# Package 'ymse'

# November 6, 2018

Title Various more or less useful functions and methods

**Description** What the package does (one paragraph).

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**Depends** R (>= 3.5.0), forecast

<b>Imports</b>	forecast
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addrows

Add rows to a data.frame

## **Description**

An "rbind for data.frames", sort of.

#### Usage

```
addrows(dtf, nrw, top = FALSE)
```

# **Arguments**

dtf data.frame; original data.frame

nrw data.frame; the new row(s) to be added

top logical; should the new rows be added to the top or the bottom (default)?

#### **Details**

Can only bind two objects at a time, but will bind data.frames with non-matching column names and -classes. In such cases the original data.frame will serve as template.

```
dtf <- data.frame(A=letters[1:5],</pre>
                   B=1:5,
                   C=as.factor(5:1),
                   D=as.Date(0:4, origin="2000-01-01"),
                   stringsAsFactors=FALSE)
nrw <- data.frame(A=letters[1:5],</pre>
                   B=4:8,
                   C=5:1,
                   D=as.Date(5:1, origin="1990-01-01"),
                   stringsAsFactors=FALSE)
str(dtf)
dtf.a <- addrows(dtf, nrw, top=FALSE)</pre>
str(dtf.a)
# adding a single row with little concern for data types and column names
b <- type.convert(beaver1[80:90,])</pre>
b$activ <- as.logical(b$activ)</pre>
addrows(b, data.frame(350, 1200, 37.02, 1))
```

arfilter 3

arfilter AR filter

# Description

Filter a time series using AR coefficients

## Usage

```
arfilter(x, mod, x.mean = mod$x.mean, init = "focb")
```

## **Arguments**

x a time seriesmod an AR model

x.mean the mean used. By default the mean of the original model. Set to zero for no

demeaning

#### See Also

armodel

## **Examples**

```
set.seed(1)
arap <- ar(AirPassengers)
spec.ar(arap)
spec.pgram(arfilter(rnorm(10000), arap), span=21, na.action=na.omit)
arm <- armodel(c(1.3, -0.4))
spec.ar(arm)
plot(x <- rnorm(200), type="1")
lines(scale(arfilter(x, arm), center=FALSE), col="red", lwd=2)</pre>
```

arfit

AR model fit

## **Description**

Fit a specified AR model to a univariate time series

# Usage

```
arfit(x, mod, x.mean = mod$x.mean)
```

#### **Arguments**

x a time seriesmod an AR model

x.mean the mean used. By default the mean of the original model. Set to zero for no

demeaning

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#### See Also

```
armodel for examples
```

## **Examples**

```
 set.seed(1) \\ x \leftarrow runif(50) + sin(1:50/10) \\ plot(x); lines(arfilter(x, armodel(c(1.5, -0.5, 0.5)), x.mean=mean(x)))
```

arimpulse

Impulse response of an AR model

# Description

Get and plot the impulse response of an AR model

## Usage

```
arimpulse(mod, pulse = 1, n.ahead = 20, plot = TRUE, ...)
```

#### **Arguments**

mod an AR model

pulse numeric vector; the initial pulse. Magnitude is added to the model mean

n.ahead the length of the computed response

plot logical; sgould the result be plotted?

