# Package 'YRmisc'

October 12, 2022

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cor.lag

Lag/Lead Correlation

# Description

Calculating correlation of two vectors with lag and lead periods. The correlations are used to determine the lag or lead effect between two variables. The correlation function uses "na.or.complete" method and calculate the Pearson's correlation.

# Usage

```
cor.lag(x,y,lag,lead)
```

# Arguments

x :the moving vector y :the fixed vector

lag :number of lag periods lead :number of lead periods 4 cv.annu.fv

## **Examples**

```
cor.lag(mtcars[,1],mtcars[,2],3,3)
```

cor.spearman

Spearman rank correlation

## **Description**

Calculate Spearman Rank Correlation, which is the nonparametric version of the Pearson product-moment correlation.

## Usage

```
cor.spearman(x,y)
```

# Arguments

x :a numeric variable y :a numeric variable

# **Examples**

```
cor.spearman(mtcars[,1], mtcars[,3])
```

cv.annu.fv

Calculate future value of annuity

# Description

Calculate future value of an ordinary annuity or an annuity due.

## Usage

```
cv.annu.fv(pmt,i,n,type = 0)
```

# Arguments

pmt : the equal amount of payment of each period

i :interest rate according to the period

n :number of periods

type : type = 0 for ordinary annuity, type = 1 for annuity due

```
cv.annu.fv(100,0.0248,10,0)
```

cv.annu.pv 5

cv.annu.pv

Calculate present value of annuity

## **Description**

Calculate present value of an ordinary annuity or an annuity due.

## Usage

```
cv.annu.pv(pmt,i,n,k)
```

## Arguments

pmt : the equal amount of payment of each period

i :interest rate according to the period

n :number of periods

k :number of periods deferred until first payment

# **Examples**

```
cv.annu.pv(100,0.0248,10,4)
```

cv.axp

Create logarithm with a random base

## **Description**

Create a new variable with the base of a random number and power of the selected variable

# Usage

```
cv.axp(dataframe, var, n, range)
```

## **Arguments**

dataframe :a data frame

var :the variable selected

n :number of new variables created

range :the range of base

```
cv.axp(mtcars,"wt",5,c(1, 2))
```

6 cv.diff

cv.bondprice

Calculate the plain vanilla bond price

# Description

Calculate the plain vanilla bond price

## Usage

```
cv.bondprice(par,c,yield,n,m)
```

## **Arguments**

par : the face value of the bond

c :the annual coupon rate of the bond yield :the annual yield to maturity of a bond

n :number of years

m :compounding period in a year

#### **Examples**

```
cv.bondprice(1000,0.0248,0.0248,10,2)
```

cv.diff

Calculating the difference of a time series

# Description

Calculate the difference of a time series, with a specific lag period. The difference is used to show the change in value over set period.

## Usage

```
cv.diff(x,n)
```

## **Arguments**

x : a numeric vector
n : number of lag periods

```
cv.diff(mtcars[,2],1)
```

cv.drawdown 7

cv.drawdown

Largest draw down of returns

# Description

Calculate largest draw down of a series of returns. This function calculates the maximum decrease in percentage over time, which can be used to test portfolio returns.

## Usage

```
cv.drawdown(x)
```

# Arguments

Χ

: a numeric vector of returns

## **Examples**

```
# rnorm() is used to simulate portfolio returns
returns <- rnorm(100)
cv.drawdown(returns)</pre>
```

cv.lag

Create a lag variable

## **Description**

Create a lag variable, with a choice of lag periods. The lag variable can be used to test lag effects between variables.

#### Usage

```
cv.lag(x,n)
```

## **Arguments**

x :a vector

n :number of lag periods

```
cv.lag(mtcars[,2],3)
data.frame(mtcars,cv.lag(mtcars[,3], 1))
```

8 cv.logs

cv.lead

Create a lead variable

# Description

Create a lead variable, with a choice of lead periods. The lead variable can be used to test lead effects between variables.

## Usage

```
cv.lead(x,n)
```

# Arguments

x :a vector

n :number of lead periods

## **Examples**

```
cv.lead(mtcars[,2],3)
data.frame(mtcars,cv.lead(mtcars[,3], 3))
```

cv.logs

Create logarithm with a random base

## Description

Create a new variable that is the logarithm of the selected variable with the base of a random number

## Usage

```
cv.logs(dataframe, var, n, range)
```

## **Arguments**

dataframe :a data frame

var : the variable selected

n :number of new variables created

range : the range of base

```
cv.logs(mtcars,"wt",5,c(1, 2))
```

cv.pctcng 9

cv.pctcng

Calculating rate of return of a vector

# Description

Calculating the percentage change of a time series vector for further analysis, including calculating beta of companies, plotting to see the trend of the stock for technical analysis.

