# Package 'provenance'

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<b>Date</b> 2019-12-05
<b>Description</b> Bundles a number of established statistical methods to facilitate the visual interpreta-
tion of large datasets in sedimentary geology. Includes functionality for adaptive kernel den-
sity estimation, principal component analysis, correspondence analysis, multidimensional scal-
ing, generalised procrustes analysis and individual differences scaling using a variety of dissim
larity measures. Univariate provenance proxies, such as single-grain ages or (isotopic) composi
tions are compared with the Kolmogorov-Smirnov, Kuiper or Sircombe-Hazelton L2 dis-
tances. Categorical provenance proxies such as chemical compositions are com-

pared with the Aitchison and Bray-Curtis distances, and point-counting data with the chi-

square distance. Also included are tools to plot compositional and point-counting data on ternary diagrams and point-counting data on radial plots, to calculate the sample size required for specified levels of statistical precision, and to assess the effects of hydraulic sorting on detrital compositions. Includes an intuitive query-based user interface for users who are not proficient in R.

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Imports MASS, methods, IsoplotR

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ALR

Additive logratio transformation

# Description

Calculates Aitchison's additive logratio transformation for a dataset of class compositional or a compositional data matrix.

# Usage

```
ALR(x, ...)
## Default S3 method:
ALR(x, inverse = FALSE, ...)
## S3 method for class 'compositional'
ALR(x, ...)
```

# Arguments

```
x an object of class compositional OR a matrix of numerical values... optional argumentsinverse perform the inverse inverse logratio transformation?
```

### Value

a matrix of ALR coordinates OR an object of class compositional (if inverse=TRUE).

```
# logratio plot of trace element concentrations:
data(Namib)
alr <- ALR(Namib$Trace)
pairs(alr[,1:5])
title('log(X/Pb)')</pre>
```

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amalgamate

Group components of a composition

# Description

Adds several components of a composition together into a single component

# Usage

```
amalgamate(X, ...)
## Default S3 method:
amalgamate(X, ...)
## S3 method for class 'compositional'
amalgamate(X, ...)
## S3 method for class 'counts'
amalgamate(X, ...)
## S3 method for class 'SRDcorrected'
amalgamate(X, ...)
```

# **Arguments**

X a compositional dataset

a series of new labels assigned to strings or vectors of strings denoting the components that need amalgamating

# Value

an object of the same class as X with fewer components

as.acomp 5

as.acomp

create an acomp object

#### **Description**

Convert an object of class compositional to an object of class acomp for use in the compositions package

# Usage

```
as.acomp(x)
```

### **Arguments**

Х

an object of class compositional

#### Value

```
a data.frame
```

# **Examples**

```
data(Namib)
qfl <- ternary(Namib$PT,c('Q'),c('KF','P'),c('Lm','Lv','Ls'))
plot(qfl,type="QFL.dickinson")
qfl.acomp <- as.acomp(qfl)
## uncomment the next two lines to plot an error
## ellipse using the compositions package:
# library(compositions)
# ellipses(mean(qfl.acomp),var(qfl.acomp),r=2)</pre>
```

as.compositional

create a compositional object

# Description

Convert an object of class matrix, data. frame or acomp to an object of class compositional

# Usage

```
as.compositional(x, method = NULL, colmap = "rainbow")
```

# **Arguments**

x an object of class matrix, data.frame or acomp

method dissimilarity measure, either 'aitchison' for Aitchison's CLR-distance or 'bray'

for the Bray-Curtis distance.

colmap the colour map to be used in pie charts.

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### Value

an object of class compositional

# Examples

```
data(Namib)
PT.acomp <- as.acomp(Namib$PT)
PT.compositional <- as.compositional(PT.acomp)
print(Namib$PT$x - PT.compositional$x)
## uncomment the following lines for an illustration of using this
## function to integrate the \code{provenance} package with \code{compositions}
# library(compositions)
# data(Glacial)
# a.glac <- acomp(Glacial)
# c.glac <- as.compositional(a.glac)
# summaryplot(c.glac,ncol=8)</pre>
```

as.counts

create a counts object

### **Description**

Convert an object of class matrix or data. frame to an object of class counts

#### Usage

```
as.counts(x, method = "chisq", colmap = "rainbow")
```

# **Arguments**

x an object of class matrix or data.frame

method either "chisq" (for the chi-square distance) or "bray" (for the Bray-Curtis dis-

tance)

colmap the colour map to be used in pie charts.

#### Value

an object of class counts

```
X <- matrix(c(0,100,0,30,11,2,94,36,0),nrow=3,ncol=3)
rownames(X) <- 1:3
colnames(X) <- c('a','b','c')
comp <- as.counts(X)
d <- diss(comp)</pre>
```

as.data.frame 7

as.data.frame

 ${\it create}\;a\;{\it data.frame}\;object$ 

### **Description**

Convert an object of class compositional to a data. frame for use in the robCompositions package

# Usage

```
## S3 method for class 'compositional'
as.data.frame(x, ...)
## S3 method for class 'counts'
as.data.frame(x, ...)
```

# **Arguments**

x an object of class compositional

... optional arguments to be passed on to the generic function

#### Value

a data.frame

# **Examples**

```
data(Namib)
Major.frame <- as.data.frame(Namib$Major)
## uncomment the next two lines to plot an error
## ellipse using the robCompositions package:
# library(robCompositions)
# plot(pcaCoDa(Major.frame))</pre>
```

botev

Compute the optimal kernel bandwidth

# **Description**

Uses the diffusion algorithm of Zdravko Botev (2011) to calculate the bandwidth for kernel density estimation

### Usage

```
botev(x)
```

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### **Arguments**

x a vector of ordinal data

#### Value

a scalar value with the optimal bandwidth

# Author(s)

Dzdravko Botev

### References

Botev, Z. I., J. F. Grotowski, and D. P. Kroese. "Kernel density estimation via diffusion." The Annals of Statistics 38.5 (2010): 2916-2957.

# **Examples**

```
fname <- system.file("DZ.csv",package="provenance")
bw <- botev(read.distributional(fname)$x$N1)
print(bw)</pre>
```

bray.diss

Bray-Curtis dissimilarity

# Description

Calculates the Bray-Curtis dissimilarity between two samples

# Usage

```
bray.diss(x, y)
```

# **Arguments**

x a vector containing the first compositional sample

y a vector of length(x) containing the second compositional sample

#### Value

a scalar value

```
data(Namib)
print(bray.diss(Namib$HM$x["N1",],Namib$HM$x["N2",]))
```

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CA

Correspondence Analysis

# Description

Performs Correspondence Analysis of point-counting data

# Usage

```
CA(x, nf = 2, ...)
```

### **Arguments**

x an object of class counts

nf number of correspondence factors (dimensions)

... optional arguments to the corresp function of the MASS package

### Value

an object of classes CA, which is synonymous to the MASS packages' correspondence class.