... further arguments to plot

## See Also

```
armodel for examples
```

armodel

Create an AR model object

# Description

Specify the characteristics of an AR model

# Usage

```
armodel(coefs, mean = 0, intercept = 0, var.pred = 1, frequency = 1,
    x.name = "Synthetic AR model")
```

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#### **Arguments**

coefs a vector of model coefficients

mean the mean of the process

intercept the intercept in the model

var.pred the portion of the variance not explained by this model

frequency the sampling frequency of the process

x.name name of the series

#### See Also

arimpulse

```
# short decay
ar.mod <- armodel(c(0.5))
arimpulse(ar.mod, pulse=1)
# long decay
ar.mod <- armodel(c(0.8))
arimpulse(ar.mod, pulse=1)
# negative second coefficient reduce damping, signal returns to normal
more quickly
ar.mod <- armodel(c(0.8, -0.1))
arimpulse(ar.mod, pulse=1)
# second coefficient reduce damping too much, overdamping, oscillations
ar.mod <- armodel(c(0.8, -0.5))
arimp <- arimpulse(ar.mod, pulse=1, n.ahead=40)$pred</pre>
polyroot(c(1, -ar.mod$ar)) # complex conjugate roots
acf(arimp) # period ~= 6?
phi1 <- ar.mod$ar[1]</pre>
phi2 <- ar.mod$ar[2]</pre>
f <- (1/(2*pi)) * acos((phi1*(phi2-1))/(4*phi2))
1/f # period = 6.78
sp <- spec.ar(ar.mod, plot=FALSE)</pre>
1/sp$freq[which.max(sp$spec)] # period = 6.79
# decaying oscillations
ar.mod1 <- armodel(c(0.8, -0.6, -0.5, 0.2, -0.2))
arimpulse(ar.mod1, n.ahead=100)
Mod(1/polyroot(c(1, -ar.mod1\$ar))) # barely inside the unit circle
# growing oscillations
ar.mod2 <- armodel(c(0.8, -0.7, -0.5, 0.2, -0.2))
arimpulse(ar.mod2, n.ahead=100)
Mod(1/polyroot(c(1, -ar.mod2\$ar))) # barely outside the unit circle
ar.mod3 <- armodel(c(1.8, -1.1, 0.2, -0.2, 0.2))
arimpulse(ar.mod3, n.ahead=100)
spec.ar(ar.mod3)
resid(arfit(rnorm(10), armodel(c(0.5, -0.1), frequency=2)))
```

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as.array.list

Coerce a list to an array

#### **Description**

Coerce a list consisting of data.frames or matrices of equal size to a 3d array

#### Usage

```
## S3 method for class 'list'
as.array(11)
```

## **Arguments**

11

a list of equal sized data.frames or matrices

## Value

A list of length l with elements of m rows and n columns will result in an  $m \times n \times l$  array.

## **Examples**

```
df1 <- data.frame(x=c(1, 2, 3), y=c(2, 3, 4), z=c(3, 4, 5))
df2 <- data.frame(x=c(4, 2, 3), y=c(2, 5, 4), z=c(3, 4, 6))
df3 <- data.frame(x=c(1, 4, 2), y=c(3, 3, 8), z=c(4, 3, 5))

1 <- list(df1, df2, df3)
as.array(1)
as.array(speedskate)</pre>
```

bartlett

Maurice Stevenson Bartlett's car data

## **Description**

This is an example data set Bartlett used for a lecture course on stochastic processes, Statistics Department, University College, London. The data represents the times, in seconds, when cars passed an observation point by a road.

Bartlett attributes the data to a Dr A. J. Miller who supplied them as a class example. According to Adery C. A. Hope the data was recorded on a rural Swedish road.

#### Usage

bartlett

# Format

A numeric vector representing time points in seconds

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#### M. S. Bartlett's notes

Analyse the above data with a view to examining:

- i whether the times of passing constitute a Poisson process;
- ii if not, whether some form of "bunching" or "clustering" seems to be present.

Possible analyses include:

- a testing the homogeneity of the consecutive random time-intervals, by means of a partitioning of the degrees of freedom for the total (approximate)  $\chi^2$ ;
- **b** testing the homogeneity of counts in consecutive fixed time-intervals, choosing an appropriate interval, and partitioning the degrees of freedom corresponding to the total dispersion by means of an analysis of variance;
- ${f c}$  testing the correlation between the consecutive random time-intervals;
- **d** examining the overall distribution of counts in fixed time-intervals;
- e examining the overall distribution of the consecutive random time-intervals

You should undertake at least sufficient of these to answer the questions asked.

