# Package 'HardyWeinberg'

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Type Package

Title Statistical Tests and Graphics for Hardy-Weinberg Equilibrium

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**Depends** R (>= 1.8.0), mice, nnet, Rsolnp, shape

Description Contains tools for exploring Hardy-Weinberg equilibrium (Hardy, 1908; Weinberg, 1908) for bi and multi-allelic genetic marker data. All classical tests (chi-square, exact, likelihood-ratio and permutation tests) with bi-allelic variants are included in the package, as well as functions for power computation and for the simulation of marker data under equilibrium and disequilibrium. Routines for dealing with markers on the X-chromosome are included (Graffelman & Weir, 2016) <doi:10.1038/hdy.2016.20>, including Bayesian procedures. Some exact and permutation procedures also work with multi-allelic variants. Special test procedures that jointly address Hardy-Weinberg equilibrium and equality of allele frequencies in both sexes are supplied, for the bi and multi-allelic case. Functions for testing equilibrium in the presence of missing data by using multiple imputation are also provided. Implements several graphics for exploring the equilibrium status of a large set of bi-allelic markers: ternary plots with acceptance regions, log-ratio plots and Q-Q plots. The functionality of the package is explained in detail in a related JSS paper <doi:10.18637/jss.v064.i03>.

License GPL (>= 2)

URL https://www.r-project.org, http://www-eio.upc.edu/~jan/

LinkingTo Rcpp

Imports Rcpp

NeedsCompilation yes

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VignetteBuilder knitr, rmarkdown

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HardyWeinberg-package Statistical Tests and Graphics for Hardy-Weinberg Equilibrium

# **Description**

Contains tools for exploring Hardy-Weinberg equilibrium (Hardy, 1908; Weinberg, 1908) for bi and multi-allelic genetic marker data. All classical tests (chi-square, exact, likelihood-ratio and permutation tests) with bi-allelic variants are included in the package, as well as functions for power computation and for the simulation of marker data under equilibrium and disequilibrium. Routines for dealing with markers on the X-chromosome are included (Graffelman & Weir, 2016) including Bayesian procedures. Some exact and permutation procedures also work with multi-allelic variants. Special test procedures that jointly address Hardy-Weinberg equilibrium and equality of allele frequencies in both sexes are supplied, for the bi and multi-allelic case. Functions for testing equilibrium in the presence of missing data by using multiple imputation are also provided. Implements several graphics for exploring the equilibrium status of a large set of bi-allelic markers: ternary plots with acceptance regions, log-ratio plots and Q-Q plots. The functionality of the package is explained in detail in a related paper (Graffelman, 2015).

#### **Details**

Package: HardyWeinberg

Type: Package Version: 1.7.7 Date: 2024-03-31

License: GPL Version 2 or later.

With function HWTernaryPlot one can create ternary plots with acceptance regions for HWE. Several routines implement statistical tests for HWE such as HWChisq, HWExact, HWLratio and HWPerm. Bayesian procedures are available using HWPosterior. Aikaike information criterion for various scenarios can be calculated with HWAIC. Power computations are possible with HWPower.

#### Author(s)

Jan Graffelman

Maintainer: Jan Graffelman < jan.graffelman@upc.edu>

### References

Weir, B.S. (1996) Genetic Data Analysis II. Sinauer Associates, Massachusetts.

Graffelman, J. and Morales, J. (2008) Graphical Tests for Hardy-Weinberg Equilibrium Based on the Ternary Plot. *Human Heredity* 65(2):77-84.

Graffelman, J. (2015) Exploring Diallelic Genetic Markers: The HardyWeinberg Package. *Journal of Statistical Software* 64(3): 1-23. doi:10.18637/jss.v064.i03.

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Graffelman, J. and Weir, B.S. (2016) Testing for Hardy-Weinberg equilibrium at bi-allelic genetic markers on the X chromosome. *Heredity* 116(6): 558-568. doi:10.1038/hdy.2016.20 Nature Publishing Group.

Hardy, G.H. (1908) Mendelian Proportions in a Mixed Population. Science 28(706): 49-50.

Weinberg, W. (1908) On the Demonstration of Heredity in Man. In: Papers on Human Genetics. S. H. Boyer (Ed.) Englewood Cliffs, NJ:Prentice Hall, translated, 1963.

# **Examples**

```
library(HardyWeinberg)

# draw random SNPs from a population that is in HWE

set.seed(123)

m <- 100 # number of markers
n <- 100 # sample size

X <- HWData(n,m)
out <- HWTernaryPlot(X,100,region=1,vertex.cex=2,signifcolour=TRUE)</pre>
```

af

Function to compute allele frequencies

# Description

Function af computes the allele frequency for a vector with autosomal or X-chromosomal genotype counts or compositions.

### Usage

af(x)

# **Arguments**

Χ

a vector or matrix with counts or compositions. x must have either three elements (autosomal) or five elements (X-chromosomal)

### **Details**

Function af calculates the A allele frequency for the given order of autosomal genotype counts (AA,AB,BB) or the given order of X chromosomal genotype counts (A,B,AA,AB,BB). The genotypes must be supplied this order. Use maf to calculate the minor allele frequency and more flexible coding.

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#### Value

a vector with allele frequencies

#### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

# See Also

```
maf, afx
```

# **Examples**

```
# A simulated random marker
X \leftarrow as.vector(rmultinom(1,100,c(0.5,0.4,0.1)))
X \leftarrow X/sum(X)
print(X)
print(af(X))
# MN blood group counts
x < -c(MM=298, MN=489, NN=213)
# Calculate frequency M allele
af(x)
# Calculate frequency N allele
af(rev(x))
1 - af(x)
# Calculated allele frequencies (p) for a matrix of genotype counts.
X \leftarrow HWData(nm=10, n=100)
Χ
p \leftarrow af(X)
```

AFtest

Exact test of equality of allele frequencies for males and females

# Description

Function AFtest tests equality of allele frequencies for males and females for bi-allelic marker data by means of a Fisher exact test.

afx 7

#### Usage

```
AFtest(x, verbose = TRUE, ...)
```

# Arguments

x a vector containg the genotypic counts c(A,B,AA,AB,BB) for a bi-allelic X-

chromosomal markers.

verbose = TRUE prints results, verbose = FALSE is silent.

... additional arguments for function fisher.test.

# **Details**

Function AFtest constructs the contingency table of sex by allele, and call fisher. test to test for equality of allele frequencies. The test assumes Hardy-Weinberg equilibrium.

#### Value

AC Two-way table of sex by allele

pval p-value of the test

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### See Also

```
HWChisq, HWExact
```

# **Examples**

```
rs5968922 <- c(A=392, B=212, AA=275, AB=296, BB=80) AFtest(rs5968922)
```

afx

Function to compute X-chromosomal allele frequencies

# **Description**

Function afx computes the allele frequency for a vector or matrix of X-chromosomal genotype counts or frequencies.

# Usage

afx(x)

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#### **Arguments**

Χ

a vector or matrix with X-chromosomal counts or compositions. x must have five elements or columns.

#### **Details**

Function afx calculates the A allele frequency for the given order of X-chromosomal genotype counts (A,B,AA,AB,BB). The genotypes must be supplied this order. Use maf to calculate the minor allele frequency and more flexible coding.

#### Value

a vector with allele frequencies

#### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### See Also

```
maf, af
```

```
# A simulated random X-chromosomal marker
#
X <- as.vector(rmultinom(1,100,c(0.3,0.2,0.1,0.15,0.25)))
X <- X/sum(X)
print(X)
print(afx(X))
#
#
#
rs5968922 <- c(A=392, B=212, AA=275, AB=296, BB=80)
afx(rs5968922)
#
#
#
#
#
y <- c(C=337, G=129, CC=277, CG=209, GG=48)
afx(y)
#
# Calculated allele frequencies (p) for a matrix of X-chromosomal genotype counts.
#
X <- HWData(nm=10,n=1000,x.linked=TRUE)
X
p <- afx(X)</pre>
```

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agcounts

Calculate allele and genotype counts for X-chromosomal markers

# Description

Function agrounts determines sample size, minor are major allele counts, allele counts in females, numbers of males and females and allele frequencies for a vector of genotypes counts of an X-chromosomal markers.

# Usage

```
agcounts(x, verbose = FALSE)
```

# **Arguments**

X	a vector of X-chromosomal genotype counts (A,B,AA,AB,BB)
verbose	print the counts if (verbose = TRUE)

# Value

n	sample size
nA	number of A alleles
nB	number of B alleles
nf	number of females
nm	number of males
nAf	number of A alleles in females
nBf	number of B alleles in females
nt	total number of alleles
fAA	number of AA females
fAB	number of AB females
fBB	number of BB females
рА	overall A allele frequency
рВ	overall B allele frequency

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### See Also

```
mac, link{maf}
```

```
rs5968922 <- c(A=392, B=212, AA=275, AB=296, BB=80) counts <- agcounts(rs5968922)
```

10 alleles

alleles

Extract alleles

# **Description**

Function alleles extracts the names of the alleles from a named genotype vector.

# Usage

```
alleles(x, fromlabels = TRUE)
```

# **Arguments**

x A named or unnamed genotype vector (e.g. c(AA=10,AB=20,BB=5))

fromlabels extract genotypes from the labels of the vector elements, or from the vector

elements themselves.

# Value

A character vector with the alleles

#### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### See Also

```
n.alleles
```

```
x <- c(AA=10,AG=10,GG=10,AT=5)
als.x <- alleles(x)
print(als.x)

y <- rep(names(x),x)
als.y <- alleles(y,fromlabels = FALSE)
print(als.y)</pre>
```

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AllelesToTriangular Calculate triangu

Calculate triangular genotype matrix for vector(s) of alleles.

# Description

AllelesToTriangular constructs a lower triangular matrix of genotype counts from one or two vectors of alleles. It is particularly useful to create genotype counts for microsatellite data (STRs).

### Usage

```
AllelesToTriangular(A1, A2 = NULL, given=NULL)
```

#### **Arguments**

A1	The first allele of each individual, or a vector with all alleles, two consecutive
	ones for each individual.
A2	The second allele of each individual (optional).

given A vector of known alleles (optional). This argument can be used to specify

alleles that may not exist in the data.

#### **Details**

If the data is a single column vector with two succesive alleles for each individual, then specify A1 only. If data consists of two columns, each holding one allele of each individual, then specify A1 and A2. Typical STR data that comes in the format of two repeat lengths for a set of individuals can be transformed into a lower triangular matrix with genotype counts. See the examples below.

### Value

A lower triangular matrix with genotype counts.

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### References

Graffelman, J. (2015) Exploring Diallelic Genetic Markers: The HardyWeinberg Package. *Journal of Statistical Software* 64(3): 1-23. doi:10.18637/jss.v064.i03.

#### See Also

toTriangular

12 CEUchr22

### **Examples**

```
data(NistSTRs)
A1 <- NistSTRs[,1]
A2 <- NistSTRs[,2]
GM <- AllelesToTriangular(A1,A2)
print(GM)</pre>
```

Alzheimer

Genotype frequencies for 70 SNPs related to Alzheimer's disease

### **Description**

The dataframe contains the genotype frequencies MM, Mm and mm for the 70 SNPs for both cases and controls. The data are taken from table 7.11 in Laird & Lange.

# Usage

```
data(Alzheimer)
```

#### **Format**

A data frame containing 70 observations.

#### **Source**

```
Laird, N. M. and Lange, C. Table 7.11, p. 124
```

### References

Laird, N. M. and Lange, C. (2011) The fundamentals of modern statistical genetics. Springer.

CEUchr22

Biallelic polymorphisms sampled from chromosome 22 of the CEU population of the 1000 Genomes project.

# Description

Matrix CEUchr22 contains 10000 single nucleotide polymorphisms sampled from chromosome 22 of sample of 99 Utah residents with Northern and Western European ancestry.

#### Usage

```
data("CEUchr22")
```

#### Format

Matrix

dgraffelmanweir.bi

#### **Details**

The polymorphisms are coded in (0,1,2) format representing the count of the reference allele.

#### Source

https://www.internationalgenome.org/

#### References

The 1000 Genomes Project Consortium (2015) A global reference for human genetic variation. Nature 526(7571), pp. 68–74.

#### **Examples**

```
data(CEUchr22)
str(CEUchr22)
```

dgraffelmanweir.bi

Calculate Graffelman-Weir exact density for bi-allelic X-chromosomal variant

# **Description**

Function dgraffelmanweir calculate the probability P(NAB=nab and MA=malNA=na) for a biallelic X-chromosomal variant.

# Usage

```
dgraffelmanweir.bi(x, y)
```

### **Arguments**

- x vector with male genotype counts (A,B)
- y vector with female genotype counts (AA,AB,BB)

#### Value

a single real number

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### References

Graffelman, J. and Weir, B.S. (2016) Testing for Hardy-Weinberg equilibrium at bi-allelic genetic markers on the X chromosome. Heredity 116(6) pp. 558–568. doi:10.1038/hdy.2016.20

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#### See Also

```
HWExact, HWExactStats
```

# **Examples**

```
males <- c(A=392, B=212)
females <- c(AA=275, AB=296, BB=80)
prob <- dgraffelmanweir.bi(males,females)
print(prob)</pre>
```

dlevene

Calculate Levene's exact density for k alleles

# **Description**

Function dlevene calculates Levene's exact density for a diploid system with k alleles.

#### Usage

dlevene(N)

# **Arguments**

Ν

A lower triangular matrix with genotype counts

#### **Details**

The supplied matrix of genotype counts should be triangular, with the homozygote counts on the diagonal, and all heterozygote counts below the diagonal.

### Value

a single real number

# Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### References

Levene, H. (1949) On a matching problem arising in genetics. Annals of Mathematical Statistics, 20, pp. 91-94.

#### See Also

**HWExact** 

dlevene.bi

# **Examples**

```
x <- c(AA=12,AB=19,AC=13,BB=7,BC=5,CC=0)
x <- toTriangular(x)
prob <- dlevene(x)
print(prob)</pre>
```

dlevene.bi

Calculate Levene's density for a bi-allelic variant

# Description

Program dlevene.bi calculates Levene's density (P(AB|A)) for a bi-allelic variant.

# Usage

```
dlevene.bi(x)
```

# Arguments

Х

a vector of genotype counts (AA,AB,BB)

# Value

a single real number

# Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

# References

Levene, H. (1949) On a matching problem arising in genetics. Annals of Mathematical Statistics, 20, pp. 91-94.

#### See Also

```
dlevene, HWExact
```

```
x <- c(AA=298,AB=489,BB=213)
prob <- dlevene.bi(x)
print(prob)</pre>
```

16 EAFExact

**EAFExact** 

Exact test for equality of allele frequencies in males and females

# **Description**

EAFExact uses a Fisher Exact test to compare allele frequencies in males and females for variants with k alleles ( $k \ge 2$ ).

#### Usage

```
EAFExact(m, f, verbose = TRUE, ...)
```

# **Arguments**

m vector or triangular matrix with male genotype counts

f vector or triangular matrix with female genotype counts

verbose print output (TRUE) or not (FALSE)
... additional arguments for fisher.test

### **Details**

For bi-allelic autosomal variants the genotype counts can be supplied as vectors ((AA,AB,BB) for males, and (AA,AB,BB) for females). For X-chromosomal bi-allelic variants the genotype counts can also supplied as vectors ((A,B) for males, and (AA,AB,BB) for females). For multi-allelic autosomal variants male and genotype counts can be supplied as vectors (AA,AB,AC,BB,BC,CC,...) or as a triangular matrix, where matrix rows and colums are labelled with the allele name (A,B,C,...). For multi-allelic X-chromosomal variants, male genotype counts must be supplied as a vector (A,B,C,...) and female genotype counts must be supplied as a triangular matrix. See the examples below.

