# Package 'LEA'

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**Title** LEA: an R package for Landscape and Ecological Association Studies

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**Depends** R (>= 3.3.0), methods, stats, utils, graphics

Suggests knitr

Description LEA is an R package dedicated to population genomics, landscape genomics and genotype-environment association tests. LEA can run analyses of population structure and genome-wide tests for local adaptation. The package includes statistical methods for estimating ancestry coefficients from large genotypic matrices and for evaluating the number of ancestral populations (snmf, pca). It performs statistical tests using latent factor mixed models for identifying genetic polymorphisms that exhibit association with environmental gradients or phenotypic traits (lfmm2). {\tt LEA} also performs imputation of missing genotypes, and computes predictive values of genetic offsets based on new or future environments. The package includes factor methods for estimating ancestry coefficients from large genotypic matrices and for evaluating the number of ancestral populations (snmf, pca). LEA is mainly based on optimized programs that can scale with the dimension of large data sets.

License GPL-3

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# Description

ancestrymap

LEA is an R package dedicated to landscape genomics and ecological association tests. LEA can run analyses of population structure and genome scans for local adaptation. It includes statistical methods for estimating ancestry coefficients from large genotypic matrices and evaluating the number of ancestral populations (snmf, pca) and identifying genetic polymorphisms that exhibit high correlation with some environmental gradient or with the variables used as proxies for ecological pressures (lfmm). LEA is mainly based on optimized C programs that can scale with the dimension of very large data sets.

#### **Details**

Package: LEA Type: Package Version: 2.0

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# Author(s)

Eric Frichot Olivier François Maintainer: Olivier François <olivier.françois@grenoble-inp.fr>

ancestrymap ancestrymap format description

## **Description**

Description of the ancestrymap format. The ancestrymap format can be used as an input format for genotypic matrices in the functions pca, 1fmm and snmf.

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# **Details**

The ancestrymap format has one row for each genotype. Each row has 3 columns: the 1st column is the SNP name, the 2nd column is the sample ID, the 3rd column is th number of alleles. Genotypes for a given SNP name are written in consecutive lines. The number of alleles can be the number of reference alleles or the number of derived alleles. Missing genotypes are encoded by the value 9.

Here is an example of a genotypic matrix using the ancestrymap format with 3 individuals and 4 SNPs:

rs0000	SAMPLE0	1
rs0000	SAMPLE1	1
rs0000	SAMPLE2	2
rs1111	SAMPLE0	0
rs1111	SAMPLE1	1
rs1111	SAMPLE2	0
rs2222	SAMPLE0	0
rs2222	SAMPLE1	9
rs2222	SAMPLE2	1
rs3333	SAMPLE0	1
rs3333	SAMPLE1	2
rs3333	SAMPLE2	1

# Author(s)

Eric Frichot

# See Also

ancestrymap2lfmm ancestrymap2geno geno lfmm.data ped vcf

ancestrymap2geno	Convert from ancestrymap to geno format
------------------	---

# Description

A function that converts from the ancestrymap format to the geno format.

# Usage

```
ancestrymap2geno(input.file, output.file = NULL, force = TRUE)
```

## **Arguments**

input.file	A character string containing a path to the input file, a genotypic matrix in the ancestrymap format.
output.file	A character string containing a path to the output file, a genotypic matrix in the geno format. By default, the name of the output file is the same name as the input file with a .geno extension.

ancestrymap2lfmm 5

force A boolean option. If FALSE, the input file is converted only if the output file

does not exist. If TRUE, convert the file anyway.

Value

output.file A character string containing a path to the output file, a genotypic matrix in the

geno format.

#### Author(s)

Eric Frichot

## See Also

ancestrymap geno read.geno ancestrymap2lfmm geno2lfmm ped2lfmm ped2geno vcf2geno lfmm2geno

# **Examples**

```
# Creation of of file called "example.ancestrymap"
# a file containing 4 SNPs for 3 individuals.
data("example_ancestrymap")
write.table(example_ancestrymap, "example.ancestrymap",
col.names = FALSE, row.names = FALSE, quote = FALSE)
# Conversion
               from the ancestrymap format ("example.ancestrymap")
#
               to the geno format ("example.geno").
# By default, the name of the output file is the same name
                as the input file with a .geno extension.
# Create file: "example.geno".
output = ancestrymap2geno("example.ancestrymap")
# Conversion
                from the ancestrymap format (example.ancestrymap)
                to the geno format with the output file called plop.geno.
# Create file: "plop.geno".
output = ancestrymap2geno("example.ancestrymap", "plop.geno")
# As force = false and the file "example.geno" already exists,
# nothing happens.
output = ancestrymap2geno("example.ancestrymap", force = FALSE)
```

ancestrymap21fmm

Convert from ancestrymap to 1fmm format

# **Description**

A function that converts from the ancestrymap format to the 1fmm format.

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#### Usage

```
ancestrymap2lfmm(input.file, output.file = NULL, force = TRUE)
```

#### **Arguments**

input.file A character string containing a path to the input file, a genotypic matrix in the

ancestrymap format.

output.file A character string containing a path to the output file, a genotypic matric in the

1fmm format. By default, the name of the output file is the same name as the

input file with a .lfmm extension.

force A boolean option. If FALSE, the input file is converted only if the output file

does not exist. If TRUE, convert the file anyway.

#### Value

output.file A character string containing a path to the output file, a genotypic matric in the

1fmm format.

# Author(s)

Eric Frichot

#### See Also

 $ancestrymap\ 1 fmm.\ data\ ancestrymap\ 2 geno\ geno\ 2 1 fmm\ ped\ 2 1 fmm\ ped\ 2 geno\ vcf\ 2 geno\ 1 fmm\ 2 geno\ 2 geno\ 2 fmm\ ped\ 2 geno\ vcf\ 2 geno\ 1 fmm\ 2 geno\ 2 fmm\ 2 fmm\ 2 geno\ 2 fmm\ 2 fmm\$ 

```
# Creation of a file called "example.ancestrymap"
# containing 4 SNPs for 3 individuals.
data("example_ancestrymap")
write.table(example_ancestrymap, "example.ancestrymap",
col.names = FALSE, row.names = FALSE, quote = FALSE)
               from the ancestrymap format ("example.ancestrymap")
# Conversion
               to the lfmm format ("example.lfmm").
# By default, the name of the output file is the same name
                as the input file with a .lfmm extension.
# Create file: "example.lfmm".
output = ancestrymap2lfmm("example.ancestrymap")
# Conversion
                from the ancestrymap format (example.ancestrymap)
                to the geno format with the output file called plop.lfmm.
# Create file: "plop.lfmm".
output = ancestrymap2lfmm("example.ancestrymap", "plop.lfmm")
# As force = false and the file "example.lfmm" already exists,
# nothing happens.
output = ancestrymap2lfmm("example.ancestrymap", force = FALSE)
```

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barchart	Bar plot representation of an snmf Q-matrix
----------	---

# Description

This function displays a bar plot/bar chart representation of the Q-matrix computed from an snmf run. The function can use a sort by Q option. See snmf.

# Usage

```
barchart (object, K, run, sort.by.Q = TRUE, lab = FALSE, ...)
```

# Arguments

object	an snmfProject object.
K	an integer value corresponding to number of ancestral populations.
run	an integer value. Usually the run number that minimizes the cross-entropy criterion.
sort.by.Q	a Boolean value indicating whether individuals should be sorted by their ancestry or not.
lab	a list of individual labels.
	other parameters of the function barplot.default.

## Value

A permutation of individual labels used in the sort.by.Q option (order). Displays the Q matrix.

