# Package 'sinkr'

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Type Package

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| <pre>URL https://github.com/menugget/sinkr, http://menugget.blogspot.com</pre> R topics documented: |   |
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addAlpha

Add alpha channel (transparency) to colors

#### **Description**

Takes a vector of colors and adds an alpha channel at the given level of transparency.

## Usage

```
addAlpha(COLORS, ALPHA)
```

# **Arguments**

COLORS Vector of any of the three kinds of R color specifications, i.e., either a color name

(as listed by colors()), a hexadecimal string of the form "#rrggbb" or "#rrggbbaa"

(see rgb), or a positive integer i meaning palette()[i].

ALPHA A value (between 0 and 1) indicating the alpha channel (opacity) value.

## **Examples**

```
# Make background image
x <- seq(-180, 180,, 30)
y <- seq(-90, 90,, 30)
grd <- expand.grid(x=x,y=y)
z <- sqrt(grd$x^2+grd$y^2)
dim(z) <- c(length(x), length(y))
pal <- colorRampPalette(c(rgb(1,1,1), rgb(0,0,0)))
COLORS <- pal(20)
image(x,y,z, col=COLORS)

# Add semi-transparent layer
z2 <- grd$x^2+grd$y
dim(z2) <- c(length(x), length(y))
pal <- colorRampPalette(c(rgb(0.5,1,0), rgb(0,1,1), rgb(1,1,1)))
COLORS <- addAlpha(pal(20), 0.4) # alpha chanel equals 0.4
image(x,y,z2, col=COLORS, add=TRUE)</pre>
```

cov4gappy

Covariance matrix calculation for gappy data

## **Description**

This function calculates a covoriance matrix for data that contain missing values ('gappy data').

## Usage

```
cov4gappy(F1, F2 = NULL)
```

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## **Arguments**

F1 A data field.

F2 An optional 2nd data field.

#### **Details**

This function gives comparable results to cov(F1, y=F2, use="pairwise.complete.obs") whereby each covariance value is divided by n number of shared values (as opposed to n-1 in the case of cov(). Futhermore, the function will return a 0 (zero) in cases where no shared values exist between columns; the advantage being that a covariance matrix will still be calculated in cases of very gappy data, or when spatial locations have accidentally been included without observations (i.e. land in fields of aquatic-related parameters).

#### Value

A matrix with covariances between columns of F1. If both F1 and F2 are provided, then the covariances between columns of F1 and the columns of F2 are returned.

## **Examples**

```
# Create synthetic data
set.seed(1)
mat <- matrix(rnorm(500, sd=10), nrow=50, ncol=10)
matg <- mat
matg[sample(length(mat), 0.5*length(mat))] <- NaN # Makes 50% missing values
matg # gappy matrix

# Calculate covariance matrix and compare to 'cov' function output
c1 <- cov4gappy(matg)
c2 <- cov(matg, use="pairwise.complete.obs")
plot(c1,c2, main="covariance comparison", xlab="cov4gappy", ylab="cov")
abline(0,1,col=8)</pre>
```

dineof

DINEOF (Data Interpolating Empirical Orthogonal Functions)

## Description

This function is based on the DINEOF (Data Interpolating Empirical Orthogonal Functions) procedure described by Beckers and Rixon (2003). The procedure has been shown to accurately determine Empirical Orthogonal Functions (EOFs) from gappy data sets (Taylor et al. 2013). Rather than directly return the EOFs, the results of the dineof function is a fully interpolated matrix which can then be subjected to a final EOF decomposition with eof, prcomp, or other function of preference.

#### Usage

```
dineof(Xo, n.max = NULL, ref.pos = NULL, delta.rms = 1e-05)
```

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#### **Arguments**

| Хо        | A gappy data field.  |
|-----------|--|
| n.max     | A maximum number of EOFs to iterate (leave equalling "NULL" if algorithm shold proceed until convergence)  |
| ref.pos   | - a vector of non-gap reference positions by which errors will be assessed via root mean squared error ("RMS"). If ref.pos = NULL, then either 30 or 1 (which ever is larger) will be sampled at random. |
| delta.rms | The threshold for RMS convergence.   |

#### Value

Results of dineof are returned as a list containing the following components:

Xa The data field with interpolated values (via EOF reconstruction) included.
 n.eof The number of EOFs used in the final solution.
 RMS A vector of the RMS values from the iteration.
 NEOF A vector of the number of EOFs used at each iteration.