# **Examples**

```
data(Namib)
plot(CA(Namib$PT))
```

central

Calculate central compositions

### **Description**

Computes the geometric mean composition of a continuous mixture of point-counting data.

# Usage

```
central(x, ...)
```

### **Arguments**

```
x an object of class counts
... optional arguments
```

#### **Details**

The central composition assumes that the observed point-counting distribution is the combination of two sources of scatter: counting uncertainty and true geological dispersion.

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### Value

an [5 x n] matrix with n being the number of categories and the rows containing:

theta the 'central' composition.

**err** the standard error for the central composition.

**sigma** the overdispersion parameter, i.e. the coefficient of variation of the underlying logistic normal distribution. central computes a continuous mixture model for each component (column) separately. Covariance terms are not reported.

- LL the lower limit of a '1 sigma' region for codetheta.
- **UL** the upper limit of a '1 sigma' region for codetheta.

mswd the mean square of the weighted deviates, a.k.a. reduced chi-square statistic.

p.value the p-value for age homogeneity

CLR

Centred logratio transformation

# **Description**

Calculates Aitchison's centered logratio transformation for a dataset of class compositional or a compositional data matrix.

# Usage

```
CLR(x, ...)
## Default S3 method:
CLR(x, inverse = FALSE, ...)
## S3 method for class 'compositional'
CLR(x, ...)
```

# **Arguments**

an object of class compositional OR a matrix of numerical valuesoptional argumentsperform the inverse logratio transformation?

#### Value

a matrix of CLR coordinates OR an object of class compositional (if inverse=TRUE)

combine 11

### **Examples**

```
# The following code shows that applying provenance's PCA function
# to compositional data is equivalent to applying R's built-in
# princomp function to the CLR transformed data.
data(Namib)
plot(PCA(Namib$Major))
dev.new()
clrdat <- CLR(Namib$Major)
biplot(princomp(clrdat))</pre>
```

combine

Combine samples of distributional data

# **Description**

Lumps all single grain analyses of several samples together under a new name

# Usage

```
combine(X, ...)
```

# Arguments

X a distributional dataset

a series of new labels assigned to strings or vectors of strings denoting the samples that need amalgamating

### Value

a distributional data object with fewer samples than X

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densities

A list of rock and mineral densities

### **Description**

List of rock and mineral densities using the following abbreviations: Q (quartz), KF (K-feldspar), P (plagioclase), F (feldspar), Lvf (felsic/porfiritic volcanic rock fragments), Lvm (microlithic / porfiritic / trachitic volcanic rock fragments), Lcc (calcite), Lcd (dolomite), Lp (marl), Lch (chert), Lms (argillaceous / micaceous rock fragments), Lmv (metavolcanics), Lmf (metasediments), Lmb (metabasites), Lv (volcanic rock fragments), Lc (carbonates), Ls (sedimentary rock fragments), Lm (metamorphic rock fragments), Lu (serpentinite), mica, opaques, FeOx (Fe-oxides), turbids, zr (zircon), tm (tourmaline), rt (rutile), TiOx (Ti-oxides), sph (titanite), ap (apatite), mon (monazite), oth (other minerals), ep (epidote), othLgM (prehnite + pumpellyite + lawsonite + carpholite), gt (garnet), ctd (chloritoid), st (staurolite), and (andalusite), ky (kyanite), sil (sillimanite), amp (amphibole), px (pyroxene), cpx (clinopyroxene), opx (orthopyroxene), ol (olivine), spinel and othHM (other heavy minerals).

# Author(s)

Alberto Resentini and Pieter Vermeesch

#### References

Resentini, A, Malusa M G and Garzanti, E. "MinSORTING: An Excel worksheet for modelling mineral grain-size distribution in sediments, with application to detrital geochronology and provenance studies." Computers & Geosciences 59 (2013): 90-97.

Garzanti, E, Ando, S and Vezzoli, G. "Settling equivalence of detrital minerals and grain-size dependence of sediment composition." Earth and Planetary Science Letters 273.1 (2008): 138-151.

#### See Also

restore, minsorting

```
data(Namib,densities)
N8 <- subset(Namib$HM,select="N8")
distribution <- minsorting(N8,densities,phi=2,sigmaphi=1,medium="air",by=0.05)
plot(distribution)</pre>
```

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diss	Calculate the dissimilarity matrix between two distributional or compositional datasets
	compositional advasers

# Description

Calculate the dissimilarity matrix between two datasets of class distributional or compositional using the Kolmogorov-Smirnov, Sircombe-Hazelton, Aitchison or Bray Curtis distance

# Usage

```
diss(x, method)
## S3 method for class 'distributional'
diss(x, method = NULL)
## S3 method for class 'compositional'
diss(x, method = NULL)
## S3 method for class 'counts'
diss(x, method = NULL)
```

# **Arguments**

```
x an object of class distributional, compositional or counts
method (optional) either "KS", "Kuiper", "SH", "aitchison", "bray" or "chisq"
```

# Value

an object of class diss

# **Examples**

```
data(Namib)
print(round(100*diss(Namib$DZ)))
```

endmembers

Petrographic end-member compositions

### **Description**

A compositional dataset comprising the mineralogical compositions of the following end-members: undissected\_magmatic\_arc, dissected\_magmatic\_arc, ophiolite, recycled\_clastic, undissected\_continental\_block, transitional\_continental\_block, dissected\_continental\_block, subcreted\_axial\_belt and subducted\_axial\_belt

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#### Author(s)

Alberto Resentini and Pieter Vermeesch

#### References

Resentini, A, Malusa M G and Garzanti, E. "MinSORTING: An Excel worksheet for modelling mineral grain-size distribution in sediments, with application to detrital geochronology and provenance studies." Computers & Geosciences 59 (2013): 90-97.

Garzanti, E, Ando, S and Vezzoli, G. "Settling equivalence of detrital minerals and grain-size dependence of sediment composition." Earth and Planetary Science Letters 273.1 (2008): 138-151.

#### See Also

minsorting

# **Examples**

```
data(endmembers,densities)
ophiolite <- subset(endmembers,select="ophiolite")
plot(minsorting(ophiolite,densities,by=0.05))</pre>
```

get.f

Calculate the largest fraction that is likely to be missed

# **Description**

For a given sample size, returns the largest fraction which has been sampled with  $(1-p) \times 100 \%$  likelihood.

# Usage

```
get.f(n, p = 0.05)
```

### **Arguments**

n the number of grains in the detrital sample

p the required level of confidence

# Value

the largest fraction that is sampled with at least (1-p) x 100% certainty

# References

Vermeesch, Pieter. "How many grains are needed for a provenance study?." Earth and Planetary Science Letters 224.3 (2004): 441-451.

get.n 15

### **Examples**

```
print(get.f(60))
print(get.f(117))
```

get.n

Calculate the number of grains required to achieve a desired level of sampling resolution

# Description

Returns the number of grains that need to be analysed to decrease the likelihood of missing any fraction greater than a given size below a given level.

