# Package 'lineup'

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ulate LOD score at physical position of each se	n a

# Description

For gene expression data with physical positions of the genes, calculate the LOD score at those positions to assess evidence for local eQTL.

# Usage

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```
calc.locallod(
  cross,
  pheno,
  pmark,
  addcovar = NULL,
  intcovar = NULL,
  verbose = TRUE,
  n.cores = 1
)
```

# Arguments

cross	An object of class "cross" containing data for a QTL experiment. See the help
	file for qtl::read.cross() in the R/qtl package (https://rqtl.org). There
	must be a phenotype named "id" or "ID" that contains the individual identifiers.
pheno	A data frame of phenotypes (generally gene expression data), stored as individuals x phenotypes. The row names must contain individual identifiers.
pmark	Pseudomarkers that are closest to the genes in pheno, as output by find.gene.pseudomarker().

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addcovar	Additive covariates passed to scanone().
intcovar	Interactive covariates passed to scanone().
verbose	If TRUE, print tracing information.
n.cores	Number of CPU cores to use in the calculations. With n.cores=0, parallel::detectCores() is used to detect the number of available cores.

#### **Details**

cross and pheno must contain exactly the same individuals in the same order. (Use findCommonID() to line them up.)

We consider the expression phenotypes in batches: those whose closest pseudomarker is the same.

We use Haley-Knott regression to calculate the LOD scores.

Actually, we use a bit of a contortion of the data to force the qtl::scanone() function in R/qtl to calculate the LOD score at a single position.

We omit any transcripts that map to the X chromosome; we can only handle autosomal loci for now.

#### Value

A vector of LOD scores. The names indicate the gene names (columns in pheno).

## Author(s)

Karl W Broman,

# See Also

```
find.gene.pseudomarker(), plotEGclass(), findCommonID(), disteg()
```

```
data(f2cross, expr1, genepos, pmap)
library(qtl)

# calc QTL genotype probabilities
f2cross <- calc.genoprob(f2cross, step=1)

# find nearest pseudomarkers
pmark <- find.gene.pseudomarker(f2cross, pmap, genepos, "prob")

# line up f2cross and expr1
id <- findCommonID(f2cross, expr1)

# calculate LOD score for local eQTL
locallod <- calc.locallod(f2cross[,id$first], expr1[id$second,], pmark)</pre>
```

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combinedist

Combine distance matrices into a single such

## **Description**

Combine multiple distance matrices into a single distance matrix providing an overall summary

# Usage

```
combinedist(..., method = c("median", "mean"))
```

## **Arguments**

... Set of distance matrices, as calculated by distee() or disteg().

method Indicates whether to summarize using the median or the mean.

## **Details**

The row and column names of the input distance matrices define the individual IDs.

If the input distance matrices all have an attribute "denom" (for denominator) and method="mean", we use a weighted mean, weighted by the denominators. This could be used to calculate an overall proportion.

#### Value

A distance matrix, with class "lineupdist". The individual IDs are in the row and column names.

#### Author(s)

## See Also

```
distee(), disteg(), summary.lineupdist()
```

```
library(qtl)
# load example data
data(f2cross, expr1, expr2, pmap, genepos)

# calculate QTL genotype probabilities
f2cross <- calc.genoprob(f2cross, step=1)

# find nearest pseudomarkers
pmark <- find.gene.pseudomarker(f2cross, pmap, genepos)</pre>
```

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```
# line up individuals
id1 <- findCommonID(f2cross, expr1)</pre>
id2 <- findCommonID(f2cross, expr2)</pre>
# calculate LOD score for local eQTL
locallod1 <- calc.locallod(f2cross[,id1$first], expr1[id1$second,], pmark)</pre>
locallod2 <- calc.locallod(f2cross[,id2$first], expr2[id2$second,], pmark)</pre>
# take those with LOD > 25
expr1s <- expr1[,locallod1>25,drop=FALSE]
expr2s <- expr2[,locallod2>25,drop=FALSE]
# calculate distance between individuals
      (prop'n mismatches between obs and inferred eQTL geno)
d1 <- disteg(f2cross, expr1s, pmark)</pre>
d2 <- disteg(f2cross, expr2s, pmark)</pre>
# combine distances
d <- combinedist(d1, d2)</pre>
# summary of problem samples
summary(d)
```

corbetw2mat

Calculate correlations between columns of two matrices

# Description

For matrices x and y, calculate the correlation between columns of x and columns of y.