#### Source

```
The Spectral Analysis of Point Processes (p. 280), M. S. Bartlett, 1963
```

Also mentioned in:

Statistical Estimation of Density Functions (p. 252), M. S. Bartlett, 1963

A Simplified Monte Carlo Significance Test Procedure (p. 583), Adery C. A. Hope, 1968

```
cpgram(diff(bartlett))
bartlett2 <- bartlett - bartlett[1]

x <- rep(0, tail(bartlett2, 1)*10)
x[bartlett2*10] <- 1

par(mfrow=c(2, 1), mar=c(2, 3, 1, 1))
plot(x, type="1", ann=FALSE)
lines(cumsum(x)/sum(x), col="red", lwd=2)

sp <- spectrum(x, main="", xlim=c(0, 0.1), ylim=c(1e-3, 0.04))
spec <- predict(loess(sp$spec[1:3000] ~ sp$freq[1:3000], span=0.15), se=TRUE)
lines(sp$freq[1:3000], spec$fit, col="red", lwd=2)
lines(sp$freq[1:3000], spec$fit - qt((0.99 + 1)/2, spec$df)*spec$se, lty=1, col="lightblue")
lines(sp$freq[1:3000], spec$fit + qt((0.99 + 1)/2, spec$df)*spec$se, lty=1, col="lightblue")</pre>
```

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binsearch

Binary search

## **Description**

Find the position of a given value in a sorted array

#### Usage

```
binsearch(val, arr, L = 1L, H = length(arr))
binclosest(val, arr, L = 1L, H = length(arr))
```

#### **Arguments**

val the value to search for

arr a sorted array to make the search in

L a lower bound
H an upper bound

#### **Details**

While both val and arr can be either integer or double, the algorithm is limited by integer storage in how long the array can be. L and H can be used to limit the range of indices to be search within. binsearch will return either the index of the exact match, or the index just below if no exact match is found. This means that if val is less than the lowest value in arr (and L=1), a 0 will be returned, which can lead to issues as such an index does not exist in R. An array indexed by 0 will return a zero length object. binclosest will return the index of the closest match, and therefore a 1 in the situation where binsearch returns a 0. If there is a tie the lower index will be returned. In either case, if there are duplicate matches, the lower index will be returned.

## Value

A single integer representing an index on the input array.

```
binsearch(15, (1:9)*3.333)
binsearch(2, (1:9)*3.333)
binclosest(2, (1:9)*3.333)
binsearch(18, seq_len(2e9))
binsearch(18, seq_len(3e9), H=2e9)
binsearch(2000, seq_len(3e7)*100 + 10.71)
set.seed(1)
x <- sort(sample(1:300, 30))
r <- sort(sample(1:300, 30))
plot(sapply(r, binsearch, x), type="1")
lines(sapply(r, binclosest, x), col="red")</pre>
```

caleidoscope 9

```
x <- c(1, 2, 3, 5, 8, 9)
binclosest(6, x)
binclosest(7, x)
binclosest(5, x)</pre>
```

caleidoscope

Caleidoscopic effect on a matrix

# Description

Flip a matrix vertically and horizontally before recombining into a new large matrix

## Usage

```
caleidoscope(m, odd = TRUE)
```

# **Arguments**

m a matrix

odd logical; should the resulting matrix have odd dimensions?

#### **Details**

Three copies of m will be made. One flipped horizontally, one flipped vertically, and one flipped both horizontally and vertically. Then they are recombined with the original matrix in the upper right corner, and the flipped copies in the upper left, lower righ and lower left corners, respectively.

# Value

A matrix of either  $2 \times$  or  $2 \times -1$  the number of rows and columns of the input matrix.

```
caleidoscope(matrix(1:4, 2), odd=FALSE)
image(caleidoscope(1:9 %0% 1:9))
image(caleidoscope(matrix(runif(180*200)^2, 180)), col=rainbow(256, start=0.58))
```

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central.tendency

Central tendency measures

## **Description**

Central tendency measures

#### Usage

```
pseudomedian(x)
cmode(x, ...)
```

# See Also

means

# **Examples**

```
xx < -c(1, 3, 4, 5, 7, 8, 9, 9, 7, 5, 4, 5, 3, 8) median(xx)#' pseudomedian(xx)
```

# Description

Compare numeric values, returning an inbetween value for ties

# Usage

```
x %tgt% y
tgt(x, y, bias = 0.5)
x %tlt% y
tlt(x, y, bias = 0.5)
```

# Arguments

x, y numeric values to be compared
bias what bias should be given to ties? 0.5, the default, is considered neutral as it's halfway between 1 and 0 (true and false).