# Usage

```
get.n(p = 0.05, f = 0.05)
```

### **Arguments**

p	the probability that all n grains in the sample have missed at least one fraction of size f
f	the size of the smallest resolvable fraction (0 <f<1)< td=""></f<1)<>
n,	the number of grains in the sample

### Value

the number of grains needed to reduce the chance of missing at least one fraction f of the total population to less than p

# References

Vermeesch, Pieter. "How many grains are needed for a provenance study?." Earth and Planetary Science Letters 224.3 (2004): 441-451.

```
# number of grains required to be 99% that no fraction greater than 5% was missed: print(get.n(0.01)) # number of grains required to be 90% that no fraction greater than 10% was missed: print(get.n(p=0.1,f=0.1))
```

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get.p

Calculate the probability of missing a given population fraction

#### **Description**

For a given sample size, returns the likelihood of missing any fraction greater than a given size

# Usage

```
get.p(n, f = 0.05)
```

# **Arguments**

n the number of grains in the detrital sample

f the size of the smallest resolvable fraction (0 < f < 1)

### Value

the probability that all n grains in the sample have missed at least one fraction of size f

#### References

Vermeesch, Pieter. "How many grains are needed for a provenance study?." Earth and Planetary Science Letters 224.3 (2004): 441-451.

# **Examples**

```
print(get.p(60))
print(get.p(117))
```

GPA

Generalised Procrustes Analysis of configurations

# Description

Given a number of (2D) configurations, this function uses a combination of transformations (reflections, rotations, translations and scaling) to find a 'consensus' configuration which best matches all the component configurations in a least-squares sense.

# Usage

```
GPA(X, scale = TRUE)
```

### **Arguments**

X a list of dissimilarity matrices

scale boolean flag indicating if the transformation should include the scaling operation

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# Value

a two column vector with the coordinates of the group configuration

#### See Also

procrustes

indscal

Individual Differences Scaling of provenance data

# Description

Performs 3-way Multidimensional Scaling analysis using Carroll and Chang (1970)'s INdividual Differences SCALing method as implemented using De Leeuw and Mair (2011)'s stress majorization algorithm.

# Usage

```
indscal(..., type = "ordinal")
```

### **Arguments**

 $\dots \qquad \qquad \text{a sequence of datasets of class distributional or compositional} \\$ 

type is either "ratio" or "ordinal"

# Value

an object of class INDSCAL, i.e. a list containing the following items:

delta: Observed dissimilarities

obsdiss: List of observed dissimilarities, normalized

confdiss: List of configuration dissimilarities conf: List of matrices of final configurations

gspace: Joint configurations aka group stimulus space

cweights: Configuration weights

stress: Stress-1 value spp: Stress per point

sps: Stress per subject (matrix) ndim: Number of dimensions model: Type of smacof model niter: Number of iterations nobj: Number of objects 18 KDE

### Author(s)

Jan de Leeuw and Patrick Mair

### References

de Leeuw, J., & Mair, P. (2009). Multidimensional scaling using majorization: The R package smacof. Journal of Statistical Software, 31(3), 1-30, < http://www.jstatsoft.org/v31/i03/>

# **Examples**

```
data(Namib)
plot(indscal(Namib$DZ,Namib$HM))
```

**KDE** 

Create a kernel density estimate

# **Description**

Turns a vector of numbers into an object of class KDE using a combination of the Botev (2010) bandwidth selector and the Abramson (1982) adaptive kernel bandwidth modifier.

# Usage

```
KDE(x, from = NA, to = NA, bw = NA, adaptive = TRUE, log = FALSE, n = 512, ...)
```

# **Arguments**

X	a vector of numbers
from	minimum age of the time axis. If NULL, this is set automatically
to	maximum age of the time axis. If NULL, this is set automatically
bw	the bandwidth of the KDE. If NULL, bw will be calculated automatically using botev()
adaptive	boolean flag controlling if the adaptive KDE modifier of Abramson (1982) is used
log	transform the ages to a log scale if TRUE
n	horizontal resolution of the density estimate
• • •	optional arguments to be passed on to density

#### Value

an object of class KDE, i.e. a list containing the following items:

x: horizontal plot coordinates

y: vertical plot coordinates

bw: the base bandwidth of the density estimate

ages: the data values from the input to the KDE function

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# See Also

**KDEs** 

# **Examples**

```
data(Namib)
samp <- Namib$DZ$x[['N1']]
dens <- KDE(samp,0,3000,kernel="epanechnikov")
plot(dens)</pre>
```

KDEs

Generate an object of class KDEs

# Description

Convert a dataset of class distributional into an object of class KDEs for further processing by the summaryplot function.

# Usage

```
KDEs(x, from = NA, to = NA, bw = NA, samebandwidth = TRUE,
   adaptive = TRUE, normalise = FALSE, log = FALSE, n = 512, ...)
```

# Arguments

X	an object of class distributional
from	minimum limit of the x-axis.
to	maximum limit of the x-axis.
bw	the bandwidth of the kernel density estimates. If $bw = NA$ , the bandwidth will be set automatically using botev()
samebandwidth	boolean flag indicating whether the same bandwidth should be used for all samples. If samebandwidth = TRUE and bw = NULL, then the function will use the median bandwidth of all the samples.
adaptive	boolean flag switching on the adaptive bandwidth modifier of Abramson (1982)
normalise	boolean flag indicating whether or not the KDEs should all integrate to the same value.
log	boolean flag indicating whether the data should by plotted on a logarithmic scale.
n	horizontal resolution of the density estimates
	optional parameters to be passed on to density

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### Value

an object of class KDEs, i.e. a list containing the following items:

kdes: a named list with objects of class KDE from: the beginning of the common time scale

to: the end of the common time scale

themax: the maximum probability density of all the KDEs

pch: the plot symbol to be used by plot.KDEs xlabel: the x-axis label to be used by plot.KDEs

# See Also

**KDE** 

### **Examples**

```
data(Namib)
KDEs <- KDEs(Namib$DZ,0,3000,pch=NA)
summaryplot(KDEs,ncol=3)</pre>
```

KS.diss

Kolmogorov-Smirnov dissimilarity

# Description

Returns the Kolmogorov-Smirnov dissimilarity between two samples

### Usage

```
KS.diss(x, y)
```

# Arguments

x the first sample as a vectory the second sample as a vector

# Value

a scalar value representing the maximum vertical distance between the two cumulative distributions

```
data(Namib)
print(KS.diss(Namib$DZ$x[['N1']],Namib$DZ$x[['T8']]))
```

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Kuiper.diss

Kuiper dissimilarity

# **Description**

Returns the Kuiper dissimilarity between two samples

# Usage

```
Kuiper.diss(x, y)
```

# Arguments

x the first sample as a vector

y the second sample as a vector

### Value

a scalar value representing the sum of the maximum vertical distances above and below the cumulative distributions of x and y