#### Value

pval p-value

tab table with allele counts

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### See Also

```
fisher.test
```

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# **Examples**

```
# bi-allelic autosomal
m < - c(AA=60, AB=96, BB=44)
f <- c(AA=44, AB=97, BB=59)
EAFtest <- EAFExact(m,f)</pre>
# bi-allelic X-chromosomal
males <-c(A=392, B=212)
females <- c(AA=275, AB=296, BB=80)
EAFtest <- EAFExact(males,females,verbose=TRUE)</pre>
# tri-allelic autosomal
males <- c(AA=20,AB=52,AC=34,BB=17,BC=51,CC=26)
females <- c(AA=28,AB=55,AC=33,BB=18,BC=50,CC=16)
EAFtest <- EAFExact(males,females,verbose=TRUE)</pre>
# tri-allelic X-chromosomal
males <-c(A=15,B=17,C=24)
females <- toTriangular(c(AA=4,AB=2,AC=13,BB=6,BC=19,CC=4))</pre>
EAFtest <- EAFExact(males,females,verbose=TRUE)</pre>
```

fisherz

Fisher's z transformation

# Description

Calculates Fisher's z transformation for a correlation coefficient

# Usage

```
fisherz(r)
```

# **Arguments**

r a correlation coefficient

18 fold

# Value

a real number

#### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### See Also

cor

#### **Examples**

```
r <- 0.5
print(fisherz(r))</pre>
```

fold

Fold a square matrix

# Description

The function fold sums corresponding below and above diagonal elements of a square matrix to form a triangular matrix.

# Usage

```
fold(X, lower = TRUE)
```

# **Arguments**

X a square matrix

lower logical. If lower=TRUE a lower triangular matrix is formed, if not an upper

triangular matrix.

# **Details**

Useful for constructing triangular matrices of genotype counts

### Value

A matrix

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

# See Also

```
lower.tri, upper.tri
```

GenerateSamples 19

# **Examples**

```
allelenames <- paste("A",11:13,sep="")
allele1 <- factor(c("A11","A11","A12","A12","A13","A12"),levels=allelenames)
allele2 <- factor(c("A11","A12","A12","A13","A13","A11"),levels=allelenames)
GC <- table(allele1,allele2)
GC <- as.matrix(unclass(GC))
GCf <- fold(GC)</pre>
```

GenerateSamples

Generate genotypic compositions

# **Description**

GenerateSamples generates all possible genotypic compositions (AA,AB,BB) for a given sample size n.

# Usage

```
GenerateSamples(n = 5)
```

#### **Arguments**

n

the desired sample size

# Value

returns a matrix with in each row a possible genotypic compostion for the given sample size.

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

```
GenerateSamples(5)
```

20 Glyoxalase

genlabels

Label genotype counts of a vector or matrix

# Description

Function genlabels sets the names of a vector or matrix of genotype counts.

# Usage

```
genlabels(X)
```

# **Arguments**

Χ

a 3 (or 5) element vector with genotype counts, a matrix of genotype counts (3 or 5 columns)

# Value

A vector or a matrix

#### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

# See Also

**HWChisq** 

# **Examples**

```
x <- c(25,50,25)
x <- genlabels(x)</pre>
```

Glyoxalase

Glyoxalase genotype data

# Description

Biallelic glyoxalase genotype data for 17 populations from India

# Usage

```
data("Glyoxalase")
```

HapMapCHBChr1 21

#### **Format**

A data frame with 17 observations on the following 3 variables.

AA number of homozygote AA individuals

AB number of heterozygote AB individuals

BB number of homozygote BB individuals

#### **Source**

```
Olson, J.M. (1993) Table 3.
```

#### References

Olson, J.M. (1993) Testing the Hardy-Weinberg law across strata. Annals of Human Genetics, 57, pp. 291-295.

### **Examples**

data(Glyoxalase)

HapMapCHBChr1	Genotype frequencies for 225 SNPs on chromosome 1 of the CHB population.
---------------	--

# **Description**

The dataframe contains the genotype frequencies in generic notation, AA, AB and BB the first 225 polymorphic SNPs without missing data on chromosome 1 of the Han Chinese in Beijing. The data are compiled from the HapMap project, phase 3.2, containing genotype information of 84 individuals.

### Usage

```
data(HapMapCHBChr1)
```

### **Format**

A matrix containing 225 rows and 3 columns (AA, AB, BB).

#### Source

http://hapmap.ncbi.nlm.nih.gov/

### References

The International HapMap Consortium (2007). A second generation human haplotype map of over 3.1 million SNPs Nature 449, pp. 851–861.

22 he

he

Calculate expected heterozygosity (He)

# Description

Function he calculates the expected heterozygosity for a biallelic genetic variant.

# Usage

```
he(x, bias.correct = TRUE)
```

# **Arguments**

```
x a vector or matrix with genotype counts.
bias.correct if bias.correct = TRUE a correction for bias will be applied.
```

#### **Details**

x can be a vector of genotype counts for a single marker (AA,AB,BB) or a three-column matrix of genotype counts for multiple markers.

# Value

a vector of expected heterozygosities.

# Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

# See Also

```
af, maf.
```

```
#
# He for a single marker
#
x <- c(MM=298,MN=489,NN=213)
he(x)
#
# He for a matrix of rmarkers
#
set.seed(123)
X <- HWData(10,100)
he(X)
he(X,bias.correct = FALSE)</pre>
```

HWABO 23

HWABO	Estimate allele frequencies and test for Hardy-Weinberg equilibrium
HWADO	Estimate attele frequencies and test for Hardy-weitherg equilibrium
	with a tri-allelic ABO system.

# Description

Function HWABO takes four genotype counts ("A","B","AB","OO") and estimates the three allele frequencies using the EM algorithm.

# Usage

```
HWABO(x, p = c(1/3, 1/3, 1/3), maxiter = 50, tol = 1e-10, verbose = TRUE)
```

# **Arguments**

X	a vector with genotype counts ("A","B","AB","OO").
p	a vector with initial allele frequencies (by default (1/3,1/3,1/3)).
maxiter	maximum number of iterations.

tol tolerance for convergence, 1e-10 by default

verbose print iteration history or not.

# Value

pn vector with estimated allele frequencies.

It.hist iteration history with log-likelihood.

expected expected genotype counts under HWE.

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### See Also

af

```
x <- c(fA=182,fB=60,nAB=17,n00=176)
al.fre <- HWABO(x)
#al2 <- HWABO(x,p=c(0.99,0.01,0.01),maxiter=25)
#al3 <- HWABO(x,p=c(0.01,0.99,0.01),maxiter=25)
#al4 <- HWABO(x,p=c(0.01,0.01,0.99),maxiter=25)</pre>
```

24 HWAIC

HWAIC	Compute Akaike's Information Criterion (AIC) for HWP and EAF models

# **Description**

Function HWAIC calculates Akaike's Information Criterion for ten different models that describe a bi-allelic genetic variant: M11: Hardy-Weinberg proportions and equality of allele frequencies in the sexes (HWP & EAF); M12: EAF and HWP in males only; M13: EAF and HWP in females only; M14: EAF and equality of inbreeding coefficients in the sexes (EIC); M15: EAF only; M21: HWP in both sexes; M22: HWP for males only; M23: HWP for females only; M24: EIC only; M25: None of the previous.

#### Usage

```
HWAIC(x, y, tracing = 0, tol = 0.000001)
```

#### Arguments

Х	Male genotype counts (AA,AB,BB)
у	Female genotype counts (AA,AB,BB)
tracing	Activate tracing in the maximization of some likelihoods (0=no tracing; 1:tracing)

mg)

tol tolerance for iterative maximization of some likelihoods

# **Details**

The log-likelihood for the six models is calculated. For two models (C and E) this is done numerically using package RSolnp.

### Value

A named vector containing 6 values for AIC

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### References

Graffelman, J. and Weir, B.S. (2018) On the testing of Hardy-Weinberg proportions and equality of allele frequencies in males and females at bi-allelic genetic markers. Genetic Epidemiology 42(1) pp. 34-48. doi:10.1002/gepi.22079

#### See Also

**HWLRtest** 

HWAlltests 25

#### **Examples**

```
males <- c(AA=11,AB=32,BB=13)
females <- c(AA=14,AB=23,BB=11)
stats <- HWAIC(males,females)
print(stats)</pre>
```

**HWAlltests** 

Perform all tests for Hardy-Weinberg equilibrium

# **Description**

HWAlltests performs all classical frequentists tests for Hardy-Weinberg equilibrium and lists their p-values.

# Usage

```
HWAlltests(x, verbose = TRUE, include.permutation.test = FALSE, x.linked = FALSE)
```

### **Arguments**

a vector with a set of genotype counts (AA, AB, BB)

verbose print output if set to TRUE

include.permutation.test

turns on the permutation test if set to TRUE

x.linked x.linked=FALSE indicates the marker is autosomal (default), and x.linked=TRUE

indicates it resides on the X-chromosome.

#### **Details**

By default the permutation test is not performed in order to reduce computing time.

#### Value

A dataframe with test statistics and p-values.

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

# See Also

```
HWLratio, HWChisq, HWExact
```

```
x <- c(298,489,213)
names(x) <- c("MM","MN","NN")
HWAlltests(x,verbose=TRUE)</pre>
```

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HWAlr

Compute additive log-ratio transformation

# Description

HWAlr computes the additive log-ratio transformation for genotype counts of bi-allelic genetic markers.

# Usage

```
HWAlr(X, zeroadj = 0.5, denominator = 2)
```

# Arguments

X A matrix of genotype counts (columns AA, AB and BB)

zeroadj A zero adjustment parameter (0.5 by default)

denominator The genotype count put in the denominator of the log-ratio (1=AA, 2=AB,

3=BB)

# Value

A matrix or vector of log-ratio coordinates

### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

# References

Graffelman, J. and Egozcue, J. J. (2011) Hardy-Weinberg equilibrium: a non-parametric compositional approach. In: Vera Pawlowsky-Glahn and Antonella Buccianti (eds.) Compositional Data Analysis: Theory and Applications, John Wiley & Sons, Ltd, pp. 207-215

#### See Also

```
HWClr, HWIlr
```

```
X <- HWData(100,100)
Y <- HWAlr(X)</pre>
```

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**HWAlrPlot** 

Plot genetic markers in additive log-ratio coordinates

# **Description**

HWAlrPlot creates a scatter plot of the log-ratio coordinates of bi-allelic genetic markers. Hardy-Weinberg equilibrium is indicated by a straigh line in the plot.

# Usage

```
HWAlrPlot(X, zeroadj = 0.5)
```

#### **Arguments**

X A matrix of genotype counts (columns AA, AB, BB)

zeroadj Zero-adjustment parameter. Zero counts in the count matrix are substituted by

zeroadj which is 0.5 by default.

#### Value

**NULL** 

### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

### References

Graffelman, J. and Egozcue, J. J. (2011) Hardy-Weinberg equilibrium: a non-parametric compositional approach. In: Vera Pawlowsky-Glahn and Antonella Buccianti (eds.) Compositional Data Analysis: Theory and Applications, John Wiley & Sons, Ltd, pp. 207-215

### See Also

```
HWClrPlot, HWIlrPlot
```

```
X <- HWClo(HWData(100,100))
HWAlrPlot(X)</pre>
```

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HWChisq Chi square tests for Hardy Weinberg equilibrium	
---	--

#### **Description**

HWChisq performs the chi-square test for Hardy-Weinberg equilibrium both for autosomal and X-chromosomal markers; it can dealt both with bi-allelic and multi-allelic variants.

#### Usage

```
HWChisq(X, cc = 0.5, verbose = TRUE, x.linked = FALSE, phifixed = NULL)
```

# Arguments

X	For bi-allelic variants, X is a vector containg the genotypic counts (AA,AB,BB for autosomal markers c(A,B,AA,AB,BB) for X-chromosomal markers). For multi-allelic variants, X is a lower triangular matrix with genotype counts, homozygotes on the diagonal and heterozygotes below the diagonal.
СС	cc the continuity correction parameter, the correction is only applied to bi-allelic markers (default $cc = 0.5$ ).
verbose	verbose = TRUE prints results, verbose = FALSE is silent.
x.linked	x.linked=FALSE indicates the marker is autosomal (default), and $x.linked=TRUE$ indicates it resides on the X-chromosome.
phifixed	(For X-chromosomal markers only) phifixed=NULL indicates that the fraction of males (females) should be estimated from the data (default). If set to any other value (e.g. phifixed=0.5) then the sample is assumed to come from a population with the specified fraction of males.

#### **Details**

HWChi sq does a chi-square test for Hardy-Weinberg equilibrium, and by default applies a continuity correction. For extreme allele frequencies, the continuity correction can lead to excessive type 1 error rates, and is better turned off in that case. The continuity correction can be turned off by specifying cc=0.

HWChisq can do the chi-square test for both autosomal and X-chrosomal markers. By setting x.linked = TRUE the marker will be assumed to be on the X-chromosome, and the count vector supplied should have 5 elements instead of 3 elements for an autosomal marker. For X-chromsomal markers argument phifixed is in general best left to its default value (NULL). Only in specific situations where the theoretical population sex ratio is known (e.g. in simulation studies where a universe with known gender ratio is sampled) phifixed could be set to the theoretical ratio of interest.

With bi-allelic variants, when alternative is set to less, a one-sided test for against a negative inbreeding coefficient (heterozygote excess) is performed. When alternative is set to greater a one-sided test for against a positive inbreeding coefficient (lack of heterozygotes) is performed.

For multi-allelic variants, which typically do have some rare alleles and rare genotypes, the asymptotic chi-square test is in general not recommended, and exact test procedures or permutation tests are recommended (see HWExact or HWPerm.mult).

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#### Value

HWChisq returns a list with the components:

chisq value of the chi-square statistic. NA is returned if the marker is monomorphic.

p-value of the chi-square test for Hardy-Weinberg equilibrium.

D Half the deviation from Hardy-Weinberg equilibrium for the AB genotype.

the allele frequency of A.the inbreeding coefficient.

expected the expected counts under Hardy-Weinberg equilibrium.

chi.contrib the contributions of the different genotypes to the chi-square statistic.

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### References

Weir, B.S. (1996) Genetic data analysis II. Sinauer Associates, Massachusetts. See Chapter3.

For the chi-square test for X-linked markers:

Graffelman, J. & Weir, B.S. (2016) Testing for Hardy-Weinberg equilibrium at bi-allelic genetic markers on the X chromosome. *Heredity* 116(6) pp. 558–568. doi:10.1038/hdy.2016.20

### See Also

```
HWLratio, HWChisqStats
```

```
#
# Test for an autosomal blood group marker
#
x <- c(MM=298,MN=489,NN=213)
HW.test <- HWChisq(x,verbose=TRUE)
#
# Same test without continuity correction
#
HW.test <- HWChisq(x,cc=0,verbose=TRUE)
#
# Test for X-chromsomal SNPs.
#
rs5968922 <- c(A=392, B=212, AA=275, AB=296, BB=80)
HW.test <- HWChisq(rs5968922,cc=0,x.linked=TRUE,verbose=TRUE)
#
#
#
y <- c(GG=48, CG=209, CC=277, G=129, C=337)
HWChisq(y,x.linked=TRUE,cc=0)
#</pre>
```

30 HWChisqMat

```
# Test a multi-allelic microsatellite
#
data(NistSTRs)
A1 <- NistSTRs[,1]
A2 <- NistSTRs[,2]
GC <- AllelesToTriangular(A1,A2)
HW.test <- HWChisq(GC)
#
# retaining only the three most common alleles
#
ii <- (A1 == 10 | A1 == 11 | A1 == 12) & (A2 == 10 | A2 == 11 | A2 == 12)
A1s <- A1[ii]
A2s <- A2[ii]
GC <- AllelesToTriangular(A1s,A2s)
HW.test <- HWChisq(GC)</pre>
```

 ${\it HWChisqMat}$ 

Matrix version of HWChisq

# **Description**

Function HWChisqMat executes the Chisquare test for HWE for each row in a matrix. This function is deprecated and it is better to use the faster HWChisqStats instead.