## References

Beckers, Jean-Marie, and M. Rixen. "EOF Calculations and Data Filling from Incomplete Oceanographic Datasets." Journal of Atmospheric and Oceanic Technology 20.12 (2003): 1839-1856.

Taylor, Marc H., Martin Losch, Manfred Wenzel, Jens Schroeter (2013). On the Sensitivity of Field Reconstruction and Prediction Using Empirical Orthogonal Functions Derived from Gappy Data. J. Climate, 26, 9194-9205.

```
# Make synthetic data field
m=50
n=100
frac.gaps <- 0.5 # the fraction of data with NaNs</pre>
N.S.ratio <- 0.1 \# the Noise to Signal ratio for adding noise to data
x \leftarrow (seq(m)*2*pi)/m
t \leftarrow (seq(n)*2*pi)/n
Xt <-
outer(sin(x), sin(t)) +
 outer(\sin(2.1*x), \sin(2.1*t)) +
 outer(\sin(3.1*x), \sin(3.1*t)) +
 outer(tanh(x), cos(t)) +
 outer(tanh(2*x), cos(2.1*t)) +
 outer(tanh(4*x), cos(0.1*t)) +
 outer(tanh(2.4*x), cos(1.1*t)) +
 tanh(outer(x, t, FUN="+")) +
 tanh(outer(x, 2*t, FUN="+")
# Color palette
pal <- colorRampPalette(c("blue", "cyan", "yellow", "red"))</pre>
# The "true" fieldd
Xt \leftarrow t(Xt)
```

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```
image(Xt, col=pal(100))
# The "noisy" field
set.seed(1)
RAND <- matrix(runif(length(Xt), min=-1, max=1), nrow=nrow(Xt), ncol=ncol(Xt))
R \leftarrow RAND * N.S.ratio * Xt
Xp \leftarrow Xt + R
image(Xp, col=pal(100))
# The "observed" gappy field field
gaps <- sample(seq(length(Xp)), frac.gaps*length(Xp))</pre>
Xo <- replace(Xp, gaps, NaN)</pre>
image(Xo, col=pal(100))
# The dineof "interpolated" field
set.seed(1)
RES <- dineof(Xo)</pre>
Xa <- RES$Xa
image(Xa, col=pal(100))
# Final comparison of all fields
ZLIM <- range(Xt, Xp, Xo, Xa, na.rm=TRUE)</pre>
op <- par(mfrow=c(2,2), mar=c(3,3,3,1))
image(z=Xt, zlim=ZLIM, main="A) True", col=pal(100), xaxt="n", yaxt="n", xlab="", ylab="")
box()
mtext("t", side=1, line=0.5)
mtext("x", side=2, line=0.5)
image(z=Xp, zlim=ZLIM, main=paste("B) True + Noise (N/S = ", N.S.ratio, ")", sep=""),
col=pal(100), xaxt="n", yaxt="n", xlab="", ylab="")
box()
mtext("t", side=1, line=0.5)
mtext("x", side=2, line=0.5)
image(z=Xo, zlim=ZLIM, main=paste("C) Observed (", frac.gaps*100, " % gaps)", sep=""),
col=pal(100), xaxt="n", yaxt="n", xlab="", ylab="")
mtext("t", side=1, line=0.5)
mtext("x", side=2, line=0.5)
image(z=Xa, zlim=ZLIM, main="D) Reconstruction", col=pal(100), xaxt="n", yaxt="n",
xlab="", ylab="")
box()
mtext("t", side=1, line=0.5)
mtext("x", side=2, line=0.5)
par(op)
```

eof

EOF (Empirical Orthogonal Functions analysis)

## **Description**

This function conducts an Empirical Orthogonal Function analysis (EOF) via a covariance matrix (cov4gappy()) and is especially designed to handle gappy data (i.e. containing missing values - NaN)

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#### **Usage**