# **Examples**

```
data(Namib)
print(Kuiper.diss(Namib$DZ$x[['N1']],Namib$DZ$x[['T8']]))
```

lines.ternary

Ternary line plotting

# **Description**

Add lines to an existing ternary diagram

# Usage

```
## S3 method for class 'ternary' lines(x, ...)
```

### **Arguments**

x an object of class ternary, or a three-column data frame or matrix

... optional arguments to the generic lines function

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# **Examples**

```
tern <- ternary(Namib$PT,'Q',c('KF','P'),c('Lm','Lv','Ls'))
plot(tern,pch=21,bg='red',labels=NULL)
middle <- matrix(c(0.01,0.49,0.01,0.49,0.98,0.02),2,3)
lines(ternary(middle))</pre>
```

MDS

Multidimensional Scaling

# **Description**

Performs classical or nonmetric Multidimensional Scaling analysis of provenance data

# Usage

```
MDS(x, ...)
## Default S3 method:
MDS(x, fn, classical = FALSE, k = 2,
   bootstrap = FALSE, nb = 10, ...)

## S3 method for class 'compositional'
MDS(x, classical = FALSE, k = 2,
   bootstrap = FALSE, nb = 10, ...)

## S3 method for class 'counts'
MDS(x, classical = FALSE, k = 2, bootstrap = FALSE,
   nb = 10, ...)

## S3 method for class 'distributional'
MDS(x, classical = FALSE, k = 2,
   bootstrap = FALSE, nb = 10, ...)

## S3 method for class 'distributional'
MDS(x, classical = FALSE, k = 2,
   bootstrap = FALSE, nb = 10, ...)
```

### **Arguments**

X	an object of class distributional, compositional, counts or diss
	optional arguments to be passed onto diss (if $x$ is of class compositional or distributional) or onto cmdscale or isoMDS (if $x$ is of class dist).
fn	a dissimilarity function. One of either ${\tt diss.compositional}, {\tt diss.counts}, {\tt or} {\tt diss.distributional}$
classical	boolean flag indicating whether classical (TRUE) or nonmetric (FALSE) MDS should be used
k	the desired dimensionality of the solution

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bootstrap resample the data to calculate confidence polygons for the MDS configuration nb number of bootstrap resamples

#### Value

an object of class MDS, i.e. a list containing the following items:

points: a two column vector of the fitted configuration

classical: a boolean flag indicating whether the MDS configuration was obtained by classical (TRUE) or nonmetric (FALSE) MDS.

diss: the dissimilarity matrix used for the MDS analysis

stress: (only if classical=TRUE) the final stress achieved (in percent)

# **Examples**

```
data(Namib)
plot(MDS(Namib$Major,classical=TRUE))
```

minsorting

Assess settling equivalence of detrital components

# Description

Models grain size distribution of minerals and rock fragments of different densities

# Usage

```
minsorting(X, dens, sname = NULL, phi = 2, sigmaphi = 1, medium = "freshwater", from = -2.25, to = 5.5, by = 0.25)
```

# **Arguments**

Х	an object of class compositional
dens	a vector of mineral and rock densities
sname	sample name if unspecified, the first sample of the dataset will be used
phi	the mean grain size of the sample in Krumbein's phi units
sigmaphi	the standard deviation of the grain size distirbution, in phi units
medium	the transport medium, one of either "air", "freshwater" or "seawater"
from	the minimum grain size to be evaluated, in phi units
to	the maximum grain size to be evaluated, in phi units
by	the grain size interval of the output table, in phi units

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#### Value

```
an object of class minsorting, i.e. a list with two tables:
mfract: the grain size distribution of each mineral (sum of the columns = 1)
mcomp: the composition of each mineral (sum of the rows = 1)
```

#### Author(s)

Alberto Resentini and Pieter Vermeesch

#### References

Resentini, A, Malusa, M G and Garzanti, E. "MinSORTING: An Excel worksheet for modelling mineral grain-size distribution in sediments, with application to detrital geochronology and provenance studies." Computers & Geosciences 59 (2013): 90-97.

Garzanti, E, Ando, S and Vezzoli, G. "Settling equivalence of detrital minerals and grain-size dependence of sediment composition." Earth and Planetary Science Letters 273.1 (2008): 138-151.

### See Also

restore

### **Examples**

Namib

An example dataset

# Description

A large dataset of provenance data from Namibia comprised of 14 sand samples from the Namib Sand Sea and 2 samples from the Orange River.

#### **Details**

Namib is a list containing the following 6 items:

DZ: a distributional dataset containing the zircon U-Pb ages for ca. 100 grains from each sample, as well as their (1-sigma) analytical uncertainties.

PT: a compositional dataset with the bulk petrography of the samples, i.e. the quartz ('Q'), K-feldspar ('KF'), plagioclase ('P'), and lithic fragments of metamorphic ('Lm'), volcanic ('Lv') and sedimentary ('Ls') origin.

HM: a compositional dataset containing the heavy mineral composition of the samples, comprised of zircon ('zr'), tourmaline ('tm'), rutile ('rt'), Ti-oxides ('TiOx'), titanite ('sph'), apatite ('ap'),

PCA 25

epidote ('ep'), garnet ('gt'), staurolite ('st'), andalusite ('and'), kyanite ('ky'), sillimanite ('sil'), amphibole ('amp'), clinopyroxene ('cpx') and orthopyroxene ('opx').

PTHM: a compositional dataset combining the variables contained in PT and HM plus 'mica', 'opaques', 'turbids' and 'other' transparent heavy minerals ('LgM'), normalised to 100.

Major: a compositional dataset listing the concentrations (in wt TiO2, P2O5 and MnO.

Trace: a compositional data listing the concentrations (in ppm) of Rb, Sr, Ba, Sc, Y, La, Ce, Pr, Nd, Sm, Gd, Dy, Er, Yb, Th, U, Zr, Hf, V, Nb, Cr, Co, Ni, Cu, Zn, Ga and Pb.

# Author(s)

Pieter Vermeesch and Eduardo Garzanti

#### References

Vermeesch, P. and Garzanti, E., Making geological sense of 'Big Data' in sedimentary provenance analysis, Chemical Geology 409 (2015) 20-27

### **Examples**

```
data(Namib)
samp <- Namib$DZ$x[['N1']]
dens <- KDE(samp,0,3000)
plot(dens)</pre>
```

PCA

Principal Component Analysis

# Description

Performs PCA of compositional data using a centred logratio distance

# Usage

```
PCA(x, ...)
```

#### **Arguments**

```
x an object of class compositional... optional arguments to R's princomp function
```

#### Value

an object of classes PCA, which is synonymous to the stats packages' prcomp class.