# Author(s)

Olivier Francois

# See Also

snmf

8 create.dataset

```
project.snmf <- snmf("genotypes.geno",</pre>
                    K = 4, entropy = TRUE,
                    repetitions = 10,
                     project = "new")
# get the cross-entropy value for each run
ce <- cross.entropy(project.snmf, K = 4)</pre>
# select the run with the lowest cross-entropy value
best <- which.min(ce)</pre>
\# plot the ancestry coefficients for the best run and K = 4
my.colors <- c("tomato", "lightblue", "olivedrab", "gold")</pre>
barchart(project.snmf, K = 4, run = best,
        border = NA, space = 0, col = my.colors,
        xlab = "Individuals", ylab = "Ancestry proportions",
        main = "Ancestry matrix") -> bp
axis(1, at = 1:length(bp$order),
      labels = bp$order, las = 3,
      cex.axis = .4)
```

create.dataset

create a data set with masked data

# Description

create.dataset creates a data set with a given percentage of masked data from the original data set. It is used to calculate the cross.entropy criterion.

# Usage

```
create.dataset (input.file, output.file, seed = -1, percentage = 0.05)
```

# **Arguments**

input.file	A character string containing a path to the input file, a genotypic matrix in the geno format.
output.file	A character string containing a path to the output file, a genotypic matrix in the geno format. The output file is the input file with masked genotypes. By default, the name of the output file is the same name as the input file with a _I.geno extension.
seed	A seed to initialize the random number generator. By default, the seed is randomly chosen.
percentage	A numeric value between 0 and 1 containing the percentage of masked genotypes.

cross.entropy 9

## **Details**

This is an internal function, automatically called by snmf with the entropy option.

#### Value

output.file A character string containing a path to the output file, a genotypic matrix in the geno format.

## Author(s)

Eric Frichot

#### See Also

```
geno snmf cross.entropy
```

# Examples

```
# Creation of tuto.geno
# A file containing 400 SNPs for 50 individuals.
data("tutorial")
write.geno(tutorial.R, "genotypes.geno")

# Creation of the masked data file
# Create file: "genotypes_I.geno"
output = create.dataset("genotypes.geno")
```

cross.entropy

Cross-entropy criterion for snmf runs

## **Description**

Return the cross-entropy criterion for runs of snmfcwith K ancestral populations. The cross-entropy criterion is based on the prediction of masked genotypes to evaluate the fit of a model with K populations. The cross-entropy criterion helps choosing the number of ancestral populations or a best run for a fixed value of K. A smaller value of cross-entropy means a better run in terms of prediction capability. The cross-entropy criterion is computed by the snmf function when the entropy Boolean option is TRUE.

## Usage

```
cross.entropy(object, K, run)
```

# **Arguments**

object A snmfProject object.

K The number of ancestral populations.

run A vector of run labels.

## Value

res

A matrix containing the cross-entropy criterion for runs with K ancestral populations.

## Author(s)

Eric Frichot

# See Also

```
geno snmf G Q
```

# **Examples**

```
### Example of analyses using snmf ###
# creation of a genotype file: genotypes.geno.
# The data contains 400 SNPs for 50 individuals.
data("tutorial")
write.geno(tutorial.R, "genotypes.geno")
################
# running snmf #
##################
\# Runs with K = 3 populations
# cross-entropy is computed for 2 runs.
project = NULL
project = snmf("genotypes.geno",
                K = 3,
                entropy = TRUE,
                repetitions = 2,
                project = "new")
\# get the cross-entropy for all runs for K = 3
ce = cross.entropy(project, K = 3)
\# get the cross-entropy for the 2nd run for K = 3
ce = cross.entropy(project, K = 3, run = 2)
```

cross.entropy.estimation

compute the cross-entropy criterion

# Description

Calculate the cross-entropy criterion. This is an internal function, automatically called by snmf. The cross-entropy criterion is a value based on the prediction of masked genotypes to evaluate the error of ancestry estimation. The criterion will help to choose the best number of ancestral population (K)

and the best run among a set of runs in snmf. A smaller value of cross-entropy means a better run in terms of prediction capacity. The cross-entropy.estimation function displays the cross-entropy criterion estimated on all data and on masked data based on the input file, the masked data file (created by create.dataset, the estimation of the ancestry coefficients Q and the estimation of ancestral genotypic frequencies, G (calculated by snmf). The cross-entropy estimation for all data is always lower than the cross-entropy estimation for masked data. The cross-entropy estimation useful to compare runs is the cross-entropy estimation for masked data. The cross-entropy criterion can also be automatically calculated by the snmf function with the entropy option.

# Usage

```
cross.entropy.estimation (input.file, K, masked.file, Q.file, G.file,
    ploidy = 2)
```

## **Arguments**

input.file	A character string containing a path to the input file without masked genotypes, a genotypic matrix in the geno format.
K	An integer corresponding to the number of ancestral populations.
masked.file	A character string containing a path to the input file with masked genotypes, a genotypic matrix in the geno format. This file can be generated with the function, create.dataset). By default, the name of the masked data file is the same name as the input file with a _I.geno extension.
Q.file	A character string containing a path to the input ancestry coefficient matrix Q. By default, the name of this file is the same name as the input file with a K.Q extension.
G.file	A character string containing a path to the input ancestral genotype frequency matrix G. By default, the name of this file is the same name as the input file with a K.G extension (input_file.K.G).
ploidy	1 if haploid, 2 if diploid, n if n-ploid.

#### Value

cross.entropy.estimation returns a list containing the following components:

masked.ce The value of the cross-entropy criterion of the masked genotypes.

The value of the cross-entropy criterion of all the genotypes.

## Author(s)

Eric Frichot

#### References

Frichot E, Mathieu F, Trouillon T, Bouchard G, Francois O. (2014). Fast and Efficient Estimation of Individual Ancestry Coefficients. Genetics, 194(4): 973–983.

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## See Also

```
geno create.dataset snmf
```

#### **Examples**

```
# Creation of tuto.geno
# A file containing 400 SNPs for 50 individuals.
data("tutorial")
write.geno(tutorial.R, "genotypes.geno")
# The following command are equivalent with
# project = snmf("genotypes.geno", entropy = TRUE, K = 3)
# cross.entropy(project)
# Creation
                of the masked data file
# Create file: "genotypes_I.geno"
output = create.dataset("genotypes.geno")
# run of snmf with genotypes_I.geno and K = 3
project = snmf("genotypes_I.geno", K = 3, project = "new")
# calculate the cross-entropy
res = cross.entropy.estimation("genotypes.geno", K = 3, "genotypes_I.geno",
    "./genotypes_I.snmf/K3/run1/genotypes_I_r1.3.Q",
    "./genotypes_I.snmf/K3/run1/genotypes_I_r1.3.G")
# get the result
res$masked.ce
res$all.ce
#remove project
remove.snmfProject("genotypes_I.snmfProject")
```

env

Environmental input file format for 1fmm

## **Description**

Description of the env format. The env format can be used as an input format for the environmental variables in the 1fmm function.

#### **Details**

The env format has one row for each individual. Each row contains one value for each environmental variable (separated by spaces or tabulations).

Here is an example of an environmental file using the env format with 3 individuals and 2 variables:

```
0.252477 0.95250639
0.216618 0.10902647
-0.47509 0.07626694
```

*G* 13

## Author(s)

Eric Frichot

#### See Also

1fmm 1fmm2 read.env write.env

G

Ancestral allele frequencies from a snmf run

# Description

Return the snmf output matrix of ancestral allele frequency matrix for the chosen run with K ancestral populations. For an example, see snmf.