```
eof(F1, centered = TRUE, scaled = FALSE, nu = NULL, method = NULL,
  recursive = FALSE)
```

#### **Arguments**

F1 A data field. The data should be arraunged as samples in the column dimension

(typically each column is a time series for a spatial location).

centered Logical (TRUE/FALSE) to define if F1 should be centered prior to the analysis.

Defaults to 'TRUE'

scaled Logical (TRUE/FALSE) to define if F1 should be scaled prior to the analysis.

Defaults to 'TRUE'

nu Numeric value. Defines the number of EOFs to return. Defaults to return the

full set of EOFs.

method Method for matrix decomposition ('svd', 'eigen', 'irlba'). Defaults to 'svd'

when method = NULL. Use of 'irlba' can dramatically speed up cumputation time when recursive = TRUE but may produce errors in computing trailing EOFs. Therefore, this option is only advisable when the field F1 is large and when only a partial decomposition is desired (i.e. nu << dim(F1)[2]). svd and eigen give similar results for non-gappy fields, but will differ slightly with gappy fields due to decomposition of a nonpositive definite covariance matrix. Specifically, eigen will produce negative eigenvalues for trailing EOFs, while

singular values derived from svd will be strictly positive.

recursive Logical. When TRUE, the function follows the method of "Recursively Sub-

tracted Empirical Orthogonal Functions" (RSEOF), which is a modification of "Least Squares Empirical Orthogonal Functions" (LSEOF) (Taylor et al., 2013).

#### Value

Results of eof are returned as a list containing the following components:

u EOFs.

Lambda Singular values.

A EOF coefficients (i.e. 'Principal Components').

F1\_dim Dimensions of field F1.

F1\_center Vector of center values from each column in field F1.

F1\_scale Vector of scale values from each column in field F1.

#### References

Bjoernsson, H. and Venegas, S.A. (1997). "A manual for EOF and SVD analyses of climate data", McGill University, CCGCR Report No. 97-1, Montreal, Quebec, 52pp.

von Storch, H, Zwiers, F.W. (1999). Statistical analysis in climate research. Cambridge University Press.

Taylor, Marc H., Martin Losch, Manfred Wenzel, Jens Schroeter (2013). On the Sensitivity of Field Reconstruction and Prediction Using Empirical Orthogonal Functions Derived from Gappy Data. J. Climate, 26, 9194-9205.

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#### **Examples**

```
# EOF of 'iris' dataset
Et <- eof(iris[,1:4])</pre>
plot(Et$A, col=iris$Species)
# Compare to results of 'prcomp'
Pt <- prcomp(iris[,1:4])</pre>
plot(Et$A, Pt$x) # Sign may be different
# Compare to a gappy dataset (sign of loadings may differ between methods)
iris.gappy <- as.matrix(iris[,1:4])</pre>
set.seed(1)
iris.gappy[sample(length(iris.gappy), 0.25*length(iris.gappy))] <- NaN</pre>
Eg <- eof(iris.gappy, method="svd", recursive=TRUE) # recursive ("RSEOF")</pre>
op <- par(mfrow=c(1,2))</pre>
plot(Et$A, col=iris$Species)
plot(Eg$A, col=iris$Species)
par(op)
# Compare Non-gappy vs. Gappy EOF loadings
op <- par(no.readonly = TRUE)</pre>
layout(matrix(c(1,2,1,3),2,2), widths=c(3,3), heights=c(1,4))
par(mar=c(0,0,0,0))
plot(1, t="n", axes=FALSE, ann=FALSE)
legend("center", ncol=4, legend=colnames(iris.gappy), border=1, bty="n",
fill=rainbow(4))
par(mar=c(6,3,2,1))
barplot(Et\$u,\ beside=TRUE,\ col=rainbow(4),\ ylim=range(Et\$u)*c(1.15,1.15))
mtext("Non-gappy", side=3, line=0)
 axis(1, labels=paste("EOF", 1:4), at=c(3, 8, 13, 18), las=2, tick=FALSE) \\ barplot(Eg$u, beside=TRUE, col=rainbow(4), ylim=range(Et$u)*c(1.15,1.15)) 
mtext("Gappy", side=3, line=0)
axis(1, labels=paste("EOF", 1:4), at=c(3, 8, 13, 18), las=2, tick=FALSE)
par(op)
```

eofRecon

EOF reconstruction (Empirical Orthogonal Functions analysis)

# **Description**

This function reconstructs the original field from an EOF object of the function eof.

# Usage