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### **Examples**

```
data(Namib)
plot(MDS(Namib$Major,classical=TRUE))
dev.new()
plot(PCA(Namib$Major),asp=1)
print("This example demonstrates the equivalence of classical MDS and PCA")
```

plot.CA

Point-counting biplot

# Description

Plot the results of a correspondence analysis as a biplot

# Usage

```
## S3 method for class 'CA'
plot(x, levels = NULL, labelcol = c("black", "blue"),
  vectorcol = "red", ...)
```

### **Arguments**

x an object of class CA

levels a vector of length nrow(dat\$x) with values that are to be used for the colour

scale

labelcol two element vector with the colours that are to be assigned to the labels of

the samples corresponding to the minimum and maximum value of levels. If

levels==NULL, then the first item of the vector is used for all sample labels.

vectorcol colour of the vector loadings for the variables

... optional arguments of the generic biplot function

# See Also

CA

```
data(Namib)
plot(CA(Namib$PT))
```

plot.compositional 27

plot.compositional

Plot a pie chart

### **Description**

Plots an object of class compositional as a pie chart

### Usage

```
## S3 method for class 'compositional'
plot(x, sname, annotate = TRUE, colmap = NULL,
    ...)
```

#### **Arguments**

x an object of class compositional

sname the sample name

annotate a boolean flag controlling if the pies of the pie-chart should be labeled

colmap an optional string with the name of one of R's built-in colour palettes (e.g.,

heat.colors, terrain.colors, topo.colors, cm.colors), which are to be used for plot-

ting the data.

... optional parameters to be passed on to the graphics object

### **Examples**

```
data(Namib)
plot(Namib$Major,'N1',colmap='heat.colors')
```

plot.distributional

Plot continuous data as histograms or cumulative age distributions

# **Description**

Plot one or several samples from a distributional dataset as a histogram or Cumulative Age Distributions (CAD).

### Usage

```
## S3 method for class 'distributional'
plot(x, snames = NULL, annotate = TRUE,
    CAD = FALSE, pch = NA, verticals = TRUE, colmap = NULL, ...)
```

28 plot.GPA

# Arguments

Х	an object of class distributional
snames	a string or a vector of string with the names of the samples that need plotting if snames is a vector, then the function will default to a CAD.
annotate	boolean flag indicating whether the x- and y-axis should be labeled
CAD	boolean flag indicating whether the data should be plotted as a cumulative age distribution or a histogram. For multi-sample plots, the function will override this value with TRUE.
pch	an optional symbol to mark the sample points along the CAD
verticals	boolean flag indicating if the horizontal lines of the CAD should be connected by vertical lines
colmap	an optional string with the name of one of R's built-in colour palettes (e.g., heat.colors, terrain.colors, topo.colors, cm.colors), which are to be used for plotting the data.
	optional arguments to the generic plot function

# **Examples**

```
data(Namib)
plot(Namib$DZ,c('N1','N2'))
```

plot.GPA

Plot a Procrustes configuration

# Description

Plots the group configuration of a Generalised Procrustes Analysis

# Usage

```
## S3 method for class 'GPA'
plot(x, pch = NA, pos = NULL, col = "black",
  bg = "white", cex = 1, ...)
```

# Arguments

X	an object of class GPA
pch	plot symbol
pos	position of the sample labels relative to the plot symbols if pch != NA
col	plot colour (may be a vector)
bg	background colour (may be a vector)
cex	relative size of plot symbols
	optional arguments to the generic plot function

plot.INDSCAL 29

### See Also

procrustes

# **Examples**

```
data(Namib)
GPA <- procrustes(Namib$DZ,Namib$HM)
coast <- c('N1','N2','N3','N10','N11','N12','T8','T13')
snames <- names(Namib$DZ)
bgcol <- rep('yellow',length(snames))
bgcol[which(snames %in% coast)] <- 'red'
plot(GPA,pch=21,bg=bgcol)</pre>
```

plot.INDSCAL

Plot an INDSCAL group configuration and source weights

# Description

Given an object of class INDSCAL, generates two plots: the group configuration and the subject weights. Together, these describe a 3-way MDS model.

# Usage

```
## S3 method for class 'INDSCAL'
plot(x, asp = 1, pch = NA, pos = NULL,
   col = "black", bg = "white", cex = 1, xlab = "X", ylab = "Y",
   xaxt = "n", yaxt = "n", ...)
```

# **Arguments**

x	an object of class INDSCAL
asp	the aspect ratio of the plot
pch	plot symbol (may be a vector)
pos	position of the sample labels relative to the plot symbols if pch $!= NA$
col	plot colour (may be a vector)
bg	background colour (may be a vector)
cex	relative size of plot symbols
xlab	a string with the label of the x axis
ylab	a string with the label of the y axis
xaxt	if = 'y', adds ticks to the x axis
yaxt	if = 'y', adds ticks to the y axis
	optional arguments to the generic plot function

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# See Also

indscal

# **Examples**

```
data(Namib)
coast <- c('N1','N2','N3','N10','N11','N12','T8','T13')
snames <- names(Namib$DZ)
pch <- rep(21,length(snames))
pch[which(snames %in% coast)] <- 22
plot(indscal(Namib$DZ,Namib$HM),pch=pch)</pre>
```

plot.KDE

Plot a kernel density estimate

# Description

Plots an object of class KDE

# Usage

```
## S3 method for class 'KDE'
plot(x, pch = "|", xlab = "age [Ma]", ylab = "", ...)
```

# Arguments

X	an object of class KDE
pch	the symbol used to show the samples. May be a vector. Set pch = $NA$ to turn them off.
xlab	the label of the x-axis
ylab	the label of the y-axis
	optional parameters to be passed on to the graphics object

### See Also

KDE

```
data(Namib)
samp <- Namib$DZ$x[['N1']]
dens <- KDE(samp,from=0,to=3000)
plot(dens)</pre>
```

plot.KDEs 31

plot.KDEs

Plot one or more kernel density estimates

### **Description**

Plots an object of class KDEs

# Usage

```
## S3 method for class 'KDEs'
plot(x, sname = NA, annotate = TRUE, pch = "|", ...)
```

### **Arguments**

x an object of class KDEs
sname optional sample name. If sname=NA, all samples are shown on a summary plot
annotate add a time axis?

pch symbol to be used to mark the sample points along the x-axis. Change to NA to
omit

optional parameters to be passed on to the summaryplot function

#### See Also

KDEs summaryplot

### **Examples**

```
data(Namib)
kdes <- KDEs(Namib$DZ)
plot(kdes,ncol=2)</pre>
```

plot.MDS

Plot an MDS configuration

# **Description**

Plots the coordinates of a multidimensional scaling analysis as an X-Y scatter plot or 'map' and, if x-classical = FALSE, a Shepard plot.