## Usage

```
cv.pctcng(x,n)
```

## **Arguments**

x :a numeric vector
n : number of lag periods

# **Examples**

```
cv.pctcng(mtcars[,1],1)
```

cv.powers

Create nth power variable

## **Description**

Create a new variable that is the nth power of the selected variable

# Usage

```
cv.powers(dataframe, var, n, range)
```

# Arguments

dataframe :a data frame

var : the variable selected

n :number of new variables created

range :the range of power

```
cv.powers(mtcars,"wt",5,c(1, 2))
```

10 df.stack

df.sortcol

Sort a data frame by a column

#### **Description**

Sort a data frame by a column of choice. The column of choice is specified by the number of the column.

## Usage

```
df.sortcol(x,n,desc)
```

## **Arguments**

x :a data frame

n :number column to sort

desc : the order of sorting, default set to TRUE; for ascending order set to FALSE

#### **Examples**

```
df.sortcol(mtcars,2,desc = TRUE)
```

df.stack

Stack data frame by one classifier

#### **Description**

Stack data frame by one classifier. This function takes the first column as a ordering variable. Then it take the variables names and repeat as the second column. The last column will be data under each variable name. This function is created to enable easier investigation with apply functions.

## Usage

```
df.stack(df,name)
```

# Arguments

df : a data frame used to stack

name : new variable names of the data frame

```
df <- data.frame(matrix(nrow=100,ncol=100))
for(i in 1:100){
    df[,i] <- rep(runif(1,1,100),100)
}
dim(df)
hdf <- df.stack(df,c("date","tkr","price"))</pre>
```

ds.corm 11

ds.corm

Correlation matrix

# Description

Calculating the correlation matrix of a data frame and return in a data frame object

# Usage

```
ds.corm(x,n)
```

# Arguments

x :a data frame

n :number of decimal points

# **Examples**

```
ds.corm(mtcars,3)
```

ds.kurt

Calculating kurtosis for numeric data.

# Description

Kurtosis

# Usage

```
ds.kurt(x)
```

## **Arguments**

X

:a numeric variable

```
ds.kurt(mtcars[,2])
```

12 ds.skew

ds.mode

Calculating mode for numeric data

# Description

Calculating mode for numeric data.

# Usage

```
ds.mode(x)
```

# Arguments

Χ

:a numeric variable

# **Examples**

```
ds.mode(mtcars[,2])
```

ds.skew

Calculating skewness for numeric data

# Description

Calculating Pearson's skewness in three types: mode, median, and mean.

## Usage

```
ds.skew(x, type = 3)
```

## **Arguments**

Χ

:a numeric variable

type

:type = 1 for mode skewness; type = 2 for median skewness; type = 3 for mean

skewness

```
ds.skew(mtcars[,1])
```

ds.summ 13

ds.summ

Descriptive statistics of a data frame

## **Description**

Calculating the descriptive statistics of a data frame and exporting in a data frame. The report data frame contains: number of observations, maximum value, minimum value, mean, median, mode, variance, standard deviation, skewness and kurtosis.

# Usage

```
ds.summ(x,n)
```

## **Arguments**

x :a data frame

n :number of decimal points rounded

## **Examples**

```
ds.summ(mtcars,3)
```

pl.2ts

Time series plot for two variables

## **Description**

Plotting two time series in one plot, with title.

#### Usage

```
pl.2ts(ts1,ts2,title)
```

# Arguments

ts1 :time series variable one ts2 :time series variable two

title :title for the plot

```
DAX <- EuStockMarkets[,1]
FTSE <- EuStockMarkets[,4]
pl.2ts(DAX,FTSE, "Times Series Plot of DAX and FTSE")</pre>
```

pl.3smoothtxt

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Time series plot for two variables with ggplot2

#### Description

Plotting two time series in one plot, with title and label. If both variables are time series object, they will be merged by time. If both variables are not time series object, they will be merged by order. The first variable is set to be a solid line and the second variable is set to be a dashed line. If the variables are of different type a warning message will be given.

## Usage

```
pl.2tsgg(ts1,ts2,title,ylab)
```

#### **Arguments**

ts1 :a time series variable or a numeric variable
ts2 :a time series variable or a numeric variable

title :title for the plot ylab :y-axis label

#### **Examples**

```
DAX <- EuStockMarkets[,1]
FTSE <- EuStockMarkets[,4]
pl.2tsgg(DAX,FTSE, "Times Series Plot of DAX and FTSE", "Index")</pre>
```

pl.3smoothtxt

Scatter smooth plot with text overlay

#### **Description**

Generate a scatter plot with text overlay, with a smooth curve fitted by loess.

#### Usage

```
pl.3smoothtxt(x,y,txt,ce)
```

# Arguments

x : a numeric vector
y : a numeric vector
txt : a vector used as labels

ce : text size, which default is set as 0.5

pl.3smoothtxtgg

## **Examples**

```
pl.3smoothtxt(mtcars[,1], mtcars[,3], row.names(mtcars))
```

pl.3smoothtxtgg

Scatter smooth plot with text overlay using ggplot2

# Description

Generate a scatter plot with text overlay, with a smooth curve fitted by loess.

