

# Package: hapassoc (via r-universe)

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**Title** Inference of Trait Associations with SNP Haplotypes and Other Attributes using the EM Algorithm

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**Depends** R (>= 2.6.0), stats

**Description** The following R functions are used for inference of trait associations with haplotypes and other covariates in generalized linear models. The functions are developed primarily for data collected in cohort or cross-sectional studies. They can accommodate uncertain haplotype phase and handle missing genotypes at some SNPs.

**License** GPL-2

**URL** <https://sfustatgen.github.io/research/hapassoc.html>

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anova.hapassoc	<i>Return likelihood ratio test of haplotype effect</i>
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### Description

This function returns the likelihood ratio test statistic comparing two nested models fit with hapassoc for cohort or cross-sectional data.

### Usage

```
## S3 method for class 'hapassoc'
anova(object, redfit, display=TRUE, ...)
```

### Arguments

object	a list of class hapassoc output by the <a href="#">hapassoc</a> function
.	
redfit	A hapassoc object resulting from fitting a reduced model
display	An indicator to suppress output displayed on screen
...	additional arguments to the summary function currently unused

### Details

See the hapassoc vignette, of the same name as the package, for details.

### Value

LRTstat	The likelihood ratio statistic comparing the two models
df	Degrees of freedom of the likelihood ratio statistic
pvalue	The p-value of the test

### References

Burkett K, McNeney B, Graham J (2004). A note on inference of trait associations with SNP haplotypes and other attributes in generalized linear models. *Human Heredity*, **57**:200-206

Burkett K, Graham J and McNeney B (2006). hapassoc: Software for Likelihood Inference of Trait Associations with SNP Haplotypes and Other Attributes. *Journal of Statistical Software*, **16**(2):1-19

### See Also

[pre.hapassoc](#), [hapassoc](#), [summary.hapassoc](#).

## Examples

```

data(hypoDatGeno)
example2.pre.hapassoc<-pre.hapassoc(hypoDatGeno, numSNPs=3, allelic=FALSE)
example2.regr <- hapassoc(affected ~ attr + hAAA+ hACA + hACC + hCAA +
pooled, example2.pre.hapassoc, family=binomial())
example2.regr2 <- hapassoc(affected ~ attr + hAAA, example2.pre.hapassoc,
family=binomial())
anova(example2.regr,example2.regr2)

# Returns:

# hapassoc: likelihood ratio test

#Full model: affected ~ attr + hAAA + hACA + hACC + hCAA + pooled
#Reduced model: affected ~ attr + hAAA

#LR statistic = 1.5433 , df = 4 , p-value = 0.8189

```

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hapassoc	<i>EM algorithm to fit maximum likelihood estimates of trait associations with SNP haplotypes</i>
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## Description

This function takes a dataset of haplotypes in which rows for individuals of uncertain phase have been augmented by “pseudo-individuals” who carry the possible multilocus genotypes consistent with the single-locus phenotypes. For cohort or cross-sectional data, the EM algorithm is used to find MLE’s for trait associations with covariates in generalized linear models. For case-control data, the algorithm solves a set of unbiased estimating equations (see **Details**).

## Usage

```

hapassoc(form,haplos.list,baseline = "missing" ,family = binomial(),
design="cohort",disease.prob=NULL,freq = NULL, maxit = 50, tol = 0.001,
start = NULL, verbose=FALSE)

```

## Arguments

form	model equation in usual R format
haplos.list	list of haplotype data from <a href="#">pre.hapassoc</a>
baseline	optional, haplotype to be used for baseline coding if the model formula either includes all haplotypes or is of the form "y~." for example. Default is the most frequent haplotype according to the initial haplotype frequency estimates returned by <a href="#">pre.hapassoc</a> .
family	binomial, poisson, gaussian or gamma are supported, default=binomial

