# Package 'featurefinder'

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Type Package

Version 1.2

Title Feature Finder

2 findFeatures

data

data

## **Description**

Sample data based on dataset EuStockMarkets in the datasets package.

## **Format**

A data frame with 1860 rows and 4 variables

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#### **Source**

```
https://stat.ethz.ch/R-manual/R-devel/library/datasets/html/00Index.html
```

## **Examples**

```
data(mycsv)
thismodel=lm(formula=DAX ~ .,data=data)
expectedprob=predict(thismodel,data)
actualprob=data$DAX
residual=actualprob-expectedprob
data=cbind(data,expectedprob, actualprob, residual)
```

findFeatures

findFeatures

## Description

Perform analysis of residuals grouped by factor to identify features which explain the target variable

## Usage

```
findFeatures(
   OutputPath,
   fcsv,
   ExclusionVars,
   FactorToNumericList,
   treeGenerationMinBucket = 50,
   treeSummaryMinBucket = 20,
   treeSummaryResidualThreshold = 0,
   treeSummaryResidualMagnitudeThreshold = 0,
   doAllFactors = TRUE,
   maxFactorLevels = 20
)
```

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#### **Arguments**

OutputPath A string containing the location of the input csv file. Results are also stored in

this location.

fcsv A string containing the name of a csv file

ExclusionVars A string consisting of a list of variable names with double quotes around each

variable

FactorToNumericList

A list of variable names as strings

treeGenerationMinBucket

Desired minimum number of data points per leaf (default 50)

treeSummaryMinBucket

Minimum number of data points in each leaf for the summary (default 20)

treeSummaryResidualThreshold

Minimum residual in the summary (default 0 for positive residuals)

treeSummaryResidualMagnitudeThreshold

Minimum residual magnitude in the summary (default 0 i.e. no restriction)

doAllFactors Flag to indicate whether to analyse the levels of all factor variables (default

TRUE)

maxFactorLevels

(maximum number of levels per factor before it is converted to numeric (default 20)

## Value

Saves residual CART trees and associated highlighted residuals for each to the path provided.

## **Examples**

```
require(featurefinder)
data(mycsv)
data$SMIfactor=paste("smi", as.matrix(data$SMIfactor), sep="")
nn=floor(length(data$DAX)/2)
# Can we predict the relative movement of DAX and SMI?
data$y=data$DAX*0
data$y[1:(nn-1)]=((data$DAX[2:nn])-(data$DAX[1:(nn-1)]))/
                                          \label{eq:continuous} $$ (data$DAX[1:(nn-1)])-(data$SMI[2:nn]-(data$SMI[1:(nn-1)]))/(data$SMI[1:(nn-1)]) $$ (data$DAX[1:(nn-1)])-(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)])/(data$SMI[1:(nn-1)]/(data
thismodel=lm(formula=y ~ .,data=data)
expected=predict(thismodel,data)
actual=data$y
residual=actual-expected
data=cbind(data,expected, actual, residual)
# setwd('.\test')
write.csv(data[(nn+1):(length(data$y)),],file='mycsv.csv',row.names=FALSE)
OutputPath="."
fcsv="mycsv.csv"
ExclusionVars="\"residual\",\"expected\", \"actual\",\"y\""
FactorToNumericList=c()
findFeatures(OutputPath, fcsv, ExclusionVars, FactorToNumericList,
                               treeGenerationMinBucket=50,
                               treeSummaryMinBucket=20)
```

4 generateTrees

```
{\tt generateResidualCutoffCode}
```

generateResidualCutoffCode

## Description

For each tree print a summary of the significant residuals as specified by the user

## Usage

```
generateResidualCutoffCode(data, filename, trees, names, runname, ...)
```

## **Arguments**

data A dataframe filename A string

trees A list of trees generated by saveTree

names A list of level names

runname A string corresponding to the name of the factor variable being analysed

... and parameters to be passed through

## Value

A list of residuals for each tree provided.

generateTrees generateTrees

## **Description**

Generate a residual tree for each level of factor mainfac

## Usage

```
generateTrees(data, vars, expr, runname, ...)
```

## **Arguments**

data A dataframe

vars A list of candidate predictors

expr A expression to be modelled by the RPART tree

runname A string corresponding to the name of the variable being modelled

... and parameters to be passed through

## Value

A list of residual trees for each level of the mainfac factor provided

getVarAv 5

## Description

This function generates a residual tree on a subset of the data

## Usage

```
getVarAv(dd, varAv, varString)
```

## **Arguments**

dd A dataframe

varAv A string corresponding to the numeric field to be averaged within each leaf node

varString A string

## Value

An average of the numeric variable varString in the segment

## Description

Extract information relating to the paths and volume of data in the leaves of the tree

## Usage

```
parseSplits(thistree)
```

## **Arguments**

thistree A tree

## Value

A list of parsed splits.

6 printResiduals

printResiduals

## Description

This function generates a residual tree on a subset of the data

## Usage

```
printResiduals(
  fileConn,
  all,
  dat,
  runname,
  levelname,
  treeSummaryResidualThreshold,
  treeSummaryResidualMagnitudeThreshold,
  ...
)
```

## Arguments

A file connection fileConn all A dataframe dat The dataset A string corresponding to the name of the factor being analysed runname levelname A string corresponding to the factor level being analysed tree Summary Residual ThresholdThe minimum residual threshold  $tree Summary {\tt MinBucket}$ The minumum volume per leaf tree Summary Residual Magnitude ThresholdMinimun residual magnitude and parameters to be passed through

## Value

Residuals are printed and also saved in a simplified format.

saveTree 7

## Description

Generate a residual tree on a subset of the data specified by the factor level mainfaclev (main factor level)

## Usage

```
saveTree(
  data,
  vars,
  expr,
  i,
  varname,
  mainfaclev,
  treeGenerationMinBucket,
  ...
)
```

## Arguments

data A dataframe containing the residual and some predictors vars A list of candidate predictors

expr A expression to be modelled by the RPART tree i An integer corresponding to the factor level

varname A string corresponding to the name of the factor variable being analysed

mainfaclev A level of the mainfac factor

 $tree {\tt Generation Min Bucket}$ 

Minimum size for tree generation

... and parameters to be passed through

## Value

A tree object

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