# Usage

```
pl.3smoothtxtgg(x,y,txt,size,title,xlab,ylab)
```

## **Arguments**

x :a numeric vector
y :a numeric vector
txt :a vector used as labels
size :text size, which default is set as 3
title :graph title
xlab :x-axis label
ylab :y-axis label

# **Examples**

```
pl.3smoothtxtgg(mtcars[,1], mtcars[,3], row.names(mtcars), 3, "MPG v. DISP", "mpg", "disp")
```

pl.3txt

Scatter plot with text overlay

# Description

Generate a scatter plot with text overlay. This plot is to better show the effect of the text variable in the domain of x and y variable.

## Usage

```
pl.3txt(x,y,txt,title)
```

pl.3txtgg

## **Arguments**

x :a numeric vector y :a numeric vector

txt :a vector used as labels

title :title of the graph

# **Examples**

```
pl.3txt(mtcars[,1], mtcars[,3], row.names(mtcars),"mpg v. cyl")
```

pl.3txtgg

Scatter plot with text overlay with ggplot2

## **Description**

Generate a scatter plot with text overlay with ggplot2. This plot is to better show the effect of the text variable in the domain of x and y variable.

## Usage

```
pl.3txtgg(x,y,txt,size,title,xlab,ylab)
```

# **Arguments**

x :a numeric vector y :a numeric vector

txt :a vector used as labels

size :text size, which default is set as 3

title :title of the graph

xlab :x-axis label ylab :y-axis label

```
pl.3txtgg(mtcars[,1], mtcars[,3], row.names(mtcars), 3,"mpg v. cyl", "mpg", "cyl")
```

pl.coplot 17

pl.coplot

Scatter plot of x and y divided by z

## **Description**

Generate 4 scatter plots of x and y divided by variable z, with a fitted line using a simple linear regression method.

## Usage

```
pl.coplot(x,y,z,varN)
```

## **Arguments**

x :x-axis value y :y-axis value

z :classification variable used to condition plots based on ascending values of z

varN :variable name of z

## **Examples**

```
pl.coplot(mtcars[,1], mtcars[,3], mtcars[,4], "hp")
```

pl.hist

Plot histograms for a data frame

## **Description**

Plotting histograms for a data frame, with titles and label numbers.

## Usage

```
pl.hist(x, l = 1)
```

## **Arguments**

x :a data frame

1 : the beginning label number in the title (default set to 1)

```
pl.hist(mtcars,1)
```

18 pl.hs

pl.histgg

Plot histograms for a data frame with ggplot2

#### Description

Plotting histograms for a data frame with 4 per page, with titles and label numbers automatically generated.

#### Usage

```
pl.histgg(x,1,bin)
```

# Arguments

x :a data frame

1 :the beginning label number in the title (default set to 1)

bin :bin width of histogram (default set to 30)

## **Examples**

```
pl.histgg(as.data.frame(EuStockMarkets),1)
```

pl.hs

Plot histograms and scatter plots for a data frame

#### **Description**

Plotting histograms or scatter plots of your choice for a data frame. Also the function will name the graphs and number them. The purpose of the function is to save time when plotting graphs for a regression analysis or other usage. The function can plot, name and number the graphs at one step.

## Usage

```
pl.hs(x,a,dependent,1)
```

## **Arguments**

x :a data frame

a : the type of graph you want; a = 1 for histograms; a = 2 for scatter plots; a = 0

for both

dependent : the dependent variable for scatterplots

1 : the beginning label number in the title (default set to 1)

```
pl.hs(mtcars,0,"mpg",1)
```

pl.hsd

pl.hsd

Plot histogram with density line for a data frame

# Description

Plotting histogram with density for a data frame, with titles and label numbers.

## Usage

```
pl.hsd(dataframe,1)
```

## **Arguments**

dataframe

:a data frame

1

: the beginning label number in the title (default set to 1)

## **Examples**

```
pl.hsd(mtcars,1)
```

pl.hsdgg

Plot histograms for a data frame with ggplot2

# Description

Plotting histograms for a data frame with 4 per page, with titles and label numbers automatically generated.

## Usage

```
pl.hsdgg(x,l,bin)
```

## **Arguments**

x :a data frame

1 :the beginning label number in the title (default set to 1)

bin :bin width of the graph

```
pl.hsdgg(as.data.frame(EuStockMarkets),1,100)
```

20 pl.s

pl.mv

Plot mean-variance simulation result

#### **Description**

This function is used to plot the result of portfolio simulation by pt.mv().

# Usage

```
pl.mv(port)
```

#### **Arguments**

port

:portfolio simulation result from pt.mv()

# **Examples**

```
set.seed(1)
rtn <- data.frame(runif(120,-1,1),runif(120,-1,1),runif(120,-1,1),runif(120,-1,1))
names(rtn) <- c("asset1","asset2","asset3","asset4")
portfolio <- pt.hismv(rtn,1000,0)
pl.mv(portfolio)</pre>
```

pl.s

Plot scatter plots for a data frame

## **Description**

Plotting scatter plots for a data frame, with titles and label numbers.