#### Usage

```
HWChisqMat(X, ...)
```

### **Arguments**

X A n times 3 matrix of genotypic counts (AA,AB,BB) ... extra arguments that are passed on to HWChisqStats

# Value

a vector with chi-square statistics or p-values

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### See Also

```
HWChisq, HWChisqStats
```

```
X <- HWData(100,100)
colnames(X) <- c("MM","MN","NN")
Results <- HWChisqMat(X)</pre>
```

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HWChisqStats Fast computation of chi-square statistics for Hardy-Weinberg equilibrium	HWChisqStats	
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# **Description**

HWChisqStats is a function for the fast computation of chi-square statistics (or the corresponding p-values) for a large set of bi-allelic markers (typically SNPs).

# Usage

```
HWChisqStats(X, x.linked = FALSE, pvalues = FALSE)
```

# Arguments

X	A matrix with genotype counts, one row per marker. X should have 5 columns for an X-chromosomal data set and 3 columns for an autosomal data set.
x.linked	$\label{logical} Logical indicating whether the markers are autosomal (x.linked=FALSE) or X-chromosomal (x.linked=TRUE).$
pvalues	Logical indicated whether chi-square statistics should be returned (pvalues=FALSE) or whether p-values should be returned (pvalues=TRUE).

#### **Details**

Matrix X should strictly comply with the following format. For an autosomal dataset is should contain the 3 genotype counts in order (AA,AB,BB). For an X-chromosomal dataset it should contain the 5 genotype counts in order (A,B,AA,AB,BB) where A and B are the male counts and AA, AB and BB the female counts.

This function was written for speed improvement, and should be much faster than looping over the rows of X with HWChisq. There is no error checking on the supplied data matrix.

#### Value

A vector of chi-square statistics

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### References

Graffelman, J. and Weir, B.S. (2016) Testing for Hardy-Weinberg equilibrium at 2 bi-allelic genetic markers on the X chromosome.

#### See Also

**HWChisq** 

32 HWClo

#### **Examples**

```
#
# Autosomal example
#
set.seed(123)
X <- HWData(1000,100)
monom <- (X[,2]==0 & X[,1]==0) | (X[,2]==0 & X[,3]==0)
X <- X[!monom,] # exclude monomorphics
Chisq.stats <- HWChisqStats(X,x.linked=FALSE,pvalues=FALSE)
Chisq.pvals <- HWChisqStats(X,x.linked=FALSE,pvalues=TRUE)
#
# X-chromosomal example
#
X <- HWData(1000,100,n.males=50,nA=75,x.linked=TRUE)
Chisq.stats <- HWChisqStats(X,x.linked=TRUE,pvalues=FALSE)
Chisq.pvals <- HWChisqStats(X,x.linked=TRUE,pvalues=TRUE)</pre>
```

HWC1o

Convert genotype counts to compositions

# Description

Function HWClo divides each row of a matrix by its total, and so produces matrix of compositions.

# Usage

HWClo(X)

### **Arguments**

Χ

A matrix of (genotype) counts

### Value

A matrix

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### See Also

```
HWAlr, HWClr, HWIlr
```

```
X <- HWData(2,100)
Y <- HWClo(X)</pre>
```

HWClr 33

HWC1r

Compute the centred log-ratio transformation

# Description

HWC1r computes the centred log-ratio transformation for genotype counts of bi-allelic genetic markers.

# Usage

```
HWClr(X, zeroadj = 0.5)
```

# Arguments

X A matrix of genotype counts (columns AA, AB and BB)

zeroadj A zero adjustment parameter (0.5 by default)

#### Value

A matrix or vector of log-ratio coordinates

#### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### References

Graffelman, J. and Egozcue, J. J. (2011) Hardy-Weinberg equilibrium: a non-parametric compositional approach. In: Vera Pawlowsky-Glahn and Antonella Buccianti (eds.) Compositional Data Analysis: Theory and Applications, John Wiley & Sons, Ltd, pp. 207-215

# See Also

```
HWAlr, HWIlr
```

```
X <- HWData(100,100)
Y <- HWClr(X)</pre>
```

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**HWClrPlot** 

Plot genetic markers in centred log-ratio coordinates

# **Description**

HWClrPlot creates a scatter plot of the centred log-ratio coordinates of bi-allelic genetic markers. Hardy-Weinberg equilibrium is indicated by a straigh line in the plot.

# Usage

```
HWClrPlot(X, zeroadj = 0.5)
```

#### **Arguments**

X A matrix of genotype counts (columns AA, AB, BB)

zeroadj Zero-adjustment parameter. Zero counts in the count matrix are substituted by

zeroadj which is 0.5 by default.

#### Value

**NULL** 

### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

### References

Graffelman, J. and Egozcue, J. J. (2011) Hardy-Weinberg equilibrium: a non-parametric compositional approach. In: Vera Pawlowsky-Glahn and Antonella Buccianti (eds.) Compositional Data Analysis: Theory and Applications, John Wiley & Sons, Ltd, pp. 207-215

#### See Also

```
HWAlrPlot, HWIlrPlot
```

```
X <- HWClo(HWData(100,100))
HWClrPlot(X)</pre>
```

HWCondProbAB 35

HWCondProbAB	Compute probability of a genotypic sample

# Description

Computes the probability of a particular genotypic sample given the allele count, sample size and number of heterozygotes.

# Usage

```
HWCondProbAB(n, nA, nAB)
```

# Arguments

n is the total sample size (total number of individuals)

nA is the number of A alleles in the sample

nAB is the number of heterozygotes in the sample

# Value

p probability of the particular sample

# Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

# See Also

**HWExact** 

```
x <- c(298,489,213)
names(x) <- c("MM","MN","NN")
n <- sum(x)
nM <- 2*x[1]+x[2]
nMN <- x[2]
p <- HWCondProbAB(n,nM,nMN)</pre>
```

36 HWD

HWD

Compute disequilibrium statistic D

# Description

Function HWD computes Weir's disequilibrium coefficient D.

# Usage

HWD(X)

# Arguments

Χ

a vector of genotype counts (AA, AB, BB)

#### Value

Returns the disequilibrium coefficient

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### References

Weir, B.S. (1996) Genetic data analysis II. Sinauer Associates, Massachusetts. See Chapter3.

### See Also

HWf HWChisq

```
x <- c(MM=298,MN=489,NN=213)
D <- HWD(x)
cat("Disequilibrium coefficient: ",D,"\n")
```

HWData 37

rium	HWData	Generate genetic marker data in or out of Hardy-Weinberg Equilibrium
------	--------	--

# Description

HWData generates samples of genotypic counts under various schemes. It mainly uses sampling from the multinomial distribution for given or random allele frequencies, either assuming Hardy-Weinberg proportions or a specified degree of inbreeding. Sampling can also be performed conditional on the allele frequency. The same procedures are also available for X linked markers.

## Usage

```
HWData(nm = 100, n = rep(100, nm), f = rep(0, nm), p = NULL, conditional = FALSE, exactequilibrium = FALSE, x.linked = FALSE, nA = NULL, n.males = round(0.5 * n), shape1 = 1, shape2 = 1, counts = TRUE)
```

## **Arguments**

nm	The number of bi-allelic markers.
n	The sample sizes.
f	The inbreeding coefficients (only for autosomal markers)
p	a vector of allele frequencies
conditional	if TRUE The Levene-Haldane distribution will be used for sampling autosomal markers, the Graffelman-Weir distribution for sampling $X$ chromosomal markers. If FALSE a multinomial distribution is used
exactequilibriu	ım
	generates data in exact HWE if set to TRUE
x.linked	$Simulated\ autosomal\ markers\ (\texttt{x.linked=FALSE}, the\ default)\ or\ X-chromosomal\ markers\ (\texttt{x.linked=TRUE})$
nA	A vector of minor allele counts, one for each marker. If not specified, it will be calculated from p
n.males	The number of males (only relevant if x.linked = TRUE)
shape1	First shape parameter of the beta distribution used to generate allele frequencies
shape2	Second shape parameter of the beta distribution used to generate allele frequencies
counts	If counts=TRUE the output are genotype counts, if counts=TRUE relative genotype frequencies $% \left( 1\right) =\left( 1\right) \left( $

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### **Details**

Option pfixed is deprecated and replaced by conditional Option pdist is deprecated and replaced by parameters shape1 and shape2

HWData returns a matrix of genotype counts, nm by 3 for autsomal markers or nm by 5 for X-chromosomal markers.

If the inbreeding coefficient is specified (f) it will only take effect for autosomal markers (x.linked=FALSE) and multinomial sampling (conditional=FALSE).

#### Value

X A matrix containing the genotype counts.

### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### See Also

**HWTernaryPlot** 

## **Examples**

```
# Generate 100 SNPs with uniform allele frequency under the equilibrium assumption.
out <- HWData(nm=100, n=100)
# Generate genotype frequencies of 100 SNPs with uniform allele frequency assuming exact equilbrium.
X <- HWData(nm=100, exactequilibrium = TRUE, counts = FALSE)</pre>
#
# Generate 100 SNPs (as counts), all having an expected A allele frequency of 0.50
X <- HWData(nm=100, p=0.5)
# Generate 100 SNPs, 50 with A allele frequency 0.25 and 50 with A allele frequency 0.75,
  assuming fixed allele frequencies.
X \leftarrow HWData(nm=100,p=rep(c(0.25,0.75),50), conditional = TRUE)
#
   Generate 100 SNPs with a skewed (beta) distribution of the allele frequency,
   rich in variants with a low minor allele frequency.
X <- HWData(nm=100, shape1=1, shape2=10)</pre>
# Generate 100 X chromosomal SNPs with uniformly distributed allele frequency.
X <- HWData(nm=100, x.linked = TRUE)</pre>
```

HWEM 39

HWEM	Estimation of contributions of two populations to a sample of genotype
	frequencies with the EM algorithm.

## **Description**

Function HWEM estimates the relative contributions of two populations with different allele frequencies that contributed to the genotype frequencies of sample which is a mixture of these two populations.

## Usage

```
HWEM(x, p = NULL, G = NULL, delta.init = c(0.5, 0.5), itmax = 50, eps = 1e-06, verbose = FALSE)
```

## **Arguments**

x	a 3x1 vector of genotype frequencies of the mixed sample, either counts or proportions.
р	a vector with two allele frequencies.
G	a 3x2 matrix with genotype frequencies as columns.
delta.init	initial values for the contributions of each populations
itmax	maximum number of iterations
eps	tolerance criterion for convergence
verbose	verbose = TRUE prints results, verbose = FALSE is silent.

### **Details**

HWEM employs the EM algorithm to estimate the contribution of two populations to the genotype frequencies of a bi-allelic marker, assumed to be typed for a mixed sample with individuals coming from two different populations with different allele frequencies. To estimate the contributions, both the mixed sample genotype frequencies (x) and either the genotype frequencies (G) or the allele frequencies (p) of the original populations must be specified. In case only allele frequencies are specified, Hardy-Weinberg proportions are assumed for the genotype frequencies.

### Value

a vector with two proportions, ordered according to the specified allele or genotype frequencies.

### Author(s)

```
Jan Graffelman < jan.graffelman@upc.edu>
```

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### References

Dempster, A.P., Laird, N.M. and Rubin, D.B. (1977) Maximum Likelihood from Incomplete Data via the EM Algorithm. Journal of the Royal Statistical Society. Series B (Methodological) 39(1) pp. 1–38.

## **Examples**

```
# genotype frequencies population 1
g1 <- c(0.034, 0.330, 0.636)
# genotype frequencies population 2
g2 <- c(0.349, 0.493, 0.158)
# sample from the mixed population
x <- c(0.270, 0.453,0.277)
#
# estimation based on genotype frequencies
#
G <- cbind(g1,g2)
contributions <- HWEM(x,G=G)
#
# estimation based on allele frequencies
#
p <- c(af(g1),af(g2))
contributions <- HWEM(x,p=p)</pre>
```

**HWExact** 

Exact test for Hardy-Weinberg equilibrium

## **Description**

HWExact performs an exact test for Hardy-Weinberg equilibrium

### Usage

```
HWExact(X, alternative = "two.sided", pvaluetype = "selome", eps=1e-10, x.linked =
FALSE, verbose = TRUE)
```

### **Arguments**

X vector with the genotype counts AA, AB, BB

alternative two.sided (default) will perform a two-sided test where both an excess and a

dearth of heterozygotes count as evidence against HWE. less is a one-sided test where only dearth of heterozygotes counts a evidence against HWE, greater is a one-sided test where only excess of heterozygotes counts as evidence against

HWE.

pvaluetype if pvaluetype is set to dost then the p-value of a two-sided test is computed

as twice the tail area of a one-sided test. When set to selome, the p-value is computed as the sum of the probabilities of all samples less or equally likely

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as the current sample. When set to midp, the p-value is computed as half the probability of the current sample + the probabilities of all samples that are more

extreme.

x.linked x.linked=FALSE indicates the marker is autosomal (default), and x.linked=TRUE

indicates it resides on the X-chromosome.

eps a tolerance that can be set for comparing probabilities in order to include tied

outcomes

verbose print results or not.

### **Details**

HWExact uses the recursion equations described by Wigginton et. al.

For testing large sets of bi-allelic variants, use the faster code in HWExactStats.

For large samples, HWExact may give the error message: "evaluation nested too deeply: infinite recursion". This can usually be resolved by increasing R's limit on nested expressions with options(expressions=10000) or a higher limit. With higher limits, the error message "protect(): protection stack overflow" can occur. This error can usually be resolved by increasing R's protection stack with the command line option --max-ppsize 100000 or higer values. However, with such large samples the exact test will give virtually the same result as a chi-square test, and it may be easier to use HWChisq in these circumstances.

### Value

pval p-value of the exact test

prob probabilities of all possible samples with the same sample size and minor allele

count

pofthesample probability of the observed sample

## Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

### References

Weir, B.S. (1996) Genetic data analysis II. Sinauer Associates, Massachusetts. See Chapter3.

Wigginton, J.E., Cutler, D.J. and Abecasis, G.R. (2005) A note on exact tests of Hardy-Weinberg equilibrium, American Journal of Human Genetics (76) pp. 887-893.

Graffelman, J. and Moreno, V. (2013) The mid p-value in exact tests for Hardy-Weinberg equilibrium, Statistical Applications in Genetics and Molecular Biology 12(4) pp. 433-448.