### Usage

```
## S3 method for class 'MDS'
plot(x, nnlines = FALSE, pch = NA, pos = NULL,
   cex = 1, col = "black", bg = "white", oma = rep(1, 4),
   mar = rep(2, 4), mgp = c(2, 1, 0), xpd = NA, ...)
```

32 plot.minsorting

### **Arguments**

x	an object of class MDS
nnlines	if TRUE, draws nearest neighbour lines
pch	plot character (see ?plot for details). May be a vector.
pos	position of the sample labels relative to the plot symbols if pch != NA
cex	relative size of plot symbols (see ?par for details)
col	plot colour (may be a vector)
bg	background colour (may be a vector)
oma	A vector of the form $c(bottom, left, top, right)$ giving the size of the outer margins in lines of text.
mar	A numerical vector of the form c(bottom,left,top,right) that gives the number of lines of margin to be specified on the four sides of the plot.
mgp	The margin line (in mex units) for the axis title, axis labels and axis line. See ?par for further details.
xpd	A logical value or NA. See ?par for further details.
	optional arguments to the generic plot function

# See Also

**MDS** 

# Examples

```
data(Namib)
mds <- MDS(Namib$DZ)
coast <- c('N1','N2','N3','N10','N11','N12','T8','T13')
snames <- names(Namib$DZ)
bgcol <- rep('yellow',length(snames))
bgcol[which(snames %in% coast)] <- 'red'
plot(mds,pch=21,bg=bgcol)</pre>
```

plot.minsorting

Plot inferred grain size distributions

# Description

Plot the grain size distributions of the different minerals under consideration

# Usage

```
## S3 method for class 'minsorting'
plot(x, cumulative = FALSE, components = NULL,
    ...)
```

plot.PCA 33

# Arguments

x an object of class minsorting
cumulative boolean flag indicating whether the grain size distribution should be plotted as a
density or cumulative probability curve.

components string or list of strings with the names of a subcomposition that needs plotting
optional parameters to be passed on to graphics::matplot (see ?par for details)

### See Also

minsorting

# **Examples**

plot.PCA

Compositional biplot

# Description

Plot the results of a principal components analysis as a biplot

### Usage

```
## S3 method for class 'PCA'
plot(x, levels = NULL, labelcol = c("black", "blue"),
  vectorcol = "red", choices = 1L:2L, scale = 1, pc.biplot = FALSE,
  ...)
```

# **Arguments**

X	an object of class PCA
levels	a vector of length $nrow(dat$x)$ with values that are to be used for the colour scale
labelcol	two element vector with the colours that are to be assigned to the labels of the samples corresponding to the minimum and maximum value of levels. If levels==NULL, then the first item of the vector is used for all sample labels.
vectorcol	colour of the vector loadings for the variables
choices	see the help pages of the biplot function.
scale	see the help pages of the biplot function.
pc.biplot	see the help pages of the biplot function.
	optional arguments of the generic biplot function

34 plot.ternary

# See Also

**PCA** 

# **Examples**

```
data(Namib)
plot(PCA(Namib$Major))
```

plot.ternary

Plot a ternary diagram

# Description

Plots triplets of compositional data on a ternary diagram

# Usage

```
## S3 method for class 'ternary'
plot(x, type = "grid", pch = NA, pos = NULL,
  labels = names(x), showpath = FALSE, bg = NA,
  col = "cornflowerblue", ticks = seq(0, 1, 0.25), ticklength = 0.02,
  lty = 2, lwd = 1, ...)
```

# Arguments

х	an object of class ternary, or a three-column data frame or matrix
type	adds annotations to the ternary diagram, one of either empty, $\tt grid, QFL.descriptive, QFL.folk$ or $\tt QFL.dickinson$
pch	plot character, see ?par for details (may be a vector)
pos	position of the sample labels relative to the plot symbols if pch != NA
labels	vector of strings to be added to the plot symbols
showpath	if x has class SRDcorrected, and showpath==TRUE, the intermediate values of the SRD correction will be plotted on the ternary diagram as well as the final composition
bg	background colour for the plot symbols (may be a vector)
col	colour to be used for the background lines (if applicable)
ticks	vector of tick values between 0 and 1
ticklength	number between 0 and 1 to mark the length of the ticks
lty	line type for the annotations (see type)
lwd	line thickness for the annotations

optional arguments to the generic points function

points.ternary 35

### See Also

ternary

# **Examples**

```
data(Namib)
tern <- ternary(Namib$PT,'Q',c('KF','P'),c('Lm','Lv','Ls'))
plot(tern,type='QFL.descriptive',pch=21,bg='red',labels=NULL)</pre>
```

points.ternary

Ternary point plotting

# Description

Add points to an existing ternary diagram

# Usage

```
## S3 method for class 'ternary'
points(x, ...)
```

### **Arguments**

x an object of class ternary, or a three-column data frame or matrix optional arguments to the generic points function

# **Examples**

```
tern <- ternary(Namib$PT,'Q',c('KF','P'),c('Lm','Lv','Ls'))
plot(tern,pch=21,bg='red',labels=NULL)
# add the geometric mean composition as a yellow square:
gmean <- ternary(exp(colMeans(log(tern$x))))
points(gmean,pch=22,bg='yellow')</pre>
```

procrustes

Generalised Procrustes Analysis of provenance data

# **Description**

Given a number of input datasets, this function performs an MDS analysis on each of these and the feeds the resulting configurations into the GPA() function.

### Usage

```
procrustes(...)
```

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# Arguments

... a sequence of datasets of classes distributional, n counts and compositional

### Value

```
an object of class GPA, i.e. a list containing the following items:
points: a two column vector with the coordinates of the group configuration
labels: a list with the sample names
```

### Author(s)

Pieter Vermeesch

#### References

```
Gower, J.C. (1975). Generalized Procrustes analysis, Psychometrika, 40, 33-50.
```

#### See Also

**GPA** 

# **Examples**

```
data(Namib)
gpa <- procrustes(Namib$DZ,Namib$HM)
plot(gpa)</pre>
```

provenance

Menu-based interface for provenance

# Description

For those less familiar with the syntax of the R programming language, the provenance() function provides a user-friendly way to access the most important functionality in the form of a menu-based query interface. Further details and examples are provided on <a href="http://provenance.london-geochron.com">http://provenance.london-geochron.com</a>

provenance provides statistical tools to interpret large amounts of distributional (single grain analyses) and compositional (mineralogical and bulk chemical) data from the command line, or using a menu-based user interface.

# Usage

```
provenance()
```

radialplot 37

#### **Details**

A list of documented functions may be viewed by typing help(package='provenance'). Detailed instructions are provided at http://provenance.london-geochron.com and in the Sedimentary Geology paper by Vermeesch, Resentini and Garzanti (2016).

#### Author(s)

Pieter Vermeesch

Maintainer: Pieter Vermeesch <p.vermeesch@ucl.ac.uk>

## References

Vermeesch, P., Resentini, A. and Garzanti, E., an R package for statistical provenance analysis, Sedimentary Geology, doi:10.1016/j.sedgeo.2016.01.009.