## Usage

```
pl.s(x,dependent,1)
```

# Arguments

x :a data frame, which includes the dependent variable

dependent : the dependent variable for scatter plot

1 : the beginning label number in the title (default set to 1)

```
pl.s(mtcars, "mpg", 1)
```

pl.sgg 21

pl.sgg

Plot scatter plots for a data frame using ggplot2

#### **Description**

Plotting scatter plots for a data frame using ggplot2, with titles and label numbers. The output will be 4 graphs per page.

#### Usage

```
pl.sgg(x, dependent, 1)
```

## **Arguments**

x :a data frame, which includes the dependent variable

dependent : the dependent variable for scatter plot

1 : the beginning label number in the title (default set to 1)

# **Examples**

```
pl.sgg(mtcars,"mpg",1)
```

pl.sm

Plot scatter smooth plots for a data frame

# Description

Plotting scatter smooth plots for a data frame, with titles and label numbers.

## Usage

```
pl.sm(x,dependent,1)
```

## **Arguments**

x :a data frame, which includes the dependent variable dependent :the dependent variable for scatter smooth plots

 $1 \hspace{1.5cm} : the \ beginning \ label \ number \ in \ the \ title \ (default \ set \ to \ 1)$ 

```
pl.sm(mtcars, "mpg", 1)
```

pl.ts

pl.smgg

Plot scatter plots with smooth line for a data frame using ggplot2

#### **Description**

Plotting scatter plots for a data frame using ggplot2, with titles and label numbers. A smooth line will be added using a chosen method. The output will be 4 graphs per page.

## Usage

```
pl.smgg(x,dependent,1,mtd)
```

## Arguments

x :a data frame, which includes the dependent variable

dependent : the dependent variable for scatter plot

1 :the beginning label number in the title (default set to 1)

mtd :sommthing method to use, accepts either a character vector or a function, e.g.

MASS::rlm, base::loess, mgcv::gam

#### **Examples**

```
pl.smgg(mtcars,"mpg",1,lm)
pl.smgg(mtcars,"mpg",1,loess)
```

pl.ts

Plot time series plots for a data frame

## Description

Plotting time series plots for a data frame, with titles and label numbers.

#### Usage

```
pl.ts(x, l = 1)
```

## **Arguments**

x :a data frame

1 : the beginning label number in the title (default set to 1)

```
pl.ts(mtcars,1)
```

pl.tsgg 23

pl.tsgg

Plot times series plot for a data frame with ggplot2

# Description

Plotting time series plot for a data frame with 4 per page, with titles and label numbers automatically generated.

## Usage

```
pl.tsgg(x,1)
```

## **Arguments**

x :a data frame

1 : the beginning label number in the title (default set to 1)

## **Examples**

```
pl.tsgg(as.data.frame(EuStockMarkets),1)
```

pl.tss

Time series plot with multiple variables

# Description

This function will return a time series plot with up to 6 variables, each with different line type.

## Usage

```
pl.tss(dataframe,ylb,title)
```

## **Arguments**

dataframe :a data frame
ylb :y-axis label
title :plot title

```
pl.tss(EuStockMarkets, "Price", "Daily Closing Prices of Major European Stock Indices")
```

24 pt.annexrtn

pt.alpha

Stock return alpha

## Description

Alpha is the intercept of a fitted line when dependent variable is the benchmark return and independent variable is a asset return of the same period. It is a measure of the active return on an investment. Alpha, along with beta, is one of the two key coefficients in the CAPM used modern portfolio theory.

## Usage

```
pt.alpha(ar,br)
```

## Arguments

ar :a vector of a risk asset return br :a vector of benchmark return

## **Examples**

```
brtn <- runif(100, -1, 1)
artn <- runif(100, -1, 1)
pt.alpha(artn,brtn)</pre>
```

pt.annexrtn

Annualized excess return

## **Description**

Annualized excess return is the difference between the annualized and cumulative return of the two series. Usually, one series are portfolio returns and the other is a benchmark returns.

## Usage

```
pt.annexrtn(ar,br)
```

#### **Arguments**

```
ar :a vector of a risk asset return
br :a vector of benchmark return
```

```
artn <- runif(100, -1, 1)
brtn <- runif(100, -1, 1)
pt.annexrtn(artn, brtn)</pre>
```

pt.annrtn 25

pt.annrtn

Annualized return

## **Description**

This function takes a series of annual returns and calculate the annualized return.

# Usage

```
pt.annrtn(r,n)
```

# Arguments

r :annual returns
n :number of years

# **Examples**

```
r <- runif(100,-1,1) # generate random number to simulate returns annualizedreturn <- pt.annrtn(r,100)
```

pt.annsd

Annualized standard deviation

## **Description**

The annualized standard deviation is the standard deviation multiplied by the square root of the number of periods in one year.

