Package 'findIPs'

October 19, 2024

Type Package

Title Influential Points Detection for Feature Rankings

Version 1.1.0 **Date** 2024-04-03

Description Feature rankings can be distorted by a single case in the context of high-dimensional data. The cases exerts abnormal influence on feature rankings are called influential points (IPs). The package aims at detecting IPs based on case deletion and quantifies their effects by measuring the rank changes (DOI:10.48550/arXiv.2303.10516). The package applies a novel rank comparing measure using the adaptive weights that stress the top-ranked important features and adjust the weights to ranking properties.

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URL https://github.com/ShuoStat/findIPs

BugReports https://github.com/ShuoStat/findIPs

Depends graphics, R (>= 4.4.0)

Imports Biobase, BiocParallel, parallel, stats, SummarizedExperiment, survival, utils

Suggests BiocStyle, knitr, rmarkdown, testthat

VignetteBuilder knitr

biocViews GeneExpression, DifferentialExpression, Regression, Survival

Encoding UTF-8 **LazyData** FALSE

Roxygen list(markdown = TRUE)

RoxygenNote 7.3.1

git_url https://git.bioconductor.org/packages/findIPs

git_branch devel

git_last_commit fda12ef

git_last_commit_date 2024-04-30 Repository Bioconductor 3.20

Date/Publication 2024-10-18

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findIPs

Function to detect influential points for feature rankings

Description

findIPs employs two important functions: getdrop1ranks and sumRanks. getdrop1ranks can calculate the original feature ranking and leave-one-out feature rankings. The outputs are subsequently taken to sumRanks, which computes the overall rank changes for each observation, indicating their influence on feature rankings.

Usage

```
findIPs(
   X,
   y,
   fun,
   decreasing = FALSE,
   topN = 100,
   method = "adaptive",
   nCores = NULL
)
```

Arguments

X A data matrix, with rows being the variables and columns being samples.

y Groups or survival object (for cox regression).

fun can either be a character or a function. fun should be one of the 't.test', 'cox',

'log2fc', and 'kruskal.test' when it is a character. findIPs() incorporates four widely used ranking criteria: t-test, univariate cox model, log2fc, and kruskal test, whose outputs are p values except log2fc (absolute log2 fold changes). The features would be ordered by specifying the argument decreasing. For instance, if fun = 't.test', the decreasing = F, such that features are order

by the pvalues of t.test in a increasing manner.

fun can also be a function to obtain ranking criteria with x and y being the only input and the ranking criteria, such as p-values being the only output.

decreasing logical. How the rank criteria are ordered? For instance, p-value should be

ordered increasingly, while fold-change should be ordered decreasingly.

topN the number of important features included for comparison.

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method method to summarize rank changes. It should be one of the 'adaptive', 'weight-

edSpearman', and 'unweighted'. Both 'adaptive' and 'weightedSpearman' are weighted rank comparison method, but former employs the weight that are adaptive to the distribution of rank changes. 'unweighted' denotes a direct compari-

son of ranks without considering weights.

nCores the number of CPU cores used for parallel running. If nCores = NULL, a single

core is used.

Value

The weight function's shape is controlled by kappa, which ranges from 0 to 1.

Weighted rank changes are calculated using kappa, with higher values indicating

more weight on top features.

score The influence of each observation on feature rankings, with larger values indi-

cating more influence.

origRank The original ranking. origRank is exactly the input. Here it is re-output for

visualization purposes.

drop1Rank The leave-one-out rankings.

origRankWeighted

The weighted original ranking

drop1RankWeighted

The weighted leave-one-out rankings

```
data(miller05)
X <- miller05$X
y <- miller05$y
obj <- findIPs(X, y,</pre>
               fun = 't.test',
               decreasing = FALSE,
               topN = 100,
               method = 'adaptive')
par(mfrow = c(1, 3), mar = c(4, 4, 2, 2))
plotRankScatters(obj, top = TRUE)
plotAdaptiveWeights(kappa = obj$kappa,
                    n = nrow(obj$drop1Rank),
                     type = 'line',
                     ylim = NULL)
plotIPs(obj, topn = 5, ylim = NULL)
## Interop with ExpressionSet class
library(Biobase)
data(sample.ExpressionSet)
design <- phenoData(sample.ExpressionSet)$type</pre>
IPs <- findIPs(exprs(sample.ExpressionSet), design, fun = "t.test",</pre>
               method = "adaptive")
plotIPs(IPs)
## Interop with SummarizedExperiment class
library(SummarizedExperiment)
## Make a SummarizedExperiment class
```

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getdrop1ranks

Derive ranking lists including original and leave-one-out rankings

Description

This function calculates the original and leave-one-out feature rankings using a predefined rank method

Usage

```
getdrop1ranks(X, y, fun, decreasing = FALSE, topN = 100, nCores = NULL)
```

Arguments

X A data matrix, with rows being the variables and columns being samples.

y Groups or survival object (for cox regression)

fun fun can either be a character or a function. fun should be one of the 't.test', 'cox',