```
eofRecon(EOF, n_pc = NULL)
```

# **Arguments**

**EOF** An object resulting from the function eof.

The number of principal components (PCs) to use in the reconstruction (defaults n\_pc

to the fullset of PCs)

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#### **Examples**

```
set.seed(1)
iris.gappy <- as.matrix(iris[,1:4])</pre>
iris.gappy[sample(length(iris.gappy), \ 0.25*length(iris.gappy))] <- \ NaN
Er <- eof(iris.gappy, method="svd", recursive=TRUE) # recursive (RSEOF)</pre>
{\tt Enr <- \ eof(iris.gappy, \ method="svd", \ recursive=FALSE) \ \# \ non-recursive \ (LSEOF)}
iris.gappy.recon.r <- eofRecon(Er)</pre>
iris.gappy.recon.nr <- eofRecon(Enr)</pre>
# Reconstructed values vs. observed values
op <- par(mfrow=c(1,2))
lim <- range(iris.gappy, na.rm=TRUE)</pre>
plot(iris.gappy, iris.gappy.recon.r,
col=c(2:4)[iris$Species], main="recursive=TRUE", xlim=lim, ylim=lim)
abline(0, 1, col=1, lwd=2)
plot(iris.gappy, iris.gappy.recon.nr,
col=c(2:4)[iris$Species], main="recursive=FALSE", xlim=lim, ylim=lim)
abline(0, 1, col=1, lwd=2)
par(op)
# Reconstructed values from gappy data vs. all original values
op <- par(mfrow=c(1,2))</pre>
plot(as.matrix(iris[,1:4]), iris.gappy.recon.r,
col=c(2:4)[iris$Species], main="recursive=TRUE")
abline(0, 1, col=1, lwd=2)
plot(as.matrix(iris[,1:4]), iris.gappy.recon.nr,
col=c(2:4)[iris$Species], main="recursive=FALSE")
abline(0, 1, col=1, lwd=2)
```

expmat

Exponentiation of a matrix

## **Description**

The expmat function performs can calculate the pseudoinverse (i.e. "Moore-Penrose pseudoinverse") of a matrix (EXP=-1) and other exponents of matrices, such as square roots (EXP=0.5) or square root of its inverse (EXP=-0.5). The function arguments are a matrix (MAT), an exponent (EXP), and a tolerance level for non-zero singular values. The function follows three steps: 1) Singular Value Decomposition (SVD) of the matrix; 2) Exponentiation of the singular values; 3) Re-calculation of the matrix with the new singular values

## Usage

```
expmat(MAT, EXP, tol = NULL)
```

## **Arguments**

| MAT | A matrix.                                     |
|-----|---|
| EXP | An exponent to apply to the matrix MAT.       |
| tol | Tolerance level for non-zero singular values. |

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```
# Example matrix from Wilks (2006)
A \leftarrow matrix(c(185.47,110.84,110.84,77.58),2,2)
solve(A) #inverse
expmat(A, -1) # pseudoinverse
expmat(expmat(A, -1), -1) #inverse of the inverse -return to original A matrix
expmat(A, 0.5) # square root of a matrix
expmat(A, -0.5) # square root of its inverse
expmat(expmat(A, -1), 0.5) # square root of its inverse (same as above)
# Pseudoinversion of a non-square matrix
D <- matrix(round(runif(24, min=1, max=100)), 4, 6)</pre>
expmat(D, -1)
expmat(t(D), -1)
# Pseudoinversion of a square matrix
set.seed(1)
D <- matrix(round(runif(25, min=1, max=100)), 5, 5)
solve(D)
expmat(D, -1)
solve(t(D))
expmat(t(D), -1)
### Examples from "corpcor" package manual
# a singular matrix
m = rbind(
 c(1,2),
 c(1,2)
# not possible to invert exactly
# solve(m) # produces an error
p <- expmat(m, -1)</pre>
# characteristics of the pseudoinverse
zapsmall( m \%*\% p \%*\% m ) == zapsmall( m )
zapsmall(p \%*\% m \%*\% p) == zapsmall(p)
zapsmall(p \%*\% m) == zapsmall(t(p \%*\% m))
zapsmall( m \% * \% p ) == zapsmall( t(m \% * \% p ) )
# example with an invertable matrix
m2 = rbind(
 c(1,1),
  c(1,0)
zapsmall( solve(m2) ) == zapsmall( expmat(m2,-1) )
```

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## **Description**

The imageScale function is wrapper for imageScale and accepts the same arguments. It converts a vector of values (z) to a vector of color levels. One must define the number of colors. The limits of the color scale ("zlim") or the break points for the color changes("breaks") can also be defined. When breaks and zlim are defined, breaks overrides zlim. All arguments are similar to those in the image function. Appearance is best when incorporated with layout.

