to class j , target $[\mathrm{i}, \mathrm{j}]=0$ otherwise.
predicted a numeric matrix with predicted values (scores): rows correspond to examples and columns to classes.
$\mathrm{g} \quad$ a graph of class graphNEL (package graph) of the classes. If g is missing no per.level results are computed
$\mathrm{n} \quad$ number of negatives $(\operatorname{def}=50)$
root the name of the root node (def. "00")
y a list of lists. The components of the outer list is a list returned from the function
AUC.single.over.classes

## Details

AUC. single computes the AUC for a single class.
AUC.single.over.classes computes AUC for a set of classes, including their average values across classes and the average values across the levels of the hierarchy (if any); level 1 classes are at distance 1 from the root, level 2 the second level, till to last level correponding to the leaves. Note that if the argument g is missing no per-level values are computed.

AUC.n.single computes the AUCn for a single class, i.e. the AUC by considering only the first n top ranked negatives, where n is the absolute number of negative examples receiving the highest scores.

AUC.n.single.over.classes computes AUCn for a set of classes, including their average values across classes and the average values across the levels of the hierarchy (if any); level 1 classes are at distance 1 from the root, level 2 the second level, till to last level correponding to the leaves. Note that if the argument g is missing no per-level values are computed.
compute.mean.AUC.single. over.classes compute means across folds of AUC.single.over.classes. It can be used to automatically computed average values (for each class, level, or average across classes) across folds.

## Value

AUC. single returns a numeric value corresponding to the AUC.
AUC. single.over.classes returns a list with three elements:

| - average | the average AUC across classes |
| :---: | :---: |
| - per.level | a named vector with average AUC for each level of the hierarchy; names correspond to levels |
| - per.class | a named vector with AUC for each class; names correspond to classes |
| AUC.n.single returns a numeric value corresponding to the AUCn. |  |
| AUC.n.single.over.classes returns a list with three elements: |  |
| - average | the average AUCn across classes |
| - per.leve | a named vector with average AUCn for each level of the hierarchy; names correspond to levels |
| - per.class | a named vector with AUCn for each class; names correspond to classes |
| compute.mean.AUC.single.over.classes returns a list obtained by averaging the results across folds of the input y . The components are: |  |
| - average | the average AUC across classes |
| - per.level | a named vector with average AUC for each level of the hierarchy; names correspond to levels |
| - per.class | a named vector with AUC for each class; names correspond to classes |

## See Also

F.measures, PXR

## Examples

```
# preparing pseudo.random scores and target-labels for examples: 100 examples
# and 10 classes
Scores <- matrix(runif(1000),nrow=100);
Targets <- matrix(integer(1000), nrow=100);
Targets[Scores>0.5] <- 1;
# adding noise to scores
Scores <- Scores + matrix(rnorm(1000, sd=0.3), nrow=100);
colnames(Scores) <-colnames(Targets) <- LETTERS[1:10];
# getting scores and labels of class "A"
scores <- Scores[,"A"];
labels <- Targets[,"A"];
# AUC for a single class
AUC.single(scores,labels);
# AUC for the 10 classes
AUC.single.over.classes(Targets, Scores);
# AUCn for a single class considering only the first top scored negatives
AUC.n.single(scores,labels, n=20);
# AUCn for the 10 classes considering only the first top scored negatives
AUC.n.single.over.classes(Targets, Scores, n=20);
```


## AUPRC

## Description

Functions to compute the Area Under the Precision Recall Curve (AUPRC) and the Area Under the F-score Recall Curve (AUFRC)

## Usage

AUPRC(z, comp.precision=TRUE)
trap.rule.integral( $\mathrm{x}, \mathrm{y}$ )

## Arguments

z
a list of lists. The components of the outer list is a list returned from the function precision.at.all.recall.levels that reports precision, recall and f-score results at different levels for different methods or tasks.
comp.precision boolean. It TRUE (default) the AUPRC is computed otherwise the area under the F-score curve is computed
$x \quad$ vector of the $x$ values in increasing order
$y \quad$ vector of the corresponding $y=f(x)$ values

## Details

AUPRC computes the Area Under the Precision Recall Curve or the Area Under the F-score Recall Curve (AUFRC) for multiple curves by using the output of the function precision. at.all. recall.levels.

