Package 'iQTL'

July 27, 2010

Title Genome-wide QTL Mapping via H-likelihood				
Version 2010.1				
Date 2010-07-26				
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Description This package provides an R implementation of the double hierarchical generalized linear model (DHGLM) particularly for analyzing genome-wide marker data.				
License Unlimited				
LazyLoad yes				
Depends VGAM				
R topics documented:				
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iQTL-package Genome-wide QTL Mapping via H-likelihood				

Description

Type Package

This package provides an R implementation of the double hierarchical generalized linear model (DHGLM) particularly for analyzing genome-wide marker data.

Details

iQTL-package

Package: iQTL
Type: Package
Version: 2010.1
Date: 2010-07-26
License: Unlimited
LazyLoad: yes
Depends: VGAM

Author(s)

Xia Shen

Maintainer: Xia Shen <xia.shen@lcb.uu.se>

References

Xia Shen, Lars Ronnegard and Orjan Carlborg (2010): Hierarchical likelihood opens a new way of estimating genetic values using dense marker maps. (14th QTLMAS, Poznan, Poland) SUBMITTED.

See Also

h.GWAS

Examples

```
## Not run:
## data
data(snpid)
data(pt)
data(gt)
data(pedi)
data(info)
marker.id <- info[snpid,1]</pre>
## set parameters
rho <- .9
## create spatial correlation matrix
## note that there are 5 chromosomes
border.idx <- c(0, 1971, 4032, 6080, 8068, 10029)
nmarker.chr <- numeric(5)</pre>
for (i in 1:5) nmarker.chr[i] <- sum((marker.id > border.idx[i]) *
(marker.id <= border.idx[i + 1]))</pre>
ac.mat <- matrix(0, dim(gt)[2], dim(gt)[2])</pre>
blockborder <- c(0, cumsum(nmarker.chr))</pre>
for (i in 1:5) ac.mat[(blockborder[i] + 1):blockborder[i + 1],
(blockborder[i] + 1):blockborder[i + 1]] <-</pre>
rho**(toeplitz(1:nmarker.chr[i]) - 1)
```

gt 3

```
## fixed effect design matrix

X <- model.matrix(~factor(pedi[,4]))
dimnames(X) <- dimnames(pt) <- dimnames(gt) <- c(NULL, NULL)

## analyze QT

GWAS <- h.GWAS(y = pt[,2], X = X, Z = gt, rho = rho, ac.mat = ac.mat,
phi.start = 2, plotting = FALSE)

par(mfrow = c(2, 1))
plot(marker.id, GWAS$v, xlab = 'Marker index', ylab = 'Marker effect',
type = 'p', cex = .6, col = 4, bty = 'n')
abline(v = border.idx, lty = 2, col = 'green4')
plot(marker.id, GWAS$lambda, xlab = 'Marker index',
ylab = 'Marker dispersion', type = 'l', col = 2, bty = 'n')
abline(v = border.idx, lty = 2, col = 'green4')
## End(Not run)</pre>
```

Genotypes of The Example Data

Description

gt

The genotypes of the example data with in the coded format.

Usage

```
data(gt)
```

Format

The matrix is 2326 (individuals) by 1628 (SNPs).

Details

The SNP markers are selected from the original dataset simulated for the 14th QTLMAS workshop, Poznan, Poland, 2010. See 'Source' and 'References'.

Source

```
http://jay.up.poznan.pl/qtlmas2010/dataset.html
```

References

Xia Shen, Lars Ronnegard and Orjan Carlborg (2010): Hierarchical likelihood opens a new way of estimating genetic values using dense marker maps. (14th QTLMAS, Poznan, Poland) SUBMITTED.

h.GWAS

h.GWAS	Genome-wide Association via H-likelihood

Description

This is an R implementation of the double hierarchical generalized linear model (DHGLM) particularly for analyzing genome-wide marker data.