#### Usage

```
imageScale(z, zlim, col = heat.colors(12), breaks, axis.pos = 1,
  add.axis = TRUE, xlim = NULL, ylim = NULL, ...)
```

## **Arguments**

| Z        | A vector or matrix of values.   |
|----------|---|
| zlim     | Limits of the color scale values.   |
| col      | Vector of color values (default is 12 colors from the heat.colors palette).   |
| breaks   | Break points for color changes. If breaks is specified then zlim is unused and the algorithm used follows cut, so intervals are closed on the right and open on the left except for the lowest interval which is closed at both ends. |
| axis.pos | Position of the axis (1=bottom, 2=left, 3=top, 4=right) (default = 1).  |
| add.axis | Logical (TRUE/FALSE). Defines whether the axis is added (default: TRUE).  |
| xlim     | Limits for the x-axis.  |
| ylim     | Limits for the y-axis.  |
|          | Additional graphical parameters to pass to the image() function.  |

```
# Make color palettes
pal.1=colorRampPalette(c("green4", "orange", "red", "white"), space="rgb", bias=0.5)
pal.2=colorRampPalette(c("blue", "cyan", "yellow", "red", "pink"), space="rgb")
# Make images with corrsponding scales
op <- par(no.readonly = TRUE)</pre>
layout(matrix(c(1,2,3,0,4,0),\ nrow=2,\ ncol=3),\ width s=c(4,4,1),\ heights=c(4,1))
#layout.show(4)
#1st image
breaks <- seq(min(volcano), max(volcano),length.out=100)</pre>
par(mar=c(1,1,1,1))
image(seq(dim(volcano)[1]), seq(dim(volcano)[2]), volcano,
col=pal.1(length(breaks)-1), breaks=breaks-1e-8, xaxt="n", yaxt="n", ylab="", xlab="")
#Add additional graphics
levs <- pretty(range(volcano), 5)</pre>
contour(seq(dim(volcano)[1]), seq(dim(volcano)[2]), volcano, levels=levs, add=TRUE)
#Add scale
par(mar=c(3,1,1,1))
imageScale(volcano, col=pal.1(length(breaks)-1), breaks=breaks-1e-8,axis.pos=1)
abline(v=levs)
box()
breaks <- c(0,100, 150, 170, 190, 200)
```

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```
par(mar=c(1,1,1,1))
image(seq(dim(volcano)[1]), seq(dim(volcano)[2]), volcano,
col=pal.2(length(breaks)-1), breaks=breaks-1e-8, xaxt="n", yaxt="n", ylab="", xlab="")
#Add additional graphics
levs=breaks
contour(seq(dim(volcano)[1]), seq(dim(volcano)[2]), volcano, levels=levs, add=TRUE)
#Add scale
par(mar=c(1,1,1,3))
imageScale(volcano, col=pal.2(length(breaks)-1), breaks=breaks-1e-8,
axis.pos=4, add.axis=FALSE)
axis(4,at=breaks, las=2)
box()
abline(h=levs)
par(op)
```

plotStacked

Stacked plot

#### **Description**

plotStacked makes a stacked plot where each y series is plotted on top of each other using filled polygons.

## Usage

```
plotStacked(x, y, order.method = "as.is", ylab = "", xlab = "",
border = NULL, lwd = 1, col = rainbow(length(y[1, ])), ylim = NULL,
...)
```

## **Arguments**

| x            | A vector of values   |
|--------------|--|
| у            | A matrix of data series (columns) corresponding to x   |
| order.method | Method of ordering y plotting order. One of the following: $c("as.is", "max", "first")$ . "as.is" - plot in order of y column. "max" - plot in order of when each y series reaches maximum value. "first" - plot in order of when each y series first value > 0. |
| ylab         | y-axis labels  |
| xlab         | x-axis labels  |
| border       | Border colors for polygons corresponding to y columns (will recycle) (see ?polygon for details)  |
| lwd          | Border line width for polygons corresponding to y columns (will recycle)   |
| col          | Fill colors for polygons corresponding to y columns (will recycle).  |
| ylim         | y-axis limits. If ylim=NULL, defaults to c(0, 1.2*max(apply(y,1,sum).  |
|              | Other plot arguments   |

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#### **Examples**