#### Usage

```
pt.annsd(r,n)
```

## **Arguments**

```
r :a vector of a risk asset return
n :number of periods in a year
```

```
rtn <- runif(30, -1, 1)
n <- 30
pt.annsd(rtn,n)</pre>
```

26 pt.bias

pt.beta

Stock return beta

# Description

Beta is the slope of a fitted line when dependent variable is the benchmark return and independent variable is an asset return of the same period. It is a measure the risk arising from exposure to general market movements.

## Usage

```
pt.beta(ar,br)
```

# Arguments

ar :a vector of a risk asset return br :a vector of benchmark return

### **Examples**

```
brtn <- runif(100, -1, 1)
artn <- runif(100, -1, 1)
pt.beta(artn, brtn)</pre>
```

pt.bias

Bias ratio

# Description

The bias ratio is an indicator used in finance analyze the returns of a portfolio, and in performing due diligence.

# Usage

```
pt.bias(r)
```

#### **Arguments**

r

:a vector of a risk asset return

pt.btavg 27

pt.btavg

Batting average

## **Description**

The batting average of the asset is the ratio between the number of periods where the asset outperforms a benchmark and the total number of periods.

# Usage

```
pt.btavg(ar,br)
```

## **Arguments**

ar :a vector of a risk asset return
br :a vector of a benchmark return

## **Examples**

```
artn <- runif(100,-1,1)
brtn <- runif(100,-1,1)
pt.btavg(artn,brtn)</pre>
```

pt.cmexrtn

Cumulative excess return

## Description

Cumulative return is the compounded return in a given period. The excess return is the difference between the cumulative return of a risky asset and the cumulative return of a benchmark.

## Usage

```
pt.cmexrtn(ar,br)
```

## **Arguments**

ar :a vector of risky asset returns
br :a vector of benchmark returns

```
brtn <- runif(12, -1, 1)
artn <- runif(12, -1, 1)
pt.cmexrtn(artn,brtn)</pre>
```

28 pt.dalpha

pt.cmrtn

Cumulative return

# Description

Cumulative return is the compounded return in a given period.

# Usage

```
pt.cmrtn(r)
```

#### **Arguments**

r

:a vector of periodic returns

# **Examples**

```
rt <- runif(12,-1,1) # generate random number to simulate returns pt.cmrtn(rt)
```

pt.dalpha

Dual-alpha

# Description

Dual-alpha method is to divide market alpha into downside beta and upside alpha. The principle behind is that upside and downside alphas are not the same.

## Usage

```
pt.dalpha(ar,mr,rf)
```

# Arguments

ar :a vector of a risk asset return
mr :a vector of market return

rf :risk free rate

```
artn <- runif(24,0,1) # generate random number to simulate returns mrtn <- runif(24,-1,1) pt.dalpha(artn,mrtn,0.024)
```

pt.dbeta 29

pt.dbeta Dual-beta

# Description

Dual-beta method is to divide market beta into downside beta and upside beta. The principle behind is that upside and downside betas are not the same.

# Usage

```
pt.dbeta(ar,mr,rf)
```

# Arguments

ar :a vector of a risk asset return
mr :a vector of market return
rf :risk free rate

## **Examples**

```
artn <- runif(24,0,1) # generate random number to simulate returns mrtn <- runif(24,-1,1) pt.dbeta(artn,mrtn,0.024)
```

pt.exploss

Expected loss

## **Description**

This function give the expected loss of given asset returns.

# Usage

```
pt.exploss(r,p)
```

## **Arguments**

r :a vector of periodic returns
p :target return

```
rt <- runif(12,-1,1) # generate random number to simulate returns
pt.exploss(rt,0)
pt.exploss(rt,1)</pre>
```

pt.info

## **Description**

This function will perform portfolio simulation with historical average returns and standard deviations. Mean-variance model, or modern portfolio theory, is a mathmatical framework for accessing a portfolio. It uses the variance of asset returns as a risk proxy. This function will return a number of simulated portfolio with different weights.

#### Usage

```
pt.hismv(r,n,mini)
```

#### **Arguments**

r :a data frame of asset returns n :number of portfolio simulated

mini :minimal weight; choose 0 if long only; choose 1 for possible short position

## **Examples**

```
set.seed(20)
rtn <- data.frame(runif(120,-1,1),runif(120,-1,1),runif(120,-1,1),runif(120,-1,1))
names(rtn) <- c("asset1","asset2","asset3","asset4")
portfolio <- pt.hismv(rtn,1000,0)
plot(portfolio[,6], portfolio[,5], xlab = "standart deviation", ylab = "expected return")</pre>
```

pt.info Information ratio

## **Description**

The information ratio of asset's returns versus benchmark returns, is the quotient of the annualized excess return and the annualized standard deviation of the excess return.