'log2fc', and 'kruskal.test' when it is a character. findIPs() incorporates four widely used ranking criteria: t-test, univariate cox model, log2fc, and kruskal test, whose outputs are p values except log2fc (absolute log2 fold changes). The features would be ordered by specifying the argument decreasing. For instance, if fun = 't.test', the decreasing = F, such that features are order

by the pvalues of t.test in the increasing manner.

fun can also be a function to obtain ranking criteria with x and y being the only

input and the ranking criteria, such as p-values being the only output.

decreasing logical. How the rank criteria are ordered? For instance, p-value should be

ordered increasingly, while fold-change should be ordered decreasingly.

topN the number of important features included for comparison. The top n features in

the original ranking list.

nCores the number of CPU cores used for parallel running. If nCores = NULL, a single

core is used.

Value

orig vector:, original ranking

drop1rank matrix, Leave-one-out rankings

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Examples

miller05

miller05 data

Description

miller05 is gene expression data with 1000 genes randomly sampled from 22283 genes and 236 samples since removing the case with missing response. The data has binary and survival response. The binary response contains 58 case with p53 mutant and 193 wild type mutant. The survival response has a total of 55 events.

Usage

```
data(miller05)
```

Format

a list

Value

miller05 data, a list containing 1000 genes and binary and survival response.

References

Miller, Lance D., et al. 'An expression signature for p53 status in human breast cancer predicts mutation status, transcriptional effects, and patient survival.' Proceedings of the National Academy of Sciences 102.38 (2005): 13550-13555.doi:10.1073pnas.0506230102

```
data(miller05)
```

6 plotIPs

plotAdaptiveWeights Visualize the weight function for adaptive weights

Description

Plot the weight function for the adaptive weights with given kappa and the list length (n).

Usage

```
plotAdaptiveWeights(kappa, n, type = c("line", "points"), ylim = NULL)
```

Arguments

kappa a shape parameter of the weight function.

n the length list.

type draw line or points. Both line and points will be plotted if type = c('line',

'points').

ylim y coordinates ranges.

Value

plot based on basic graph

Examples

```
par(mfrow = c(1, 2), mar = c(4, 4, 2, 2))
plotAdaptiveWeights(kappa = 0.01, n = 100, type = 'line', ylim = c(0, 0.025))
plotAdaptiveWeights(kappa = 0.02, n = 100, type = 'line', ylim = c(0, 0.025))
```

plotIPs

Visualize the influential scores

Description

Visualize influential score using lollipop plot. The function uses the output obtained from rank.compare or findIPs function.

Usage

```
plotIPs(obj, topn = 5, ylim = NULL, ...)
```

Arguments

obj	the object obtained from rank.compare or findIPs function.
topn	the top n most influential points to be labelled in the plot.
ylim	y coordinates ranges
	other arguments

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Value

plot based on basic graph

Examples

plotRankScatters

Visualize the unweighted rank changes

Description

Visualize the unweighted rank changes using scatter plot. The plot displays the original ranking and leave-one-out rankings.

Usage

```
plotRankScatters(obj, top = TRUE, points.arg = list(), top.arg = list())
```

Arguments

obj	the objective obtained from findIPs() or sumRanks() functions
top	logical, whether the most influential case needs to be plot in black
points.arg	a list. Arguments in graphics::points() can be used to define the points.
top.arg	a list. Arguments in $graphics::points()$ can be used to define the top points.

Value

a plot based on basic graphic.

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sumRanks

Summarize the weighted rank changes caused by case-deletion

Description

This function measures the overall rank changes due to case deletion. A large rank changes indicates more influence of the deleted case on feature rankings. sumRanks() provides three methods to compute the overall rank changes: unweighted, weighted Spearman, and adaptive weights.

Usage

```
sumRanks(origRank, drop1Rank, topN = NULL, method = "adaptive", ...)
```

Arguments

or igRank vectors, reference rankings. For influential observation detection, origRank de-

notes the original ranking obtained using the whole data.

drop1Rank matrix or data.frame, Each column is a feature list with a case removed.

topN the top n features in origRank will be used for rank comparison. If null, include

all features.

method method to summarize rank changes. It should be one of the 'adaptive', 'weight-

edSpearman', and 'unweighted'. Both 'adaptive' and 'weightedSpearman' are weighted rank comparison method, but former employs the weight that are adaptive to the distribution of rank changes. 'unweighted' denotes a direct compari-

son of ranks without considering weights.

... other arguments

Value

kappa The weight function's shape is controlled by kappa, which ranges from 0 to 1.

Weighted rank changes are calculated using kappa, with higher values indicating

more weight on top features.

score The influence of each observation on feature rankings, with larger values indi-

cating more influence.

origRank The original ranking. origRank is exactly the input. Here it is re-output for

visualization purposes.

drop1Rank The leave-one-out rankings.

origRankWeighted

The weighted original ranking. origRankWeighted will be returned when method

= 'adaptive'.

drop1RankWeighted

The weighted leave-one-out rankings. drop1RankWeighted will be returned

when method = 'adaptive'.

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