#### See Also

HWLratio, HWChisq, HWExactStats

42 HWExactMat

## **Examples**

```
# Example for an autosomal marker using the standard exact p-value
#
x <- c(298,489,213)
names(x) <- c("MM","MN","NN")
HW.test <- HWExact(x,verbose=TRUE)
#
# Example for an autosomal marker using the mid p-value
#
HW.test <- HWExact(x,verbose=TRUE,pvaluetype="midp")
#
# Example x-linked markers
#
rs5968922 <- c(A=392, B=212, AA=275, AB=296, BB=80 )
HWExact(rs5968922,x.linked=TRUE,verbose=TRUE)
#
#
#
y <- c(GG=48, CG=209, CC=277, G=129, C=337)
HWExact(y,x.linked=TRUE)
#
#
#</pre>
```

**HWExactMat** 

Matrix version of HWExact

# Description

Function HWExactMat is deprecated; use HWExactStats instead.

### Usage

```
HWExactMat(X, ...)
```

### **Arguments**

X A n times 3 matrix of genotypic counts (AA,AB,BB)
... extra arguments that are passed on to HWExact

### Value

pvalvec Vector with the p-values of each test

## Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

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### See Also

HWExact HWExactStats

# Examples

```
X <- HWData(100,100)
colnames(X) <- c("MM","MN","NN")
Results <- HWExactMat(X)
Output <- cbind(X,Results)
print(Output)</pre>
```

**HWExactPrevious** 

Exact test for Hardy-Weinberg equilibrium

# Description

HWExactPrevious performs an exact test for Hardy-Weinberg equilibrium

### Usage

```
HWExactPrevious(X, alternative = "two.sided", pvaluetype = "selome",
x.linked = FALSE, verbose = FALSE)
```

## **Arguments**

X vector with the genotype counts AA, AB, BB

alternative two.sided (default) will perform a two-sided test where both an excess and a

dearth of heterozygotes count as evidence against HWE. less is a one-sided test where only dearth of heterozygotes counts a evidence against HWE, greater is a one-sided test where only excess of heterozygotes counts as evidence against

HWE.

pvaluetype if pvaluetype is set to dost then the p-value of a two-sided test is computed

as twice the tail area of a one-sided test. When set to selome, the p-value is computed as the sum of the probabilities of all samples less or equally likely as the current sample. When set to midp, the p-value is computed as half the probability of the current sample + the probabilities of all samples that are more

extreme.

x.linked x.linked=FALSE indicates the marker is autosomal (default), and x.linked=TRUE

indicates it resides on the X-chromosome.

verbose print results or not.

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### **Details**

HWExactPrevious uses the recursion equations described by Wigginton et. al.

For large samples, HWExactPrevious may give the error message: "evaluation nested too deeply: infinite recursion". This can usually be resolved by increasing R's limit on nested expressions with options(expressions=10000) or a higher limit. With higher limits, the error message "protect(): protection stack overflow" can occur. This error can usually be resolved by increasing R's protection stack with the command line option --max-ppsize 100000 or higer values. However, with such large samples the exact test will give virtually the same result as a chi-square test, and it may be easier to use HWChisq in these circumstances.

### Value

pval p-value of the exact test

prob probabilities of all possible samples with the same sample size and minor allele

count

pofthesample probability of the observed sample

### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### References

Weir, B.S. (1996) Genetic data analysis II. Sinauer Associates, Massachusetts. See Chapter3.

Wigginton, J.E., Cutler, D.J. and Abecasis, G.R. (2005) A note on exact tests of Hardy-Weinberg equilibrium, American Journal of Human Genetics (76) pp. 887-893.

## See Also

```
HWLratio, HWChisq
```

### **Examples**

```
#
# Example autosomal marker
#
x <- c(298,489,213)
names(x) <- c("MM","MN","NN")
## Not run: HW.test <- HWExactPrevious(x,verbose=TRUE)
#
# Example x-linked marker
#
rs5968922 <- c(A=392, B=212, AA=275, AB=296, BB=80)
## Not run: HWExactPrevious(rs5968922,x.linked=TRUE,verbose=TRUE)</pre>
```

HWExactStats 45

HWExactStats	Computation of Exact p-values for Hardy-Weinberg equilibrium for sets of SNPs

## Description

HWExactStats is a function for the computation of Exact p-values for a large set of bi-allelic markers (typically SNPs).

## Usage

```
HWExactStats(X, x.linked = FALSE, plinkcode = TRUE, midp = FALSE,...)
```

### **Arguments**

X	A matrix with genotype counts, one row per marker. X should have 5 columns for an X-chromosomal data set and 3 columns for an autosomal data set.
x.linked	Logical indicating whether the markers are autosomal (x.linked=FALSE) or X-chromosomal (x.linked=TRUE).
plinkcode	Logical indicating whether to use faster C++ code from the PLINK software.
midp	Logical indicating whether to use the mid p-value for the C++ code or not
	extra arguments that are passed on to HWExact

### **Details**

Matrix X should strictly comply with the following format. For an autosomal dataset is should contain the 3 genotype counts in order (AA,AB,BB). For an X-chromosomal dataset it should contain the 5 genotype counts in order (A,B,AA,AB,BB) where A and B are the male counts and AA, AB and BB the female counts.

Argument plinkcode=TRUE (the default) will use C++ code for faster calculation (functions SNPHWE2 and SNPHWEX) with larger datasets. The C++ code was generously shared by Christopher Chang, and the same code is used in the program PLINK (2.0).

### Value

A vector of p-values

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu > (R code) and Christopher Chang < chrchang523@gmail.com > (C++ code)

#### References

Graffelman, J. and Weir, B.S. (2016) Testing for Hardy-Weinberg equilibrium at bi-allelic genetic markers on the X chromosome. Heredity 116(6) pp. 558–568. doi:10.1038/hdy.2016.20

Purcell et al. (2007) PLINK: A Toolset for Whole-Genome Association and Population-Based Linkage Analysis. American Journal of Human Genetics 81(3) pp. 559–575.

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### See Also

**HWExact** 

## **Examples**

```
#
# Autosomal example
#
set.seed(123)
X <- HWData(1000,100)
monom <- (X[,2]==0 & X[,1]==0) | (X[,2]==0 & X[,3]==0)
X <- X[!monom,] # exclude monomorphics
Exact.pvalues <- HWExactStats(X,x.linked=FALSE)
#
# X-chromosomal example
#
X <- HWData(1000,100,n.males=50,nA=75,x.linked=TRUE)
Exact.pvalues <- HWExactStats(X,x.linked=TRUE)</pre>
```

HWf

Computation of inbreeding coefficient

## **Description**

HWf computes the inbreeding coefficient for sample of genotype counts, or a matrix of genotype counts.

# Usage

HWf(X)

# Arguments

Χ

a vector or matrix of genotype counts (AA, AB, BB)

### **Details**

For monomorphic markers a warning is issued, and the estimate for the inbreeding coefficient is NaN.

## Value

Returns a single inbreeding coefficient (intraclass correlation coefficient), if X is a single sample, or a vector of inbreeding coefficients, if X is a matrix with genotype counts.

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

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### References

Crow, J. F. and Kimura, M. (1970) An introduction to population genetics theory. Harper & Row, publishers, New York

### See Also

```
HWChisq
```

# **Examples**

```
#
# A single sample
#
x <- c(MM=298,MN=489,NN=213)
fhat <- HWF(x)
fhat
#
# Multiple samples
#
set.seed(123)
X <- HWData(nm=100,n=1000)
fhat <- HWF(X)</pre>
```

HWGenotypePlot

Scatter plot of the genotype frequencies

# Description

HWGenotypePlot makes a scatterplots of the AB or BB frequency versus the AA frequency and represents a blue curve indicating the Hardy-Weinberg equilibrium condition.

## Usage

```
\label{eq:hwgenotypePlot} \begin{split} &\text{HWGenotypePlot(X, plottype = 1, xlab = expression(f[AA]), ylab = } \\ &\text{ifelse(plottype == 1, expression(f[AB]), expression(f[BB])), asp = 1,} \\ &\text{pch = 19, xlim = c(0, 1), ylim = c(0, 1), cex = 1, cex.axis = 2, cex.lab = 2, ...)} \end{split}
```

## **Arguments**

,	X	A matrix of genotype counts or frequencies with three columns (AA, AB, BB)
1	plottype	plottype=1 produces a plot of AB versus AA, plottype=2 produced a plot of BB versus AA.
:	xlab	A label for the x axis
	ylab	A label for the y axis
	asp	Aspec ratio (1 by default)
ı	pch	Plotting charachter (19 by default)

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xlim	Limits for the x axis (0-1 by default)
ylim	Limits for the y axis (0-1 by default)
cex	Character expansion factor (1 by default)
cex.axis	Character expansion factor for the axes (2 by default)
cex.lab	Character expansion factor for labels of axis (2 by default)
	Additional arguments for the plot function

### Value

NULL

## Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

## See Also

**HWTernaryPlot** 

## **Examples**

```
n <- 100 # sample size
m <- 100 # number of markers
Xc <- HWClo(HWData(n,m))
HWGenotypePlot(Xc,plottype=1,main="Heterozygote-homozygote scatterplot")</pre>
```

HWIlr

Compute isometric log ratio coordinates.

## **Description**

HWI1r computes isometric log ratio coordinates for genotypic compositions (AA, AB, BB)

## Usage

```
HWIlr(X, zeroadj = 0.5)
```

## **Arguments**

X A matrix of genotype counts, markers in rows, counts for AA, AB and BB in

three columns

zeroadj Adjustment for zeros (0.5 by defaults)

## Value

A matrix of log ratio coordinates.

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### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### References

Egozcue, J.J., Pawlowsky-Glahn, V., Mateu-Figueras, G. and Barcelo-Vidal, C. (2003) Isometric Logratio Transformations for Compositional Data Analysis. Mathematical Geology 35(3), pp. 279-300.

Graffelman, J. and Egozcue, J. J. (2011) Hardy-Weinberg equilibrium: a non-parametric compositional approach. In: Vera Pawlowsky-Glahn and Antonella Buccianti (eds.) Compositional Data Analysis: Theory and Applications, John Wiley & Sons, Ltd, pp. 207-215

### See Also

```
HWAlr, HWClr
```

### **Examples**

```
X <- HWData(100,100)
Y <- HWIlr(X)</pre>
```

**HWIlrPlot** 

Plot bi-allelic genetic markers in isometric log ratio coordinates

## **Description**

HWI1rPlot makes a scatter plot of the isometric log ratio coordinates for bia-llelic markers.

### Usage

```
HWIlrPlot(X, zeroadj = 0.5, ...)
```

## **Arguments**

X Matrix of genotype counts, one marker per row, AA, AB and BB in three columns
 zeroadj Adjustment for zero values (0.5 by default)
 ... Additional arguments for function plot

### Value

A matrix of log ratio coordinates.

## Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

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### References

Graffelman, J. and Egozcue, J. J. (2011) Hardy-Weinberg equilibrium: a nonparametric compositional approach. In Pawlowsky-Glahn, V. and Buccianti A., editors, Compositional Data Analysis: Theory and Applications, pages 208-217, John Wiley & Sons, Ltd.

#### See Also

```
HWAlrPlot, HWClrPlot
```

## **Examples**

```
X <- HWClo(HWData(100,100))
HWIlrPlot(X)</pre>
```

**HWLindley** 

Calculate a posteriori density for Lindley's alpha

## **Description**

Function HWLindley calculates the posterior density for disequilibrium measure alpha, as defined by Lindley (1988).

### Usage

```
HWLindley(alphaseq = seq(-3, 3, by = 0.01), x)
```

## Arguments

alphaseq a single value or a sequence of values for alpha x the genotype count vector in format (AA,AB,BB)

#### **Details**

Numerical integration is used to compute the density.

## Value

a vector with values of the density for each value in alphaseq

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

## References

Lindley, D.V. (1988) Statistical Inference Concerning Hardy-Weinberg Equilibrium. In: Bernardo, J.M., DeGroot, M.H., Lindley, D.V. and Smith, A.F.M. Bayesian Statistics, 3, pp. 307-326. Oxford University Press.

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## See Also

**HWPosterior** 

### **Examples**

HWLindley.cri

Calculate a credible interval for Lindley's alpha for HWE,

# **Description**

Function HWLindley.cri calculates a Bayesian credible interval using Lindley's posterior density for equilibrium paramater alpha.

## Usage

```
HWLindley.cri(x, verbose = TRUE, limits = c(0.025, 0.975))
```

## **Arguments**

x a vector of three genotype counts in order (AA,AB,BB).

verbose print output (verbose=TRUE) or be silent

limits upper and lower probability limits of the interval

## **Details**

The limits are found by numerical integration over Lindley's density.

### Value

a vector with the lower and upper limit of the credible interval

### Author(s)

```
Jan Graffelman < jan.graffelman@upc.edu>
```

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### References

Lindley, D.V. (1988) Statistical Inference Concerning Hardy-Weinberg Equilibrium. In: Bernardo, J.M., DeGroot, M.H., Lindley, D.V. and Smith, A.F.M. Bayesian Statistics, 3, pp. 307-326. Oxford University Press.

Graffelman, J. (2020) Statistical tests for the Hardy-Weinberg equilibrium. Wiley StatsRef: Statistics Reference Online doi:10.1002/9781118445112.stat08274.

### See Also

```
HWLindley, HWPosterior,
```

## **Examples**

```
#
# MN blood group data
#
x <- c(MM=298,MN=489,NN=213)
#
# credible interval of 95%
#
HWLindley.cri(x)
#
# credible interval of 90%
#
HWLindley.cri(x,limits=c(0.05,0.95))</pre>
```

**HWLRAllTests** 

Perform most relevant likelihood ratio test for Hardy-Weinberg equilibrium and equality of allele frequencies

# **Description**

Function HWLRAllTests performs a set of likelihood ratio tests in relation with Hardy-Weinberg proportions (HWP) and equality of allele frequencies (EAF) for autosomal bi-allelic genetic variants.

# Usage

```
HWLRAllTests(x, y)
```

## **Arguments**

```
x Male genotype counts (AA,AB,BB)
```

y Female genotype counts (AA,AB,BB)

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### **Details**

Function HWLRA11Tests calls HWLRtest and calculates the p-value of six different tests: 1) joint HWP and EAF (A-F); 2) EAF irrespective of HWP (C-F); 3) HWP irrespective of EAF (D-F); 4) HWP versus EIC (given EAF) (A-B); 5) EIC irrespective of EAF (E-F) and 6) HWP versus EIC. Letters refer to scenarios described by Graffelman & Weir (2018).

#### Value

A named vector with six p-values

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### References

Graffelman, J. and Weir, B.S. (2018) On the testing of Hardy-Weinberg proportions and equality of allele frequencies in males and females at bi-allelic genetic markers. Genetic Epidemiology 42(1): pp. 34–48. doi:10.1002/gepi.22079

#### See Also

**HWLRtest** 

### **Examples**

```
males <- c(AA=11,AB=32,BB=13)
females <- c(AA=14,AB=23,BB=11)
pvalues <- HWLRAllTests(males,females)
print(pvalues)</pre>
```

**HWLratio** 

Likelihood ratio test for Hardy Weinberg equilibrium

## **Description**

HWLratio performs the Likelihood ratio test for Hardy Weinberg equilibrium, both for autosomal and X-chromosomal markers.