The function trap.rule.integral implements the trapezoidal rule of integration and can be used to compute the integral of any empirical function expressed as a set of pair values (a vector of $x$ values and a vector of $y=f(x)$ values). In particular if $x$ is the recall (with values in ascending order) and $y$ the corresponding precision, trap. rule.integral copmutes the AUPRC.

## Value

AUPRC returns the value of the AUPRC (if the argument comp. precision = TRUE), otherwise the value of the AUFRC.
trap.rule.integral returns the value of the integral.

## See Also

AUC.measures, PXR

## Examples

```
# loading matrices of scores an correponding table of classes
data(T);
data(Scores);
res=list();
classes=1:10
# computing precision recall values
for (j in classes) res=c(res, list(precision.at.all.recall.levels(Scores[,j], T[,j])));
names(res)<-seq(0.1, 1, by=0.1);
# computing AUPRC
AUPRC (res, comp.precision=TRUE);
# computing AU F-score recall curve
AUPRC (res, comp.precision=TRUE);
# Loading precision at given recall levels for different methods
data(PrecRec);
# computing AUPRC for different methods
x <- seq(0.1, 1, by=0.1);
res <- numeric(nrow(PrecRec));
names(res) <- rownames(PrecRec);
for (i in 1:nrow(PrecRec))
    res[i] <- trap.rule.integral(x, PrecRec[i,]);
print(res);
```

example.data

## Description

Collection of datasets used in the examples of the package

## Usage

data(Scores)
data(T)
data(PrecRec)

## Details

The T data is a named 1901 X 10 matrix whose rows correspondes to yest genes, while columns correspond to 10 FunCat (Functional Categories) classes. If $\mathrm{T}_{-} \mathrm{ij}=1$ gene i belong to class j , if $\mathrm{T}_{-} \mathrm{ij}=0$ gene i does not belong to class j . The Scores data is a named 1901 X 10 matrix representing scores (likelihood) that a given gene belongs to a given class: higher the value higher the likelihood. PrecRec is a matrix representing precision at 10 different recall values of 7 methods for gene function prediction.

## F.measures F-measures

## Description

Set of functions to compute the F-measure, precision, recall, specificity and $0 / 1$ loss accuracy.

## Usage

F.measure.single(pred, labels)
F.measure.single.over.classes(target, predicted, g, root = "00")
compute.mean.F.measure.single.over.classes(y)

## Arguments

pred vector of the predicted labels. 0 stands for negative and 1 for positive
labels vector of the true labels. 0 stands for negative and 1 for positive
target matrix with the target multilabels. 0 stands for negative and 1 for positive. Rows correspond to examples and columns to classes.
predicted matrix with the predicted multilabels. 0 stands for negative and 1 for positive. Rows correspond to examples and columns to classes.
g graph of the classes (object of class graphNEL, package graph). If missing, no per level results are computed.
root the name of the root node (def. " 00 ") of the graph g .
$y \quad a$ list of lists. The components of the outer list is a list returned from the function F.measure.single.over.classes

## Details

F.measure.single computes the F.score, precision, recall, specificity and accuracy for a single class.
F.measure.single.over.classes computes precision, recall, specificity, accuracy and F-measure for a set of classes. In particualr it computes the correponding average values across classes, the average values across levels of the hierarchy of the classes (if any), and the values of the measures for each class. Note that if there is no hierarchy between classes (represented by the graph g), you can miss the g parameter and no per-level values are computed.
compute.mean.F.measure.single.over.classes computes means across folds of F.measure.single.over.classes. This function could be useful in cross-validated or multiple holdout experimental settings.