# See Also

```
Comparison, tied_triple_test
```

dput2

#### **Examples**

```
1:5 %tlt% 3
1:5 %tgt% 3

c(1, 4, 3, 1) %tlt% c(1, 3, 3, 2)
c(1, 4, 3, 1) %tgt% c(1, 3, 3, 2)
```

dput2

Write an Object to console

#### **Description**

Writes an ASCII text representation of an R object to the console for easy copy/paste sharing

## Usage

```
dput2(x, width = 65, assign = c("front", "end", "none"),
  breakAtParen = FALSE)
```

## **Arguments**

x an object

width integer; column width

assign character; should assignment be included?

breakAtParen logical; should lines break at parenthesis begins (default FALSE)

#### **Details**

This is similar to the way dput is used to print ASCII representations of objects to the console. The differences are that dput2 lets you specify the width of the resulting column, and assignment of the object to the name used in the call will by default be included. Line breaks are by default only done on whitespace, but can be set to happen at parenthesis begins as well. This should not break code and can make for a more compact representation, but it can also make the code harder to read.

#### See Also

```
dput, deparse
```

```
xmpl <- faithful[sort(sample(1:nrow(faithful), 50)), ]

dput(xmpl)
cat(deparse(xmpl, width.cutoff=65), sep='\n')

dput2(xmpl, 65)
dput2(xmpl, 65, assign="end")
dput2(xmpl, 80, assign="none")
dput2(xmpl[1:10,], 10, "none")

# no line breaks on whitespaces or parens within character strings</pre>
```

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dusd

Discrete (Uniform) Sum Distributions

## **Description**

Generate distributions of the sum of discrete (uniform) random variables. Two different approaches.

## Usage

```
dusd1(xr = 1:6, n = 2)

dusd2(xi = rep(1, 6), n = 2, round, zero.index = FALSE, limit = 1e-13)
```

# **Arguments**

xr	numeric vector; a vector of equiprobable values
n	integer; the number of distributions to be summed
xi	numeric vector; a vector of probabilities, with indices representing values
round	integer; number of digits to round to after each convolution
zero.index	logical; should the index of xi start at zero?
limit	numeric; values (frequencies or counts) less than this will be omitted.

#### **Details**

dusd1 works by recursively taking the outer sum of xr, while dusd2 recursively convolves xi. Although convolution is more efficient, it can introduce small errors, and with repeated convolutions those errors can compound. By rounding to a slightly lower precision after each convolution the generation of spurious singletons and general imprecicions can be mitigated.

## Value

dusd1 returns an array of size  $length(xr)^n$  representing every possible outcome. dusd2 returns a probability mass function in the form of a table.

```
# five coin flips
plot(table(dusd1(0:1, 5)))
plot(dusd2(c(1, 1), 5, zero.index=TRUE))
plot(dbinom(0:5, 5, 0.5), type="h", lwd=2)
# ten flips with a loaded coin
```