Vermeesch, P., Resentini, A. and Garzanti, E., 2016, An R package for statistical provenance analysis, Sedimentary Geology, 336, 14-25.

#### See Also

```
\verb|http://provenance.london-geochron.com|\\
```

Useful links:

• http://provenance.london-geochron.com

radialplot

Visualise point-counting data on a radial plot

## Description

Implementation of a graphical device developed by Rex Galbraith to display several estimates of the same quantity that have different standard errors.

#### Usage

```
radialplot(x, num = 1, den = 2, from = NA, to = NA, t0 = NA,
  sigdig = 2, show.numbers = FALSE, pch = 21, levels = NA,
  clabel = "", bg = c("white", "red"), title = TRUE, ...)
```

## **Arguments**

х	an object of class counts
num	index or name of the numerator variable
den	index or name of the denominator variable
from	minimum limit of the radial scale
to	maximum limit of the radial scale

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t0 central value

sigdig the number of significant digits of the numerical values reported in the title of

the graphical output.

show.numbers boolean flag (TRUE to show sample numbers)

pch plot character (default is a filled circle)

levels a vector with additional values to be displayed as different background colours

of the plot symbols.

clabel label of the colour legend

bg a vector of two background colours for the plot symbols. If levels=NA, then

only the first colour is used. If levels is a vector of numbers, then bg is used to

construct a colour ramp.

title add a title to the plot?

... additional arguments to the generic points function

#### **Details**

The radial plot (Galbraith, 1988, 1990) is a graphical device that was specifically designed to display heteroscedastic data, and is constructed as follows. Consider a set of dates  $\{t_1, ..., t_i, ..., t_n\}$  and uncertainties  $\{s[t_1], ..., s[t_i], ..., s[t_n]\}$ . Define  $z_i = z[t_i]$  to be a transformation of  $t_i$  (e.g.,  $z_i = log[t_i]$ ), and let  $s[z_i]$  be its propagated analytical uncertainty (i.e.,  $s[z_i] = s[t_i]/t_i$  in the case of a logarithmic transformation). Create a scatterplot of  $(x_i, y_i)$  values, where  $x_i = 1/s[z_i]$  and  $y_i = (z_i - z_o)/s[z_i]$ , where  $z_o$  is some reference value such as the mean. The slope of a line connecting the origin of this scatterplot with any of the  $(x_i, y_i)$ s is proportional to  $z_i$  and, hence, the date  $t_i$ . These dates can be more easily visualised by drawing a radial scale at some convenient distance from the origin and annotating it with labelled ticks at the appropriate angles. While the angular position of each data point represents the date, its horizontal distance from the origin is proportional to the precision. Imprecise measurements plot on the left hand side of the radial plot, whereas precise age determinations are found further towards the right. Thus, radial plots allow the observer to assess both the magnitude and the precision of quantitative data in one glance.

#### References

Galbraith, R.F., 1988. Graphical display of estimates having differing standard errors. Technometrics, 30(3), pp.271-281.

Galbraith, R.F., 1990. The radial plot: graphical assessment of spread in ages. International Journal of Radiation Applications and Instrumentation. Part D. Nuclear Tracks and Radiation Measurements, 17(3), pp.207-214.

Galbraith, R.F. and Laslett, G.M., 1993. Statistical models for mixed fission track ages. Nuclear Tracks and Radiation Measurements, 21(4), pp.459-470.

```
data(Namib)
radialplot(Namib$PT,components=c('Q','P'))
```

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read.compositional

Read a .csv file with compositional data

# Description

Reads a data table containing compositional data (e.g. chemical concentrations)

## Usage

```
read.compositional(fname, method = NULL, colmap = "rainbow", ...)
```

## **Arguments**

fname	a string with the path to the .csv file
method	either "bray" (for the Bray-Curtis distance) or "aitchison" (for Aitchison's central logratio distance). If omitted, the function defaults to 'aitchison', unless there are zeros present in the data.
colmap	an optional string with the name of one of R's built-in colour palettes (e.g., heat.colors, terrain.colors, topo.colors, cm.colors), which are to be used for plotting the data.
	optional arguments to the built-in read.csv function

#### Value

```
an object of class compositional, i.e. a list with the following items:

x: a data frame with the samples as rows and the categories as columns

method: either "aitchison" (for Aitchison's centred logratio distance) or "bray" (for the Bray-Curtis distance)

colmap: the colour map provided by the input argument
```

```
fname <- system.file("Major.csv",package="provenance")
Major <- read.compositional(fname)
plot(PCA(Major))</pre>
```

40 read.densities

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read	. 1	ററ	H.	ın	т.	S

Read a .csv file with point-counting data

# Description

Reads a data table containing point-counting data (e.g. petrographic, heavy mineral, palaeontological or palynological data)

#### Usage

```
read.counts(fname, method = "chisq", colmap = "rainbow", ...)
```

# **Arguments**

fname a string with the path to the .csv file

method either "chisq" (for the chi-square distance) or "bray" (for the Bray-Curtis dis-

tance)

colmap an optional string with the name of one of R's built-in colour palettes (e.g.,

heat.colors, terrain.colors, topo.colors, cm.colors), which are to be used for plot-

ting the data.

... optional arguments to the built-in read.csv function

#### Value

an object of class counts, i.e. a list with the following items:

x: a data frame with the samples as rows and the categories as columns
colmap: the colour map provided by the input argument

#### **Examples**

```
fname <- system.file("HM.csv",package="provenance")
Major <- read.counts(fname)
#plot(PCA(HM))</pre>
```

read.densities

Read a .csv file with mineral and rock densities

#### **Description**

Reads a data table containing densities to be used for hydraulic sorting corrections (minsorting and srd functions)

#### Usage

```
read.densities(fname, ...)
```

read.distributional 41

## **Arguments**

fname a string with the path to the .csv file

... optional arguments to the built-in read.csv function

#### Value

a vector with mineral and rock densities

# **Examples**

```
data(Namib,densities)
N8 <- subset(Namib$HM,select="N8")
distribution <- minsorting(N8,densities,phi=2,sigmaphi=1,medium="air",by=0.05)
plot(distribution)</pre>
```

read.distributional

Read a .csv file with continuous (detrital zircon) data

# Description

Reads a data table containing continuous data (e.g. detrital zircon ages)

#### Usage

```
read.distributional(fname, errorfile = NA, method = "KS",
    xlab = "age [Ma]", colmap = "rainbow", ...)
```

#### **Arguments**

fname	the path of a .csv file with the input data, arranged in columns.
errorfile	the (optional) path of a .csv file with the standard errors of the input data, arranged by column in the same order as fname. Must be specified if the data are to be compared with the Sircombe-Hazelton dissimilarity.
method	an optional string specifying the dissimilarity measure which should be used for comparing this with other datasets. Should be one of either "KS" (for Kolmogorov-Smirnov) or "SH" (for Sircombe and Hazelton). If method = "SH", then errorfile should be specified. If method = "SH" and errorfile is unspecified, then the program will default back to the Kolmogorov-Smirnov dissimilarity.
xlab	an optional string specifying the nature and units of the data. This string is used to label kernel density estimates.
colmap	an optional string with the name of one of R's built-in colour palettes (e.g., heat.colors, terrain.colors, topo.colors, cm.colors), which are to be used for plotting the data.
	optional arguments to the built-in read.csv function