## Value

F.measure.single returns a named numeric vector with six elements:

- P precision

| -R | recall (sensitivity) |
| :--- | :--- |
| - S | specificity |
| - F | F measure |
| - A | $0 / 1$ loss accuracy |
| - npos | number of positive examples |
| F.measure.single.over.classes returns a list with three elements: |  |
| - average | a named vector with the average precision, recall, specificity, F-measure, accu- <br> racy and average number of positive examples across classes. |
| - per.level | a named matrix with average precision, recall, specificity, F-measure and accu- <br> racy for each level of the hierarchy. Named rows correspond to levels, named <br> columns correspond respectively to precision, recall, specificity, F-measure, ac- <br> curacy and number of positive examples. |
| - per.class | a named matrix with precision, recall, specificity, F-measure, accuracy and num- <br> ber of positive examples for each class. Named rows correspond to classes, <br> named columns correspond respectively to precision, recall, specificity, F-measure <br> accuracy and and number of positive examples. |
| compute.mean.F.measure. single.over.classes returns a list obtained by averaging the results |  |
| across folds of the input y. The components are: |  |

## See Also

AUC.measures, PXR

## Examples

```
# preparing pseudo-random predictions and target-labels for examples: 100 examples
# and 10 classes
Scores <- matrix(runif(1000),nrow=100);
Targets <- Pred <- matrix(integer(1000),nrow=100);
Targets[Scores>0.5] <- 1;
# adding noise to scores
Scores <- Scores + matrix(rnorm(1000, sd=0.3), nrow=100);
Pred[Scores>0.5] <- 1;
colnames(Pred) <-colnames(Targets) <- LETTERS[1:10];
# getting predictions and labels of class "A"
pred <- Pred[,"A"];
```

```
labels <- Targets[,"A"];
# F.score and other metrics for a single class
F.measure.single(pred,labels);
# F.score and other metrics for the 10 classes
F.measure.single.over.classes(Targets, Pred);
```

```
get.all.nodes.by.depth
```

Getting nodes by their depth

## Description

Grouping classes by level in a given hierarchy.

## Usage

get.all.nodes.by.depth(g, root $=" 00 ")$

## Arguments

g graph of the classes (object of class graphNEL, package graph).
root name of the root node (def. 00)

## Details

The minimum paths between the "root" and all the other classes/nodes are computed. Levels are numbered from 1 in increasing order by their distance from the "root" class.

## Value

a list of the nodes, grouped w.r.t. the distance from the root. The first element of the list corresponds to the nodes at distance 1 , the second to nodes at distance 2 and so on.

```
graphics Graphics function to plot precision/recall or f.score/recall curves
```


## Description

Function to plot multiple precision/recall or f.score/recall curves

## Usage

```
precision.recall.curves.plot(y, range=seq(from=0, to=1, by=0.1),
curve.names=1:length(y), cex.val=0.6, f="", height=9, width=11,
col=c("black", "red1", "blue1", "green1", "darkgrey", "brown1", "yellow1", "orange1",
"red4", "blue4", "green4", "lightgrey", "brown4", "yellow4", "orange4"),
line.type=1, leg=TRUE, pos=c(range[length(range)-2], range[length(range)]), do.grid=TRUE,
plot.precision=TRUE, trap.rule=TRUE)
performance.curves.plot(m, x.range=seq(from=0.1, to=1, by=0.1),
y.range=c(0,1), curve.names=1:nrow(m), cex.val=0.6, f="", height=9, width=11,
col=c("black", "red1", "blue1", "green1", "darkgrey", "brown1", "yellow1", "orange1",
"red4", "blue4","green4","lightgrey","brown4","yellow4","orange4"), line.type=1,
patch.type=1:16, leg=TRUE, pos=c(x.range[length(x.range)-2], y.range[2]), do.grid=TRUE,
x.label="Recall", y.label="Precision")
```


