

Package ‘Pursuit’

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Type Package

Title Projection Pursuit

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Imports graphics, grDevices, MASS, stats

Description Projection pursuit (PP) with 17 methods and grand tour with 3 methods. Being that projection pursuit searches for low-dimensional linear projections in high-dimensional data structures, while grand tour is a technique used to explore multivariate statistical data through animation.

License GPL-3

Encoding UTF-8

NeedsCompilation yes

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GrandTour

*Animation technique Grand Tour.***Description**

Performs the exploration of the data through the technique of animation Grand Tour.

Usage

```
GrandTour(data, method = "Interpolation", title = NA, xlabel = NA,
          ylabel = NA, size = 1.1, grid = TRUE, color = TRUE, linlab = NA,
          class = NA, classcolor = NA, posleg = 2, boxleg = TRUE,
          axesvar = TRUE, axes = TRUE, numrot = 200, choicerot = NA,
          savptc = FALSE, width = 3236, height = 2000, res = 300)
```

Arguments

data	Numerical data set.
method	Method used for rotations: "Interpolation" - Interpolation method (default), "Torus" - Torus method, "Pseudo" - Pseudo Grand Tour method.
title	Titles of the graphics, if not set, assumes the default text.
xlabel	Names the X axis, if not set, assumes the default text.
ylabel	Names the Y axis, if not set, assumes the default text.
size	Size of the points in the graphs.
grid	Put grid on graphs (default = TRUE).
color	Colored graphics (default = TRUE).
linlab	Vector with the labels for the observations.
class	Vector with names of data classes.
classcolor	Vector with the colors of the classes.
posleg	0 with no caption, 1 for caption in the left upper corner, 2 for caption in the right upper corner (default), 3 for caption in the right lower corner, 4 for caption in the left lower corner.
boxleg	Puts the frame in the caption (default = TRUE).
axesvar	Puts axes of rotation of the variables (default = TRUE).
axes	Plots the X and Y axes (default = TRUE).
numrot	Number of rotations (default = 200). If method = "Interpolation", numrot represents the angle of rotation.

choicerot	Choose specific rotation and display on the screen, or save the image if savptc = TRUE.
savptc	Saves graphics images to files (default = FALSE).
width	Graphics images width when savptc = TRUE (default = 3236).
height	Graphics images height when savptc = TRUE (default = 2000).
res	Nominal resolution in ppi of the graphics images when savptc = TRUE (default = 300).

Value

Graphs with rotations.

proj.data	Projected data.
vector.opt	Vector projection.
method	method used on Grand Tour.

Author(s)

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References

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Examples

```

data(iris) # database

res <- GrandTour(iris[,1:4], method = "Torus", title = NA, xlabel = NA, ylabel = NA,
  color = TRUE, linlab = NA, class = NA, posleg = 2, boxleg = TRUE,
  axesvar = TRUE, axes = FALSE, numrot = 10, choicerot = NA,
  savptc = FALSE, width = 3236, height = 2000, res = 300)

print("Projected data:"); res$proj.data
print("Projection vectors:"); res$vector.opt
print("Grand Tour projection method:"); res$method

res <- GrandTour(iris[,1:4], method = "Interpolation", title = NA, xlabel = NA, ylabel = NA,
  color = TRUE, linlab = NA, posleg = 2, boxleg = FALSE, axesvar = FALSE,
  axes = FALSE, numrot = 10, choicerot = NA, class = iris[,5],
  classcolor = c("goldenrod3", "gray53", "red"), savptc = FALSE,
  width = 3236, height = 2000, res = 300)

print("Projected data:"); res$proj.data
print("Projection vectors:"); res$vector.opt
print("Grand Tour projection method:"); res$method

```

LocLab

Function for better position of the labels in the graphs.

Description

Function for better position of the labels in the graphs.

Usage

```

LocLab(x, y = NULL, labels = seq(along = x), cex = 1,
  method = c("SANN", "GA"), allowSmallOverlap = FALSE,
  trace = FALSE, shadotext = FALSE,
  doPlot = TRUE, ...)