# Usage

```
G(object, K, run)
```

## **Arguments**

object A snmfProject object.

K The number of ancestral populations.

run A chosen run.

# Value

res A matrix containing the ancestral allele frequencies for a run with K ancestral

populations.

## Author(s)

Eric Frichot

# See Also

```
geno snmf Q cross.entropy
```

#################

```
### Example of analyses using snmf ###

# creation of a genotype file: genotypes.geno.
# The data contain 400 SNPs for 50 individuals.
data("tutorial")
write.geno(tutorial.R, "genotypes.geno")
```

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genetic.offset

Population genetic offset under new environments.

# **Description**

The function returns genetic offset estimates computed from user-specified population labels and new environments based on predictions of an 1fmm2 model. It takes as input an object of class 1fmm2Class together with the data that were used to adjust the LFMM, and a matrix of new environmental variables in the same format as the original ones.

# Usage

```
genetic.offset (input, env, new.env, pop.labels, K, pca, candidate.loci)
```

# Arguments

input	A genotypic matrix or a character string containing a path to the input file. The genotypic matrix must be in the lfmm{lfmm_format} format without missing values (9 or NA). See impute for completion based on nonnegative matrix factorization and consider R packages for reading large matrices.
env	A matrix of environmental covariates or a character string containing a path to the environmental file. The environment matrix must be in the env format without missing values. All variables must be encoded as numeric.
new.env	A matrix of new environmental covariates or a character string containing a path to the new environmental data file. The new environmental matrix must be in the env format without missing values, and of same dimension as the env matrix. All variables must be encoded as numeric.
pop.labels	A numeric or character vector providing population labels for all rows (individuals) of the response matrix.
K	An integer corresponding to the number of latent factors.
pca	A boolean value indicating whether genetic offsets are computed from a PCA approximation or not (default value: FALSE).
candidate.loci	A vector specifying which loci (column label) in the genotype matrix are included in the computation of the genetic offset (default value: all positions).

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#### Value

offset

A matrix or vector (depending on the pca parameter) of genetic offset values computed for every population in pop.labels.

## Author(s)

Olivier Francois, Clement Gain

#### References

Gain C, Francois O. (2021). LEA 3: Factor models in population genetics and ecological genomics with R. Molecular Ecology Resources. doi.org/10.1111/1755-0998.13366.

## See Also

```
1fmm.data1fmm2
```

```
### Example of offset prediction using lfmm2 ###
# Simulation with 100 target loci
# Effect sizes ranging between -10 an 10
\# n = 100 individuals and L = 1000 loci
X <- as.matrix(rnorm(100)) # environmental variable</pre>
B < - rep(0, 1000)
target <- sample(1:1000, 100) # target loci</pre>
B[target] <- runif(100, -10, +10) # effect sizes
# Creating hidden factors and their loadings
U \leftarrow t(tcrossprod(as.matrix(c(-1.25, 0.5, 1.25)), X)) +
     matrix(rnorm(300), ncol = 3)
V \leftarrow matrix(rnorm(3000), ncol = 3)
# Simulating a binarized matrix of haploid genotypes
# Simulation performed with a generative LFMM
Y <- tcrossprod(as.matrix(X), B) + tcrossprod(U, V) +
     matrix(rnorm(100000, sd = .5), nrow = 100)
Y \leftarrow matrix(as.numeric(Y > 0), ncol = 1000)
# Fitting an LFMM with K = 3 factors #
# Computing genetic offset statistics for 2 populations
# defined from PCA
```

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geno

Input file for snmf

# Description

Description of the geno format. The geno format can be used as an input format for genotypic matrices in the functions snmf, 1fmm, and pca.

# **Details**

The geno format has one row for each SNP. Each row contains 1 character for each individual: 0 means zero copy of the reference allele. 1 means one copy of the reference allele. 2 means two copies of the reference allele. 9 means missing data.

Here is an example of a genotypic matrix using the geno format with 3 individuals and 4 loci:

112

010

091

121

## Author(s)

Eric Frichot

## See Also

geno2lfmm lfmm2geno ancestrymap2geno ped2geno vcf2geno read.geno write.geno

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geno2lfmm	Convert from geno to 1fmm format	
-----------	----------------------------------	--

## **Description**

A function that converts from the geno format to the 1fmm format.

## Usage

```
geno2lfmm(input.file, output.file = NULL, force = TRUE)
```

# **Arguments**

input.file A character string containing a path to the input file, a genotypic matrix in the

geno format.

output.file A character string containing a path to the output file, a genotypic matrix in the

1fmm format. By default, the name of the output file is the same name as the

input file with a .lfmm extension.

force A boolean option. If FALSE, the input file is converted only if the output file

does not exist. If TRUE, convert the file anyway.

#### Value

output.file A character string containing a path to the output file, a genotypic matrix in the

1fmm format.

## Author(s)

Eric Frichot

## See Also

 $1 fmm. \, data \, geno \, ancestry map 2 lfmm \, ancestry map 2 geno \, ped 2 lfmm \, ped 2 geno \, vcf 2 geno \, lfmm 2 geno \, read. \, geno \, write. \, geno \, description \, des$ 

```
# Creation of a file called "genotypes.geno" in the working directory
# with 400 SNPs for 50 individuals.
data("tutorial")
write.geno(tutorial.R, "genotypes.geno")

# Conversion from the geno format ("genotypes.geno")
# to the lfmm format ("genotypes.lfmm").
# By default, the name of the output file is the same name
# as the input file with a .lfmm extension.
# Create file: "genotypes.lfmm".
output = geno2lfmm("genotypes.geno")
```

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```
# Conversion from the geno format ("genotypes.geno")
# to the lfmm format with the output file called "plop.lfmm".
# Create file: "plop.lfmm".
output = geno2lfmm("genotypes.geno", "plop.lfmm")

# As force = false and the file "genotypes.lfmm" already exists,
# nothing happens.
output = geno2lfmm("genotypes.geno", force = FALSE)
```

impute

Impute missing genotypes using an snmf object

## **Description**

Impute missing genotypes in a genotype file (.lfmm) by using ancestry and genotype frequency estimates from an snmf run. The function generates a new 1fmm file. See 1fmm and 1fmm2.

## Usage

```
impute (object, input.file, method, K, run)
```

## **Arguments**

object An snmfProject object.

input.file A path (character string) to an input file in lfmm format with missing genotypes.

The same input data must be used when generating the snmf object.

method A character string: "random" or "mode". With "random", imputation is per-

formed by using the genotype probabilities. With "mode", the most likely geno-

type is used for matrix completion.

K An integer value. The number of ancestral populations.

run An integer value. A particular run used for imputation (usually the run number

that minimizes the cross entropy criterion).

#### Value

NULL The function writes the imputed genotypes in an output file having the "\_im-

puted.lfmm" suffix.

#### Author(s)

Olivier Francois

## References

Gain C, Francois O. (2021). LEA 3: Factor models in population genetics and ecological genomics with R. Molecular Ecology Resources, doi.org/10.1111/1755-0998.13366.

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#### See Also