Usage

```
h.GWAS(y, X, Z, family = 'normal', link = 'logit', X.disp = NULL,
    ac.mat = NULL, rho = 0, phi.start = 1, lambda.start = 0.1,
    alpha.start = 1, conv.crit = 1e-05, max.iter = 200, plotting = TRUE)
```

Arguments

У	a numeric vector of the phenotypes.
X	a design matrix of the fixed effects.
Z	an incidence matrix of the markers across the genome. Each element in the matrix gives the genotype of the corresponding individual at the corresponding marker. See 'Details' for more about the coding.
family	a string indicating the distribution family of the phenotypes. 'normal' and 'binary' is available. See 'Details' for more information.
link	a string giving the link function for binary phenotypes. 'logit', 'probit', and 'cloglog' are implemented and only available when family = 'binary'
X.disp	a design matrix of the fixed effects of the dispersion part of the model, i.e. for modeling the marker-specific variance.
ac.mat	an autocorrelation matrix for the markers, which is used for smoothing the marker specific variances. If NULL, an spatial correlation matrix is created with the parameter rho. See 'Details' for more information.
rho	a numeric value in $[0, 1)$ defining the spatial correlation for two adjacent markers. Only activated when ac.mat = NULL. See Details for more information.
phi.start	a numeric value giving the IWLS starting value of the dispersion parameter of the phenotypes. When $family = 'binary'$, this parameter is always fixed to be 1.
lambda.start	a numeric value giving the IWLS starting value of the dispersion parameter of the marker effects.
alpha.start	a numeric value giving the IWLS starting value of the dispersion parameter of the marker-specific variances. See 'References'.
conv.crit	a numeric value giving the convergence tolerance of the IWLS algorithm.
max.iter	an integer restricting the maximum number of IWLS interations.
plotting	a logic value specifying whether a figure will be produced after each IWLS iteration. The figures record the convergence behaviors of marker effects and marker-specific variances and will be saved in a single PDF file.

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Details

Typically, for an intercross, at a particular marker, there are three kinds of genotypes: AA, AB, and BB. Coding of $\mathbb Z$ can be simply set to be the number of 'A' copies, i.e. 2 for 'AA', 1 for 'AB', and 0 for 'BB'. For a backcross, 1 and 0 can be used for coding heterozygotes and homozygotes, respectively.

When family = 'binary', a link function is applied for linearizing the binary phenotypes to a continuous scale. The algorithm is not easy to converge depending on the starting values. plotting = TRUE is recommended to observe results for a certain number of iterations.

It is recommended that ac.mat is created according to the linkage disequilibium for each pair of markers across the genome. If ac.mat = NULL, the algorithm will create a spatial correlation matrix using argument rho. By default, rho = 0 indicating no autocorrelation for the marker-specific variances. See 'References'.

Value

phi	the dispersion parameter of the phenotypes. When family = $'$ binary', this parameter is always fixed to be 1.
alpha	the dispersion parameter of the marker-specific variances. See 'References'.
lambda	the marker-specific variances.
beta	the estimated fixed effects of the phenotypes.
gamma	the estimated fixed effects of the marker-specific variances.
V	the marker effects.
b	the estimated random effects of marker-specific variances.
niter	the number of iterations taken in the IWLS algorithm.

Author(s)

Xia Shen

References

Xia Shen, Lars Ronnegard and Orjan Carlborg (2010): Hierarchical likelihood opens a new way of estimating genetic values using dense marker maps. (14th QTLMAS, Poznan, Poland) SUBMITTED.

See Also

```
iQTL-package
```

Examples

```
## Not run:
## data

data(snpid)
data(pt)
data(gt)
data(pedi)
data(info)
marker.id <- info[snpid,1]</pre>
```