## Usage

```
HWLratio(X, verbose = TRUE, x.linked = FALSE)
```

# **Arguments**

```
    X a vector containing the genotypic counts (AA,AB,BB).
    verbose verbose = TRUE prints results, verbose = FALSE is silent.
    x.linked = TRUE prints results, verbose = FALSE is silent.
    x.linked = TRUE indicates the marker is autosomal (default), and x.linked = TRUE indicates it resides on the X-chromosome.
```

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### Value

HWLratio returns a list with the components:

Lambda the likelihood ratio

G2 -2\*log(Lambda)

pval the p-value

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### References

Weir, B.S. (1996) Genetic data analysis II. Sinauer Associates, Massachusetts. See Chapter 3.

### See Also

```
HWChisq
```

### **Examples**

```
x <- c(298,489,213)
names(x) <- c("MM","MN","NN")
HW.test <- HWLratio(x,verbose=TRUE)
#
# Test for X-chromsomal SNPs.
#
rs5968922 <- c(A=392, B=212, AA=275, AB=296, BB=80)
HW.test <- HWLratio(rs5968922,x.linked=TRUE,verbose=TRUE)
#
#
#
y <- c(GG=48, CG=209, CC=277, G=129, C=337)
HWLratio(y,x.linked=TRUE)
#
#</pre>
```

**HWLRtest** 

Perform likelihood ratio test comparing two nested scenarios for a biallelic genetic variant, distinguishing the two sexes.

## **Description**

Program HWLRtest performs a likelihood ratio test comparing two scenarios for an autosomal biallelic genetic variant. The scenarios concern Hardy-Weinberg proportions (HWP) and equality of allele frequencies (EAF) in both sexes. The different scenarios are described by Graffelman & Weir (2017).

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## Usage

```
HWLRtest(x, y, scene.null = "S1", scene.alt = "S6", verbose = TRUE, tracing = 0)
```

### **Arguments**

x Male genotype countsy Female genotype counts

scene.null Scenario under the null hypothesis (E.g. "S1")

scene.alt Scenario under the alternative hypothesis (E.g. "S6")

verbose print output or not

tracing Show tracing of the numeric likelihood maximization (1) or not (0).

### **Details**

The different scenarios are indicated with S1, S2, S3, S4, S6 and S6. S1 refers to Hardy-Weinber proportions and equality of allele frequencies. S2 refers to equality of allele frequencies and equality of inbreeding coefficients for the two sexes. S3 refers to equality of allele frequencies irrespective of HWP. S4 refers to HWP irrespective of allele frequencies. S5 refers to equality of inbreeding coefficients irrespective of allele frequencies. S6 is unrestricted.

### Value

G2 Likelihood ratio statistic

df Degrees of freedom of the likelihood ratio statistic

pval p-value

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### References

Graffelman, J. and Weir, B.S. (2018) On the testing of Hardy-Weinberg proportions and equality of allele frequencies in males and females at bi-allelic genetic markers. Genetic Epidemiology 42(1) pp. 34–48 doi:10.1002/gepi.22079

## See Also

**HWAIC** 

## **Examples**

```
males <- c(AA=11,AB=32,BB=13)
females <- c(AA=14,AB=23,BB=11)
#
# test EAF
#</pre>
```

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```
lr1.out <- HWLRtest(males,females,scene.null="S3",scene.alt="S6")
#
# test EIC given EAF
#
lr2.out <- HWLRtest(males,females,scene.null="S2",scene.alt="S3")
#
# test HWP versus EIC, given EAF.
#
lr3.out <- HWLRtest(males,females,scene.null="S1",scene.alt="S2")</pre>
```

**HWMissing** 

Test a bi-allelic marker for Hardy-Weinberg equilibrium in the presence of missing genotype information.

## **Description**

Function HWMissing imputes missing genotype data with a multinomial logit model that uses information from allele intensities and/or neighbouring markers. Multiple imputation algorithms implemented in the Mice package are used to obtain imputed data sets. Inference for HWE is carried out by estimating the inbreeding coefficient or exact p-values for each imputed data set, and by combining all estimates using Rubin's pooling rules.

### Usage

```
HWMissing(X, imputecolumn = 1, m = 50, coding = c(0,1,2), verbose = FALSE, alpha = 0.05, varest = "oneovern", statistic = "chisquare", alternative = "two.sided", ...)
```

### Arguments

X	An input data frame. By default, the first column should contain the SNP with missing values.
imputecolumn	Indicates which column of the supplied data frame is to be imputed (by default, the first colum, imputecolumn=1
m	The number of imputations (50 by default)
coding	Indicates how the genotype data is coded (e.g. 0 for AA, 1 for AB, and 2 for BB).
verbose	verbose = TRUE prints results, verbose = FALSE is silent.
alpha	significance level (0.05 by default) used when computing confidence intervals

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varest Estimator for the variance of the inbreeding coefficient. varest="oneovern" is

the default and sets the variance under the null (1/n). varest="bailey" uses an

approximation (see details).

statistic If statistic = "chisquare" then inbreeding coefficients (equivalent to chisquare

statistics) will be computed for each imputed data set and then combined. If statistic = "exact" then one-sided exact tests will be computed for each im-

puted data set and the resulting p-values will be combined.

alternative two.sided (default) will perform a two-sided test where both an excess and a

dearth of heterozygotes count as evidence against HWE. less is a one-sided test where only dearth of heterozygotes counts a evidence against HWE, greater is a one-sided test where only excess of heterozygotes counts as evidence against

HWE.

... additional options for function mice of the Mice package

#### **Details**

The function HWMissing tests one genetic marker (e.g. a SNP) with missings for HWE. By default, this marker is supposed to be the first column of dataframe X. The other columns of X contain covariates to be used in the imputation model. Covariates will typically be other, correlated markers or allele intensities of the SNP to be imputed. Covariate markers should be coded as factor variables whereas allele intensities should be numerical variables. By default, a polytomous regression model will be used to impute the missings. If the covariates also contain missings, an imputation method for each column of X can be specified by using the method of mice (see example below).

If there are no covariates, missings can be imputed under the MCAR assumption. In that case, missings are imputed by taking a random sample from the observed data. This is what HWMissing will do if no covariates are supplied, X being a single factor variable.

Several estimators for the variance of the inbreeding coefficient have been described in the literature. The asymptotic variance of the inbreeding coefficient under the null hypothesis is 1/n, and is used if varest = "oneovern" is used. This is the recommended option. Alternatively, the approximation described in Weir (p. 66) can be used with varest = "bailey".

### Value

Res A vector with the inbreeding coefficient, a confidence interval for the inbreeding

coefficient, a p-value for a HWE test and missing data statistics.

Xmat A matrix with the genotypic composition of each of the m imputed data sets.

## Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### References

Little, R. J. A. and Rubin, D. B. (2002) Statistical analysis with missing data. Second edition, New York, John Wiley & sons.

Graffelman, J., S\'anchez, M., Cook, S. and Moreno, V. (2013) Statistical inference for Hardy-Weinberg proportions in the presence of missing genotype information. PLoS ONE 8(12): e83316. doi:10.1371/journal.pone.0083316

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Graffelman, J. (2015) Exploring Diallelic Genetic Markers: The HardyWeinberg Package. *Journal of Statistical Software* 64(3): 1-23. doi:10.18637/jss.v064.i03.

### See Also

```
HWChisq
```

### **Examples**

```
data(Markers)
## Not run:
set.seed(123)
Results <- HWMissing(Markers[,1],m=50,verbose=TRUE)$Res # no covariates, imputation assuming MCAR.
set.seed(123)
Results <- HWMissing(Markers[,1:3],m=50,verbose=TRUE)$Res # impute with two allele intensities.
set.seed(123)
Results <- HWMissing(Markers[,c(1,4,5)],m=50,verbose=TRUE)$Res # impute with two covariate SNPs
## End(Not run)

HWNetwork

Autosomal and X-chromosomal exact tests for HWE via a Network
algorithm</pre>
```

## **Description**

Program HWNetwork implements a network algorithm for efficient calculation of exact test p-values in HWE tests with multiple alleles.

### Usage

```
HWNetwork(a1, a2, ma = NULL, fe = NULL, gender = NULL, verbose = TRUE)
```

# Arguments

a1	the first allele (expressed as a number)
a2	the second allele (expressed as a number; NA if the variant is X chromosomal)
ma	alternative format: vector of male X chromosomal allele counts.
fe	triangular matrix of female genotype counts
gender	gender of the individual (1=male; 2=female)
verbose	be silent (verbose=FALSE) or informative (verbose=TRUE)

### **Details**

Function HWNetwork accepts data in two formats. Original genotype data (e.g. repeat numbers of microsatellites) can be supplied, or the data can be supplied in summarized form as a male genotype count vector and a female genotype count matrix. If one of the two male alleles is missing (NA) the variant will taken to be X-chromosomal. If all males have two alleles, the variant will taken to be autosomal.

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### Value

the exact p-value of the test.

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### References

Aoki, S. (2003) Network algorithm for the Exact Test of Hardy-Weinberg Proportion for Multiple Alleles. *Biometrical Journal* 45(4), pp. 471-490.

Engels, W. R. (2009) Exact Tests for Hardy-Weinberg Proportions. Genetics 183, pp. 1431-1441.

Graffelman, J. (2015) Exploring Diallelic Genetic Markers: The HardyWeinberg Package. *Journal of Statistical Software* 64(3): 1-23. doi:10.18637/jss.v064.i03.

### See Also

```
HWExact, HWExactStats
```

## **Examples**

```
#
# From vectors with counts of genotypes
#
data(TSIXTriAllelics)
ma <- as.matrix(TSIXTriAllelics[1,2:4])
names(ma) <- c("A","B","C")

fe <- TSIXTriAllelics[1,5:10]
names(fe) <- c("AA","AB","AC","BB","BC","CC")

fe <- HardyWeinberg:::toTriangularfixed(fe)

HWNetwork(ma=ma,fe=fe)</pre>
```

**HWPerm** 

Permutation test for Hardy-Weinberg equilibrium

## **Description**

Function HWPerm does a permutation test for Hardy-Weinberg equilibrium using a user-supplied test statistic.

### Usage

```
HWPerm(x, nperm = 17000, verbose = TRUE, x.linked = FALSE,
FUN = ifelse(x.linked,Chisquare.x,Chisquare), eps=1e-10, ...)
```

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## **Arguments**

X	A vector of genotype counts (AA,AB,BB)
nperm	The number of permutations
verbose	verbose = TRUE will print results, verbose = FALSE is silent.
x.linked	x.linked=FALSE indicates the marker is autosomal (default), and $x.linked=TRUE$ indicates it resides on the $X$ -chromosome.
FUN	An function call for calculating the test statistic for HWE (see examples below)
eps	Tolerance for comparison of floating point numbers (1e-10 by default)
	Additional parameters for the function call argument FUN

## **Details**

The set of alleles for the observed sample is permuted. Consequently, the test is conditional on allele frequency.

### Value

HWPerm returns a list with the components:

stat value of the chosen test statistic for the observed sample.

pval p-value of the permutation test.

## Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### References

Ziegler, A. & Konig, I.R. (2006) A statistical approach to genetic epidemiology. Wiley.

### See Also

HWChisq,HWExact,HWLratio

# **Examples**

```
x <- c(MM=298,MN=489,NN=213)
## Not run:
#
# use a default chi-square statistic
#
HW.test <- HWPerm(x,nperm=10000,verbose=TRUE)
#
# use a chi-square statistic with continuity correction.
#
HW.test <- HWPerm(x,nperm=10000,verbose=TRUE,
FUN=function(z) HWChisq(z,verbose=FALSE)$chisq,cc=0.5) #
#</pre>
```

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```
# use a likelihood ratio statistic.
#
HW.test <- HWPerm(x,nperm=10000,verbose=TRUE,
FUN=function(y) HWLratio(y,verbose=FALSE)$G2)
#
# use an exact test p-value
#
HWPerm(x,nperm=10000,verbose=TRUE,FUN=function(y) 1-HWExact(y,verbose=FALSE)$pval)
#
#
# Permutation test for a marker on the X chromosome
#
rs5968922 <- c(A=392, B=212, AA=275, AB=296, BB=80)
HW.test <- HWPerm(rs5968922,nperm=10000,x.linked=TRUE,verbose=TRUE)
## End(Not run)</pre>
```

HWPerm.mult

Permutation tests fo variants with multiple alleles

## **Description**

Function HWPerm.mult implements permutation tests for Hardy-Weinberg equilibrium for autosomal and X-chromosomal variants.

## Usage

```
HWPerm.mult(x, y = NULL, nperm = 17000, eps = 1e-10, verbose = TRUE, ...)
```

### **Arguments**

X	vector or triangular matrix with male genotype counts
У	vector or triangular matrix with female genotype counts
nperm	number of permutations (17.000 by default)
eps	a tolerance for the comparison of floating point numbers
verbose	print output or not
	addtional arguments

### Details

This function approximates exact test probabilities for joint tests for HWE and equality of allele frequencies for variants with multiple alleles. For purely bi-allelic variant HWPerm can be used which allows for more statistics than just probabilities.

If argument y is not specified, gender is considered irrelevant, and x contains total genotype counts. If x and y are specified, x should contain male genotype counts and y female genotype counts. x and y can be vectors if the variant is bi-allelic, but are assumed lower triangular if there are more than two alleles. x is still a vector if there are multiple alleles but the variant is X-chromosomal. See the examples given below.

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## Value

```
pofthesample probability of the observed sample pseudodist probabilities of simulated samples pval p-value
```

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### References

Graffelman, J. and Weir, B.S. (2017) Multi-allelic exact tests for Hardy-Weinberg equilibrium that account for gender. Molecular Ecology Resources. 18(3) pp. 461–473. doi:10.1111/1755-0998.12748

## See Also

HWPerm

## **Examples**

```
# bi-allelic autosomal
x1 < -c(AA=298, AB=489, BB=213)
## Not run:
out <- HWPerm.mult(x1)</pre>
## End(Not run)
# bi-allelic X-chromosomal
x2.m < -c(A=39, B=21)
x2.f <- toTriangular(c(AA=28, AB=30, BB=8))</pre>
## Not run:
out <- HWPerm.mult(x2.m,x2.f)</pre>
## End(Not run)
# autosomal k alleles not accounting for gender
x3 < -c(AA=12,AB=19,AC=13,BB=7,BC=5,CC=0)
x3 <- toTriangular(x3)
## Not run:
out <- HWPerm.mult(x3)</pre>
```

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```
## End(Not run)
# X-chromosomal k alleles
x4.m < - c(A=15, B=17, C=24)
x4.f <- toTriangular(c(AA=4,AB=2,AC=13,BB=6,BC=19,CC=4))
## Not run:
out <- HWPerm.mult(x4.m,x4.f)</pre>
## End(Not run)
# Autosomal k alleles accounting for gender
x5.m <- toTriangular(c(AA=12,AB=19,AC=13,BB=7,BC=5,CC=0))</pre>
x5.f <- toTriangular(c(AA=8,AB=12,AC=13,BB=8,BC=7,CC=0))
## Not run:
out <- HWPerm.mult(x5.m,x5.f)</pre>
## End(Not run)
# Autosomal STR with multipe alleles
data(NistSTRs)
A1 <- NistSTRs[,1]
A2 <- NistSTRs[,2]
GenotypeCounts <- AllelesToTriangular(A1,A2)</pre>
print(GenotypeCounts)
## Not run:
out <- HWPerm.mult(GenotypeCounts)</pre>
## End(Not run)
```

**HWPosterior** 

Calculation of posterior probabilities and Bayes factors for Hardy-Weinberg tests at X-chromosomal variants.

# Description

Function HWPosterior calculates posterior probabilities and Bayes factors for tests for Hardy-Weinberg equilibrium of autosomal and X-chromosomal variants.

#### Usage