## Arguments

| $y$ | a list of lists. Each component list is a list returned from precision. at. all.recall.levels that reports precision and recall results at different levels for different methods or tasks |
| :---: | :---: |
| range | numeric vector of the precision/recall values to be represented (def: values between 0 and 1 step 0.1) |
| curve.names | names of the compared methods to be reported in the legenda (def: numbers) |
| cex.val | magnification value for characters (def. 0.6) |
| f | file name. If is given, an encapsulated postscript file is created, otherwise the output is rendered on a window. |
| height | relative heigth of the graph (def. 9) |
| width | relative width of the graph (def. 11) |
| col | colors of the lines. 14 different colors are given as default, but any vector of color from colors() (package graphics) can be used. Colors are recycled if length(col) $<$ length(y). |
| line.type | type of the line. Any valid vector of integer can be assigned (values between 1 and 6 , see lty in par(), package graphics for details). Values are recycled if length(line.type) < length(y). Def.: 1 (solid lines). |
| leg | boolean: if TRUE (def.) a legend is depicted. |
| pos | coordinates of the position of the legend. |
| plot.precision | boolean: if TRUE (def.) precision/recall curves are plotted, othewise f-score/recall curves. |
| trap.rule | boolean: if TRUE (def.) the integral of the curves are computed. |
| m | a numeric matrix. Rows correspond to different methods and columns to precision or f-score at given recall values |
| $x$. range | vector of the recall values to be represented |
| $y$. range | vector with 2 elements: range of the precision/f-score to be represented |

patch.type numeric vector corresponding to the symbols to be plotted for different recall values (def. 1:16)
do.grid a bollen value indicating whether a grid will be plotted or not (def: TRUE)
x.label label of the abscissa (def: Recall)
$y . l a b e l$ label of the ordinate (def: Precision)

## Details

The function precision. recall.curves.plot plots multiple precision/recall curves (or f-score/recall curves, if the argument plot.precision=FALSE) by using the output of the function precision.at.all.recall.levels, that compute several precison/recall pairs by moving the threshold from the lowest to the highest score achieved by each example.
The function performance.curves.plot plots precision of F-score/recall curves ofr a predefined set of recall levels. Thsi function can be useful to plot and to compare the average results between methods across multiple classes.

The curves can differ by color, type of line and for performance.curves.plot for each recall value a symbol can be also plotted. A legend can be automatically constructed.

## Value

The functions output a graphic file either on a window or on an encapsulated postscript file. The function precision.recall.curves.plot, if the parameter trap.rule = TRUE (default), outputs a vector with the AUPRC (or the Area Under the F-score Curve if the parameter plot. precision=FALSE) for each curve.

## Examples

```
# loading an example matrix of scores an the correponding table of classes
data(T);
data(Scores);
res=list();
classes=c(1,2,7,8)
# computing precison recall values
for (j in classes) res=c(res, list(precision.at.all.recall.levels(Scores[,j], T[,j])));
# plotting precision/recall curves
precision.recall.curves.plot(res,curve.names=colnames(T)[classes],
pos=c(0.7,1), plot.precision=TRUE);
# black and white version
precision.recall.curves.plot(res,curve.names=colnames(T)[classes], pos=c(0.7,1),
plot.precision=TRUE, line.type=1:4, col=1);
# plotting f-score/recall curves
precision.recall.curves.plot(res,curve.names=colnames(T)[classes], pos=c(0.7,1),
plot.precision=FALSE);
# black and white version
precision.recall.curves.plot(res,curve.names=colnames(T)[classes], pos=c(0.7,1),
plot.precision=TRUE, line.type=1:4, col=1);
```


## Description

Set of functions to compute the precision at fixed recall levels.