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```
## set parameters
rho <- .9
## create spatial correlation matrix
## note that there are 5 chromosomes
border.idx <- c(0, 1971, 4032, 6080, 8068, 10029)
nmarker.chr <- numeric(5)</pre>
for (i in 1:5) nmarker.chr[i] <- sum((marker.id > border.idx[i]) *
(marker.id <= border.idx[i + 1]))</pre>
ac.mat <- matrix(0, dim(gt)[2], dim(gt)[2])
blockborder <- c(0, cumsum(nmarker.chr))</pre>
for (i in 1:5) ac.mat[(blockborder[i] + 1):blockborder[i + 1],
(blockborder[i] + 1):blockborder[i + 1]] <-</pre>
rho**(toeplitz(1:nmarker.chr[i]) - 1)
## fixed effect design matrix
X <- model.matrix(~factor(pedi[,4]))</pre>
dimnames(X) <- dimnames(pt) <- dimnames(gt) <- c(NULL, NULL)</pre>
## analyze QT
GWAS \leftarrow h.GWAS(y = pt[,2], X = X, Z = gt, rho = rho, ac.mat = ac.mat,
phi.start = 2, plotting = FALSE)
par(mfrow = c(2, 1))
plot(marker.id, GWAS$v, xlab = 'Marker index', ylab = 'Marker effect',
type = 'p', cex = .6, col = 4, bty = 'n')
abline(v = border.idx, lty = 2, col = 'green4')
plot(marker.id, GWAS$lambda, xlab = 'Marker index',
ylab = 'Marker dispersion', type = 'l', col = 2, bty = 'n')
abline(v = border.idx, lty = 2, col = 'green4')
## End(Not run)
```

info

Marker Information of The Example Data

Description

The marker information of the example data.

Usage

```
data(info)
```

Format

Column 1: marker ID. Column 2: chromosome. Column 3: chromosomal position in base pairs.

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Details

The markers are from the original dataset simulated for the 14th QTLMAS workshop, Poznan, Poland, 2010. See 'Source' and 'References'.

Source

```
http://jay.up.poznan.pl/qtlmas2010/dataset.html
```

References

Xia Shen, Lars Ronnegard and Orjan Carlborg (2010): Hierarchical likelihood opens a new way of estimating genetic values using dense marker maps. (14th QTLMAS, Poznan, Poland) SUBMITTED.

pedi

The Pedigree Structure of The Example Data

Description

The pedigree information of the example data.

Usage

```
data(pedi)
```

Format

Column 1: individual ID. Column 2: sire ID. Column 3: dam ID. Column 4: sex.

Details

The pedigree is from the original dataset simulated for the 14th QTLMAS workshop, Poznan, Poland, 2010. See 'Source' and 'References'.

Source

```
http://jay.up.poznan.pl/qtlmas2010/dataset.html
```

References

Xia Shen, Lars Ronnegard and Orjan Carlborg (2010): Hierarchical likelihood opens a new way of estimating genetic values using dense marker maps. (14th QTLMAS, Poznan, Poland) SUBMITTED.

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pt

Phenotypes of The Example Data

Description

The phenotypes of the example data.

Usage

```
data(pt)
```

Format

Column 1: individual ID. Column 2: the simulated quantitative trait. Column 3: the simulated binary trait.

Details

The phenotypes are from the original dataset simulated for the 14th QTLMAS workshop, Poznan, Poland, 2010. See 'Source' and 'References'.

Source

```
http://jay.up.poznan.pl/qtlmas2010/dataset.html
```

References

Xia Shen, Lars Ronnegard and Orjan Carlborg (2010): Hierarchical likelihood opens a new way of estimating genetic values using dense marker maps. (14th QTLMAS, Poznan, Poland) SUBMITTED.

snpid

Indices of The Selected SNPs of The Example Data

Description

The indices of the selected SNPs of the example data.

Usage

```
data(snpid)
```

Format

a numeric vector containing SNP indices.

Details

Every 6 SNPs were selected from the original dataset simulated for the 14th QTLMAS workshop, Poznan, Poland, 2010. See 'Source' and 'References'.

snpid 9

Source

http://jay.up.poznan.pl/qtlmas2010/dataset.html

References

Xia Shen, Lars Ronnegard and Orjan Carlborg (2010): Hierarchical likelihood opens a new way of estimating genetic values using dense marker maps. (14th QTLMAS, Poznan, Poland) SUBMITTED.

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