design	study design. Default is “cohort” for cohort or cross-sectional sampling. Users may optionally specify “cc” for case-control or retrospective sampling of exposures (i.e. genotypes and non-genetic attributes) conditional on disease status. When design=“cc”, family=binomial() is assumed and the robust MPSE estimator of the regression parameters (Spinka et al., 2005) is returned; see <b>Details</b> for more information.
disease.prob	marginal disease probability [P(D=1)] to use in the MPSE estimator, if design=“cc”. If disease.prob=NULL (the default value), a rare disease is assumed. This argument is ignored if design=“cohort”.
freq	initial estimates of haplotype frequencies, default values are calculated in <a href="#">pre.hapassoc</a> using standard haplotype-counting (i.e. EM algorithm without adjustment for non-haplotype covariates)
maxit	maximum number of iterations of the EM algorithm; default=50
tol	convergence tolerance in terms of either the maximum difference in parameter estimates between iterations or the maximum relative difference in parameter estimates between iterations, which ever is larger.
start	starting values for parameter estimates in the risk model
verbose	should the iteration number and value of the coverage criterion be printed at each iteration of the EM algorithm? Default=FALSE

### Details

See the hapassoc vignette, of the same name as the package, for details.

When the study design is case-control, i.e. genotypes and non-genetic attributes have been sampled retrospectively given disease status, naive application of prospective maximum likelihood methods can yield biased inference (Spinka et al., 2005, Chen, 2006). Therefore, when design=“cc”, the algorithm solves the modified prospective score equations or MPSE (Spinka et al. 2005) for regression and haplotype frequency parameters. The implementation in **hapassoc** is due to Chen (2006). In general, the MPSE approach requires that the marginal probability of disease, P(D=1), be known. An exception is when the disease is rare; hence, when disease.prob=NULL (the default) a rare disease is assumed. The variance-covariance matrix of the regression parameter and haplotype frequency estimators is approximated as described in Chen (2006). Limited simulations indicate that the resulting standard errors for regression parameters perform well, but not the standard errors for haplotype frequencies, which should be ignored. For case-control data, we hope to implement the variance-covariance estimator of Spinka et al. (2005) in a future version of **hapassoc**.

### Value

it	number of iterations of the EM algorithm
beta	estimated regression coefficients
freq	estimated haplotype frequencies
fits	fitted values of the trait
wts	final weights calculated in last iteration of the EM algorithm. These are estimates of the conditional probabilities of each multilocus genotype given the observed single-locus genotypes.

var	joint variance-covariance matrix of the estimated regression coefficients and the estimated haplotype frequencies
dispersion	maximum likelihood estimate of dispersion parameter (to get the moment estimate, use <a href="#">summary.hapassoc</a> ) if applicable, otherwise 1
family	family of the generalized linear model (e.g. binomial, gaussian, etc.)
response	trait value
converged	TRUE/FALSE indicator of convergence. If the algorithm fails to converge, only the converged indicator is returned.
model	model equation
loglik	the log-likelihood evaluated at the maximum likelihood estimates of all parameters if design="cohort", or NA if design="cc"
call	the function call

## References

- Burkett K, McNeney B, Graham J (2004). A note on inference of trait associations with SNP haplotypes and other attributes in generalized linear models. *Human Heredity*, **57**:200-206
- Burkett K, Graham J and McNeney B (2006). hapassoc: Software for Likelihood Inference of Trait Associations with SNP Haplotypes and Other Attributes. *Journal of Statistical Software*, **16(2)**:1-19
- Chen, Z. (2006): Approximate likelihood inference for haplotype risks in case-control studies of a rare disease, Masters thesis, Statistics and Actuarial Science, Simon Fraser University, available at <https://www.stat.sfu.ca/content/dam/sfu/stat/alumnitheses/MiscellaneousTheses/Chen-2006.pdf>.
- Spinka, C., Carroll, R. J. & Chatterjee, N. (2005). Analysis of case-control studies of genetic and environmental factors with missing genetic information and haplotype-phase ambiguity. *Genetic Epidemiology*, **29**, 108-127.

## See Also

[pre.hapassoc](#), [summary.hapassoc](#), [glm](#), [family](#).

