

Package ‘pcaBootPlot’

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Title Create 2D Principal Component Plots with Bootstrapping

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Description Draws a 2D principal component plot using the first 2 principal components from the original and bootstrapped data to give some sense of variability.

Depends R (>= 3.0.2)

License GPL-2

LazyData true

Imports FactoMineR, RColorBrewer

Suggests knitr

VignetteBuilder knitr

URL <https://github.com/starmerj/pcaBootPlot>

NeedsCompilation no

Repository CRAN

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pcaBootPlot

*Create 2D PCA Plots with Bootstrapping***Description**

pcaBootPlot draws a 2D PCA plot using the first 2 principal components using the original and bootstrapped data to give some sense of variability.

Usage

```
pcaBootPlot(data = NULL, groups = NULL, min.value = 1,
  all.min.value = FALSE, num.boot.samples = 100, log2.transform = TRUE,
  pdf.filename = NULL, pdf.width = 6, pdf.height = 6,
  draw.legend = FALSE, legend.names = NULL, legend.x = NULL,
  legend.y = NULL, transparency = 77, min.x = NULL, max.x = NULL,
  min.y = NULL, max.y = NULL, correct.inversions = TRUE,
  confidence.regions = FALSE, confidence.size = 0.95, step.size = 0.1,
  trim.proportion = 0, return.samples = FALSE, use.pcomp = FALSE)
```

Arguments

data A data.frame where the first column is named "ID" and contains IDs for each item measured. Measurements for each sample are in subsequent columns.

groups The default value is NULL.

If you want use different colors and shapes to deliniate the samples into groups, you can specify the grouping with this argument. Currently there is a limit of 9 different groups.

For example, if you have three consecutive columns of "untreated" samples followed by three consecutive columns of "treated" samples, you can set this argument to c(1,1,1,2,2,2), and the untreated samples will be red circles and the treated samples will be blue triangles.

min.value The default value is **1**.

This allows you to filter out rows (entries) that will not contribute to the PCA. For example, if you are performing PCA on RNA-seq data, you may wish to filter out genes with less than 1 read per sample, 1 read per group or 1 read overall. If you set all.min.value to TRUE, it will filter entries where at least one sample has less than min.value. If you do not set all.min.value to TRUE, then filtering will be performed by group if groups are specified. In this case, an entry will be filtered out if one or more groups have less than min.value.

If groups are not specified, then only entries where all samples have less than min.value will be removed from the analysis.

	groups will also effect filtering based on <code>min.value</code> . See that part of the documentation for details.
<code>all.min.value</code>	This parameter, set to either <code>TRUE</code> or <code>FALSE</code> , affects <code>min.value</code> . See the documentation for <code>min.value</code> for more details.
<code>num.boot.samples</code>	The default value is 100 . The number of bootstrap iterations to be performed.
<code>log2.transform</code>	The default value is <code>TRUE</code> . Should the data be log2 transformed or not?
<code>pdf.filename</code>	If you wish to save the the graph as a PDF, you may use this argument to specify the filename.
<code>pdf.width</code>	If you specify a value for <code>pdf.filename</code> , you can specify a width for the saved graph. The default value is 6 inches.
<code>pdf.height</code>	If you specify a value for <code>pdf.filename</code> , you can specify a height for the saved graph. The default value is 6 inches.
<code>draw.legend</code>	The default value is <code>FALSE</code> . Should there be a legend in the graph?
<code>legend.names</code>	If <code>draw.legend</code> is <code>TRUE</code> , you can specify the names of the groups listed in the legend.
<code>legend.x, legend.y</code>	If <code>draw.legend</code> is <code>TRUE</code> , you can specify the x and y axis coordinate for its top left corner.
<code>transparency</code>	The default value is 77 . This allow you to set how transparent the bootstrapped symbols are in the graph. Values range from <code>00</code> to <code>FF</code> .
<code>min.x, min.y, max.x, max.y</code>	By default, <code>pcaBootPlot</code> automatically determines limits for the x and y axes. Use this option to override this behavior.
<code>correct.inversions</code>	The default value is <code>TRUE</code> . Some of the bootstrapped PCAs may have their axes inverted. <code>pcaBootPlot</code> can try to correct for this by ensuring that the PCA loading values are positively correlated with the original dataset.
<code>confidence.regions</code>	The default value is <code>FALSE</code> . This option will draw circles that contain <code>confidence.size</code> of the bootstrapped values.
<code>confidence.size</code>	The default value is 0.95 . A value between 0 and 1 - the proportion of bootstrapped points that need to be within the confidence regions.
<code>step.size</code>	The default value is 0.1 . This option determines how the radii for confidence regions are increased each iteration when trying to contain <code>confidence.size</code> of the bootstrapped samples.
<code>trim.proportion</code>	The default value is 0.0 . This is the proportion of entries that should be removed from the plot based on the size of the the confidence regions. This should be a value between 0 and 1. For example, if you set it to 0.1, then the top 10 regions will be removed from the plot.
<code>return.samples</code>	The default value is <code>FALSE</code> . If this is set to <code>TRUE</code> then the program will return the names of the samples that were included in the plot. This can be useful if <code>trim.proportion > 0</code> .

`use.pcomp` The default value is FALSE. Usually, `pcaBootPlot` uses `FactoMineR` to process samples. However, this can be unnecessarily slow if there are less than 50 samples. By setting `use.pcomp` to TRUE, it will use `prcomp()` to process samples and will, most likely, run much faster.

Examples

```
sample1=rnorm(n=100, mean=100, sd=10)
sample2=jitter(sample1, factor=10, amount=10)
sample3=rnorm(n=100, mean=100, sd=10)

data <- data.frame(ID=c(1:100), sample1, sample2, sample3)

pcaBootPlot(data, log2.transform = FALSE)
```

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