```
HWPosterior(males, females, verbose = TRUE, prior.af = c(0.5,0.5), prior.gf = c(0.333,0.333,0.333), x.linked = FALSE, precision = 0.05)
```

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## **Arguments**

males	A vector of male genotype counts. For autosomal variants, this should be a three-element named vector like (AA,AB,BB); for X-chromosomal variants it should be a two-element vector giving the counts of the hemizygous genotypes like (A,B).
females	A vector of female genotype counts. This should be a three-element named vector like $(AA,\!AB,\!BB)$
verbose	prints results if verbose = TRUE
prior.af	Beta prior parameters for male and female allele frequencies
prior.gf	Dirichlet prior parameters for female genotype frequencies
x.linked	logical indicating whether the variant is autosomal or X-chromosomal
precision	precision parameter for marginal likelihoods that require numeric integration

#### **Details**

For X-chromosomal variants, four possible models are considered, and the posterior probabilities and Bayes factors for each model are calculated.

For autosomal variants, ten possible scenarios are considered, and the posterior probabilities for all models are calculated.

In general, default Dirichlet priors are used for genotype frequencies, and beta prior are used for allele frequencies.

### Value

For X-chromosomal variants, a matrix with posterior probabilities and Bayes factors will be produced. For autosomal variants, a vector of posterior probabilities is produced.

#### Author(s)

### References

Puig, X., Ginebra, J. and Graffelman, J. (2017) A Bayesian test for Hardy-Weinberg equilibrium of bi-allelic X-chromosomal markers. Heredity 119(4):226–236. doi:10.1038/hdy.2017.30.

Puig, X., Ginebra, J. and Graffelman, J. (2019) Bayesian model selection for the study of Hardy-Weinberg proportions and homogeneity of gender allele frequencies. Heredity 123(5), pp. 549-564. doi:10.1038/s4143701902320

#### See Also

HWChisq, HWExact, HWExactStats

HWPower 65

## **Examples**

```
#
# An X-chromosomal example
#
males <- c(A=43,B=13)
females <- c(AA=26,AB=19,BB=3)
out <- HWPosterior(males,females,verbose=TRUE,x.linked=TRUE)
#
# An autosomal example
#
data(JPTsnps)
males <- JPTsnps[1,1:3]
females <- JPTsnps[1,4:6]
post.prob <- HWPosterior(males,females,x.linked=FALSE)</pre>
```

**HWPower** 

Compute the power of a test for Hardy-Weinberg equilibrium.

## **Description**

HWPower is a function that computes the power of a test for Hardy-Weinberg equilibrium.

## Usage

```
HWPower(n = 100, nA = 100, pA = 0.5, y = c(AA=25,AB=50,BB=25), alpha = 0.05, theta = 4, f = NULL, test = "exact", alternative = "two.sided", pvaluetype = "selome", cc = 0.5)
```

# Arguments

n	The sample size
nA	The minor allele count
pA	The minor allele frequency
У	A sample of genotype counts (AA,AB,BB)
alpha	The significance level (0.05 by default)
theta	The degree of disequilibrium (theta = $4$ is equilibrium, theta > $4$ is heterozygote excess, theta < $4$ is heterozygote dearth)
f	The inbreeding coefficient. Overrules theta if specified.
test	The type of test for which power is to be computed. Can be "exact" (default) or "chisq" (chi-square)
alternative	The nature of the alternative hypothesis ("two.sided" (default), "greater" or "less")
pvaluetype	The type of p-value used in an exact test ("selome", "dost" or "midp")
сс	Continuity correction parameter for the chi-square test (0.5 by default)

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### **Details**

HWPower uses the Levene-Haldane distribution (distribution of the number of heterzygotes given the minor allele count) for computing power.

HWPower can be used in three different way. In principle, the power is calcuted on the basis of the sample size (n) and the minor allele count (nA). Alternatively, the user may specify sample size (n) and minor allele frequency (pA). Finally, power can also be calculated directly from a sample of genotype counts. In that case the calculated power is the power for a sample of the given sample size and minor allele count. The three ways to use HWPower are illustrated in the example section.

### Value

if test = "exact" the power of the exact test is computed for the given significance level and minor allele count.

if test = "chisq" the power of the chi-square test is computed for the given significance level and minor allele count.

### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### References

Graffelman, J. and Moreno, V. (2013) The Mid p-value in exact tests for Hardy-Weinberg proportions. *Statistical Applications in Genetics and Molecular Biology* 12(4): 433-448.

Graffelman, J. (2015) Exploring Diallelic Genetic Markers: The HardyWeinberg Package. *Journal of Statistical Software* 64(3): 1-23. doi:10.18637/jss.v064.i03.

### See Also

**HWExact** 

### **Examples**

```
pw.chisq <- HWPower(n=100,nA=100,alpha=0.05,test="chisq",theta=16)
print(pw.chisq)
pw.exact <- HWPower(n=100,nA=100,alpha=0.05,test="exact",theta=16,pvaluetype="selome")
print(pw.exact)
pw.exact <- HWPower(n=100,nA=100)
print(pw.exact)
pw.exact <- HWPower(n=100,pA=0.5)
print(pw.exact)
pw.exact <- HWPower(y=c(AA=25,AB=50,BB=25))
print(pw.exact)</pre>
```

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$ \begin{tabular}{ll} HWQqplot & A Q-Q plot for Hardy-Weinberg equilibrium \\ \end{tabular} $	
---	--

## **Description**

HWQqplot creates a Q-Q plot for the p-values obtained in an Exact test for Hardy-Weinberg equilibrium. Empirical p-values are plotted against multiple simulated quantiles of the theoretical p-value distribution.

## Usage

```
HWQqplot(X, nsim = 100, fit = "curve", logplot = FALSE,
main = "Q-Q plot for HWE", mm = NULL, pvaluetype = "selome", ...)
```

## **Arguments**

X	Data matrix with genotype counts, one row for each sample, 3 columns
nsim	Number of samples drawn from the null distribution (100 by default)
fit	If fit is set to "line" straight lines will be fitted to the simulated samples, if set to "curve", ascending curves will be shown.
logplot	If logplot is set to true, then the $\log 10$ of the p-values will be used in the plot. If not, untransformed p-values will be used.
main	Title for the plot
mm	Maximal value for x and y axis in the plot
pvaluetype	Type of p-value to be used in an exact test. Can be "selome" (default), "midp" or "dost".
	Any additional arguments for the plot instruction

## **Details**

HWQqplot constructs a Q-Q plot of the p-values of an exact test for Hardy-Weinberg equilibrium. Under the null, this p-value is not uniform. HWQqplot samples from the theoretical null distribution, taking into account that markers may vary in allele frequency and in sample size (due to missing values). For each simulated sample a grey curve or line is shown. A green reference line with intercept 0 and slope 1 is also shown in the plot.

## Value

**NULL** 

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

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### References

Rohlfs, R.V. and Weir, B.S. (2008) Distributions of Hardy-Weinberg equilibrium test statistics. Genetics 180, pp. 1609-1616.

## See Also

```
HWTernaryPlot, HWExact, qqplot
```

## **Examples**

```
## Not run:
set.seed(1234)
n <- 200 # sample size
m <- 100 # number of markers
X <- HWData(n,m)
HWQqplot(X,logplot=TRUE,pvaluetype="selome",main="Q-Q Plot for HWE")
## End(Not run)</pre>
```

HWStr

Testing a set of microsatellites (STRs) for Hardy-Weinberg equilibrium

## **Description**

Function HWStr is a wrapper function arount HWPerm.mult and HWChisq in order to test a set of STRs for Hardy-Weinberg equilibrium.

# Usage

```
HWStr(X, test = "permutation", verbose = FALSE, ...)
```

# Arguments

X	A data matrix with STRs in columns, with there two alleles coded in successive columns.
test	permutation for a permutation test, or "chisq" for a chisquare test
verbose	be silent if set to FALSE or print info for each STR if TRUE.
	possible extra parameters to be passed on to HWPerm.mult such as nperm or eps.

### **Details**

It is recommended to test by a permutation test. By default, 17.000 permutations are used. Exact testing is not implemented. Missing genotypes are excluded on a per STR basis.

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## Value

A data frame with the

strnames STR name

N sample size

Nt number of alleles

MinorAF minor allele frequency

MajorAF major allele frequency

Ho observed heterozygosity

He expected heterozygosity (without bias correction)

Hp Shannon index of allele frequencies

pval p-value of the HWE test

### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### See Also

```
HWPerm.mult, HWChisq
```

## **Examples**

```
data(NistSTRs)
## Not run:
Results <- HWStr(NistSTRs,test="permutation")
## End(Not run)</pre>
```

**HWStrata** 

Asymptotic test for HWE across strata for a single biallelic marker

## **Description**

Function HWStrata implements Olson's asymptotic test for HWE for a stratified sample of single biallelic polymorphism.

## Usage

```
HWStrata(X, verbose = TRUE)
```

# Arguments

X A three-column matrix of genotype counts (e.g. with columns AA, AB, BB)

verbose print output if verbose = TRUE

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### **Details**

See the references for the related homogeneity assumption.

#### Value

T2 The test statistic pval The p-value

## Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

### References

Olson J.M. (1993) Testing the Hardy-Weinberg law across strata. Annals of Human Genetics 57(4):291-295.

Olson, J.M. and Foley, M. (1996) Testing for homogeneity of Hardy-Weinberg disequilibrium using data sampled from several populations. Biometrics 52(3) pp. 971-979.

### See Also

HWExactStats, HWChisqStats

## **Examples**

```
#
# Test across strata
#
data("Glyoxalase")
Glyoxalase <- as.matrix(Glyoxalase)
HWStrata(Glyoxalase)
#
# Stratified exact testing, testing each sample
#
HWExactStats(Glyoxalase)</pre>
```

HWTernaryPlot

Ternary plot with the Hardy-Weinberg acceptance region

# Description

HWTernaryPlot is a routine that draws a ternary plot for three-way genotypic compositions (AA,AB,BB), and represents the acceptance region for different tests for Hardy-Weinberg equilibrium (HWE) in the plot. This allows for graphical testing of a large set of markers (e.g. SNPs) for HWE. The (non) significance of the test for HWE can be inferred from the position of the marker in the ternary plot. Different statistical tests for HWE can be done graphically with this routine: the ordinary chisquare test, the chisquare test with continuity correction and the Haldane's exact test.

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### Usage

```
HWTernaryPlot(X, n = NA, addmarkers = TRUE, newframe = TRUE, hwcurve = TRUE, vbounds = FALSE, mafbounds = FALSE, mafvalue = 0.05, axis = 0, region = 1, vertexlab = colnames(X), alpha = 0.05, vertex.cex = 1, pch = 19, cc = 0.5, markercol = "black", markerbgcol = "black", cex = 0.75, axislab = "", verbose = FALSE, markerlab = NULL, markerpos = NULL, mcex = 1, connect = FALSE, curvecols = rep("black",5), signifcolour = TRUE, patternsigsymbol = 19, curtyp = "solid", ssf = "max", pvaluetype = "selome", grid = FALSE, gridlabels = TRUE, patternramp = FALSE, axisticklabels = FALSE, ...)
```

### **Arguments**

X a matrix of n genotypic compositions or counts. If it is a matrix of compositions,

X should have (n rows that sum 1, and 3 columns, with the relative frequencies of AA, AB and BB respectively. Argument n should be supplied as well. If X is a matrix of raw genotypic counts, it should have 3 columns with the absolute counts of AA, AB and BB respectively. Argument n may be supplied and will be used for painting acceptance regions. If not supplied n is computed from the

data in X.

n the samples size (for a complete composition with no missing data).

addmarkers represent markers by dots in the triangle (addmarkers=TRUE) or not

(addmarkers=FALSE).

newframe allows for plotting additional markers in an already existing ternary plot. Over-

plotting is achieved by setting newframe to FALSE. Setting newframe =  $\mathsf{TRUE}$ 

(default) will create a new ternary plot.

hwcurve draw the HW parabola in the plot (hwcurve=TRUE) or not (hwcurve=FALSE).

vbounds indicate the area corresponding to expected counts > 5 (vbounds=TRUE) or not

(vbounds=FALSE).

mafbounds indicate the area corresponding to MAF < mafvalue.

mafvalue a critical value for the minor allele frequency (MAF).

axis draw a vertex axis

0 = no axis is drawn 1 = draw the AA axis 2 = draw the AB axis 3 = draw the BB axis

4 = draw tick on the A-B axis

region the type of acceptance region to be delimited in the triangle

0 = no acceptance region is drawn

1 = draw the acceptance region corresponding to a Chi-square test

2 = draw the acceptance region corresponding to a Chi-square test with continuity correction

 $3=\mbox{draw}$  the acceptance region corresponding to a Chi-square test with continuity correction for D>0

4 = draw the acceptance region corresponding to a Chi-square test with continuity

ity correction for D < 0

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5 = draw the acceptance regions for all preceding tests simultaneously

6 = draw the acceptance region corresponding to a Chi-square test with continuity correction with the upper limit for D > 0 and the lower limit for D < 0 7 = draw the acceptance region corresponding to a two-sided exact test

vertexlab labels for the three vertices of the triangle

alpha significance level (0.05 by default)

vertex.cex character expansion factor for the labels of the vertices of the triangle.

pch the plotting character used to represent the markers.

value for the continuity correction parameter (0.5 by default).

wector with colours for the marker points in the triangle.

markerbgcol vector with background colours for the marker points in the triangle.

cex expansion factor for the marker points in the triangle.

axislab a label to be put under the horizontal axis.

verbose print information on the numerically found cut-points between curves of the

acceptance region and the edges of the triangle.

markerlab labels for the markers in the triangle.

markerpos positions for the marker labels in the triangle (1,2,3 or 4).

mcex character expansion factor for the labels of the markers in the ternary plot.

connect the represented markers by a line in the ternary plot.

curvecols a vector with four colour specifications for the different curves that can be used

to delimit the HW acceptance region. E.g. curvecols=c("red", "green", "blue", "black", "purple") will paint the Hardy-Weinberg curve red, the limits of the acceptance region for

will paint the Hardy-Weinberg curve red, the limits of the acceptance region for an ordinary chi-square test for HWE green, the limits of the acceptance region for a chi-square test with continuity correction when D > 0 blue and the limits of the acceptance region for a chi-square test with continuity correction when D

< 0 black, and the limits of the exact acceptance region purple.

signifcolour colour the marker points automatically according to the result of a signifance

test (green markers non-siginficant, red markers significant). signifcolour

only takes effect if region is set to 1, 2 or 7.

patternsigsymbol

plotting character used to represent significant makers when patternramp=TRUE

curtyp style of the drawn curves ("dashed", "solid", "dotted", ...)

ssf sample size function ("max", "min", "mean", "median", ...). Indicates how

the sample size for drawing acceptance regions is determined from the matrix

of counts.

pvaluetype method to compute p-values in an exact test ("dost" or "selome") draw a reference grid for genotype frequencies at (0.2,0.4,0.6,0.8)

gridlabels if grid=TRUE plot labels at the grid

patternramp paint a colour ramp for patterns of genotype frequencies. This will ignore

signifcolour, and use colour to represent marker frequencies. The ramp is

a green-red gradient for the relative frequency of the pattern.

axisticklabels place numeric labels for allele frequencies on the A-B axis.

.. other arguments passed on to the plot function (e.g. main for a main title).

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#### **Details**

HWTernaryPlot automatically colours significant markers in red, and non-significant markers in green if region is set to 1, 2 or 7.

#### Value

minp minimum allele frequency above which testing for HWE is appropriate (ex-

pected counts exceeding 5).

maxp maximum allele frequency below which testing for HWE is appropriate.

inrange number of markers in the appropriate range.

percinrange percentage of markers in the appropriate.

nsignif number of significant markers (only if region equals 1,2 or 7.)

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### References

Graffelman, J. and Morales, J. (2008) Graphical Tests for Hardy-Weinberg Equilibrium Based on the Ternary Plot. *Human Heredity* 65(2):77-84.

Graffelman, J. (2015) Exploring Diallelic Genetic Markers: The HardyWeinberg Package. *Journal of Statistical Software* 64(3): 1-23. doi:10.18637/jss.v064.i03.

# See Also

**HWChisq** 