```
#Create data
set.seed(1)
m < -500
n <- 30
x < - seq(m)
y <- matrix(0, nrow=m, ncol=n)
colnames(y) <- seq(n)
for(i in seq(ncol(y))){
mu <- runif(1, min=0.25*m, max=0.75*m)</pre>
 SD <- runif(1, min=5, max=20)</pre>
TMP <- rnorm(1000, mean=mu, sd=SD)</pre>
HIST <- hist(TMP, breaks=c(0,x), plot=FALSE)</pre>
fit <- smooth.spline(HIST$counts ~ HIST$mids)</pre>
y[,i] <- fit$y
}
y <- replace(y, y<0.01, 0)
#Ex.1 : Color by max value)
pal <- colorRampPalette(c(rgb(0.85,0.85,1), rgb(0.2,0.2,0.7)))
BREAKS <- pretty(apply(y,2,max),8)</pre>
LEVS <- levels(cut(1, breaks=BREAKS))</pre>
COLS <- pal(length(BREAKS )-1)</pre>
z <- val2col(apply(y,2,max), col=COLS)</pre>
#Create stacked plot (plot order = "as.is")
plotStacked(x,y, xlim=c(100, 400), ylim=c(0, 1.2*max(apply(y,1,sum), na.rm=TRUE)),
yaxs="i", col=z, border="white", lwd=0.5)
#Create stacked plot (plot order = "max")
plotStacked(x,y, xlim=c(100, 400), ylim=c(0, 1.2*max(apply(y,1,sum), na.rm=TRUE)),\\
order.method="max", yaxs="i", col=z, border="white", lwd=0.5)
#Ex. 2 : Color by first value
ord <- order(apply(y, 2, function(r) min(which(r>0))))
y2 \leftarrow y[, ord]
pal <- colorRampPalette(c("blue", "cyan", "yellow", "red"))</pre>
z <- pal(ncol(y2))</pre>
#Create stacked plot (plot order = "as.is")
plotStacked(x,y2, xlim=c(100, 400), ylim=c(0, 1.2*max(apply(y2,1,sum), na.rm=TRUE)),\\
yaxs="i", col=z, border=1, lwd=0.25)
#Create stacked plot (plot order = "max")
plotStacked(x,y2,\ xlim=c(100,\ 400),\ ylim=c(0,\ 1.2*max(apply(y2,1,sum),\ na.rm=TRUE)),
order.method="max", yaxs="i", col=z, border=1, lwd=0.25)
```

plotStream

Stream plot

#### **Description**

plotStream makes a "stream plot" where each y series is plotted as stacked filled polygons on alternating sides of a baseline. A random wiggle is applied through the arguments frac.rand and spar such that each plot will be different unless preceded by a random seed (e.g. set.seed(1)).

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## Usage

```
plotStream(x, y, order.method = "as.is", frac.rand = 0.1, spar = 0.2,
  center = TRUE, ylab = "", xlab = "", border = NULL, lwd = 1,
  col = rainbow(length(y[1, ])), ylim = NULL, ...)
```

## **Arguments**

A vector of values Χ A matrix of data series (columns) corresponding to x У order.method Method of ordering y plotting order. One of the following: c("as.is", "max", "first"). "as.is" - plot in order of y column. "max" - plot in order of when each y series reaches maximum value. "first" - plot in order of when each y series first value > 0. frac.rand Fraction of the overall data "stream" range used to define the range of random wiggle (uniform distrubution) to be added to the baseline go Setting for smooth.spline function to make a smoothed version of baseline "g0" spar Logical. If TRUE, the stacked polygons will be centered so that the middle, i.e. center baseline (g0), of the stream is approximately equal to zero. Centering is done before the addition of random wiggle to the baseline. ylab y-axis labels x-axis labels xlab border Border colors for polygons corresponding to y columns (will recycle) (see ?polygon for details) Border line width for polygons corresponding to y columns (will recycle) lwd Fill colors for polygons corresponding to y columns (will recycle). col y-axis limits. If ylim=NULL, defaults to c(-0.7, 0.7)\*max(apply(y,1,sum))ylim

## **Examples**