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#### Value

```
an object of class distributional, i.e. a list with the following items:
```

x: a named list of vectors containing the numerical data for each sample

err: an (optional) named list of vectors containing the standard errors of x

method: either "KS" (for Kolmogorov-Smirnov), "Kuiper" (for the Kuiper statistic) or "SH" (for Sircombe Hazelton)

breaks: a vector with the locations of the histogram bin edges

xlab: a string containing the label to be given to the x-axis on all plots

colmap: the colour map provided by the input argument

#### **Examples**

```
agefile <- system.file("DZ.csv",package="provenance")
errfile <- system.file("DZerr.csv",package="provenance")
DZ <- read.distributional(agefile,errfile)
plot(KDE(DZ$x$N1))</pre>
```

restore

Undo the effect of hydraulic sorting

#### **Description**

Restore the detrital composition back to a specified source rock density (SRD)

#### Usage

```
restore(X, dens, target = 2.71)
```

## **Arguments**

X an object of class compositional dens a vector of rock and mineral densities

target the target density (in g/cm3)

#### Value

an object of class SRDcorrected, i.e. an object of class compositional which is a daughter of class compositional containing the restored composition, plus one additional member called restoration, containing the intermediate steps of the SRD correction algorithm.

#### Author(s)

Alberto Resentini and Pieter Vermeesch

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#### References

Garzanti E, Ando, S and Vezzoli, G. "Settling equivalence of detrital minerals and grain-size dependence of sediment composition." Earth and Planetary Science Letters 273.1 (2008): 138-151.

#### See Also

minsorting

# Examples

SH.diss

Sircombe and Hazelton distance

## **Description**

Calculates Sircombe and Hazelton's L2 distance between the Kernel Functional Estimates (KFEs, not to be confused with Kernel Density Estimates!) of two samples with specified analytical uncertainties

#### Usage

```
SH.diss(x, i, j, c.con = 0)
```

#### **Arguments**

X	an object of class distributional
i	index of the first sample

j index of the second sample

c.con smoothing bandwidth of the kernel functional estimate

#### Value

a scalar value expressing the L2 distance between the KFEs of samples i and j

#### Author(s)

Keith Sircombe and Martin Hazelton

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#### References

Sircombe, K. N., and M. L. Hazelton. "Comparison of detrital zircon age distributions by kernel functional estimation." Sedimentary Geology 171.1 (2004): 91-111.

# See Also

KS.diss

# **Examples**

```
datfile <- system.file("DZ.csv",package="provenance")
errfile <- system.file("DZerr.csv",package="provenance")
DZ <- read.distributional(datfile,errfile)
d <- SH.diss(DZ,1,2)
print(d)</pre>
```

subset

Get a subset of provenance data

## **Description**

Return a subset of provenance data according to some specified indices

# Usage

```
## S3 method for class 'distributional'
subset(x, subset = NULL, select = NULL, ...)
## S3 method for class 'compositional'
subset(x, subset = NULL, components = NULL,
    select = NULL, ...)
## S3 method for class 'counts'
subset(x, subset = NULL, components = NULL,
    select = NULL, ...)
```

## **Arguments**

x	an object of class distributional
subset	logical expression indicating elements or rows to keep: missing values are taken as false.
select	a vector of sample names
	optional arguments for the generic subset function
components	categories to keep

summaryplot 45

#### Value

```
an object of class distributional
```

#### See Also

read.distributional

#### **Examples**

```
data(Namib)
coast <- c("N1","N2","T8","T13","N12","N13")
ZTRcoast <- subset(Namib$HM,select=coast,components=c('gt','cpx','ep'))
DZcoast <- subset(Namib$DZ,select=coast)
summaryplot(ZTRcoast,KDEs(DZcoast),ncol=2)</pre>
```

summaryplot

Joint plot of several provenance datasets

#### **Description**

Arranges kernel density estimates and pie charts in a grid format

#### Usage

```
summaryplot(..., ncol = 1, pch = NA)
```

#### Arguments

... a sequence of datasets of class compositional, KDEs, or distributional

ncol the number of columns

pch (optional) symbol to be used to mark the sample points along the x-axis of the

KDEs (if appropriate).

#### Value

a summary plot of all the data comprised of KDEs for the datasets of class KDEs, pie charts for those of class compositional and histograms for those of class distributional.

## See Also

**KDEs** 

```
data(Namib)
KDEs <- KDEs(Namib$DZ,0,3000)
summaryplot(KDEs,Namib$HM,Namib$PT,ncol=2)</pre>
```

46 ternary

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Define a ternary composition

# Description

Create an object of class ternary

# Usage

```
ternary(X, x = 1, y = 2, z = 3)
```

# Arguments

X	an object of class compositional OR a matrix or data frame with numerical data
X	string/number or a vector of strings/numbers indicating the variables/indices making up the first subcomposition of the ternary system.
у	second (set of) variables
z	third (set of) variables

## Value

```
an object of class ternary, i.e. a list containing:

x: a three column matrix (or vector) of ternary compositions.

and (if X is of class SRDcorrected)

restoration: a list of intermediate ternary compositions inherited from the SRD correction
```

## See Also

restore

```
data(Namib)
tern <- ternary(Namib$PT,c('Q'),c('KF','P'),c('Lm','Lv','Ls'))
plot(tern,type="QFL")</pre>
```

ternary.ellipse 47

ternary.ellipse

Ternary confidence ellipse

## **Description**

plot a  $100(1-\alpha)\%$  confidence region around the data or around its mean.

# Usage

## **Arguments**

an object of class ternary
 optional formatting arguments
 cutoff level for the confidence ellipse
 population
 show the standard deviation of the entire population or the standard error of the mean?

```
data(Namib)
tern <- ternary(Namib$Major,'CaO','Na2O','K2O')
plot(tern)
ternary.ellipse(tern)</pre>
```

48 text.ternary

text.ternary

Ternary text plotting

# Description

Add text an existing ternary diagram

# Usage

```
## S3 method for class 'ternary'
text(x, labels = 1:nrow(x$x), ...)
```

# Arguments

an object of class ternary, or a three-column data frame or matrix
 a character vector or expression specifying the text to be written
 optional arguments to the generic text function

```
data(Namib)
tern <- ternary(Namib$Major,'CaO','Na2O','K2O')
plot(tern,pch=21,bg='red',labels=NULL)
# add the geometric mean composition as a text label:
gmean <- ternary(exp(colMeans(log(tern$x))))
text(gmean,labels='geometric mean')</pre>
```

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