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```
plot(table(dusd1(c(1, 1, 2), 10)))
plot(dusd2(c(2, 1), 10))
plot(dbinom(0:10, 10, 1/3), type="h", lwd=2)
# sample from a multi-roll d4 distribution
sample(dusd1(1:4, 5), 20, replace=TRUE)
plot(ecdf(dusd1(1:4, 5)))
tt <- dusd2(xi=rep(1, 4), n=3)
plot(tt)
tt <- tt/sum(tt)</pre>
rr <- replicate(50000, sample(names(tt), prob=tt))</pre>
barplot(apply(rr, 1, table), beside=TRUE)
# distribution of the sum of three d6 rolls
plot(table(dusd1(xr=1:6, 3)))
plot(dusd2(xi=rep(1, 6), n=3))
# D6 die with faces 2, 3, 5, 7, 11, 13 (prime numbers)
plot(table(dusd1(xr=c(2, 3, 5, 7, 11, 13), 3)))
# Loaded die
p \leftarrow c(0.5, 1, 1, 1, 1, 1.5); sum(p)
plot(dusd2(xi=p*3, n=2))
# A loaded die with prime number faces
s <- vector(length=13)</pre>
s[c(2, 3, 5, 7, 11, 13)] \leftarrow c(0.5, 1, 1, 1, 1, 1.5)
plot(dusd2(xi=s, n=3))
# tricky to do with dusd2
plot(table(dusd1(xr=c(0.1105, 2, exp(1)), 10)))
# Demonstrating CLT
# dusd1 struggles with many iterations
# remember it returns an array of size length(xr)^n
plot(table(dusd1(xr=c(1, 2, 9), 12)))
s <- vector(length=9)</pre>
s[c(1, 2, 9)] <- 1
plot(dusd2(xi=s, 12, round=9)) # much quicker
plot(dusd2(xi=s/sum(s), 12)) # for frequencies instead of counts
# Impossible with dusd1
clt <- dusd2(xi=s, 15, round=9)</pre>
plot(clt, lwd=0.5, col="#00000088")
# small floating-point errors from convolution.
tail(dusd2(xi=s, 15))
# dusd2 isn't always quicker
plot(table(dusd1(xr=c(1, 220, 3779), 12)), lwd=1)
s2 <- vector(length=3779)</pre>
s2[c(1, 220, 3779)] <- 1
plot(dusd2(xi=s2, 12, round=8), lwd=1)
```

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```
# making sure the length of xi is highly composite (or more precicely 'smooth')
# improves speed
# 3779 is prime, 3780 == 2*2*3*3*3*5*7
s3 <- vector(length=3780)
s3[c(1, 220, 3779)] <- 1
plot(dusd2(xi=s3, 12, round=9), lwd=1)</pre>
```

entropy

Information entropy

## **Description**

Computes the information entropy (also called Shannon entropy) of a set of discrete values, or a tabulated such set.

## Usage

```
entropy(...)
## S3 method for class 'table'
entropy(x, base = 2)
## S3 method for class 'data.frame'
entropy(x, base = 2)
## S3 method for class 'matrix'
entropy(x, base = 2)
## Default S3 method:
entropy(x, base = 2)
```

## **Arguments**

Χ

a vector, table, data.frame or matrix. In the case of table, data.frame and matrix each row is treated as a separate set of counts or proportions, with columns representing species, types, categories etc.

base

the log base to be used.

```
entropy(c(5, 5, 4, 4, 2, 3, 5)) # default is unit bits
entropy(c(5, 5, 4, 4, 2, 3, 5), base=exp(1)) # unit nats
entropy(rep(1:4, 1:4), 4)
entropy(rep(1:4, 1), 4)
entropy(as.factor(c(1, 1, 2, 3, 4, 4)))
entropy(as.character(c(1, 1, 2, 3, 4, 4)))
mtctab <- table(mtcars$cyl, mtcars$carb)
entropy(mtctab, 6)</pre>
```

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fitrange

Fit to a range

# Description

Linearly shift and scale a numeric vector so that it fits to a given range.

# Usage

```
fitrange(W, lower = -1, upper = 1)
```

## **Arguments**

W a numeric vector

lower the lower bound of the new vector upper the upper bound of the new vector

# See Also

norma

## **Examples**

```
range(fitrange(runif(9, -2, 0.1), 0, 1))
fitrange(c(2, 3, 5, 7, 4), 1, 0)
# same, but without warning
1 - fitrange(c(2, 3, 5, 7, 4), 0, 1)
```

forecast.test

Compare forecast accuracies

## **Description**

Test the efficacy of time series models by comparing forecast with actual data

# Usage

```
## S3 method for class 'test'
forecast(m, y = NULL, holdout = NULL)
```

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#### **Arguments**

```
    m a list of models to compare
    y a monovariate time series; the data to train and test the models on holdout
    single integer; the last n points will be forecasted
```

