

# Package ‘PPtree’

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**Title** Projection pursuit classification tree

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**Imports** MASS (>= 3.1-20), penalizedLDA (>= 1.0)

**Description** Projection pursuit classification tree using LDA, Lr or PDA projection pursuit index

**License** LGPL-2.1

**NeedsCompilation** yes

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ClassPP *Projection Pursuit for Supervised Classification*

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## Description

This package is for projection pursuit method for supervised classification.

**Author(s)**

Eun-kyung Lee

**References**Lee, E., Cook, D., and Klinke, S.(2002) *Projection Pursuit indices for supervised classification***See Also**

PPindex.class \ PPindex.LDA \ PPindex.Lp \ PPindex.PDA

PP.optimize.anneal

PP.optimize.plot

PP.Tree

PP.classify

---

LDA.Tree

---

*Find PP tree structure using LDA*


---

**Description**

Find tree structure using linear discriminant in each split.

**Usage**

LDA.Tree( i.class, i.data, weight = TRUE, ...)

**Arguments**

i.data            A training data without class information

i.class           class information

weight           weight flag using in LDA index

...                ...

**Value**

Tree.Struct      Tree structure

Alpha.Keep      1D projection of each split

C.Keep           splitting rule for each split

**Author(s)**

Eun-kyung Lee

**References**

Lee, E., Cook, D., and Klink, S.(2002) *Projection Pursuit indices for supervised classification*

**See Also**

[PPindex.class](#), [PP.optimize](#)

**Examples**

```
data(iris)
n <- nrow(iris)
tot <- c(1:n)
n.train <- round(n*0.9)
train <- sample(tot,n.train)
test <- tot[-train]

Tree.result <- LDA.Tree(iris[train,5],iris[train,1:4])
Tree.result
```

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PDA.Tree

*Find PP tree structure using PDA*


---

**Description**

Find tree structure using projection pursuit in each split.

**Usage**

```
PDA.Tree(i.class, i.data, weight = TRUE, lambda=1, ...)
```

**Arguments**

<code>i.data</code>	A training data without class information
<code>i.class</code>	class information
<code>weight</code>	weight flag using in LDA index
<code>lambda</code>	a parameter for PDA index
<code>...</code>	...

**Value**

<code>Tree.Struct</code>	Tree structure
<code>Alpha.Keep</code>	1D projection of each split
<code>C.Keep</code>	splitting rule for each split

**Author(s)**

Eun-kyung Lee

**References**Lee, E., Cook, D., and Klinke, S.(2002) *Projection Pursuit indices for supervised classification***See Also**[PPindex.class](#), [PP.optimize](#)**Examples**

```

data(iris)
n <- nrow(iris)
tot <- c(1:n)
n.train <- round(n*0.9)
train <- sample(tot,n.train)
test <- tot[-train]

Tree.result <- PDA.Tree(iris[train,5],iris[train,1:4])
Tree.result

```

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PP.classify

---

*Predict class for the test set and calculate prediction error*


---

**Description**

After finding tree structure, predict class for the test set and calculate prediction error.

**Usage**

```
PP.classify(test.data, true.class, Tree.result, Rule, ...)
```

**Arguments**

test.data	the test dataset
true.class	true class of test dataset if available
Tree.result	the result of PP.Tree
Rule	split rule
	1 - mean of two group means
	2 - weighted mean of two group means
	3 - mean of max(left group) and min(right group)
	4 - weighted mean of max(left group) and min(right group)
...	...

**Value**

A list with components:

```
predict.class  predicted class
predict.error  prediction error
```

**Author(s)**

Eun-kyung Lee

**References**

Lee E., Cook D., and Klinke, S. (2002) *Projection Pursuit indices for supervised classification*

**See Also**

[PPindex.class](#), [PP.optimize](#), [PP.Tree](#)

**Examples**

```
data(iris)
n <- nrow(iris)
n.train <- round(n*0.9)
train <- sample(n, n.train)

Tree.result <- PP.Tree("LDA", iris[train,5], iris[train, 1:4])
tree.train <- PP.classify(iris[train, 1:4], iris[train, 5], Tree.result,
                          Rule=1)

tree.train
tree.test <- PP.classify(iris[-train, 1:4], iris[-train, 5],
                          Tree.result, Rule=1)

tree.test
```

---

PP.optimize

*Find optimal projection by maximizing selected PP index*

---

**Description**

Find optimal projection using PP index.