## Usage

```
corbetw2mat(
   x,
   y,
   what = c("paired", "bestright", "bestpairs", "all"),
   corthresh = 0.9
)
```

## Arguments

x A numeric matrix.

y A numeric matrix with the same number of rows as x.

what Indicates which correlations to calculate and return. See value, below.

corthresh Threshold on correlations if what="bestpairs".

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#### **Details**

Missing values (NA) are ignored, and we calculate the correlation using all complete pairs, as in stats::cor() with use="pairwise.complete.obs".

#### Value

If what="paired", the return value is a vector of correlations, between columns of x and the corresponding column of y. x and y must have the same number of columns.

If what="bestright", we return a data frame of size ncol(x) by 3, with the *i*th row being the maximum correlation between column *i* of x and a column of y, and then the y-column index and y-column name with that correlation. (In case of ties, we give the first one.)

If what="bestpairs", we return a data frame with five columns, containing all pairs of columns (with one in x and one in y) with correlation  $\geq$  corthresh. Each row corresponds to a column pair, and contains the correlation and then the x- and y-column indices followed by the x- and y-column names.

If what="all", the output is a matrix of size ncol(x) by ncol(y), with all correlations between columns of x and columns of y.

#### Author(s)

Karl W Broman,

## See Also

```
distee(), findCommonID()
```

```
data(expr1, expr2)

# correlations with paired columns
r <- corbetw2mat(expr1, expr2)
# top 10, by absolute value
r[order(abs(r), decreasing=TRUE)[1:10]]

# all pairs of columns with correlation >= 0.8
r_allpairs <- corbetw2mat(expr1, expr2, what="bestpairs", corthresh=0.6)

# for each column in left matrix, most-correlated column in right matrix
r_bestright <- corbetw2mat(expr1, expr2, what="bestright")</pre>
```

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distee

Calculate distance between two gene expression data sets

# Description

Calculate a distance between all pairs of individuals for two gene expression data sets

# Usage

```
distee(
   e1,
   e2 = NULL,
   d.method = c("rmsd", "cor"),
   labels = c("e1", "e2"),
   verbose = TRUE
)
```

# Arguments

e1	Numeric matrix of gene expression data, as individuals x genes. The row and column names must contain individual and gene identifiers.
e2	(Optional) Like e1. An appreciable number of individuals and genes must be in common.
d.method	Calculate inter-individual distance as RMS difference or as correlation.
labels	Two character strings, to use as labels for the two data matrices in subsequent output.
verbose	if TRUE, give verbose output.

## **Details**

We calculate the pairwise distance between all individuals (rows) in e1 and all individuals in e2. This distance is either the RMS difference (d.method="rmsd") or the correlation (d.method="cor").

## Value

A matrix with nrow(e1) rows and nrow(e2) columns, containing the distances. The individual IDs are in the row and column names. The matrix is assigned class "lineupdist".

# Author(s)

```
pulldiag(), omitdiag(), summary.lineupdist(), plot2dist(), disteg(), corbetw2mat()
```

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## **Examples**

```
# load the example data
data(expr1, expr2)
# find samples in common
id <- findCommonID(expr1, expr2)</pre>
# calculate correlations between cols of x and cols of y
thecor <- corbetw2mat(expr1[id$first,], expr2[id$second,])</pre>
# subset at genes with corr > 0.8 and scale values
expr1s <- expr1[,thecor > 0.8]/1000
expr2s <- expr2[, thecor > 0.8]/1000
# calculate distance (using "RMS difference" as a measure)
d1 <- distee(expr1s, expr2s, d.method="rmsd", labels=c("1","2"))</pre>
# calculate distance (using "correlation" as a measure...really similarity)
d2 <- distee(expr1s, expr2s, d.method="cor", labels=c("1", "2"))</pre>
# pull out the smallest 8 self-self correlations
sort(pulldiag(d2))[1:8]
# summary of results
summary(d1)
summary(d2)
# plot histograms of RMS distances
plot(d1)
# plot histograms of correlations
plot(d2)
# plot distances against one another
plot2dist(d1, d2)
```

disteg

Calculate distance between two gene expression data sets

# **Description**

Calculate a distance between all pairs of individuals for two gene expression data sets