```
snmf lfmm lfmm2
```

#### **Examples**

```
### Example of analysis ###
data("tutorial")
# creation of a genotype file with missing genotypes
# The data contain 400 SNPs for 50 individuals.
dat = as.numeric(tutorial.R)
dat[sample(1:length(dat), 100)] <- 9</pre>
dat <- matrix(dat, nrow = 50, ncol = 400 )</pre>
write.lfmm(dat, "genotypes.lfmm")
#################
# running snmf #
#################
project.snmf = snmf("genotypes.1fmm", K = 4,
        entropy = TRUE, repetitions = 10,
        project = "new")
# select the run with the lowest cross-entropy value
best = which.min(cross.entropy(project.snmf, K = 4))
# Impute the missing genotypes
impute(project.snmf, "genotypes.lfmm", method = 'mode', K = 4, run = best)
# Compare with truth
# Proportion of correct imputation results:
mean( tutorial.R[dat == 9] == read.lfmm("genotypes.lfmm_imputed.lfmm")[dat == 9] )
```

1fmm

Fitting Latent Factor Mixed Models (MCMC algorithm)

#### **Description**

Latent Factor Mixed Models (LFMMs) are factor regression models in which the response variable is a genotypic matrix, and the explanatory variables are environmental measures of ecological interest or trait values. The 1fmm function estimates latent factors and effect sizes based on an MCMC algorithm. The resulting object can be used in the function 1fmm.pvalues to identify genetic polymorphisms exhibiting association with ecological gradients or phenotypes, while correcting for unobserved confounders. An exact and computationally efficient least-squares method is implemented in the function 1fmm2 which should be the prefered option.

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#### Usage

```
lfmm(input.file, environment.file, K,
    project = "continue",
    d = 0, all = FALSE,
    missing.data = FALSE, CPU = 1,
    iterations = 10000, burnin = 5000,
    seed = -1, repetitions = 1,
    epsilon.noise = 1e-3, epsilon.b = 1000,
    random.init = TRUE)
```

## **Arguments**

input.file A character string containing a path to the input file, a genotypic matrix in the

1fmm{1fmm\_format} format. The matrix must not contain missing values. See

impute for completion based on nonnegative matrix factorization.

environment.file

A character string containing a path to the environmental file, an environmental

data matrix in the env format.

K An integer corresponding to the number of latent factors.

project A character string among "continue", "new", and "force". If "continue", the

results are stored in the current project. If "new", the current project is removed and a new project is created. If "force", the results are stored in the current project even if the input file has been modified since the creation of the project.

d An integer corresponding to the fit of an 1fmm model with the d-th variable only

from environment.file. By default (if NULL and all are FALSE), 1fmm fits

each variable from environment. file sequentially and independently.

all A Boolean option. If TRUE, 1fmm fits all variables from the environment.file

at the same time. This option is not compatible with the d option.

missing.data A Boolean option. If TRUE, the input.file contains missing genotypes. Cau-

tion: 1fmm requires imputed genotype matrices. See impute.

CPU A number of CPUs to run the parallel version of the algorithm. By default, the

number of CPUs is 1.

iterations The total number of cycles for the Gibbs Sampling algorithm.

burnin The burnin number of cycles for the Gibbs Sampling algorithm.

seed A seed to initialize the random number generator. By default, the seed is ran-

domly chosen. The seed is initialized in each run of the program. For modifying

the default setting, provide one seed per run.

repetitions A number of replicate runs for the Gibbs Sampler algorithm.

epsilon.noise A prior parameter for variances.

epsilon.b A prior parameter for the variance of correlation coefficients.

random.init A Boolean option. If TRUE, the Gibbs Sampler is initiliazed randomly. Other-

wise, it is initialized with zero values.

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#### Value

1fmm returns an object of class 1fmmProject.

The following methods can be applied to an object of class 1fmmProject:

show Display information about all analyses.

summary Summarize analyses.

z.scores Return the 1fmm output vector of z.scores for some runs.

1fmm.pvalues Return the vector of adjusted p-values for a combination of runs with K latent

factors, and for the d-th predictor.

load.lfmmProject (file = "character")

Load the file containing an lfmmProject objet and show the object.

remove.lfmmProject (file = "character")

Erase a 1fmmProject object. Caution: All the files associated with the object

will be removed.

export.lfmmProject(file.lfmmProject)

Create a zip file containing the full lfmmProject object. It allows users to move the project to a new directory or a new computer (using import). If you want to

overwrite an existing export, use the option force == TRUE.

import.lfmmProject(file.lfmmProject)

Import and load an lfmmProject object from a zip file (made with the export function) into the chosen directory. If you want to overwrite an existing project,

use the option force == TRUE.

combine.lfmmProject(file.lfmmProject, toCombine.lfmmProject)

Combine to.Combine.1fmmProject into file.1fmmProject. Caution: Only projects with runs coming from the same input file can be combined. If the same input file has different names in the two projects, use the option force == TRUE.

## Author(s)

Eric Frichot Olivier François

#### References

Frichot E, Schoville SD, Bouchard G, Francois O. (2013). *Testing for associations between loci and environmental gradients using latent factor mixed models*. Molecular biology and evolution, 30(7), 1687-1699.

## See Also

lfmm.dataz.scoreslfmm.pvaluespcalfmm tutorial

```
### Example of analysis using lfmm ###
data("tutorial")
# creation of a genotype file: genotypes.lfmm.
```

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```
# The file contains 400 SNPs for 50 individuals.
write.lfmm(tutorial.R, "genotypes.lfmm")
# Creation of a phenotype/environment file: gradient.env.
# One environmental predictor for 40 individuals.
write.env(tutorial.C, "gradients.env")
#################
# running lfmm #
##################
# main options, K: (the number of latent factors),
            CPU: the number of CPUs.
# Runs with K = 6 and 5 repetitions.
# runs with 6000 iterations
# including 3000 iterations for burnin.
# Around 30 seconds per run.
project = lfmm( "genotypes.lfmm",
                "gradients.env",
                 K = 6,
                 repetitions = 5,
                 project = "new")
# get adjusted p-values using all runs
pv = lfmm.pvalues(project, K = 6)
# Evaluate FDR and POWER (TPR)
for (alpha in c(.05,.1,.15,.2)) {
    # expected FDR
   print(paste("expected FDR:", alpha))
   L = length(pv$pvalues)
   # Benjamini-Hochberg's mehod for an expected FDR = alpha.
   w = which(sort(pv$pvalues) < alpha * (1:L)/L)</pre>
   candidates = order(pv$pvalues)[w]
    # estimated FDR and True Positive Rate
    # The targets SNPs are loci 351 to 400
   Lc = length(candidates)
    estimated.FDR = length(which(candidates <= 350))/Lc</pre>
   estimated.TPR = length(which(candidates > 350))/50
   print(paste("FDR:", estimated.FDR, "True Positive Rate:", estimated.TPR))
}
# Post-treatments #
#######################
# show the project
show(project)
# summary of the project
summary(project)
```

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```
# get the z-scores for the 2nd run for K = 6
z = z.scores(project, K = 6, run = 2)
\# get the p-values for K = 6 and run 2
p = lfmm.pvalues(project, K = 6, run = 2)
##############################
# Manage an lfmm project #
# All the runs of lfmm for a given file are
# automatically saved into an lfmm project directory and a file.
# The name of the lfmmProject file is the concatenation of
# the name of the input file and the environment file
# with a .lfmmProject extension ("genotypes_gradient.lfmmProject").
\ensuremath{\text{\#}} The name of the lfmmProject directory is the same name as
# the lfmmProject file with a .lfmm extension ("genotypes_gradient.lfmm/")
# There is a unique lfmm Project for each input file.
# An lfmmProject can be loaded in an R session as follows
project = load.lfmmProject("genotypes_gradients.lfmmProject")
# An lfmmProject can be exported to be imported in another directory
# or in another computer as follows
export.lfmmProject("genotypes_gradients.lfmmProject")
dir.create("test", showWarnings = TRUE)
#import
newProject = import.lfmmProject("genotypes_gradients_lfmmProject.zip", "test")
# combine projects
combinedProject <- combine.lfmmProject(</pre>
                  "genotypes_gradients.lfmmProject",
                  "test/genotypes_gradients.lfmmProject"
                  )
# remove
remove.lfmmProject("test/genotypes_gradients.lfmmProject")
# An lfmmProject can be removed as follows.
# Caution: All the files associated with the project will be removed.
remove.lfmmProject("genotypes_gradients.lfmmProject")
```

24 lfmm.pvalues

## **Description**

Description of the 1fmm format. The 1fmm format can be used as an input format for genotypic matrices in the functions snmf, 1fmm, 1fmm2, and pca.

# **Details**

The 1fmm format has one row for each individual. Each row contains one value at each loci (separated by spaces or tabulations) corresponding to the number of alleles. The number of alleles corresponds to the number of reference alleles or the number of derived alleles. Missing genotypes are encoded by the value -9 or the value 9.

For the use of functions 1fmm and 1fmm2 missing genotypes must be removed or imputed with the function impute.

Here is an example of a genotypic matrix using the 1fmm format with 3 individuals and 4 loci:

```
1 0 0 1
1 1 9 2
2 0 1 1
```

#### Author(s)

Eric Frichot

#### See Also

1fmm 1fmm2 geno21fmm 1fmm2geno ancestrymap21fmm ped21fmm read.1fmm write.1fmm

lfmm.pvalues

P-values from lfmm runs

# **Description**

Returns a vector of p-values computed from a combination of 1fmm runs. For an example, see 1fmm.

## Usage

```
lfmm.pvalues (object, genomic.control, lambda, K, d, all, run)
```

#### **Arguments**

```
object An lfmmProject object. genomic.control
```

A Boolean value. If TRUE, the p-values are automatically calibrated using genomic control. If FALSE, the p-values are calculated by rescaling the chi-squared test statistics using the lambda parameter.

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lambda	A numeric value. The lambda value is used as inflation factor to rescale the chi-squared statistics in the computation of p-values. This option requires that genomic.control = FALSE. The default value of lambda is equal to 1.0 (no rescaling).
K	An integer value. The number of latent factors used in the model.
d	An integer value. Computes the p-values for the d-th covariable in the model.
all	A Boolean value. Each variable is considered separately (Obscure parameter).
run	An integer vector representing a list of runs to be combined in the computation of p-values (by default, all runs).

## Value

pvalues A vector of combined p-values for each locus.

GIF The inflation factor value used for correcting the test statistics.

# Author(s)

Eric Frichot Olivier Francois

#### See Also

```
1fmm.data1fmm
```

```
### Example of analyses using 1fmm ###
data("tutorial")
# creation of a genotype file, "genotypes.lfmm".
# The data contain 400 SNPs for 50 individuals.
write.lfmm(tutorial.R, "genotypes.lfmm")
# creation of an environmental variable file, "gradient.env".
# The data contain one environmental variable measured for 50 individuals.
write.env(tutorial.C, "gradients.env")
################
# 1fmm runs
################
# main options, K: (the number of latent factors),
           CPU: the number of CPUs.
\# runs with K = 3 and 2 repetitions.
# around 15 seconds per run.
project = NULL
project = lfmm("genotypes.lfmm", "gradients.env", K = 3, repetitions = 2,
    iterations = 6000, burnin = 3000, project = "new")
# get adjusted p-values using the genomic control method
p = lfmm.pvalues(project, K = 3)
```

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```
hist(p$pvalues, col = "yellow3")
# get adjusted p-values using lambda = 0.6
p = lfmm.pvalues(project, genomic.control = FALSE,
    lambda = 0.6, K = 3)
hist(p$pvalues, col = "yellow3")
```

1fmm2

Fitting Latent Factor Mixed Models (Least squares algorithm)

#### **Description**

Latent Factor Mixed Models (LFMMs) are factor regression models in which the response variable is a genotypic matrix, and the explanatory variables are environmental measures of ecological interest or trait values. The 1fmm2 function estimates latent factors based on an exact least-squares approach. The resulting object can be used by the function 1fmm2.test to identify genetic polymorphisms exhibiting association with ecological gradients or phenotypes, while correcting for unobserved confounders. An MCMC estimation algorithm is implemented in the function 1fmm, but this version should be prefered.

#### Usage

```
lfmm2 (input, env, K, lambda)
```

#### **Arguments**

input	A genotypic matrix or a character string containing a path to the input file. The genotypic matrix must be in the lfmm{lfmm_format} format without missing values (9 or NA). See impute for completion based on nonnegative matrix factorization and consider R packages for reading large matrices.
env	A matrix of environmental covariates or a character string containing a path to the environmental file. The environment matrix must be in the env format without missing values. Response variables must be encoded as numeric.
K	An integer corresponding to the number of latent factors.
lambda	A positive numeric value for a ridge regularization parameter. The default value is set to 1e-5.

## Value

 $1 fmm2\ returns\ an\ object\ of\ class\ 1 fmm2Class\ that\ contains\ $K\$\ estimated\ latent\ factors\ @U\ and\ their\ loadings\ @V.$ 

The following method can be applied to an object of class 1fmm2Class:

1fmm2. test P-values adjusted for latent factors computed by lfmm2.

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## Author(s)

Olivier François

#### References

Caye K, Jumentier B, Lepeule J, Francois O. (2019). LFMM 2: fast and accurate inference of gene-environment associations in genome-wide studies. Molecular biology and evolution, 36(4), 852-860.

Gain C, Francois O. (2021). LEA 3: Factor models in population genetics and ecological genomics with R. Molecular Ecology Resources. doi: 10.1111/1755-0998.13366

#### See Also

1fmm.data impute 1fmm2.test pca 1fmm tutorial

```
### Example of analysis using lfmm2 ###
# Simulation with 10 target loci, with effect sizes ranging between -10 an 10
# n = 100 individuals and L = 1000 loci
X <- as.matrix(rnorm(100)) # causal environmental variable</pre>
B < - rep(0, 1000)
target <- sample(1:1000, 10) # target loci</pre>
B[target] <- runif(10, -10, +10) # effect sizes
# Creating hidden factors and loadings
U \leftarrow t(tcrossprod(as.matrix(c(-1,0.5,1.5)), X)) + matrix(rnorm(300), ncol = 3)
V \leftarrow matrix(rnorm(3000), ncol = 3)
# Simulating a binarized matrix containing haploid genotypes
# Simulation performed with the generative LFMM
Y \leftarrow tcrossprod(as.matrix(X), B) + tcrossprod(U, V) + matrix(rnorm(100000, sd = .5), nrow = 100)
Y \leftarrow matrix(as.numeric(Y > 0), ncol = 1000)
# Fitting an LFMM with K = 3 factors #
mod2 \leftarrow 1fmm2(input = Y, env = X, K = 3)
# Computing P-values and plotting their minus log10 values
# Target loci are highlighted
pv <- lfmm2.test(object = mod2, input = Y, env = X, linear = TRUE)</pre>
plot(-log10(pv$pvalues), col = "grey", cex = .4, pch = 19)
points(target, -log10(pv$pvalues[target]), col = "red")
```

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lfmm2.	test
--------	------

*P-values adjusted for latent factors computed by* 1fmm2.

#### **Description**

The function returns a vector of p-values for association between loci and environmental variables adjusted for latent factors computed by 1fmm2. It takes an object of class 1fmm2Class with the data that were used to adjust the LFMM.