```
#
# Ternary plot with 1000 SNPs and HWE curve
#

X <- HWData(nm=1000,n=100)
HWTernaryPlot(X,100,region=0,vertex.cex=2,pch=1)

#
# Genotype frequency pattern of simulated SNPs with uniform
# allele frequency distribution
#

X <- HWData(nm=1000,n=100)
HWTernaryPlot(X,patternramp = TRUE)

#
# Genotype frequency pattern of simulated SNPs with skewed
# allele frequency distribution
#</pre>
```

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```
X <- HWData(nm=1000, n=100, shape1=1, shape2=20)</pre>
HWTernaryPlot(X,patternramp = TRUE)
# Genotype frequency pattern of SNPs simulated under HWE with
# allele frequency of 0.50
X <- HWData(nm=1000, n=100, p=0.25)</pre>
HWTernaryPlot(X,patternramp = TRUE)
# Genotype frequency pattern of SNPs simulated under HWE with
# allele frequency of 0.25
X <- HWData(nm=1000, n=100, p=0.25)</pre>
HWTernaryPlot(X,patternramp = TRUE)
# Genotype frequency pattern of SNPs simulated under HWE with
\# allele frequency of 0.25, using a triangle to mark significant markers
X <- HWData(nm=1000, n=100, p=0.25)</pre>
HWTernaryPlot(X,patternramp = TRUE, region=1, patternsigsymbol = 2)
# Ternary plot of 1000 SNPs simulated under HWE and uniform allele frequency, with
# acceptance region for a chi-square test and significant markers in red.
X <- HWData(nm=1000,n=100)</pre>
HWTernaryPlot(X)
# Ternary plot of 100 SNPs simulated under HWE and uniform allele frequency, with
# acceptance region for an exact test and significant markers in red.
X <- HWData(nm=100,n=100)</pre>
## Not run: HWTernaryPlot(X,region=7)
```

**HWTriExact** 

Exact test for Hardy-Weinberg equilibrium and equality of allele frequencies for tri-allelic variants.

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#### **Description**

Function HWTriExact does a standard exact test for Hardy-Weinberg equilibrium of a tri-allelic variant, and also does joint exact tests for equilibrium and equality of allele frequencies if the genotype counts are given separately for both sexes

# Usage

```
HWTriExact(x, y = NULL, eps = 1e-10, nperm = 17000, verbose = TRUE)
```

#### Arguments

x vector with 6 genotype counts (AA,AB,AC,BB,BC,CC)

y vector with 6 or 3 genotype counts (AA,AB,AC,BB,BC,CC) or (A,B,C)

eps a tolerance that can be set for comparing exact probabilities

nperm number of permutations (only relevant for autosomal stratified by gender)

verbose print output or not

#### **Details**

If only x is specified, an exact test for an autosomal variant with three alleles will be performed.

If both x and y are supplied as vectors with 6 elements, a permutation test for HWE and equality of allele frequencies (EAF) for an autosomal variant is performed, using nperm permutations. The distribution of the probabilities is returned in pseudodist. The computational cost of a completed enumeration algorithm can be prohibitive in this case.

If x is supplied as a length 6 vector, and y as a length 3 vector, the variant is assumed to be X-chromosomal, x containing female genotype counts and y containing male genotype counts. In this case a joint exact test for HWE and EAF for an X-chromosomal tri-allelic variant is executed.

See the examples in the example section below.

#### Value

pval The p-value of the sample

pseudodist Distribution of probabilities obtained by simulation

pofthesample The probability of the observed sample

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

# References

Graffelman, J. and Weir, B.S. (2017) Multi-allelic exact tests for Hardy-Weinberg equilibrium that account for gender. Molecular Ecology Resources. 18(3) pp. 461–473. doi:10.1111/1755-0998.12748.

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# See Also

```
HWPerm.mult
```

# **Examples**

```
# # Autosomal tri-allelic (not accounting for gender)
#

x <- c(AA=20,AB=31,AC=26,BB=15,BC=12,CC=0)
## Not run: out <- HWTriExact(x)

# # Autosomal tri-allelic accounting for gender
# males <- c(A=1,B=21,C=34)
females <- c(AA=0,AB=1,AC=0,BB=8,BC=24,CC=15)
## Not run: out <- HWTriExact(females,males)

# # X-chromosomal tri-allelic accounting for gender
# males <- c(A=1,B=21,C=34)
females <- c(AA=0,AB=1,AC=0,BB=8,BC=24,CC=15)
## Not run: out <- HWTriExact(females,males)</pre>
```

ifisherz

Inverse Fisher z transformation

# **Description**

Calculates the inverse of Fisher's z transformation

# Usage

```
ifisherz(y)
```

#### **Arguments**

```
y a real number
```

# Value

```
a correlation coefficient in the range (-1,1)
```

# Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

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#### See Also

cor

#### **Examples**

```
r <- 0.5
print(ifisherz(fisherz(r)))</pre>
```

is.mono

Detects autosomal and X-chromosomal monomorphic variants

# **Description**

Function is. mono tests if a bi-allelic variant is monomorphic or not

#### Usage

```
is.mono(x)
```

#### **Arguments**

Х

a vector of three or five genotype counts ((AA,AB,BB) or (A,B,AA,AB,BB)), or a three-column or five-column matrix with genotype counts (variants in rows, columns)

# **Details**

is mono assumes autosomal variants are coded in three-element vectors or three-column matrices, whereas X-chromosomal variants are coded in five-element vectors or five-column matrices.

#### Value

A logical or vector of logicals

# Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

```
#
# a polymorphic autosomal marker
#
x <- c(AA=10,AB=20,BB=10)
print(is.mono(x))
#
# a monomorphic autosomal marker
#
x <- c(AT=0,AA=100,TT=0)</pre>
```

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```
print(is.mono(x))
# an autosomal marker with only heterozygotes
x <- c(AT=100, AA=0, TT=0)
print(is.mono(x))
# a matrix with low maf autosomal markers
set.seed(123)
X <- HWData(100,50,shape1=1,shape2=20)</pre>
number.monomorphics <- sum(is.mono(X))</pre>
print(number.monomorphics)
# a polymorphic X chromosomal marker
x <- c(G=24,C=26,GG=12,CC=13,GC=25)
print(is.mono(x))
# another polymorphic X chromosomal marker
x <- c(G=24,C=1,GG=25,CC=0,GC=0)
is.mono(x)
# a monomorphic X chromosomal marker
x <- c(G=24, C=0, GG=12, CC=0, GC=0)
is.mono(x)
# a matrix with low maf X-chromosomal markers
set.seed(123)
Y <- HWData(100,50,shape=1,shape2=20,x.linked = TRUE)
number.monomorphics <- sum(is.mono(Y))</pre>
print(number.monomorphics)
```

JPTmultiallelicsChr7 Multi-allelic autosomal variants of the Japanese population of the 1000 genomes project

# Description

JPTtriallelicsChrX contains three selected multi-allelic variants on chromosome 7 from the Japanese sample of the 1000 genomes project.

# Usage

```
data("JPTmultiallelicsChr7")
```

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#### **Format**

List object with fields m4,f4; m5,f5; m6,f6;

#### **Details**

The list object contains male and female genotype counts for 3 multi-allelic variants on chromosome 7 of the JPT sample of the 1000 genomes project.

#### Source

The The 1000 genomes project.

#### References

Graffelman, J. and Weir, B.S. (2017) Multi-allelic exact tests for Hardy-Weinberg equilibrium that account for gender. doi: 10.1101/172874. Table 10.

# **Examples**

```
data(JPTmultiallelicsChr7)
str(JPTmultiallelicsChr7)
```

JPTmultiallelicsChrX Multi-allelic X-chromosomal variants of the Japanese population of the 1000 genomes project

#### **Description**

JPTtriallelicsChrX contains four selected multi-allelic variants on the X chromosome from the Japanese sample of the 1000 genomes project.

#### Usage

```
data("JPTmultiallelicsChrX")
```

# **Format**

List object with fields m4,f4; m5,f5; m6,f6; m7,f7

#### Details

The list object contains male and female genotype counts for four multi-allelic variants of the JPT sample of the 1000 genomes project.

# Source

The The 1000 genomes project.

JPTsnps

#### References

Graffelman, J. and Weir, B.S. (2017) Multi-allelic exact tests for Hardy-Weinberg equilibrium that account for gender. doi: 10.1101/172874. Table 7.

#### **Examples**

```
data(JPTmultiallelicsChrX)
m4 <- JPTmultiallelicsChrX$m4
f4 <- JPTmultiallelicsChrX$f4</pre>
```

**JPTsnps** 

Bi-allelic SNPs from a Japanese population stratified by gender

# **Description**

JPTsnps contains genotype counts for the two sexes of ten single nucleotide polymorphisms of the Japanese (JPT) sample of the 1000 Genomes project, consisting of 56 males and 48 females.

The first three columns contain the male genotype counts, and the last three columns contain the female genotype counts.

# Usage

```
data("JPTsnps")
```

#### **Format**

data frame

#### Source

The The 1000 genomes project.

#### References

Puig, X., Ginebra, J. and Graffelman, J. (2019) Bayesian model selection for the study of Hardy-Weinberg proportions and homogeneity of gender allele frequencies.

```
data(JPTsnps)
```

JPTtriallelicsChr7 81

	Tri-allelic variants on chromosome 7 of the Japanese (JPT) sample of the 1000 genomes project
--	---

#### **Description**

JPTtriallelics contains six selected tri-allelic variants on chromosome 7 from the Japanese sample of the 1000 genomes project.

# Usage

```
data("JPTtriallelicsChr7")
```

#### **Format**

A data frame with 6 observations on the following 14 variables.

```
id RS identifier
```

pos position in base pairs

mAA number of AA males

mAB number of AB males

mAC number of AC males

mBB number of BB males

mBC number of BC males

mCC number of CC males

fAA number of AA females

fAB number of AB females

fAC number of AC females

fBB number of BB females

fBC number of BC females

fCC number of CC females

#### **Source**

The The 1000 genomes project.

#### References

Graffelman, J. and Weir, B.S. (2017) Multi-allelic exact tests for Hardy-Weinberg equilibrium that account for gender. doi: 10.1101/172874. Table 9.

```
data(JPTtriallelicsChr7)
str(JPTtriallelicsChr7)
```

82 JPTtriallelicsChrX

ple of the 1000 genomes project	JPTtriallelicsChrX	Tri-allelic variants on the X-chromosome of the Japanese (JPT) sample of the 1000 genomes project
---------------------------------	--------------------	---

# **Description**

JPTtriallelicsChrX contains five selected tri-allelic variants on the X chromosome from the Japanese sample of the 1000 genomes project.

# Usage

```
data("JPTtriallelicsChrX")
```

#### **Format**

A data frame with 5 observations on the following 12 variables.

id Identifier of the polymorphism

pos Position of the polymorphism in base pairs

chr Chromosome

A Number of males with A genotype

B Number of males with B genotype

C Number of males with C genotype

AA Number of AA females

AB Number of AB females

AC Number of AC females

BB Number of BB females

BC Number of BC females

CC Number of CC females

#### **Source**

The The 1000 genomes project.

#### References

Graffelman, J. and Weir, B.S. (2017) Multi-allelic exact tests for Hardy-Weinberg equilibrium that account for gender. doi: 10.1101/172874. Table 6.

```
data(JPTtriallelicsChrX)
str(JPTtriallelicsChrX)
```

mac 83

mac

Compute the minor allele count.

# Description

mac computes the smallest allele count for a given vector of genotype counts.

# Usage

mac(X)

#### **Arguments**

Χ

a vector or matrix with genotype counts (AA, AB, BB)

# Value

a vector of the minor allele counts

#### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### See Also

maf

# Examples

```
X <- as.vector(rmultinom(1,100,c(0.5,0.4,0.1)))
names(X) <- c("AA","AB","BB")
print(X)
print(mac(X))</pre>
```

maf

Function to compute minor allele frequencies

# Description

Function maf computes the minor allele frequency for a matrix or vector of genotype counts.

# Usage

```
maf(x, option = 1, verbose = FALSE)
```

84 maf

# **Arguments**

x a vector or matrix of with genotype counts

option determines output that is returned. option=1 calculates the minor allele fre-

quency; option=2 returns all allele frequencies; option=3 returns the allele

counts.

verbose be silent (verbose=FALSE) or not.

#### Value

a vector of minor allele frequencies or minor allele counts.

## Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

# See Also

```
mac, MakeCounts
```

```
# MAF of a single random marker
set.seed(123)
X \leftarrow as.vector(rmultinom(1,100,c(0.5,0.4,0.1)))
names(X) <- c("AA", "AB", "BB")</pre>
print(X)
print(maf(X))
# MAF of MN bloodgroup counts
x < -c(MM=298, MN=489, NN=213)
maf(x)
# Both allele frquuencies, ordered from minor to major
maf(x,2)
# allele counts of MN bloodgroup counts, order from minor to major
maf(x,3)
# MAF of single triallelic marker in triangular format
x <- c(AA=20, AB=52, AC=34, BB=17, BC=51, CC=26)
print(x)
GT <- toTriangular(x)</pre>
print(GT)
maf(GT)
#
```

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```
# extract all allele frequencies
#
maf(GT,option=2)
#
# extract all allele counts
#
maf(GT,option=3)
#
# Calculate the MAF for 10 random SNPs under HWE
#
set.seed(123)
Z <- HWData(nm=10)
print(Z)
#
# vector with minor allele frequencies, one per marker
#
maf(Z)
#
# Matrix with minor and major allele frequencies, one row per marker
#
maf(Z,2)
#
# Matrix with minor and major allele count, one row per marker
#
maf(Z,3)</pre>
```

MakeCounts

Create genotype counts from bi-allelic marker data

# Description

MakeCounts creates a matrix of genotype counts, with one row for each bi-allelic marker, containing 4 columns with the counts AA, AB, BB and NA (missings) respectively

# Usage

```
MakeCounts(X, alleles, pos1 = 1, pos2 = 3, coding = c(AA=0,AB=1,BB=2), sep = "")
```

# Arguments

X	A matrix or dataframe with bi-allelic genotyping information, markers in columns, individuals in rows
alleles	a vector of alleles for each marker (e.g. c("A/T","A/G",)). Only relevanit if X is a matrix with text entries.
pos1	position of the first allele in the allele string (1 by default)
pos2	position of the second allele in the allele string (3 by default)
coding	indicates how homozygotes and heterozygote are coded as numbers. Only relevant if X is a matrix with numeric entries.
sep	allele separator character for genotype data in text format ("" for AA; "/" for "A/A")

86 MakeFactor

#### **Details**

MakeCounts is thought for bi-allelic marker data only. Missings are should be coded by NA. It produces the right input for HWTernaryPlot.