**Usage**

```
PP.optimize.random(PPmethod, projdim, data, class, std=TRUE,
                   cooling=0.99, temp=1, r=NULL, lambda=NULL, weight=TRUE, ...)
PP.optimize.anneal(PPmethod, projdim, data, class, std=TRUE,
                   cooling=0.999, temp=1, energy=0.01,
                   r=NULL, lambda=NULL, weight=TRUE, ...)
PP.optimize.Huber(PPmethod, projdim, data, class, std=TRUE,
```

```

        cooling=0.99, temp=1, r=NULL, lambda=NULL,
        weight=TRUE, ...)
PP.optimize.plot(PP.opt, data, class, std=TRUE)

```

### Arguments

PPmethod	Selected PP index "LDA" - LDA index "Lp" - Lp index; "PDA" - PDA index
projdim	dimension of projection that you want to find
data	data without class information
class	class information
std	decide whether data will be standardized or not before applying projection pursuit
weight	weight flag using in LDA index
cooling	parameter for optimization
temp	inital temperature for optimization
energy	parameter for simulated annealing optimization
r	a parameter for $L_r$ index
lambda	a parameter for PDA index
PP.opt	the optimal projection
...	...

### Value

index.best	PP index for optimal projected data
proj.best	optimal projection

### Author(s)

Eun-kyung Lee

### References

Lee E., Cook D., and Klinke, S. (2002) *Projection Pursuit indices for supervised classification*

### See Also

{PPindex.class}

**Examples**

```

data(iris)

PP.opt<-PP.optimize.random("LDA",1,iris[,1:4],iris[,5],cooling=0.999,temp=1)

PP.opt$index.best
PP.optimize.plot(PP.opt,iris[,1:4],iris[,5])

PP.opt<-PP.optimize.anneal("LDA",1,iris[,1:4],iris[,5],cooling=0.999,temp=1,energy=0.01)
PP.opt$index.best

PP.optimize.plot(PP.opt,iris[,1:4],iris[,5])

PP.opt<-PP.optimize.Huber("LDA",2,iris[,1:4],iris[,5],cooling=0.999,r=1)
PP.opt$index.best
PP.optimize.plot(PP.opt,iris[,1:4],iris[,5])

```

---

 PP.Tree

*Find PP tree structure*


---

**Description**

Find tree structure using projection pursuit in each split.

**Usage**

```

PP.Tree(PPmethod, i.class, i.data, weight = TRUE, r = NULL,
        lambda = NULL, cooling = 0.999, temp = 1, energy = 0.01, ...)

```

**Arguments**

PPmethod	Selected PP index "LDA" - LDA index "Lp" - Lp index; "PDA" - PDA index
i.data	A training data without class information
i.class	class information
weight	weight flag using in LDA index
r	a parameter for $L_r$ index
lambda	a parameter for PDA index
cooling	parameter for optimization
temp	inital temperature for optimization
energy	parameter for simulated annealing optimization
...	...

**Value**

Tree.Struct	Tree structure
Alpha.Keep	1D projection of each split
C.Keep	splitting rule for each split

**Author(s)**

Eun-kyung Lee

**References**

Lee, E., Cook, D., and Klinke, S.(2002) *Projection Pursuit indices for supervised classification*

**See Also**

[PPindex.class](#), [PP.optimize](#)

**Examples**

```
data(iris)
n <- nrow(iris)
tot <- c(1:n)
n.train <- round(n*0.9)
train <- sample(tot,n.train)
test <- tot[-train]

Tree.result <- PP.Tree("LDA",iris[train,5],iris[train,1:4])
Tree.result
```

---

PPindex.class

*Calculate Projection Pursuit index*

---

**Description**

For given projected data and class information, calculate projeciton pursuit index.

**Usage**

```
PPindex.class(PPmethod, data, class, weight=TRUE, r=NULL, lambda=NULL, ...)
PPindex.LDA(data, class, weight=TRUE, ...)
PPindex.Lp(data, class, r, ...)
PPindex.PDA(data, class, lambda, ...)
```



**Arguments**

PPmethod	Selected PP index "LDA" - LDA index "Lp" - Lp index "PDA" - Entropy-class index
data	A data without class information
class	class information
weight	weight flag using in LDA index
r	a parameter for $L^r$ index
lambda	a parameter for PDA index
...	...

**Value**

The value is an projection pursuit index for given data.

**Author(s)**

Eun-kyung Lee

**References**

Lee, E., Cook, D., and Klinke, S.(2002) *Projection Pursuit indices for supervised classification*

**See Also**

[PP.optimize](#)

**Examples**

```
data(iris)

PPindex.class("LDA",iris[,1:2],iris[,5])
PPindex.class("LDA",iris[,1:2],iris[,5],weight=FALSE)
PPindex.class("Lp",iris[,1:2],iris[,5],r=1)
PPindex.class("PDA",iris[,1:2],iris[,5],lambda=0.1)
```

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