#### Usage

```
lfmm2.test (object, input, env, genomic.control, linear, family)
```

# **Arguments**

object An object of class 1fmm2Class.

input A genotypic matrix or a character string containing a path to the input file. The

genotypic matrix must be in the lfmm{lfmm\_format} format without missing values (9 or NA). See impute for completion based on nonnegative matrix fac-

torization and consider R packages for reading large matrices.

env A matrix of environmental covariates or a character string containing a path

to the environmental file. The environment matrix must be in the env format

without missing values. Variables must be encoded as numeric.

genomic.control

A logical value. If TRUE, the p-values are recalibrated by using genomic control

after correction for confounding.

linear A logical value indicating whether linear or generalized linear models should be

used to perform the association tests. If FALSE, family should be provided in

the next argument.

family a family for generalized linear models used in the association tests. The default

is binomial(link = "logit")), which requires that y is between 0 and 1.

#### Value

pvalues A matrix of p-values for each locus and each environmental variable.

zscores A matrix of z-scores for each locus and each environmental variable.

gif A vector of genomic inflation factors computed for each environmental variable.

#### Author(s)

Olivier Francois

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## References

Caye K, Jumentier B, Lepeule J, Francois O. (2019). LFMM 2: fast and accurate inference of gene-environment associations in genome-wide studies. Molecular biology and evolution, 36(4), 852-860.

#### See Also

```
1fmm.data1fmm2
```

```
### Example of analysis using lfmm2 ###
# Simulation with 10 target loci, with effect sizes ranging between -10 an 10
\# n = 100 individuals and L = 1000 loci
X <- as.matrix(rnorm(100)) # environmental variable</pre>
B \leftarrow rep(0, 1000)
target <- sample(1:1000, 10) # target loci</pre>
B[target] <- runif(10, -10, +10) # effect sizes
# Creating hidden factors and loadings
U \leftarrow t(tcrossprod(as.matrix(c(-1,0.5,1.5)), X)) + matrix(rnorm(300), ncol = 3)
V \leftarrow matrix(rnorm(3000), ncol = 3)
# Simulating a binarized matrix containing haploid genotypes
# Simulation performed with the generative LFMM
Y \leftarrow tcrossprod(as.matrix(X), B) + tcrossprod(U, V) + matrix(rnorm(100000, sd = .5), nrow = 100)
Y \leftarrow matrix(as.numeric(Y > 0), ncol = 1000)
# Fitting an LFMM with K = 3 factors #
mod2 \leftarrow 1fmm2(input = Y, env = X, K = 3)
# Computing P-values and plotting their minus log10 values
# Target loci are highlighted
pv <- lfmm2.test(object = mod2, input = Y, env = X, linear = TRUE)</pre>
plot(-log10(pv$pvalues), col = "grey", cex = .4, pch = 19)
points(target, -log10(pv$pvalues[target]), col = "red")
```

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## **Description**

A function that converts from the 1fmm format to the geno format.

## Usage

```
lfmm2geno(input.file, output.file = NULL, force = TRUE)
```

# **Arguments**

input.file A character string containing a path to the input file, a genotypic matrix in the

1fmm format.

output.file A character string containing a path to the output file, a genotypic matrix in the

geno format. By default, the name of the output file is the same name of the

input file with a .geno extension.

force A boolean option. If FALSE, the input file is converted only if the output file

does not exist. If TRUE, convert the file anyway.

#### Value

output.file A character string containing a path to the output file, a genotypic matrix in the

geno format.

## Author(s)

Eric Frichot

# See Also

lfmm.datagenoancestrymap2lfmmancestrymap2genogeno2lfmmped2lfmmped2genovcf2geno

```
# Creation of a file called "genotypes.lfmm" in the working directory,
# with 400 SNPs for 50 individuals.
data("tutorial")
write.lfmm(tutorial.R, "genotypes.lfmm")
# Conversion
               from the lfmm format ("genotypes.lfmm")
               to the geno format ("genotypes.geno").
# By default, the name of the output file is the same name
               as the input file with a .geno extension.
# Create file: "genotypes.geno".
output = lfmm2geno("genotypes.lfmm")
# Conversion
                from the lfmm format ("genotypes.lfmm")
                to the geno format with the output file called "plop.geno".
# Create file: "plop.geno".
output = lfmm2geno("genotypes.lfmm", "plop.geno")
# As force = false and the file "genotypes.geno" already exists,
```

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```
# nothing happens.
output = lfmm2geno("genotypes.lfmm", force = FALSE)
```

pca

Principal Component Analysis

#### **Description**

The pca function performs a principal component analysis of a genotypic matrix encoded in one of the following formats: 1fmm, geno, ancestrymap, ped or vcf. The pca function computes eigenvalues, eigenvectors, and standard deviations for all principal components and the projections of individuals on each component. Thepca function returns an object of class "pcaProject" containing the output data and the input parameters.

## Usage

```
pca (input.file, K, center = TRUE, scale = FALSE)
```

# **Arguments**

input.file A character string containg the path to the genotype input file, a genotypic matrix

in the 1fmm format.

K An integer corresponding to the number of principal components calculated. By

default, all principal components are calculated.

center A boolean option. If TRUE, the data matrix is centered (default: TRUE).

scale A boolean option. If TRUE, the data matrix is centered and scaled (default:

FALSE).

#### Value

pca returns an object of class pcaProject containing the following components:

eigenvalues The vector of eigenvalues.

eigenvectors The matrix of eigenvectors (one column for each eigenvector).

sdev The vector of standard deviations.

projections The matrix of projections (one column for each projection).

The following methods can be applied to the object of class pcaProject returned by pca:

plot Plot the eigenvalues.

show Display information on analysis.

summary Summarize analysis.

tracy.widom Perform Tracy-Widom tests for eigenvalues.

load.pcaProject(file.pcaProject)

Load the file containing a pcaProject object and return the pcaProject object.

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```
remove.pcaProject(file.pcaProject)
```

Erase a pcaProject object. Caution: All the files associated with the pcaProject object will be removed except the genotype file.

```
export.pcaProject(file.pcaProject)
```

Create a zip file containing the full pcaProject object. It allows users to move the project to a new directory or a new computer (using import). If you want to overwrite an existing export, use the option force == TRUE.

```
import.pcaProject(file.pcaProject)
```

Import and load an pcaProject object from a zip file (made with the export function) into the chosen directory. If you want to overwrite an existing project, use the option force == TRUE.

## Author(s)

Eric Frichot

#### See Also

```
1fmm.data snmf 1fmm 1fmm2 tutorial
```

```
# Create a genotype file "genotypes.lfmm"
# with 1000 SNPs for 165 individuals.
data("tutorial")
write.lfmm(tutorial.R, "genotypes.lfmm")
#################
# Perform PCA
##################
# run PCA
# Available options: K (the number of PCs),
                     center and scale.
# Creation of genotypes.pcaProject - the pcaProject object.
               a directory genotypes.pca containing:
#
# genotypes.eigenvalues - eigenvalues,
# genotypes.eigenvectors - eigenvectors,
# genotypes.sdev - standard deviations,
# genotypes.projections - projections,
# Create a pcaProject object: pc.
pc <- pca("genotypes.lfmm", scale = TRUE)</pre>
##########################
# Display information #
##########################
# Display information on analysis.
show(pc)
```