## **Examples**

```
library(forecast)
extr <- aggregate(sunspot.month, nfrequency=2, mean)</pre>
extr <- ts(extr, f=21)
extr <- head(extr, 342)
mod1 <- StructTS(extr)</pre>
mod2 <- ar(extr)</pre>
mod3 <- nnetar(extr)</pre>
mod4 <- arfima(extr)</pre>
mod5 <- Arima(extr, order=c(3, 0, 1))</pre>
mod6 \leftarrow Arima(extr, order=c(2, 0, 2), seasonal=c(2, 1, 0))
mod.1 \leftarrow list(mod1, mod2, mod3, mod4, mod5, mod6)
1 <- forecast.test(mod.1, extr, 21)</pre>
diffs <- sapply(1, function(y) y[["fcast"]] - y[["test"]])</pre>
matplot(diffs, type="l",
  col=c("red", "lightgreen", "blue", "orange", "pink", "cyan"), lty=1)
par(mfrow=c(3,\ 2),\ mar=c(3,\ 3,\ 2,\ 1),\ mgp=c(2,\ 0.6,\ 0),\ oma=c(0,\ 0,\ 0,\ 0))
invisible(lapply(l, function(x) {
  plot(x$fcast.obj, shaded=FALSE, PI=FALSE, include=66, type="1", cex.main=0.9, xpd=NA)
  lines(x$test, col="#00FF4488")
  }
))
summary(1)
head(forecasts(1))
```

incdiff

Increase difference

#### **Description**

Rearrange a sorted numeric sequence so that the difference between subsequent elements is increased

#### Usage

```
incdiff(x, step = 2)
```

## Arguments

```
x a numeric sequence
step how long a step the difference is considered for.
```

index.value 17

#### **Details**

With step=2 (default) only the difference between immediate neighbours are considered; the difference between every second element will remain small, or rather reduced, compared to the original sequence. With step=3 say, differences of both lag 1 and 2 is increased, but the difference of lag 1 will be less than if a step of 2 was used.

#### **Examples**

```
x <- 1:100
diff(x)

diff(incdiff(x, 2))
diff(incdiff(x, 3))

diff(incdiff(x, 2), 2)
diff(incdiff(x, 3), 2)

# incdiff will introduce a periodicity equal to the step length acf(incdiff(x, 10))

# useful for making a sequence of colours more distinct y <- seq(0.4, 1, 1=18)
cols1 <- hsv(y, 1, y)
cols2 <- hsv(y, 1, incdiff(y, 3))

plot(y, col=cols1, pch=16, cex=5, ylim=c(0.4, 1.5))
points(y+0.5, col=cols2, pch=16, cex=5)</pre>
```

index.value

Index-value representation of arrays

#### **Description**

Represent an array as columns of dimensional indices and value

#### Usage

```
index.value(x, reverse = FALSE)
```

## **Arguments**

x an array or something that can be coerced into an array

reverse logical; convert from Index-value representation to regular array representation?

#### **Details**

An n-dimensional array will be unfolded to a n+1-column data.frame where the first n columns represent the indices of the n dimensions, and the last column gives the value found at each index tuple. The reverse process can also be performed.

# See Also

```
latin.sq
```

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#### **Examples**

```
arr <- array(1:(2*3*4), dim=c(2, 3, 4))
arr.is <- index.value(arr)</pre>
# can be used to permutate an array
index.value(arr.is[,c(2, 1, 3, 4)], rev=TRUE)
aperm(arr, c(2, 1, 3))
\# can interpret values (symbols) as dimensional indices and permute them as well
arr2 <- array(rep(1:6, 4), dim=c(2, 3, 4))
arr2.is <- index.value(arr2)</pre>
index.value(arr2.is[,c(1, 2, 4, 3)], rev=TRUE)
# a latin square will produce an "orthogonal array"
set.seed(1)
lsq <- latin.sq(5)</pre>
iv <- index.value(lsq)</pre>
i٧
\mbox{\tt\#} any permutation of a latin square is also a latin square
index.symbol(iv[, c(1, 3, 2)], reverse=TRUE)
```

isPrime

Primality check

## **Description**

Test an integer for whether it is prime or not

## Usage

```
isPrime(x)
```

#### **Arguments**

n

integer; one or more prime candidates

#### See Also

primes

latin.sq

Latin square

## **Description**

Generate latin squares, either randomly or ordered

#### Usage

```
latin.sq(n, random = TRUE, reduce = TRUE)
```

mean.pca 19

#### **Arguments**

n integer; number of unique values (aka. symbols)
random logical; should the square be generated randomly?
redude logical; should the square be in reduced form?