```
#Create data
set.seed(1)
m <- 500
n <- 30
x < - seq(m)
y <- matrix(0, nrow=m, ncol=n)</pre>
colnames(y) <- seq(n)
for(i in seq(ncol(y))){
 mu <- runif(1, min=0.25*m, max=0.75*m)</pre>
 SD <- runif(1, min=5, max=20)</pre>
 TMP <- rnorm(1000, mean=mu, sd=SD)</pre>
 HIST <- hist(TMP, breaks=c(0,x), plot=FALSE)</pre>
 fit <- smooth.spline(HIST$counts ~ HIST$mids)</pre>
y[,i] <- fit$y
}
y <- replace(y, y<0.01, 0)
#Ex.1 : Color by max value)
pal <- colorRampPalette(c(rgb(0.85,0.85,1), rgb(0.2,0.2,0.7)))</pre>
BREAKS <- pretty(apply(y,2,max),8)</pre>
```

Other plot arguments

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```
LEVS <- levels(cut(1, breaks=BREAKS))
COLS <- pal(length(BREAKS )-1)
z <- val2col(apply(y,2,max), col=COLS)</pre>
#Plot order = "as.is"
plotStream(x,y, xlim=c(100, 400), center=TRUE, spar=0.3, frac.rand=0.2,
col=z, border="white", lwd=0.5)
#Plot order = "max"
plotStream(x,y, xlim=c(100, 400), center=TRUE, order.method="max", spar=0.3.
frac.rand=0.2, col=z, border="white", lwd=0.5)
#Ex. 2 : Color by first value
ord <- order(apply(y, 2, function(r) min(which(r>0))))
y2 \leftarrow y[, ord]
pal <- colorRampPalette(c("blue", "cyan", "yellow", "red"))</pre>
z <- pal(ncol(y2))</pre>
#Plot order = "as.is"
plotStream(x,y2, xlim=c(100, 400), center=FALSE, spar=0.1, frac.rand=0.05,
col=z, border=1, lwd=0.25)
#Plot order = "max"
plotStream(x,y2, xlim=c(100, 400), center=FALSE, order.method="max", spar=0.1,
frac.rand=0.05, col=z, border=1, lwd=0.25)
#Extremely wiggly, no borders, no box, no axes, black background
op <- par(bg=1, mar=c(0,0,0,0))
plotStream(x,y2, xlim=c(100, 400), center=FALSE, spar=0.3, frac.rand=1, col=z,
border=NA, axes=FALSE)
par(op)
```

val2col

Convert values to color levels

# Description

The val2col function converts a vector of values("z") to a vector of color levels. One must define the number of colors. The limits of the color scale ("zlim") or the break points for the color changes("breaks") can also be defined. When breaks and zlim are defined, breaks overrides zlim. All arguments are similar to those in the image function.

#### **Usage**

```
val2col(z, zlim, col = heat.colors(12), breaks)
```

## Arguments

z A vector of values (default is 12 colors from the heat.colors palette).

zlim Limits of the color scale values.

col Vector of color values

breaks Break points for color changes. If breaks is specified then zlim is unused and the algorithm used follows cut, so intervals are closed on the right and open on

the left except for the lowest interval which is closed at both ends.

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```
set.seed(1)
n <- 250
x \leftarrow seq(n)
y <- rnorm(n)
# Use all levels, evenly distributed breaks
Col <- val2col(y, col=rainbow(20))</pre>
plot(x,y, pch=21, bg=Col)
# Use limits, evenly distributed breaks
pal <- colorRampPalette(c("blue", "cyan", "yellow", "red"))</pre>
Col <- val2col(y, zlim=c(-1,1), col=pal(20))</pre>
plot(x,y, pch=21, bg=Col)
abline(h=c(-1,1), col=8, lty=2)
# Use custom breaks (break vector must have one more break than color)
Col \leftarrow val2col(y, col=topo.colors(6), breaks=c(-Inf, -2, -1, 0, 1, 2, Inf))
plot(x,y, pch=21, bg=Col)
abline(h=c(-Inf, -2, -1, 0, 1, 2, Inf), col=8, lty=2)
```

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