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```
# Summarize analysis.
summary(pc)
# Graphical outputs #
####################################
par(mfrow=c(2,2))
# Plot eigenvalues.
plot(pc, lwd=5, col="blue", cex = .7, xlab=("Factors"), ylab="Eigenvalues")
# PC1-PC2 plot.
plot(pc$projections)
# PC3-PC4 plot.
plot(pc$projections[,3:4])
# Plot standard deviations.
plot(pc$sdev)
# Perform Tracy-Widom tests #
# Perfom Tracy-Widom tests for all eigenvalues.
# Create file: genotypes.tracyWidom - tracy-widom test information,
          in the directory genotypes.pca/.
tw <- tracy.widom(pc)</pre>
# Plot the percentage of variance explained by each component.
plot(tw$percentage)
# Show p-values for the Tracy-Widom tests.
tw$pvalues
# Manage a pca project #
# All the project files for a given input matrix are
# automatically saved into a pca project directory.
# The name of the pcaProject file is the same name as
# the name of the input file with a .pcaProject extension
# ("genotypes.pcaProject").
# The name of the pcaProject directory is the same name as
# the name of the input file with .pca extension ("genotypes.pca/")
# There is only one pca Project for each input file including all the runs.
# An pcaProject can be load in a different session.
project = load.pcaProject("genotypes.pcaProject")
# An pcaProject can be exported to be imported in another directory
# or in another computer
```

ped ped

```
export.pcaProject("genotypes.pcaProject")

dir.create("test", showWarnings = TRUE)
#import
newProject = import.pcaProject("genotypes_pcaProject.zip", "test")
# remove
remove.pcaProject("test/genotypes.pcaProject")

# A pcaProject can be erased.
# Caution: All the files associated with the project will be removed.
remove.pcaProject("genotypes.pcaProject")
```

ped

ped format description

# **Description**

Description of the ped format. The ped format can be used as an input format for genotypic matrices in the functions snmf, 1fmm, and pca.

# **Details**

The ped format has one row for each individual. Each row contains 6 columns of information for each individual, plus two genotype columns for each SNP. Each column must be separated by spaces or tabulations. The genotype format must be either 0ACGT or 01234, where 0 means missing genotype. The first 6 columns of the genotype file are: the 1st column is the family ID, the 2nd column is the sample ID, the 3rd and 4th columns are the sample IDs of parents, the 5th column is the gender (male is 1, female is 2), the 6th column is the case/control status (1 is control, 2 is case), the quantitative trait value or the population group label.

The ped format is described here.

Here is an example with 3 individuals and 4 SNPs:

```
1 SAMPLE0 0 0 2 2 1 2 3 3 1 1 2 1
2 SAMPLE1 0 0 1 2 2 1 1 3 0 4 1 1
3 SAMPLE2 0 0 2 1 2 2 3 3 1 4 1 2
```

## Author(s)

Eric Frichot

## See Also

ped2lfmm ped2geno geno lfmm.data ancestrymap vcf

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ped2geno	Convert from ped to geno format

## **Description**

A function that converts from the ped format to the geno format.

# Usage

```
ped2geno(input.file, output.file = NULL, force = TRUE)
```

# **Arguments**

input.file A character string containing a path to the input file, a genotypic matrix in the

ped format.

output.file A character string containing a path to the output file, a genotypic matrix in the

geno format. By default, the name of the output file is the same name as the

input file with a .geno extension.

force A boolean option. If FALSE, the input file is converted only if the output file

does not exist. If TRUE, convert the file anyway.

## Value

output.file A character string containing a path to the output file, a genotypic matrix in the

geno format.

# Author(s)

Eric Frichot

## See Also

ped geno ancestrymap2lfmm ancestrymap2geno geno2lfmm ped2lfmm vcf2geno lfmm2geno

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```
# Conversion from the ped format ("example.ped")
# to the geno format with the output file called "plop.geno".
# Create file: "plop.geno".
output = ped2geno("example.ped", "plop.geno")

# As force = false and the file "example.geno" already exists,
# nothing happens.
output = ped2geno("example.ped", force = FALSE)
```

ped21fmm

Convert from ped to 1fmm format

#### **Description**

A function that converts from the ped format to the 1fmm format.

## Usage

```
ped2lfmm(input.file, output.file = NULL, force = TRUE)
```

# Arguments

input.file A character string containing a path to the input file, a genotypic matrix in the

ped format.

output.file A character string containing a path for the output file, a genotypic matricx in

the 1fmm format. By default, the name of the output file is the same name as the

input file with a .lfmm extension.

force A boolean option. If FALSE, the input file is converted only if the output file

does not exist. If TRUE, convert the file anyway.

#### Value

output.file A character string containing a path for the output file, a genotypic matricx in

the 1fmm format.

# Author(s)

Eric Frichot

#### See Also

pedlfmm.dataancestrymap2lfmmancestrymap2genogeno2lfmmped2genovcf2genolfmm2geno

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#### **Examples**

```
# Creation of a file called "example.ped"
# with 4 SNPs for 3 individuals.
data("example_ped")
write.table(example_ped, "example.ped",
    col.names = FALSE, row.names = FALSE, quote = FALSE)
# Conversion
               from the ped format ("example.ped")
                to the lfmm format ("example.lfmm").
# By default, the name of the output file is the same name
                as the input file with a .lfmm extension.
# Create file: "example.lfmm".
output = ped21fmm("example.ped")
# Conversion
                from the ped format ("example.ped")
                to the geno format with the output file called "plop.lfmm".
# Create file: "plop.lfmm".
output = ped2lfmm("example.ped", "plop.lfmm")
# As force = false and the file "example.lfmm" already exists,
# nothing happens.
output = ped2lfmm("example.ped", force = FALSE)
```

Admixture coefficients from a snmf run

# Description

Q

Return the snmf output matrix of admixture coefficients for the chosen run with K ancestral populations. For an example, see snmf.

#### Usage

```
Q(object, K, run)
```

#### **Arguments**

object A snmfProject object.

K The number of ancestral populations.

run A chosen run.

#### Value

res A matrix containing the admixture coefficients for the chosen run with K ances-

tral populations.

#### Author(s)

Eric Frichot

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#### See Also

```
geno snmf G cross.entropy
```

#### **Examples**

```
### Example of analysis using snmf ###
# Creation of the genotype file: genotypes.geno.
# The data contain 400 SNPs for 50 individuals.
data("tutorial")
write.geno(tutorial.R, "genotypes.geno")
#################
# running snmf #
#################
project.snmf <- snmf("genotypes.geno",</pre>
                K = 3,
                repetitions = 2,
                project = "new")
\# get the ancestry coefficients for the 2nd run for K = 3.
Q.3 \leftarrow Q(project.snmf, K = 3, run = 2)
# cluster assignment for each individual
cluster <- apply( Q.3, 1, which.max)</pre>
table(cluster)
```

read.env

Read environmental file in the envformat

# **Description**

Read a file in the env format.

#### Usage

```
read.env(input.file)
```

# **Arguments**

input.file

A character string containing a path to the input file, an environmental data matrix in the env format.

# Value

R

A matrix containing the environmental variables with one line for each individual and one column for each environmental variable.

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#### Author(s)

Eric Frichot

#### See Also

```
env write.env lfmm
```

#### **Examples**

```
# Creation of an environmental matrix, C
# containing 2 environmental variables for 3 individuals.
# C contains one line for each individual and one column for each variable.
C = matrix(runif(6), ncol=2, nrow=3)

# Write C in a file called "example.env".
# Create file: "example.env".
write.env(C, "example.env")

# Read the file "example.env".
C = read.env("example.env")
```

read.geno

read a file in the geno format

#### **Description**

Read a file in the geno format.

#### Usage

```
read.geno(input.file)
```

#### **Arguments**

input.file A character string containing a path to the input file, a genotypic matrix in the

geno format.

# Value

R A matrix containing the genotypes with one line for each individual and one

column for each SNP.