## Examples

```
data(hypoDat)
example.pre.hapassoc<-pre.hapassoc(hypoDat, 3)

example.pre.hapassoc$initFreq # look at initial haplotype frequencies
#   h000   h001   h010   h011   h100   h101   h110
#0.25179111 0.26050418 0.23606001 0.09164470 0.10133627 0.02636844 0.01081260
#   h111
#0.02148268

names(example.pre.hapassoc$haploDM)
# "h000" "h001" "h010" "h011" "h100" "pooled"

# Columns of the matrix haploDM score the number of copies of each haplotype
```

```

# for each pseudo-individual.

# Logistic regression for a multiplicative odds model having as the baseline
# group homozygotes '001/001' for the most common haplotype

example.regr <- hapassoc(affected ~ attr + h000+ h010 + h011 + h100 + pooled,
                        example.pre.hapassoc, family=binomial())

# Logistic regression with separate effects for 000 homozygotes, 001 homozygotes
# and 000/001 heterozygotes

example2.regr <- hapassoc(affected ~ attr + I(h000==2) + I(h001==2) +
                        I(h000==1 & h001==1), example.pre.hapassoc, family=binomial())

```

---

hypoDat

*Simulated data for a hypothetical binary trait*

---

## Description

Simulated binary trait data used to illustrate the hapassoc package.

## Usage

```
data(hypoDat)
```

## Format

Matrix with columns:\

[,1]	affected	numeric	affection status (1=yes, 0=no)
[,3]	attr	numeric	simulated quantitative attribute
[,5]	M1.1	numeric	the first allele of hypothetical SNP M1
[,6]	M1.2	numeric	the second allele of hypothetical SNP M1
[,5]	M2.1	numeric	the first allele of hypothetical SNP M2
[,6]	M2.2	numeric	the second allele of hypothetical SNP M2
[,7]	M3.1	numeric	the first allele of hypothetical SNP M3
[,8]	M3.2	numeric	the second allele of hypothetical SNP M3

---

hypoDatGeno	<i>Simulated data for a hypothetical genetic SNPs</i>
-------------	---

---

**Description**

Simulated genetic SNPs data used to illustrate the hapassoc package.

**Usage**

```
data(hypoDatGeno)
```

**Format**

Matrix with columns:\

[,1]	affected	numeric	affection status (1=yes, 0=no)
[,2]	attr	numeric	simulated quantitative attribute
[,3]	M1	numeric	hypothetical SNP M1
[,4]	M2	numeric	hypothetical SNP M2
[,5]	M3	numeric	hypothetical SNP M3

---

logLik.hapassoc	<i>Return log-likelihood</i>
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---

**Description**

This function is used to return the log-likelihood at the maximum likelihood estimates computed by hapassoc and to return the number of parameters fit by hapassoc (i.e. the degrees of freedom in R) for cohort or cross-sectional data.

**Usage**

```
## S3 method for class 'hapassoc'
logLik(object, ...)
```

**Arguments**

object	a list of class hapassoc output by the <a href="#">hapassoc</a> function
...	additional arguments to the summary function (currently unused)

**Details**

See the hapassoc vignette, of the same name as the package, for details.

**Value**

logLik	log-likelihood computed at the maximum likelihood estimates if design="cohort", or NA if design="cc"
df	number of parameters in the model (i.e. regression coefficients, any dispersion parameters and haplotype frequencies). This is not the residual degrees of freedom, which is the number of subjects minus the number of parameters estimated.

**References**

Burkett K, McNeney B, Graham J (2004). A note on inference of trait associations with SNP haplotypes and other attributes in generalized linear models. *Human Heredity*, **57**:200-206

Burkett K, Graham J and McNeney B (2006). hapassoc: Software for Likelihood Inference of Trait Associations with SNP Haplotypes and Other Attributes. *Journal of Statistical Software*, **16(2)**:1-19

**See Also**

[pre.hapassoc](#), [hapassoc](#), [summary.hapassoc](#).

**Examples**

```
data(hypoDatGeno)
example2.pre.hapassoc<-pre.hapassoc(hypoDatGeno, numSNPs=3, allelic=FALSE)
example.regr <- hapassoc(affected ~ attr + hAAA+ hACA + hACC + hCAA +
pooled, example2.pre.hapassoc, family=binomial())
logLik(example.regr)

# Returns:
# Log Lik: -322.1558 (df=14)
```

---

```
pre.hapassoc
```

*Pre-process the data before fitting it with hapassoc*

---

**Description**

This function takes as an argument a dataframe with non-SNP and SNP data and converts the genotype data at single SNPs (the single-locus genotypes) into haplotype data. The rows of the input data frame should correspond to subjects. Single-locus SNP genotypes may be specified in one of two ways: (i) as pairs of columns, with one column for each allele of the single-locus genotypes (“allelic format”), or (ii) as columns of two-character genotypes (“genotypic format”). The SNP data should comprise the last 2\*numSNPs columns (allelic format) or the last numSNPs columns (genotypic format) of the data frame.