# Usage

```
disteg(
```

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```
cross,
pheno,
pmark,
min.genoprob = 0.99,
k = 20,
min.classprob = 0.8,
classprob2drop = 1,
repeatKNN = TRUE,
max.selfd = 0.3,
phenolabel = "phenotype",
weightByLinkage = FALSE,
map.function = c("haldane", "kosambi", "c-f", "morgan"),
verbose = TRUE
)
```

## **Arguments**

cross	An object of class	cross"	containing d	lata for a	QTL	experiment.	See the help

file for qtl::read.cross() in the R/qtl package (https://rqtl.org). There must be a phenotype named "id" or "ID" that contains the individual identifiers.

pheno A data frame of phenotypes (generally gene expression data), stored as individ-

uals x phenotypes. The row names must contain individual identifiers.

pmark Pseudomarkers that are closest to the genes in pheno, as output by find.gene.pseudomarker().

min.genoprob Threshold on genotype probabilities; if maximum probability is less than this,

observed genotype taken as NA.

k Number of nearest neighbors to consider in forming a k-nearest neighbor clas-

sifier.

min.classprob Minimum proportion of neighbors with a common class to make a class predic-

tion.

classprob2drop If an individual is inferred to have a genotype mismatch with classprob > this

value, treat as an outlier and drop from the analysis and then repeat the KNN

construction without it.

repeatKNN If TRUE, repeat k-nearest neighbor a second time, after omitting individuals

who seem to not be self-self matches

max.selfd Min distance from self (as proportion of mismatches between observed and pre-

dicted eQTL genotypes) to be excluded from the second round of k-nearest

neighbor.

phenolabel Label for expression phenotypes to place in the output distance matrix.

weightByLinkage

If TRUE, weight the eQTL to account for their relative positions (for example,

two tightly linked eQTL would each count about 1/2 of an isolated eQTL)

map.function Used if weightByLinkage is TRUE

verbose if TRUE, give verbose output.

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#### **Details**

We consider the expression phenotypes in batches, by which pseudomarker they are closest to. For each batch, we pull the genotype probabilities at the corresponding pseudomarker and use the individuals that are in common between cross and pheno and whose maximum genotype probability is above min.genoprob, to form a classifier of eQTL genotype from expression values, using k-nearest neighbor (the function class::knn()). The classifier is applied to all individuals with expression data, to give a predicted eQTL genotype. (If the proportion of the k nearest neighbors with a common class is less than min.classprob, the predicted eQTL genotype is left as NA.)

If repeatKNN is TRUE, we repeat the construction of the k-nearest neighbor classifier after first omitting individuals whose proportion of mismatches between observed and inferred eQTL genotypes is greater than max.selfd.

Finally, we calculate the distance between the observed eQTL genotypes for each individual in cross and the inferred eQTL genotypes for each individual in pheno, as the proportion of mismatches between the observed and inferred eQTL genotypes.

If weightByLinkage is TRUE, we use weights on the mismatch proportions for the various eQTL, taking into account their linkage. Two tightly linked eQTL will each be given half the weight of a single isolated eQTL.

#### Value

A matrix with nind(cross) rows and nrow(pheno) columns, containing the distances. The individual IDs are in the row and column names. The matrix is assigned class "lineupdist".

The names of the genes that were used to construct the classifier are saved in an attribute "retained".

The observed and inferred eQTL genotypes are saved as attributes "obsg" and "infg".

The denominators of the proportions that form the inter-individual distances are in the attribute "denom".

#### Author(s)

Karl W Broman,

## See Also

```
distee(), summary.lineupdist(), pulldiag(), omitdiag(), findCommonID(), find.gene.pseudomarker(),
calc.locallod(), plot.lineupdist(), class::knn(), plotEGclass()
```

```
library(qtl)
# load example data
data(f2cross, expr1, pmap, genepos)