# Usage

```
pt.info(ar,br,n)
```

#### **Arguments**

ar :a vector of a risk asset return br :a vector of benchmark return

n :number of years

pt.jalpha 31

#### **Examples**

```
brtn <- runif(100, -1, 1)
artn <- runif(100, 0, 1)
pt.info(artn,brtn,100)</pre>
```

pt.jalpha

Jensen's alpha

# Description

Jensen's alpha is a financial statistic used to quantify the abnormal return of a security or portfolio over the theoretical expected return. Unlike, standard alpha, it uses theoretical performance return instead of a market return.

#### Usage

```
pt.jalpha(pr,mr,rf,beta)
```

## **Arguments**

```
pr :portfolio return
mr :market return
rf :risk free rate
beta :portfolio beta
```

## **Examples**

```
prtn <- runif(24, -1, 1)
mrtn <- runif(24, -1, 1)
rf <- 0.024
pt.jalpha(mean(prtn), mean(mrtn), rf, pt.beta(prtn,mrtn))</pre>
```

pt.m2

Modigliani risk-adjusted performance

## **Description**

Modigliani risk-adjusted performance is a financial measure of risk-adjusted returns of a portfolio. It measures the returns of the portfolio after adjusting it relative to some benchmark.

#### Usage

```
pt.m2(pr,br,rf)
```

32 pt.probloss

# Arguments

pr :portfolio return
br :benchmark return
rf :risk free rate

# **Examples**

```
prtn <- runif(12,-1,1)
brtn <- runif(12,-1,1)
rf <- 0.024
pt.m2(prtn,brtn,rf)</pre>
```

pt.probloss

Probability of loss

# Description

This function give the probability of loss of given asset returns.

# Usage

```
pt.probloss(r,p)
```

# **Arguments**

r :a vector of periodic returns

p :target return

```
rt <- runif(12,-1,1) # generate random number to simulate returns pt.probloss(rt,0) pt.probloss(rt,0.05)
```

pt.roy 33

pt.roy

Roy's safety-first criterion

## **Description**

Roy's safety-first criterion is a risk management technique that allows to choose a portfolio based on the criterion that the probability of the portfolio's return falling below a minimum desired threshold is minimized.

#### Usage

```
pt.roy(r,mar)
```

#### **Arguments**

r :a vector of a risk asset return
mar :minimum acceptable return

## **Examples**

```
r \leftarrow runif(100,0,1) # generate random number to simulate returns pt.roy(r,0.024)
```

pt.sdexrtn

Standard deviation of excess return

## **Description**

The standard deviation of excess return is simply the standard deviation of the asset return over the benchmark return.

## Usage

```
pt.sdexrtn(ar,br)
```

#### **Arguments**

ar :a vector of a risk asset return
br :a vector of benchmark return

```
artn <- runif(12, -1, 1)
brtn <- runif(12, -1, 1)
pt.sdexrtn(artn,brtn)</pre>
```

pt.sharp

pt.semivar

Semivariance of loss

## **Description**

This function give the semivariance of a losing scenario.

#### Usage

```
pt.semivar(r,p)
```

## Arguments

r :a vector of periodic returns

p :target return

#### **Examples**

```
rt <- runif(12,-1,1) # generate random number to simulate returns pt.semivar(rt,0) pt.semivar(rt,0.03)
```

pt.sharp

Sharp ratio

# Description

The Sharpe Ratio of an asset return is the quotient of the annualized excess return of the asset minus the annualized risk-free rate over the annualized standard deviation of the asset return.

#### Usage

```
pt.sharp(r,n,m,rf)
```

#### **Arguments**

r :a vector of asset returns

n :number of years

m :number of periods in a year; m = 12 if r is monthly returns

rf :annulized risk-free rate

```
set.seed(20)  \verb|rtn| <- \verb|runif| (12,-0.5,1) | # | generate | random | number | to | simulate | monthly | returns | rfr| <- 0.024 | # | set | risk | free | rate | at | 2.4% | annual | pt.sharp(rtn,1,12,rfr) | # | the | return | is | for | one | year | | |
```

pt.sortino 35

pt.sortino

Sortino ratio

# Description

The Sortino ratio is an analog to the sharp ratio, with standard deviation replaced by the downside deviation.

#### Usage

```
pt.sortino(r,p,n,rf)
```

## **Arguments**

r :a vector of a risk asset return

p :target return, aka minimum acceptable return(MAR)

n :number of years of asset return, used to calculate annualized return

rf :risk free rate

## **Examples**

```
rtn <- runif(12, -1, 1)
pt.sortino(rtn,0.3,1,0.024)
```

pt.te

Tracking error

## **Description**

Tracking error, in finance, is a measure of risk in a portfolio that is due to active management decisions made by the manager. It indicates how closely the portfolio follows the benchmark of choosing.

#### Usage

```
pt.te(pr,br)
```

## Arguments

pr :portfolio return br :benchmark return

```
prtn <- runif(12,-1,1)
brtn <- runif(12,-1,1)
pt.te(prtn,brtn)</pre>
```

pt.udrtn

pt.treynor

Treynor ratio

## **Description**

The Treynor ratio is an analog to the sharp ratio, with standard deviation replaced by the asset beta to benchmark.