Heterozygotes may be coded in the data as "AB" or "BA". Both entries will be counted as a heterozygote.

#### Value

A matrix of 4 columns

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### See Also

**HWTernaryPlot** 

# **Examples**

MakeFactor

Make factors from genotyping data

# **Description**

MakeFactor converts bi-allelic genetic marker data, whether coded numerically as (0,1,2) or as (GG,GT,TT), etc. into standard factors coded as AA, AB, BB.

#### Usage

```
MakeFactor(x, coding = c(0, 1, 2))
```

#### **Arguments**

x A vector containing genotyping results

coding Describes the numerical coding of the genotype data in order AA, AB and BB.

Only relevant if x is numerical

Markers 87

#### **Details**

If x is a factor, it will be coerced to a factor with levels AA, AB and BB. Important detail: the produced factors will have only those levels that are observed in the data. E.g., if genotyping results only consist of (0,1), then the resulting factor will not have level BB (which would be an empty category)

#### Value

A factor variable

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### See Also

MakeCounts

#### **Examples**

```
y <- c(1,1,0,0,2,2)
data.frame(y,MakeFactor(y))

y <- c(2,2,3,3,1,1)
data.frame(y,MakeFactor(y,coding=c(1,2,3)))

data(Markers)
data.frame(Markers[,1],MakeFactor(Markers[,1],coding=c(1,2,3)))</pre>
```

Markers

SNP data and intensities

#### **Description**

The dataframe contains the genotypes of 3 SNPs and two allele intensities of 146 individuals. The first column is a GT polymorphism that has missing values for several individuals. The second and third column (iG and iG) are the allele intensities of this polymorphism. Column 4 and 5 are covariate SNPs (an AC and an AG polymorphism) that have no missing values.

#### Usage

```
data(Markers)
```

#### **Format**

A data frame containing 146 rows and 5 columns

n.alleles

#### References

Graffelman, J. (2015) Exploring Diallelic Genetic Markers: The HardyWeinberg Package. Journal of Statistical Software, 64(3), 1-23. doi:10.18637/jss.v064.i03.

Mourant

Genotype frequencies for blood group locus MN

# Description

The dataframe contains the genotype frequencies MM, MN and NN for the MN blood group locus for 216 populations. The data are taken from table 2.5 in Mourant et al., using only entries with a sample size of at least 500.

#### Usage

```
data(Mourant)
```

#### **Format**

A data frame containing 216 observations.

#### **Source**

Mourant et al, Table 2.5

#### References

Mourant, A. E. and Kope\'c, A. C. and Domaniewska-Sobczak, K. (1976) The Distribution of the Human Blood Groups and other Polymorphisms. Second edition. Oxford University Press, London.

n.alleles

Number of alleles

#### **Description**

Function n. alleles determines the number of alleles in a named genotype vector.

#### Usage

```
n.alleles(x, ...)
```

# Arguments

x A named genotype vector (e.g. c(AA=10,AB=20,BB=5))

... extra arguments that are passed on to alleles

NistSTRs 89

# Value

integer

#### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### See Also

alleles

# **Examples**

```
x <- c(AA=25,AB=50,BB=25)
k <- n.alleles(x)
print(k)</pre>
```

NistSTRs

NIST autosomal STR data

# **Description**

NistSTRs contains 29 autosomal microsatellites (STRs) of individuals of Caucasian ancestry. The two alleles of an individual are separated into two columns for each STR.

#### Usage

```
data("NistSTRs")
```

# **Format**

A data frame with 361 observations on the following 58 variables.

```
'CSF1PO-1' First allele of CSF1PO
```

'CSF1P0-2' Second allele of CSF1PO

'D10S1248-1' First allele of D10S1248

'D10S1248-2' Second allele of D10S1248

'D12S391-1' First allele of D12S391

'D12S391-2' Second allele of D12S391

'D13S317-1' First allele of D13S317

'D13S317-2' Second allele of D13S317

'D16S539-1' First allele of D16S539

'D16S539-2' Second allele of D16S539

'D18S51-1' First allele of D18S51

'D18S51-2' Second allele of D18S51

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- 'D19S433-1' First allele of D19S433
- 'D19S433-2' Second allele of D19S433
- 'D1S1656-1' First allele of D1S1656
- 'D1S1656-2' Second allele of D1S1656
- 'D21S11-1' First allele of D21S11
- 'D21S11-2' Second allele of D21S11
- 'D22S1045-1' First allele of D22S1045
- 'D22S1045-2' Second allele of D22S1045
- 'D2S1338-1' First allele of D2S1338
- 'D2S1338-2' Second allele of D2S1338
- 'D2S441-1' First allele of D2S441
- 'D2S441-2' Second allele of D2S441
- 'D3S1358-1' First allele of D3S1358
- 'D3S1358-2' Second allele of D3S1358
- 'D5S818-1' First allele of D5S818
- 'D5S818-2' Second allele of D5S818
- 'D6S1043-1' First allele of D6S1043
- 'D6S1043-2' Second allele of D6S1043
- 'D7S820-1' First allele of D7S820
- 'D7S820-2' Second allele of D7S820
- 'D8S1179-1' First allele of D8S1179
- 'D8S1179-2' Second allele of D8S1179
- 'F13A01-1' First allele of F13A01
- 'F13A01-2' Second allele of F13A01
- 'F13B-1' First allele of F13B
- 'F13B-2' Second allele of F13B
- 'FESFPS-1' First allele of FESFPS
- 'FESFPS-2' Second allele of FESFPS
- 'FGA-1' First allele of FGA
- 'FGA-2' Second allele of FGA
- 'LPL-1' First allele of LPL
- 'LPL-2' Second allele of LPL
- 'Penta\_C-1' First allele of Penta\_C
- 'Penta\_C-2' Second allele of Penta\_C
- 'Penta\_D-1' First allele of Penta\_D
- 'Penta\_D-2' Second allele of Penta D
- 'Penta\_E-1' First allele of Penta\_E

order.auto 91

```
'Penta_E-2' Second allele of Penta_E
'SE33-1' First allele of SE33
'SE33-2' Second allele of SE33
'TH01-1' First allele of TH01
'TH01-2' Second allele of TH01
'TPOX-1' First allele of TPOX
'TPOX-2' Second allele of TPOX
'vWA-1' First allele of vWA
'vWA-2' Second allele of vWA
```

#### **Source**

http://strbase.nist.gov

# References

Hill, C. et al. (2013) U.S. population data for 29 autosomal STR loci. Forensic science international: Genetics 497 7(3):e82-3.

# **Examples**

data(NistSTRs)

order.auto

Reordering of autosomal genotype counts

# **Description**

Function order auto reorders a named vector of three genotype counts, such that the sequence (minor homozygote, heterozygote, major heterzygote) is establised.

# Usage

```
order.auto(X)
```

# Arguments

Χ

a named vector of genotype counts (e.g. c(AA=25,AB=50,BB=25))

#### Value

a vector

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

92 order.x

#### References

Graffelman, J. (2015) Exploring Diallelic Genetic Markers: The HardyWeinberg Package. *Journal of Statistical Software* 64(3): 1-23. doi:10.18637/jss.v064.i03.

#### See Also

```
order.x
```

#### **Examples**

```
x <- c(MN=489,MM=298,NN=213)
print(x)
y <- order.auto(x)
print(y)</pre>
```

order.x

Reordering of X-chromosomal genotype counts

# **Description**

Function order.x reorders a named vector of three genotype counts, such that the sequence (minor hemizygote, major hemizygote, minor homozygote, heterozygote, major heterzygote) is establised.

# Usage

```
order.x(X)
```

# **Arguments**

Χ

a named vector of genotype counts (e.g. c(A=10,B=10,AA=25,AB=50,BB=25))

# Value

a vector

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### References

Graffelman, J. (2015) Exploring Diallelic Genetic Markers: The HardyWeinberg Package. *Journal of Statistical Software* 64(3): 1-23. doi:10.18637/jss.v064.i03.

#### See Also

```
order.auto
```

qqunif 93

# **Examples**

```
x <- c(A=392, B=212, AA=275, AB=296, BB=80)
print(x)
y <- order.x(x)
print(y)</pre>
```

qqunif

Q-Q plot for a uniform distribution.

# Description

qqunif makes a Q-Q plot against a uniform distribution for the supplied data vector.

# Usage

#### **Arguments**

x	The data vector
logplot	If logplot is set to true, then the log10 of the p-values will be used in the plot. If not, untransformed p-values will be used.
lbs	A vector of labels for the points in the Q-Q plot
texton	Logical indicating whether labels should be plotted or not
xylim	Shared upper limit x and y axis in the plot
main	Title for the plot
plotline	Setting plotline=0 plots a (0,1) line; plotline=1 plots a robust fit,
xlab	label for the x axis
ylab	label for the y axis
colvec	vector with colours for the points in the QQ plot
colline	colour for the line in the plot
	Any additional arguments for the plot instruction

#### Value

```
pvals observed probabilities epvals expected probabilities
```

94 recode

#### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### See Also

qqnorm

# **Examples**

```
x <- runif(1000)
z <- qqunif(x)</pre>
```

recode

Recode genotype information

# Description

function recode recodes bi-allelic genetic marker information expressed as strings (e.g. "AA", "AB", "BB") into numerical form.

# Usage

```
recode(X, alleles, values = c(0, 1, 2), pos1 = 1, pos2 = 3, minor = FALSE, verbose = FALSE)
```

# **Arguments**

X	A matrix or dataframe of bi-allelic markers, individuals in rows, markers in columns
alleles	a vector with the alleles for each marker (e.g. c("A/T", "A/G", etc))
values	a vector of numerica values for AA, AB and BB, ((0,1,2) by default).
pos1	position of the first allele in the allele string (1 by default).
pos2	position of the second allele in the allele string (3 by default).
minor	coding is according to the number of copies of the minor allele. if minor = TRUE, the value of 2 reflects two copies of the minor allele, and the value 0 reflects no copies of the minor allele.
verbose	print progress on the conversion or not.

# **Details**

recode is written for bi-allelic marker data only. Heterozygotes may be coded both as AB or BA. By default, the second allele specified (e.g. "T" in "A/T") is counted in the recoding, and homozygotes AA are coded as 0 and homozygotes TT as 2.

shannon 95

# Value

A numerical matrix, individuals in rows, markers in columns

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### See Also

MakeCounts

# **Examples**

shannon

Shannon index

# Description

Function shannon calculates the Shannon index and its variance for a vector of counts.

# Usage

```
shannon(x)
```

# Arguments

x a vector of counts

#### Value

Hp the sample Shannon index

VHp the sample variance of the Shannon index

#### Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

96 strsort

# **Examples**

```
#
# Shannon index for allele frequencies of a biallelic MN blood group polymorphism
#
x <- c(MM=298,MN=489,NN=213)
p <- af(x)
shannon(c(p,1-p))$Hp
#
# Shannon index for the allele frequencies of an STR
#
data("NistSTRs")
AlleleTable <- table(c(NistSTRs[,1],NistSTRs[,2]))
AlleleFreq <- AlleleTable/sum(AlleleTable)
shannon(AlleleFreq)$Hp</pre>
```

strsort

Sort tokens of a set of strings

# Description

Function strsort collapses all tokens of a vector of strings in a single string with sorted tokens

#### Usage

```
strsort(s)
```

# **Arguments**

a vector of character strings

#### Value

a string

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

# See Also

alleles

```
x <- c("AA","AB","BB","AC","CC")
print(strsort(x))</pre>
```

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ThetatoF

Convert theta to an inbreeding coefficient

# Description

Function ThetatoF converts disequilibrium measure theta to an inbreeding coefficient.

# Usage

```
ThetatoF(p, theta = 4)
```

# Arguments

p the allele frequency

theta the disequilibrium parameter

#### Value

a real number

# Author(s)

Jan Graffelman (jan.graffelman@upc.edu)

#### References

Rohlfs, R.V. and Weir, B.S. (2008) Distributions of Hardy-Weinberg Equilibrium Test Statistics, Genetics, 180(3) pp. 1609-1616.

#### See Also

HWf

```
f \leftarrow ThetatoF(0.5,4)
```

98 TSIXTriAllelics

toTriangular

Convert a vector of genotype counts to triangular format

# **Description**

Function toTriangular converts a named vector of genotype counts into a triangular matrix format, with homozygotes on the diagonal and heterozygotes below the diagonal.

# Usage

```
toTriangular(x)
```

#### **Arguments**

Х

A vector of genotype counts

#### Value

a matrix

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

# **Examples**

```
x <- c(AA=20,AB=52,AC=34,BB=17,BC=51,CC=26)
print(x)
X <- toTriangular(x)
print(X)</pre>
```

TSIXTriAllelics

Tri-allelic polymorphisms on the X chromosome of the TSI population

# **Description**

This dataframe contains genotype counts for six three-allelic polymorphisms (A,B,C) on chromosome X of a sample of individuals from the TSI population (Tuscany, Italy) of the 1,000 genomes project.

# Usage

```
data(TSIXTriAllelics)
```

#### **Format**

- A data frame with 6 observations on the following 10 variables.
- ID Identifier of the polymorphism
- A Male A genotype count
- B Male B genotype count
- C Male C genotype count
- AA Female AA genotype count
- AB Female AB genotype count
- AC Female AC genotype count
- BB Female BB genotype count
- BC Female BC genotype count
- CC Female CC genotype count

#### **Source**

Data taken from the 1,000 genomes project at www.internationalgenome.org

#### References

Graffelman, J. and Ortoleva, L. (2020) A network algorithm for the X chromosomal exact test for Hardy-Weinberg equilibrium with multiple alleles. Under review.

#### **Examples**

```
data(TSIXTriAllelics)
```

UniqueGenotypeCounts Extract unique genotypic compositions from a matrix

#### **Description**

Function UniqueGenotypeCounts creates a matrix containing only the unique rows in the given matrix, together with their frequency of occurrence

#### Usage

```
UniqueGenotypeCounts(X, verbose = TRUE)
```

#### **Arguments**

Χ A n by 3 matrix with genotypic counts (AA,AB,BB)

verbose If TRUE then print some statistics 100 vaf

# Value

A matrix with 4 columns, AA, AB, BB, and frequency of occurrence

#### Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

#### See Also

```
GenerateSamples
```

# **Examples**

```
set.seed(123)
X <- HWData(n=100,nm=100)
print(nrow(X))
Y <- UniqueGenotypeCounts(X)
print(nrow(Y))
print(sum(Y$w))</pre>
```

vaf

Computes the sample variance of the allele frequency for a biallelic marker.

# **Description**

Function vaf computes the sample variance of the allele frequencies of a single sample or a matrix of samples.

# Usage

```
vaf(X, hw = FALSE)
```

# Arguments

X vector or matrix with genotype counts (AA,AB,BB)

hw assume Hardy-Weinberg proportions (hw=TRUE) or not (hw=FALSE)

#### **Details**

For biallelic markers the variance of the minor allele frequency equals the variance of the major allele frequency.

#### Value

a numeric vector of variances.

vaf

# Author(s)

Jan Graffelman < jan.graffelman@upc.edu>

# References

Weir, B.S. (1996) Genetic data analysis II. Sinauer Associates, Massachusetts. See Chapter 2.

# See Also

```
af, maf
```

```
x <- c(MM=298,MN=489,NN=213)
pA <- af(x)
vA <- vaf(x)
cat("allele frequency:",pA,"\n")
cat("sample variance allele frequency:",vA,"\n")</pre>
```

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