If the haplotypes for a subject cannot be inferred from his or her genotype data, “pseudo-individuals” representing all possible haplotype combinations consistent with the single-locus genotypes are considered. Missing single-locus genotypes, up to a maximum of maxMissingGenos (see below), are allowed, but subjects with missing data in more than maxMissingGenos, or with missing non-SNP data, are removed. Initial estimates of haplotype frequencies are then obtained using the EM



algorithm applied to the genotype data pooled over all subjects. Haplotypes with frequencies below a user-specified tolerance (zero.tol) are assumed not to exist and are removed from further consideration. (Pseudo-individuals having haplotypes of negligible frequency are deleted and the column in the design matrix corresponding to that haplotype is deleted.) For the remaining haplotypes, those with non-negligible frequency below a user-defined pooling tolerance (pooling.tol) are pooled into a single category called “pooled” in the design matrix for the risk model. However, the frequencies of each of these pooled haplotypes are still calculated separately.

### Usage

```
pre.hapassoc(dat,numSNPs,maxMissingGenos=1,pooling.tol = 0.05,
             zero.tol = 1/(2 * nrow(dat) * 10), allelic=TRUE, verbose=TRUE)
```

### Arguments

dat	the non-SNP and SNP data as a data frame. The SNP data should comprise the last 2*numSNPs columns (allelic format) or last numSNPs columns (genotypic format). Missing allelic data should be coded as NA or "" and missing genotypic data should be coded as, e.g., "A" if one allele is missing and "" if both alleles are missing.
numSNPs	number of SNPs per haplotype
maxMissingGenos	maximum number of single-locus genotypes with missing data to allow for each subject. (Subjects with more missing data, or with missing non-SNP data are removed.) The default is 1.
pooling.tol	pooling tolerance – by default set to 0.05
zero.tol	tolerance for haplotype frequencies below which haplotypes are assumed not to exist – by default set to $\frac{1}{2*N*10}$ where N is the number of subjects
allelic	TRUE if single-locus SNP genotypes are in allelic format and FALSE if in genotypic format; default is TRUE.
verbose	indicates whether or not a list of the genotype variables used to form haplotypes and a list of other non-genetic variables should be printed; default is TRUE.

### Details

See the hapassoc vignette, of the same name as the package, for details.

### Value

haplotest	logical, TRUE if some haplotypes had frequency less than zero.tol and are assumed not to exist
initFreq	initial estimates of haplotype frequencies
zeroFreqHaplos	list of haplotypes assumed not to exist
pooledHaplos	list of haplotypes pooled into a single category in the design matrix
haploDM	Haplotype portion of the data frame <b>augmented</b> with pseudo-individuals. Has $2^{numSNPs}$ columns scoring number of copies of each haplotype for each pseudo-individual

nonHaploDM	non-haplotype portion of the data frame <b>augmented</b> with pseudo-individuals
haploMat	matrix with 2 columns listing haplotype labels for each pseudo-individual
wt	vector giving initial weights for each pseudo-individual for the EM algorithm
ID	index for each individual in the original data frame. Note that all pseudo-individuals have the same ID value

## References

Burkett K, McNeney B, Graham J (2004). A note on inference of trait associations with SNP haplotypes and other attributes in generalized linear models. *Human Heredity*, **57**:200-206

Burkett K, Graham J and McNeney B (2006). hapassoc: Software for Likelihood Inference of Trait Associations with SNP Haplotypes and Other Attributes. *Journal of Statistical Software*, **16**(2):1-19

## See Also

[hapassoc,summary.hapassoc.](#)