#### Usage

```
pt.treynor(ar,br,n,rf)
```

## **Arguments**

ar :a vector of a risk asset return
br :a vector of benchmark return

n :number of years of asset return, used to calculate annualized return

rf :risk free rate

## **Examples**

```
rtn <- runif(24, -1, 1)
brtn <- runif(24,-1,1)
pt.treynor(rtn,brtn,2,0.024)</pre>
```

pt.udrtn

Average up and down returns

## **Description**

This function calculates the average up and down returns from a series of returns.

## Usage

```
pt.udrtn(r)
```

#### **Arguments**

r

:a vector of periodic returns

```
r <- runif(100,-1,1) # generate random number to simulate returns pt.udrtn(r)
```

pt.updwcap 37

# Description

The up and down capture is a measure of how an asset was able to improve on benchmark returns or how it underperforms over the benchmark.

## Usage

```
pt.updwcap(ar,br,n)
```

#### **Arguments**

ar :a vector of a risk asset return
br :a vector of benchmark return

n :number of years of asset return, used to calculate annualized return

## **Examples**

```
artn <- runif(12, -1, 1)
brtn <- runif(12,-1,1)
pt.updwcap(artn,brtn,1)</pre>
```

reg.adj.r.squared

Adjusted R-squared for lm.fit

## **Description**

Calculate Adjusted R-squared for the outcome of lm.fit. This function is built for reg.linreg() for higher efficiency only. It can't be used for calculating Adjusted R-squared in general operation.

# Usage

```
reg.adj.r.squared(r,n,p)
```

# Arguments

r :R-squared for regression

n :number of observations aka. sample size

p :number of explanatory variables in the model

38 reg.bic

#### **Examples**

```
X <- as.matrix(cbind(1,EuStockMarkets[,1:2])) # create the design matrix
Y <- as.data.frame(EuStockMarkets)$FTSE
fit <- lm.fit(x = X, y = Y)
SSR <- sum((fit$fitted.values - mean(Y))^2)
SSTO <- sum((Y - mean(Y))^2)
r <- reg.r.squared(SSR,SSTO)
n <- dim(X)[1]; p <- dim(X)[2]
reg.adj.r.squared(r,n,p)</pre>
```

reg.aic

AIC for lm.fit

## **Description**

Calculate AIC for the outcome of AIC. This function is built for reg.linreg for higher efficiency only. It can't be used for calculating AIC in general operation.

#### Usage

```
reg.aic(fit,w)
```

#### **Arguments**

fit :the outcome of lm.fit

w :wright

## **Examples**

```
X <- as.matrix(cbind(1,EuStockMarkets[,1:2])) # create the design matrix
Y <- as.data.frame(EuStockMarkets)$FTSE
fit <- lm.fit(x = X, y = Y)
w <- rep(1,length(Y))
reg.aic(fit,w)</pre>
```

reg.bic

BIC for lm.fit

#### **Description**

Calculate BIC for the outcome of lm.fit This function is built for reg.linreg() for higher efficiency only. It can't be used for calculating BIC in general operation.

#### Usage

```
reg.bic(fit,w)
```

reg.dof

## **Arguments**

fit :the outcome of lm.fit

w :wright

#### **Examples**

```
X <- as.matrix(cbind(1,EuStockMarkets[,1:2])) # create the design matrix
Y <- as.data.frame(EuStockMarkets)$FTSE
fit <- lm.fit(x = X, y = Y)
w <- rep(1,length(Y))
reg.bic(fit,w)</pre>
```

reg.dof

Degree of freedom for lim.fit

# Description

Calculate degree of freedom for the outcome of lm.fit(). This function is built for reg.linreg for higher efficiency only. It can't be used for calculating degree of freedom in general operation.

## Usage

```
reg.dof(fit)
```

## **Arguments**

fit

:outcome of lm.f

## **Examples**

```
X <- as.matrix(cbind(1,EuStockMarkets[,1:2])) # create the design matrix
Y <- as.data.frame(EuStockMarkets)$FTSE
fit <- lm.fit(x = X, y = Y)
reg.dof(fit)</pre>
```

reg.dw

**Durbin-Watson Test** 

#### **Description**

Performs the Durbin-Watson Test for a regression model

## Usage

```
reg.dw(fit)
```

40 reg.model

## Arguments

fit :a lm object

#### **Examples**

```
fit <- lm(mpg~wt, mtcars, na.action = na.omit)
reg.dw(fit)</pre>
```

reg.linreg

Linear regression processor

# **Description**

This function will take a data frame and the dependent variable and fit all possible combinations of models. The result will be a data frame of models and test statistics for all the models possible. The test statistics are current set and contain all the following: R-squared, Adjusted R-squared, Degree of freedom, Residual standard error, AIC, BIC, Durbin-Watson statistic.