```

Arguments

x	Coordinate x
y	Coordinate y
labels	The labels
cex	cex
method	Not used
allowSmallOverlap	Boolean


```

tit <- c(NA,"Graph example") # titles for the graphics

Plot.PP(res, titles = tit, posleg = 1, boxleg = FALSE, color = TRUE,
        classcolor = c("blue3","red","goldenrod3"), linlab = NA,
        axesvar = TRUE, axes = TRUE, savptc = FALSE, width = 3236,
        height = 2000, res = 300, casc = FALSE)

# Example 3 - Without the classes in the data, but informing
#           the classes in the plot function
res <- PP_Optimizer(data = data, class = NA, findex = "Moment",
                  optmethod = "GTSA", dimproj = 2, sphere = sphere,
                  weight = TRUE, lambda = 0.1, r = 1, cooling = 0.9,
                  eps = 1e-3, maxiter = 500, half = 30)

lin <- c(rep("a",50),rep("b",50),rep("c",50)) # data class

Plot.PP(res, titles = NA, posleg = 1, boxleg = FALSE, color = TRUE,
        linlab = lin, axesvar = TRUE, axes = TRUE, savptc = FALSE,
        width = 3236, height = 2000, res = 300, casc = FALSE)

# Example 4 - With the classes in the data, but not informed in plot function
class <- iris[,5] # data class

dim <- 2 # dimension of data projection

findex <- "lda" # index function

res <- PP_Optimizer(data = data, class = class, findex = findex,
                  optmethod = "GTSA", dimproj = dim, sphere = sphere,
                  weight = TRUE, lambda = 0.1, r = 1, cooling = 0.9,
                  eps = 1e-3, maxiter = 500, half = 30)

tit <- c("",NA) # titles for the graphics

Plot.PP(res, titles = tit, posleg = 1, boxleg = FALSE, color = TRUE,
        linlab = NA, axesvar = TRUE, axes = TRUE, savptc = FALSE,
        width = 3236, height = 2000, res = 300, casc = FALSE)

```

Description

Function used to find Projection Pursuit indexes (PP).

Usage

```
PP_Index(data, class = NA, vector.proj = NA,
         findex = "HOLES", dimproj = 2, weight = TRUE,
         lambda = 0.1, r = 1, ck = NA)
```

Arguments

<code>data</code>	Numeric dataset without class information.
<code>class</code>	Vector with names of data classes.
<code>vector.proj</code>	Vector projection.
<code>findex</code>	Projection index function to be used: "lda" - LDA index, "pda" - PDA index, "lr" - Lr index, "holes" - Holes index (default), "cm" - Central Mass index, "pca" - PCA index, "friedmantukey" - Friedman Tukey index, "entropy" - Entropy index, "legendre" - Legendre index, "laguerrefourier" - Laguerre Fourier index, "hermite" - Hermite index, "naturalhermite" - Natural Hermite index, "kurtosismax" - Maximum kurtosis index, "kurtosismin" - Minimum kurtosis index, "moment" - Moment index, "mf" - MF index, "chi" - Chi-square index.
<code>dimproj</code>	Dimension of data projection (default = 2).
<code>weight</code>	Used in index LDA, PDA and Lr to weight the calculations for the number of elements in each class (default = TRUE).
<code>lambda</code>	Used in the PDA index (default = 0.1).
<code>r</code>	Used in the Lr index (default = 1).
<code>ck</code>	Internal use of the CHI index function.

Value

<code>num.class</code>	Number of classes.
<code>class.names</code>	Class names.
<code>findex</code>	Projection index function used.
<code>vector.proj</code>	Projection vectors found.
<code>index</code>	Projection index found in the process.

Author(s)

Paulo Cesar Ossani
Marcelo Angelo Cirillo

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See Also

[PP_Optimizer](#) and [Plot.PP](#)

Examples

```

data(iris) # data set

data <- iris[,1:4]

# Example 1 - Without the classes in the data
ind <- PP_Index(data = data, class = NA, vector.proj = NA,
               findex = "moment", dimproj = 2, weight = TRUE,
               lambda = 0.1, r = 1)

print("Number of classes:"); ind$num.class
print("class Names:"); ind$class.names
print("Projection index function:"); ind$findex
print("Projection vectors:"); ind$vector.proj
print("Projection index:"); ind$index

# Example 2 - With the classes in the data
class <- iris[,5] # data class

findex <- "pda" # index function

sphere <- TRUE # spherical data

res <- PP_Optimizer(data = data, class = class, findex = findex,
                  optmethod = "SA", dimproj = 2, sphere = sphere,
                  weight = TRUE, lambda = 0.1, r = 1, cooling = 0.9,
                  eps = 1e-3, maxiter = 1000, half = 30)

# Comparing the result obtained
if (match(toupper(findex),c("LDA", "PDA", "LR"), nomatch = 0) > 0) {
  if (sphere) {
    data <- apply(predict(prcomp(data)), 2, scale) # spherical data
  }
} else data <- as.matrix(res$proj.data[,1:Dim])

ind <- PP_Index(data = data, class = class, vector.proj = res$vector.opt,
               findex = findex, dimproj = 2, weight = TRUE, lambda = 0.1,
               r = 1)

print("Number of classes:"); ind$num.class
print("class Names:"); ind$class.names
print("Projection index function:"); ind$findex
print("Projection vectors:"); ind$vector.proj
print("Projection index:"); ind$index
print("Optimized Projection index:"); res$index[length(res$index)]

```

Description

Optimization function of the Projection Pursuit index (PP).