#### Details

Computation time increses rapidly with n. On my computer generating a random square with n=12 takes about ten minutes, marking the upper limit of practicability, or even stretching it a little. A latin square in reduced form will have elements in the first row and the first column in a sorted order. By setting reduced=TRUE the first row and the first column will always be 1:n.

#### Value

A square integer matrix of size n^2

#### See Also

```
index.symbol
```

# **Examples**

```
set.seed(1)
ls <- latin.sq(10, reduce=TRUE)
image(ls, col=randcolours(ncol(ls)))

# The more "classic" representation with latin capital letters
ls[] <- LETTERS[ls]
ls</pre>
```

mean.pca

PCA mean

# **Description**

Takes the average of several PCA objects

## Usage

```
## S3 method for class 'pca'
mean(...)
```

#### **Arguments**

... prcomp, princomp or factanal objects, or a single list of such objects

## **Details**

I don't know if this kind of calculation has any sort of merit. It was written more as an impromptu challenge than as a solution to any problem

20 means

#### See Also

```
prcomp, princomp, factanal
```

## **Examples**

means

Generalized means

#### **Description**

Harmonic, geometric, quadratic, cubic, power and Lehmer means.

## Usage

```
harm(x, na.rm = TRUE)
geom(x, zero.rule = c("1p", "rm", "1"), na.rm = TRUE)
quad(x, na.rm = TRUE)
cubi(x, na.rm = TRUE)
powr(x, p = 1.5, na.rm = TRUE)
lehm(x, p = 2, na.rm = TRUE)
```

## **Arguments**

x numeric vector of values whose \*mean is to be computed na.rm logical; should NA values be removed? (default TRUE)

zero.rule for the geometric mean, how should zeros be dealt with? Add one before, and subtract one after the calculation (see lop1p), remove all zeros, or replace all zeros with 1.

norma 21

р

exponential power. For the power mean p=-1, p=2 and p=3 gives the harmonic, quadratic and cubic means, respectively. For the Lehmer mean p=0, p=1 and p=2 gives the harmonic, arithmetic and contraharmonic means, respectively.

#### **Notice**

For some of these means zeros and/or negative values are undefined, or make otherwise little sense in context. Workarounds are given for the geometric mean, but if you end up using it on data  $\leq 0$ , the wise call would be to reconsider whether using a geometric mean really makes sense in this case.

## **Examples**

norma

Normalize

## **Description**

Linearly shift and scale a numeric vector so that it has a given range, about a given centre.

# Usage

```
norma(W, c = 0, r = 2)
```

# Arguments

W a numeric vector

c the centre (as in the midrange) for the new vector

r the range of the new vector

#### See Also

fitrange

22 plot.stl

#### **Examples**

```
range(norma(runif(9, -2, 0.1), 0, 2))
```

plot.histogram

Plot histogram object

#### **Description**

A a very minor modification of graphics::plot.histogram.

Only difference is that lwd now specifies the weight of the histogram bars' outline.

#### Usage

```
## S3 method for class 'histogram'
plot(x, freq = equidist, density = NULL, angle = 45,
   col = NULL, border = par("fg"), lty = NULL, lwd = 1,
   main = paste("Histogram of", paste(x$xname, collapse = "\n")),
   sub = NULL, xlab = x$xname, ylab, xlim = range(x$breaks), ylim = NULL,
   axes = TRUE, labels = FALSE, add = FALSE, ann = TRUE, ...)
```

#### See Also

stats::stl

plot.stl

Plot stl object

## **Description**

A a very minor modification of stats::stl.

Only difference is that the distance between the plotting window and the x and y labels is set by par("mgp")[1], as it is for regular plots.