# calculate QTL genotype probabilities
f2cross <- calc.genoprob(f2cross, step=1)
# find nearest pseudomarkers</pre>
```

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```
pmark <- find.gene.pseudomarker(f2cross, pmap, genepos)</pre>
# line up individuals
id <- findCommonID(f2cross, expr1)</pre>
# calculate LOD score for local eQTL
locallod <- calc.locallod(f2cross[,id$first], expr1[id$second,], pmark)</pre>
# take those with LOD > 25
expr1s <- expr1[,locallod>25,drop=FALSE]
# calculate distance between individuals
      (prop'n mismatches between obs and inferred eQTL geno)
d <- disteg(f2cross, expr1s, pmark)</pre>
# plot distances
plot(d)
# summary of apparent mix-ups
summary(d)
# plot of classifier for and second eQTL
par(mfrow=c(2,1), las=1)
plotEGclass(d)
plotEGclass(d, 2)
```

expr-data

Example gene expression data

# Description

Matrices of simulated gene expression data, each for 98 individuals at 5,000 genes. Think of expr1 and expr2 as expression data on two different tissues.

### Usage

```
data(expr1)
data(expr2)
```

#### **Format**

A matrix of integers, individuals as rows and genes as columns.

```
genepos(), f2cross(), pmap()
```

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## **Examples**

```
data(expr1)
data(expr2)

# identify the common individuals
id <- findCommonID(rownames(expr1), rownames(expr2))

# correlation between tissues for each gene
rho <- corbetw2mat(expr1[id$first,], expr2[id$second,])
hist(rho, breaks=100)</pre>
```

f2cross

Example experimental cross data

# **Description**

Simulated experimental cross data with some sample mix-ups. The only phenotype is an individual ID. There are 100 individuals genotyped at 1000 markers on 19 autosomes.

# Usage

```
data(f2cross)
```

## **Format**

An object of class "cross". See qtl::read.cross() in the R/qtl package for details.

## See Also

```
expr1(), expr2(), genepos(), pmap()
```

```
library(qt1)
data(f2cross)
summary(f2cross)
```

find.gene.pseudomarker

Find nearest peudomarker to each gene

# **Description**

Pull out the pseudomarker that is closest to the position of each of a series of genes.

## Usage

```
find.gene.pseudomarker(cross, pmap, geneloc, where = c("prob", "draws"))
```

## **Arguments**

cross	An object of class "cross" containing data for a QTL experiment. See the help file for qtl::read.cross() in the R/qtl package (https://rqtl.org).
pmap	A physical map of the markers in cross, with locations in Mbp. This is a list whose components are the marker locations on each chromosome.
geneloc	A data frame specifying the physical locations of the genes. There should be two columns, chr for chromosome and pos for position in Mbp. The rownames should indicate the gene names.
where	Indicates whether to pull pseudomarkers from the genotype probabilities (produced by qtl::calc.genoprob()) or from the imputed genotypes (produced by qtl::sim.geno()).

### **Details**

We first convert positions (by interpolation) from those contained within cross to physical coordinates contained in pmap. We then use qtl::find.pseudomarker() to identify the closest pseudomarker to each gene location.

We also include the positions of the pseudomarkers, and we print a warning message if pseudomarkers are > 2 Mbp from the respective gene.

#### Value

A data frame with columns chr (the chromosome) and pmark (the name of the pseudomarker). The third column pos contains the Mbp position of the pseudomarker. The final column is the signed distance between the gene and the pseudomarker. The rownames indicate the gene names.

#### Author(s)

Karl W Broman,

```
qtl::find.pseudomarker(),qtl::find.pseudomarkerpos(),plotEGclass(),disteg(),calc.locallod()
```

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## **Examples**

```
data(f2cross, expr1, genepos, pmap)
library(qtl)

# calc QTL genotype probabilities
f2cross <- calc.genoprob(f2cross, step=1)

# find nearest pseudomarkers
pmark <- find.gene.pseudomarker(f2cross, pmap, genepos, "prob")</pre>
```

findCommonID

Find individuals in common between a cross and a phenotype matrix

## **Description**

Identify which individuals are in common between a QTL mapping data set and a matrix of phenotypes, series of genes.

# Usage

```
findCommonID(id1, id2)
```

## **Arguments**

id1	A character vector of individual IDs. This can also be a QTL cross object (see
	qtl::read.cross()), in which case qtl::getid() is used to grab individual
	IDs, or a matrix or data frame, in which case the rownames are taken to be IDs.
id2	Like id1, can be a character vector, a cross or a matrix/data frame.