Usage

```
PP_Optimizer(data, class = NA, findex = "HOLES",
             dimproj = 2, sphere = TRUE, optmethod = "GTSA",
             weight = TRUE, lambda = 0.1, r = 1, cooling = 0.9,
             eps = 1e-3, maxiter = 3000, half = 30)
```

Arguments

data	Numeric dataset without class information.
class	Vector with names of data classes.
findex	Projection index function to be used: "lda" - LDA index, "pda" - PDA index, "lr" - Lr index, "holes" - Holes index (default), "cm" - Central Mass index, "pca" - PCA index, "friedmantukey" - Friedman Tukey index, "entropy" - Entropy index, "legendre" - Legendre index, "laguerrefourier" - Laguerre Fourier index, "hermite" - Hermite index, "naturalhermite" - Natural Hermite index, "kurtosismax" - Maximum kurtosis index, "kurtosismin" - Minimum kurtosis index, "moment" - Moment index, "mf" - MF index, "chi" - Chi-square index.
dimproj	Dimension of the data projection (default = 2).
sphere	Spherical data (default = TRUE).
optmethod	Optimization method GTSA - Grand Tour Simulated Annealing or SA - Simulated Annealing (default = "GTSA").
weight	Used in index LDA, PDA and Lr to weight the calculations for the number of elements in each class (default = TRUE).
lambda	Used in the PDA index (default = 0.1).
r	Used in the Lr index (default = 1).
cooling	Cooling rate (default = 0.9).
eps	Approximation accuracy for cooling (default = 1e-3).
maxiter	Maximum number of iterations of the algorithm (default = 3000).
half	Number of steps without incrementing the index, then decreasing the cooling value (default = 30).

Value

num.class	Number of classes.
class.names	Class names.
proj.data	Projected data.
vector.opt	Projection vectors found.
index	Vector with the projection indices found in the process, converging to the maximum, or the minimum.
findex	Projection index function used.

Author(s)

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References

COOK, D., LEE, E. K., BUJA, A., WICKHAM, H.. Grand tours, projection pursuit guided tours and manual controls. In Chen, Chunhouh, Hardle, Wolfgang, Unwin, e Antony (Eds.), *Handbook of data Visualization*, Springer Handbooks of Computational Statistics, chapter III.2, p. 295-314. Springer, 2008.

LEE, E., COOK, D., KLINKE, S., LUMLEY, T.. Projection pursuit for exploratory supervised classification. *Journal of Computational and Graphical Statistics*, 14(4):831-846, 2005.

See Also

[Plot.PP](#) and [PP_Index](#)

Examples

```
data(iris) # data set

# Example 1 - Without the classes in the data
data <- iris[,1:4]

class <- NA # data class

findex <- "kurtosismax" # index function

dim <- 1 # dimension of data projection

sphere <- TRUE # spherical data

res <- PP_Optimizer(data = data, class = class, findex = findex,
                    optmethod = "GTSA", dimproj = dim, sphere = sphere,
                    weight = TRUE, lambda = 0.1, r = 1, cooling = 0.9,
                    eps = 1e-3, maxiter = 1000, half = 30)

print("Number of classes:"); res$num.class
```

```
print("class Names:"); res$class.names
print("Projection index function:"); res$findex
print("Projected data:"); res$proj.data
print("Projection vectors:"); res$vector.opt
print("Projection index:"); res$index

# Example 2 - With the classes in the data
class <- iris[,5] # classe dos dados

res <- PP_Optimizer(data = data, class = class, findex = findex,
                    optmethod = "GTSA", dimproj = dim, sphere = sphere,
                    weight = TRUE, lambda = 0.1, r = 1, cooling = 0.9,
                    eps = 1e-3, maxiter = 1000, half = 30)

print("Number of classes:"); res$num.class
print("class Names:"); res$class.names
print("Projection index function:"); res$findex
print("Projected data:"); res$proj.data
print("Projection vectors:"); res$vector.opt
print("Projection index:"); res$index
```

Pursuit

Projection Pursuit

Description

Projection pursuit (PP) with 17 methods and grand tour with 3 methods. Being that projection pursuit searches for low-dimensional linear projections in high-dimensional data structures, while grand tour is a technique used to explore multivariate statistical data through animation.

Details

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