## Examples

```
#First example data set has single-locus genotypes in "allelic format"
data(hypoDat)
example.pre.hapassoc<-pre.hapassoc(hypoDat, numSNPs=3)

# To get the initial haplotype frequencies:
example.pre.hapassoc$initFreq
#   h000      h001      h010      h011      h100      h101      h110
#0.25179111 0.26050418 0.23606001 0.09164470 0.10133627 0.02636844 0.01081260
#   h111
#0.02148268
# The '001' haplotype is estimated to be the most frequent

example.pre.hapassoc$pooledHaplos
# "h101" "h110" "h111"
# These haplotypes are to be pooled in the design matrix for the risk model

names(example.pre.hapassoc$haploDM)
# "h000" "h001" "h010" "h011" "h100" "pooled"

####
#Second example data set has single-locus genotypes in "genotypic format"
data(hypoDatGeno)
example2.pre.hapassoc<-pre.hapassoc(hypoDatGeno, numSNPs=3, allelic=FALSE)

# To get the initial haplotype frequencies:
example2.pre.hapassoc$initFreq
#   hAAA      hAAC      hACA      hACC      hCAA      hCAC
#0.25179111 0.26050418 0.23606001 0.09164470 0.10133627 0.02636844
#   hCCA      hCCC
#0.01081260 0.02148268
# The 'hAAC' haplotype is estimated to be the most frequent
```

```

example2.pre.hapassoc$pooledHaplos
# "hCAC" "hCCA" "hCCC"
# These haplotypes are to be pooled in the design matrix for the risk model

names(example2.pre.hapassoc$haploDM)
# "hAAA" "hAAC" "hACA" "hACC" "hCAA" "pooled"

```

---

summary.hapassoc	<i>Summarize results of the hapassoc function</i>
------------------	---

---

## Description

Summary function for reporting the results of the hapassoc function in a similar style to the lm and glm summaries.

## Usage

```

## S3 method for class 'hapassoc'
summary(object, ...)

```

## Arguments

object	a list of class hapassoc output by the <a href="#">hapassoc</a> function
...	additional arguments to the summary function (currently unused)

## Details

See the hapassoc vignette, of the same name as the package, for details.

## Value

call	The function call to hapassoc
subjects	The number of subjects used in the analysis
coefficients	Table of estimated coefficients, standard errors and Wald tests for each variable
frequencies	Table of estimated haplotype frequencies and standard errors
dispersion	Estimate of dispersion parameter (Moment estimator for gamma model)

## References

Burkett K, McNeney B, Graham J (2004). A note on inference of trait associations with SNP haplotypes and other attributes in generalized linear models. *Human Heredity*, **57**:200-206

Burkett K, Graham J and McNeney B (2006). hapassoc: Software for Likelihood Inference of Trait Associations with SNP Haplotypes and Other Attributes. *Journal of Statistical Software*, **16(2)**:1-19

## See Also

[pre.hapassoc](#), [hapassoc](#).

**Examples**

```

data(hypoDat)
example.pre.hapassoc<-pre.hapassoc(hypoDat, 3)
example.regr <- hapassoc(affected ~ attr + h000+ h010 + h011 + h100 + pooled,
                        example.pre.hapassoc, family=binomial())

# Summarize the results:
summary(example.regr)

# Results:
# $coefficients
#           Estimate Std. Error      zscore Pr(>|z|)
#(Intercept) -1.24114270  0.7820977 -1.58694079 0.11252606
#attr         0.74036920  0.2918205  2.53707057 0.01117844
#h000         1.14968352  0.5942542  1.93466627 0.05303126
#h010        -0.59318434  0.6569672 -0.90291311 0.36657201
#h011        -0.03615243  0.9161959 -0.03945928 0.96852422
#h100        -0.85329292  1.0203105 -0.83630709 0.40298217
#pooled       0.38516864  0.8784283  0.43847478 0.66104215
#
# $frequencies
#           Estimate Std. Error
#f.h000 0.26716394 0.03933158
#f.h001 0.25191674 0.03866739
#f.h010 0.21997138 0.03881578
#f.h011 0.10094795 0.02949617
#f.h100 0.09507014 0.02371878
#f.h101 0.02584918 0.01411881
#f.h110 0.01779455 0.01386080
#f.h111 0.02128613 0.01247265
#
# $dispersion
#[1] 1

```

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