# Value

A list with three components:

First, a data frame with rows corresponding to all individuals (across the two sets of individual IDs) and three columns: indexInFirst and indexInSecond contain numeric indices to the locations of the individuals within cross and pheno, and inBoth is a logical vector to indicate which individuals appear in both crosses. The row names are the individual identifiers.

The second and third components are vectors of indices in id1 and id2, respectively, indicating the paired locations of the individuals that are in common.

## Author(s)

```
Karl W Broman,
```

```
calc.locallod(), corbetw2mat()
```

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## **Examples**

```
data(f2cross, expr1)
# align IDs
id <- findCommonID(f2cross, expr1)
# aligned data
f2cross_aligned <- f2cross[,id$first]
expr1_aligned <- expr1[id$second,]</pre>
```

fscale

Standardize the columns of a matrix

# **Description**

Standardize each column in a matrix, so that the columns have mean 0 and SD 1.

# Usage

```
fscale(x)
```

## **Arguments**

х

A numeric matrix.

# **Details**

Missing values (NA) are ignored and left as is.

If there is just 1 non-missing value in a column, it is left as is.

This function uses a one-pass algorithm to calculate the mean and SD, which is fast but can show a bit of round-off error.

# Value

A matrix of the same form as the input, but with columns transformed to have mean 0 and SD 1.

## Author(s)

```
Karl W Broman,
```

```
base::scale()
```

lineupversion

# **Examples**

```
x <- matrix(1:10, ncol=2)
y <- fscale(x)</pre>
```

genepos

Genomic positions of genes in simulated expression data

## **Description**

A table with the genomic positions of genes in the simulated expression data, expr1() and expr2().

# Usage

```
data(genepos)
```

## **Format**

A data frame with two columns, chromosome and physical position (in Mbp).

#### See Also

```
expr1(), expr2(), f2cross(), pmap()
```

# **Examples**

```
data(genepos)

# interplot genetic positions
library(qtl)
data(pmap)
data(f2cross)
genepos_interp <- interpPositions(genepos, pmap, pull.map(f2cross))
genepos[1:5,] # 'newpos' column is the interpolated cM position</pre>
```

lineupversion

Installed version of R/lineup

# Description

Print the version number of the currently installed version of R/lineup.

# Usage

```
lineupversion()
```

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## Value

A character string with the version number of the currently installed version of R/lineup.

## Author(s)

# **Examples**

lineupversion()

omitdiag

Replace the diagonal in a distance matrix with missing values

# **Description**

Replace the diagonal (that is, self-self distances) from a distance matrix calculated by distee() or disteg() with missing values (so that only self-nonself distances are left).

# Usage

omitdiag(d)

## **Arguments**

d

A distance matrix calculated by distee() or disteg().

## **Details**

We use the row and column names to identify which entries are self-self.

## Value

A matrix of the same form as the input, but with self-self distances replaced with NA.

## Author(s)

Karl W Broman,

```
pulldiag(), distee(), disteg(), summary.lineupdist(), plot2dist(), plot.lineupdist()
```

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## **Examples**

```
data(expr1, expr2)

# distance as RMS difference
d <- distee(expr1, expr2)

# focus on the self-nonself distances
# (replace self-self distances with NA)
d_selfnonself <- omitdiag(d)</pre>
```

plot.lineupdist

Plot summary of inter-individual distances

# Description

Plot histograms of self-self and self-nonself distances from a distance matrix calculated by distee() or disteg().

# Usage

```
## S3 method for class 'lineupdist'
plot(x, breaks = NULL, add.rug = TRUE, what = c("both", "ss", "sn"), ...)
```

# **Arguments**

X	Output of distee() or disteg().
breaks	Optional vector of breaks, passed to graphics::hist(), though if it is length 1, we interpret it as the number of breaks and ensure that both histograms use the same set of breaks.
add.rug	If true, also include graphics::rug() below histograms.
what	Indicates whether to plot both self-self and self-nonself distances (or correlations) or just one or the other. ("ss" indicates self-self and "sn" indicates self-nonself.)
	Ignored at this point.

## **Details**

We call pulldiag() and omitdiag() to get the self-self and self-nonself distances.