# Usage

```
## S3 method for class 'stl' plot(x, labels = colnames(X), set.pars = list(mar = c(0, 6, 0, 6), oma = c(6, 0, 4, 0), tck = -0.01, mfrow = c(nplot, 1)), main = NULL, range.bars = TRUE, ..., col.range = "light gray")
```

# See Also

```
stats::stl
```

primes 23

primes	Prime number generator
•	0

## **Description**

Prime generator based on the sieve of Eratosthenes

# Usage

```
primes(n)
```

## **Arguments**

n integer; all prime numbers up to this will be returned

#### **Details**

```
Effective for primes up to \sim 100,000,000.
On my lightweight laptop: 1e7 -> 0.32s, 5e7 -> 1.7s, 1e8 -> 3.7s, 2e8 -> 7.6s, 3e8 -> 15s
```

#### **Source**

https://stackoverflow.com/questions/3789968/generate-a-list-of-primes-up-to-a-certain-number/3791284#3791284

## See Also

isPrime

randcolours Random colours
----------------------------

# Description

Generate a randomly selected colour palette

# Usage

```
randcolours(n, 1 = c(0.2, 1), c1 = c(0, 1), c2 = c(0, 1), alpha = 1, space = c("Luv", "Lab"))
```

# **Arguments**

n	number of colours
1	lightness range
c1	colour channel one range
c2	colour channel two range
alpha	alpha channel range
space	should the parameters be interpreted as Luv or Lab components?

24 speedskate

#### **Details**

The range of 1, c1, c2 and alpha, will be interpreted as the wanted range of each colour component, whether their length is 1, 2, or more. Although they all should nominally lie within [0, 1], only alpha must do so to achieve a valid output. The others can exceed this range, at an icreased risk of clipping.

# **Examples**

```
set.seed(3)
n <- 20
plot(1:n, col=randcolors(n), pch=16, cex=5)</pre>
```

speedskate

2018 MarbleLympics speed skating times

# Description

Intermediate and total times for all 16 runs, arranged by lane and heat number.

#### Usage

speedskate

## **Format**

A list containing two data.frames, one for each lane. Columns are heat and rows are time checks in seconds.

#### **Source**

https://www.youtube.com/watch?v=fA-O6f\_jArk

```
tt <- t(do.call(cbind, speedskate))
pairs(tt)
cor(tt)
outer(
  colnames(tt),
  colnames(tt),
  Vectorize(function(i,j) cor.test(tt[,i],tt[,j])$p.value)
)</pre>
```

tied\_triple\_test 25

tied\_triple\_test

Tied triple test

#### **Description**

Compare numeric values, returning an inbetween value for ties

## Usage

```
x %ttt% y

ttt(x, y)

is.ttt(x)

## S3 method for class 'ttt'
print(x, symbol = TRUE, ...)

## S3 method for class 'ttt'
table(...)
```

## Arguments

х, у

numeric values to be compared

## See Also

Comparison, comparison\_with\_ties

```
1:5 %ttt% 3
ttt(1:3, 2)
print(ttt(1:3, 2), FALSE)
c(1, 6, 3, 0) %ttt% c(1, 3, 3, 2)
# Equivalent
as.integer(c(1, 6, 3, 0) ttt c(1, 3, 3, 2))
sign(c(1, 6, 3, 0) - c(1, 3, 3, 2))
# Demonstrating table method
dtf \leftarrow data.frame(x=1:5, y=3)
dtf" <- ttt(dtfx, dtfy)
dtf
x \leftarrow c(8, 4, 6, 8, 9, 6, 5, 7, 0, 3, 2, 1, 5, 6, 4, 7, 6,
       3, 1, 9, 5, 6, 7, 7, 4, 5, 8, 6, 2, 5, 9, 5, 4, 8)
y \leftarrow c(1, 3, 2, 4, 6, 0, 5, 3, 7, 5, 7, 4, 5, 6, 0, 1, 4,
       2, 4, 3, 1, 5, 3, 9, 2, 2, 4, 7, 5, 6, 8)
ou <- outer(sort(x), sort(y), "%ttt%")</pre>
```

26 ymse

```
ta <- table(ou)
pa <- capture.output(ta)

par(mar=c(1, 2, 3, 2), family="PT Mono")
image(ou, col=topo.colors(length(ta)), axes=FALSE)
title(pa)
box()</pre>
```

ymse

ymse: A collection of more or less useful functions

# Description

There is go grand "theme" to ymse, other than that none of the functions, and in some cases function groups, seemed to fit too well in any other package

# ymse functions

addrows

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