## Usage

```
reg.linreg(dataframe, dependent)
```

## **Arguments**

dataframe :a data

:a data frame, which includes the dependent variable

dependent :dependent variable

# Examples

```
reg.linreg(mtcars,"mpg")
```

reg.model

Linear model generator

## Description

This function will take a data frame and generate all the combinations of linear model

# Usage

```
reg.model(dataframe, dependent)
```

## **Arguments**

dataframe :a data frame

dependent : dependent variable

reg.r.squared 41

#### **Examples**

```
reg.model(mtcars, "mpg")
```

reg.r.squared

R-squared for lm.fit

# Description

Calculate R-squared for the outcome of lm.fit(). This function is built for reg.linreg for higher efficiency only. It can't be used for calculating R-squared in general operation.

## Usage

```
reg.r.squared(SSR,SSTO)
```

#### Arguments

SSR :regression sum of squares or explained of squares

SSTO :total sum of squares

#### **Examples**

```
X <- as.matrix(cbind(1,EuStockMarkets[,1:2])) # create the design matrix
Y <- as.data.frame(EuStockMarkets)$FTSE
fit <- lm.fit(x = X, y = Y)
me <- mean(Y)
SSR <- sum((fit$fitted.values - me)^2)
SSTO <- sum((Y - me)^2)
reg.r.squared(SSR,SSTO)</pre>
```

reg.std.err

Standard error for lim.fit

#### **Description**

Calculate standard error for the outcome of lm.fit(). This function is built for reg.linreg for higher efficiency only. It can't be used for calculating standard error in general operation.

#### Usage

```
reg.std.err(SSE,dof)
```

## Arguments

SSE :error sum of squared aka. residual sum of squared

dof :degree of freedom

42 tr.logtb

#### **Examples**

```
X <- as.matrix(cbind(1,EuStockMarkets[,1:2])) # create the design matrix
Y <- as.data.frame(EuStockMarkets)$FTSE
fit <- lm.fit(x = X, y = Y)
SSE <- sum((Y - fit$fitted.values)^2)
dof <- reg.dof(fit)
reg.std.err(SSE,dof)</pre>
```

tr.log

Sigmoid function

## **Description**

Generate sigmoid curve series, which is a specific case of logistic function, with a control of top and bottom acceleration.

## Usage

```
tr.log(x,top,a,b)
```

## **Arguments**

x :a numeric vector

top :a numeric value as vertical scaler

a :a number to control top acceleration of the curve
b :a number to control bottom acceleration of the curve

#### **Examples**

```
sigc <- round(tr.log(seq(-3, 3, 0.1), 1, -3, 3), 3)
ts.plot(sigc)</pre>
```

tr.logtb

Logistic function

# Description

Generate logistic series, with set top and bottom value and acceleration.

## Usage

```
tr.logtb(x,top,bot,a,b)
```

tr.nd 43

# Arguments

x :a vector

top :higher level y asymptote
bot :lower level y asymptote

a :a number to control top acceleration of the curve

b : a number to control bottom acceleration of the curve

# Examples

```
tr.logtb(seq(-3, 3, 0.1), 1, 0.4, -3, 3)
```

tr.nd

Normal density function

# Description

Calculate normal density function value at x with a mean of mu and standard deviation of sig.

## Usage

```
tr.nd(x,mu,sig)
```

# Arguments

x :x value

mu :mean value

sig :standard deviation

```
tr.nd(seq(-3, 3, 0.1), 0, 1)
```

44 xd.fred

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Unit normal loss integral

# Description

Compute the value of the unit normal loss integral, with discontinuity and dispersion

#### Usage

```
tr.unli(x,disc,disp)
```

## **Arguments**

```
x :a vector
disc :discontinuity
```

disp :dispersion

# **Examples**

```
tr.unli(-3:10, 1, 3)
```

xd.fred

Download data from Federal Reserve Bank of St. Louis

# Description

This function returns a data from the Federal Reserve Bank of St. Louis database. It can take one ticker or a string of tickers, which will output a merged data frame with all observations.

#### Usage

```
xd.fred(tkr, start_date, end_date)
```

# Arguments

tkr :one data ticker or a string of tickers used by the database

start\_date :starting date of the data(default is set as 1900-01-01)
end\_date :ending date of the data(default is set as 2018-01-01)

xd.fred.tickers 45

## **Examples**

```
cpi <- xd.fred("CPIAUCSL") # CPI data
head(cpi)
tail(cpi)

#Frequently used tickers:
#CPIAUCSL: Consumer Price Index for All Urban Consumers: All Items
#A191RL1Q225SBEA: Real Gross Domestic Product
#DGS10: 10-Year Treasury Constant Maturity Rate
#UNRATE: Civilian Unemployment Rate</pre>
```

xd.fred.tickers

Federal Reserve Bank of St. Louis Economic Data Tickers

# Description

This function returns a data contains information of data name, type and tickers

# Usage

```
xd.fred.tickers()
```

```
xd.fred.tickers()
```

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