If all of the self-self distances are missing, we plot just the self-nonself distances.

# Value

None.

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## Author(s)

## See Also

```
pulldiag(), distee(), plot2dist()
```

# **Examples**

```
data(expr1, expr2)

# distance as correlation
d <- distee(expr1, expr2, "cor")

# plot histograms of self-self and self-nonself correlations
plot(d)</pre>
```

plot2dist

Plot two sets of inter-individual distances against one another

# **Description**

Plot two sets of inter-individual distances against one another, colored by self and non-self distances.

# Usage

```
plot2dist(
   d1,
   d2,
   hirow = NULL,
   hicol = NULL,
   xlab = NULL,
   ylab = NULL,
   smoothScatter = FALSE,
   colself = "black",
   colnonself = "gray",
   colhirow = "green",
   colhicol = "orange",
   ...
)
```

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## **Arguments**

d1	Output of distee().
d2	Output of distee().

hirow Names of rows to highlight in green.

hicol Names of columns to highlight in orange.

xlab X-axis label (optional) ylab Y-axis label (optional)

smoothScatter If TRUE, plot non-self distances with graphics::smoothScatter(); if FALSE,

use base::plot().

colself Color to use for the self-self points. If NULL, these aren't plotted.

colnonself Color to use for the non-self points. If NULL, these aren't plotted.

colhirow Color to use for the hirow points. If NULL, these aren't plotted.

colhicol Color to use for the hicol points. If NULL, these aren't plotted.

... Passed to base::plot() and graphics::points().

#### Value

None.

## Author(s)

## See Also

```
pulldiag(), distee(), summary.lineupdist()
```

```
data(expr1, expr2)

# distances as RMS difference and correlation
d_rmsd <- distee(expr1, expr2, "rmsd")
d_cor <- distee(expr1, expr2, "cor")

# plot distances against one another
plot2dist(d_rmsd, d_cor)</pre>
```

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plotEGclass

Plot classifier of eQTL genotype from expression data

## **Description**

Diagnostic plot of one of the eQTL classifiers from the results of disteg(): generally expression phenotype against observed eQTL genotype, colored by inferred eQTL genotype.

## Usage

```
plotEGclass(
    d,
    eqt1 = 1,
    outercol = "inferred",
    innercol = "observed",
    thecolors = c("#7B68ED", "#1B9E78", "#CA3767", "#E59E00"),
    ...
)
```

## **Arguments**

d	Output of disteg().
eqtl	Numeric index or a character vector (of the form " $1@102.35$ ") indicating the eQTL to consider.
outercol	Indicates how to color the outer edge of the points: "observed" indicates to color based on observed genotypes; "inferred" indicates to color based on inferred genotypes; otherwise, give a color.
innercol	Like outercol, but indicating the interior of the points.
thecolors	The colors to use in the plot. The last element (after the number of genotypes) indicates the color to use for missing values.
	Passed to base::plot() and graphics::points().

## Details

The function produces a diagnostic plot for studying one of the k-nearest neighbor classifiers underlying the output from disteg().

In the case of one expression phenotype attached to the selected eQTL, the plot is a dot plot of gene expression against observed eQTL genotype.

In the case of two expression phenotypes, the plot is a scatterplot of the two expression phenotypes against each other.

In the case of more than two expression phenotypes, we use graphics::pairs() to produce a matrix of scatterplots.

## Value

None.

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## Author(s)

#### See Also

```
disteg(), plot.lineupdist(), plot2dist(), class::knn()
```

## **Examples**

```
library(qtl)
# load example data
data(f2cross, expr1, pmap, genepos)
# calculate QTL genotype probabilities
f2cross <- calc.genoprob(f2cross, step=1)</pre>
# find nearest pseudomarkers
pmark <- find.gene.pseudomarker(f2cross, pmap, genepos)</pre>
# line up individuals
id <- findCommonID(f2cross, expr1)</pre>
# calculate LOD score for local eQTL
locallod <- calc.locallod(f2cross[,id$first], expr1[id$second,], pmark)</pre>
# take those with LOD > 25
expr1s <- expr1[,locallod>25,drop=FALSE]
# calculate distance between individuals
      (prop'n mismatches between obs and inferred eQTL geno)
d <- disteg(f2cross, expr1s, pmark)</pre>
# plot of classifier for and second eQTL
par(mfrow=c(2,1), las=1)
plotEGclass(d)
plotEGclass(d, 2)
```

pmap

Physical map of markers

# Description

Physical map (Mbp positions) of the markers in f2cross()

## Usage

```
data(pmap)
```

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# **Format**

A list of vectors, each containing the locations of markers in Mbp. (Technically, an object of class "map".)