## Usage

precision.at.recall.level(scores, labels, rec.level = 0.2)
precision.at.recall.level.over.classes(target, predicted,
g, rec.level = 0.2, root $=" 00 "$ )
precision.at.multiple.recall.level(scores, labels,
rec.levels $=\operatorname{seq}($ from $=0.1$, to $=1$, by $=0.1)$ )
precision.at.multiple.recall.level.over.classes(target,
predicted, rec.levels $=$ seq(from $=0.1$, to $=1$, by $=0.1$ )
precision.at.all.recall.levels(scores, labels, resolution=1)

## Arguments

scores $\quad v e c t o r ~ o f ~ t h e ~ p r e d i c t e d ~ s c o r e s ~ i n ~[0,1] ~$
labels $\quad 0 / 1$ vector of the true labels
rec.level rec.level: the desired recall level (def: 0.2)
target matrix with the target multilabels; rows correspond to examples, columns to classes
predicted matrix with the predicted multilabels; rows correspond to examples, columns to classes
g graph of the classes (object of class graphNEL, package graph). If missing, no per level results are computed.
root the name of the root node (def. "00") of the graph g.
rec.levels a vector with the desired recall levels (def. 0.1 to 1 by 0.1 step)
resolution a number between 0 and 1 (def. 1). This represents the fraction of precision, recall and f-score values returned.

## Details

precision.at.recall. level computes the precision at a given recall level for a single class.
precision.at.recall.level.over.classes computes precision at a given recall level for a set of classes.
precision.at.multiple.recall.level computes the precision at multiple levels of recall for a single class.
precision.at.multiple.recall.level.over.classes computes the precision at multiple levels of recall for multiple classes.
precision.at.all.recall. levels compute the precision at all recall levels for a single class. It returns a pair of precision and recall values by moving a threshold from the lowest to the highest score: a number of precision and recall values equal to the number $n$ of available examples is returned if resolution $=1$, otherwise a number of values equal to $n *$ resolution.

## Value

precision.at.recall.level returns the precision at the requested recall
precision.at.recall.level.over.classes a list with three elements:

- average the average precision at a given recall level across classes.
- per.level a named vector with average precision at a given recall level for each level of the hierarchy; names correspond to levels
- per.class a named vector with precision at a given recall level for each class. Names correspond to classes
precision.at.multiple.recall.level a list with 2 elements:
- precisions a vector with the precision at different recall levels
- f.score a vector with the f-score at different recall levels
precision.at.multiple.recall.level.over.classes
- PXR a matrix with the precisions at different recall levels: rows are classes, columns precisions at different recall levels
- avgPXR a vector with the the average precisions at different recall levels across classes
precision.at.all.recall.levels a list with 3 elements:
- precision precision at different thresholds
- recall recall at different thresholds
- f.score f.score at different thresholds


## See Also

AUC.measures, F.measures

## Examples

```
# preparing pseudo-random predictions and target-labels for examples:
# 100 examples and 10 classes
Scores <- matrix(runif(1000),nrow=100);
Targets <- matrix(integer(1000),nrow=100);
Targets[Scores>0.5] <- 1;
# adding noise to scores
Scores <- Scores + matrix(rnorm(1000, sd=0.3), nrow=100);
colnames(Scores) <-colnames(Targets) <- LETTERS[1:10];
# getting scores and labels of class "A"
scores <- Scores[,"A"];
labels <- Targets[,"A"];
```

```
# precsion at 0.4 recall level for class A
precision.at.recall.level(scores, labels, rec.level=0.4);
# precision at 0.4 recall level for all the 10 classes
precision.at.recall.level.over.classes(Targets, Scores, rec.level=0.4);
# precision at multiple recall levels for class A
levels <- seq(from=0.1, to=1, by=0.1);
precision.at.multiple.recall.level(scores, labels, rec.levels=levels);
# precision at multiple recall levels for all the 10 classes
precision.at.multiple.recall.level.over.classes(Targets, Scores);
# precision, recall and f-score for a single class obtained
# by moving the threshold across the examples
precision.at.all.recall.levels(scores, labels);
```


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