## See Also

```
expr1(), expr2(), f2cross(), genepos()
```

## **Examples**

```
data(pmap)
summary(pmap)
plot(pmap)
```

pulldiag

Pull out the diagonal from a distance matrix

# Description

Pull out the diagonal from a distance matrix calculated by distee() (that is, self-self distances).

# Usage

```
pulldiag(d)
```

## **Arguments**

d

A distance matrix calculated by distee().

# **Details**

We use the row and column names to identify which entries are self-self.

## Value

A vector with the self-self distances.

# Author(s)

```
Karl W Broman,
```

```
omitdiag(), distee(), disteg(), summary.lineupdist(), plot2dist(), plot.lineupdist()
```

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## **Examples**

```
data(expr1, expr2)

# distance as RMS difference
d <- distee(expr1, expr2)

# pull out the self-self distances
d_selfself <- pulldiag(d)

# samples with smallest self-self correlation
sort(d_selfself)[1:10]</pre>
```

subset.lineupdist

Subsetting distance matrix

# **Description**

Pull out a specified set of rows and columns from a distance matrix calculated by distee() or disteg().

#### Usage

```
## S3 method for class 'lineupdist'
subset(x, rows = NULL, cols = NULL, ...)
## S3 method for class 'lineupdist'
x[rows = NULL, cols = NULL]
```

# Arguments

X	A distance matrix object as obtained from distee() or disteg().
rows	Optional vector of selected rows.
cols	Optional vector of selected columns.
	Ignored at this point.

## Value

The input distance matrix object, but with only the specified subset of the data.

## Author(s)

summary.lineupdist 25

## See Also

```
disteg(), distee(), pulldiag()
```

# **Examples**

```
data(expr1, expr2)

# find samples in common
id <- findCommonID(expr1, expr2)

# calculate correlations between cols of x and cols of y
thecor <- corbetw2mat(expr1[id$first,], expr2[id$second,])

expr1s <- expr1[,thecor > 0.8]/1000

expr2s <- expr2[,thecor > 0.8]/1000

# calculate correlations among samples
d <- distee(expr1s, expr2s, d.method="cor")

# pull out distances for samples 24, 92, 44, 66
samp <- c("24", "92", "44", "66")
d[samp, samp]</pre>
```

summary.lineupdist

Summarize inter-individual distances

# **Description**

Summarize the results of distee() or disteg(), with inter-individual distances between two sets of gene expression data.

## Usage

```
## S3 method for class 'lineupdist'
summary(
  object,
  cutoff = NULL,
  dropmatches = TRUE,
  reorder = c("alignmatches", "bydistance", "no"),
  ...
)
```

# **Arguments**

object

Output of distee() or disteg().

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cutoff (Optional) Cutoff on correlation/distance, with rows in the results only being

kept if the best distance/correlation is above this cutoff or the self-self result is

not missing and is above this cutoff.

dropmatches If TRUE, omit rows for which an individual's best match is itself.

reorder If "bydistance", reorder rows by increasing distance (or decreasing correla-

tion) to the best match and then by decreasing distance (or decreasing correlation) to self; if "alignmatches", group related errors together; if "no", leave as

is.

... Passed to base::print.data.frame().

## Value

A list with two components: the distances summarized by row and the distances summarized by column.

For each individual, we calculate the minimum distance to others, next-smallest distance, the self-self distance, the mean and SD of the distances to others, and finally indicate the individual (or individuals) that is closest.

## Author(s)

Karl W Broman, <brownse.edu>

#### See Also

```
pulldiag(), omitdiag(), distee(), disteg(), plot2dist(), plot.lineupdist()
```

```
data(expr1, expr2)

# distance as correlation
d <- distee(expr1, expr2, "cor")

# summary of potential problems
summary(d)</pre>
```

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