# Author(s)

Eric Frichot

#### See Also

write.geno geno snmf geno2lfmm lfmm2geno ancestrymap2geno ped2geno vcf2geno

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#### **Examples**

```
# tutorial contains a matrix of genotypes R with 1000 SNPs for 165 individuals.
# and a matrix with an environmental variable C.
data("tutorial")

# Write R in a file called "genotypes.geno".
# Create file: "genotypes.geno".
write.geno(tutorial.R, "genotypes.geno")

# Read the file "genotypes.geno".
R = read.geno("genotypes.geno")
```

read.lfmm

Read files in the 1fmm format

#### **Description**

Read a file in the 1fmm format.

#### Usage

```
read.lfmm(input.file)
```

#### **Arguments**

input.file A character string containing a path to the input file, a genotypic matrix in the

1fmm format.

#### Value

R A matrix containing the genotypes with one line per individual and one column

per SNP.

#### Author(s)

Eric Frichot

# See Also

```
write.lfmm lfmm.data lfmm geno2lfmm lfmm2geno ancestrymap2lfmm ped2lfmm
```

#### **Examples**

```
# tutorial contains a matrix of genotypes R with 1000 SNPs for 165 individuals.
# and a matrix with an environmental variable C.
data("tutorial")

# write R in a file called "genotypes.lfmm"
# Create file: "genotypes.lfmm".
```

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```
write.lfmm(tutorial.R, "genotypes.lfmm")
# read the file "genotypes.lfmm".
R = read.lfmm("genotypes.lfmm")
```

read.zscore

Read the output files of 1fmm

# **Description**

Read the output file from 1fmm. This is an internal function. Zscores of a run can be accessed using the function z.scores.

# Usage

```
read.zscore(input.file)
```

#### **Arguments**

input.file a character string containing a path to the output of 1fmm.

#### Value

R

A matrix containing the 1fmm results with one line per SNP. The first column is the zscore. The second column is the -log10(p-value). The third column is the p-value.

#### Author(s)

Eric Frichot

#### See Also

```
zscore.formatlfmm
```

#### **Examples**

#### #################

snmf

Estimates individual ancestry coefficients and ancestral allele frequencies.

# **Description**

snmf estimates admixture coefficients using sparse Non-Negative Matrix Factorization algorithms, and provides STRUCTURE-like outputs.

#### **Usage**

```
snmf (input.file, K,
    project = "continue",
    repetitions = 1, CPU = 1,
    alpha = 10, tolerance = 0.00001, entropy = FALSE, percentage = 0.05,
    I, iterations = 200, ploidy = 2, seed = -1, Q.input.file)
```

# **Arguments**

input.file	A character string containing a the path to the input file, a genotypic matrix in the geno format.
K	An integer vector corresponding to the number of ancestral populations for which the snmf algorithm estimates have to be calculated.
project	A character string among "continue", "new", and "force". If "continue", the results are stored in the current project. If "new", the current project is removed and a new one is created to store the result. If "force", the results are stored in the current project even if the input file has been modified since the creation of the project.
repetitions	An integer corresponding with the number of repetitions for each value of K.
CPU	A number of CPUs to run the parallel version of the algorithm. By default, the number of CPUs is 1.
alpha	A numeric value corresponding to the snmf regularization parameter. The results

can depend on the value of this parameter, especially for small data sets.

tolerance A numeric value for the tolerance error.

entropy A boolean value. If true, the cross-entropy criterion is calculated (see create.dataset

and cross.entropy.estimation).

percentage A numeric value between 0 and 1 containing the percentage of masked geno-

types when computing the cross-entropy criterion. This option applies only if

entropy == TRUE (see cross.entropy).

I The number of SNPs to initialize the algorithm. It starts the algorithm with a

run of snmf using a subset of nb.SNPs random SNPs. If this option is set with nb.SNPs, the number of randomly chosen SNPs is the minimum between 10000 and 10% of all SNPs. This option can considerably speeds up snmf estimation

for very large data sets.

iterations An integer for the maximum number of iterations in algorithm.

ploidy 1 if haploid, 2 if diploid, n if n-ploid.

seed A seed to initialize the random number generator. By default, the seed is ran-

domly chosen.

Q. input. file A character string containing a path to an initialization file for Q, the individual

admixture coefficient matrix.

#### Value

snmf returns an object of class snmfProject.

The following methods can be applied to the object of class snmfProject:

plot Plot the minimal cross-entropy in function of K.

show Display information about the analyses.

summary Summarize the analyses.

Q Return the admixture coefficient matrix for the chosen run with K ancestral pop-

ulations.

G Return the ancestral allele frequency matrix for the chosen run with K ancestral

populations.

cross.entropy Return the cross-entropy criterion for the chosen runs with K ancestral popula-

tions.

snmf.pvalues Return the vector of adjusted p-values for a run with K ancestral populations.

impute Return a geno or 1fmm file with missing data imputation.

barchart Return a bar plot representation of the Q matrix from a run with K ancestral

populations.

load.snmfProject(file.snmfProject)

Load the file containing an snmfProject objet and return the snmfProject object.

remove.snmfProject(file.snmfProject)

Erase a snmfProject object. Caution: All the files associated with the object

will be removed.

export.snmfProject(file.snmfProject)

Create a zip file containing the full snmfProject object. It allows to move the project to a new directory or a new computer (using import). If you want to

overwrite an existing export, use the option force == TRUE.

```
import.snmfProject(file.snmfProject)
```

Import and load an snmfProject object from a zip file (made with the export function) into the chosen directory. If you want to overwrite an existing project, use the option force == TRUE.

```
combine.snmfProject(file.snmfProject, toCombine.snmfProject)
```

Combine to Combine snmfProject into file snmfProject. Caution: Only projects with runs coming from the same input file can be combined. If the same input file has different names in the two projects, use the option force == TRUE.

#### Author(s)

Eric Frichot

#### References

Frichot E, Mathieu F, Trouillon T, Bouchard G, Francois O. (2014). Fast and Efficient Estimation of Individual Ancestry Coefficients. Genetics, 194(4): 973–983.

#### See Also

```
geno pca lfmm Q barchart tutorial
```

#### **Examples**

```
### Example of analysis using snmf ###
# Creation of the genotype file: genotypes.geno.
# The data contain 400 SNPs for 50 individuals.
data("tutorial")
write.geno(tutorial.R, "genotypes.geno")
##################
# running snmf #
#################
project.snmf = snmf("genotypes.geno",
                K = 1:10,
                entropy = TRUE,
                repetitions = 10,
                project = "new")
# plot cross-entropy criterion of all runs of the project
plot(project.snmf, cex = 1.2, col = "lightblue", pch = 19)
# get the cross-entropy of the 10 runs for K = 4
ce = cross.entropy(project.snmf, K = 4)
# select the run with the lowest cross-entropy for K = 4
best = which.min(ce)
# display the Q-matrix
```

```
my.colors <- c("tomato", "lightblue",</pre>
              "olivedrab", "gold")
barchart(project.snmf, K = 4, run = best,
        border = NA, space = 0, col = my.colors,
        xlab = "Individuals", ylab = "Ancestry proportions",
        main = "Ancestry matrix") -> bp
axis(1, at = 1:length(bp$order),
    labels = bp$order, las = 3, cex.axis = .4)
######################
# Post-treatments #
##############################
# show the project
show(project.snmf)
# summary of the project
summary(project.snmf)
\# get the cross-entropy for all runs for K = 4
ce = cross.entropy(project.snmf, K = 4)
# get the cross-entropy for the 2nd run for K = 4
ce = cross.entropy(project.snmf, K = 4, run = 2)
\# get the ancestral genotype frequency matrix, G, for the 2nd run for K = 4.
freq = G(project.snmf, K = 4, run = 2)
# Advanced snmf run options #
#####################################
# Q.input.file: init a run with a given ancestry coefficient matrix Q.
# Here it is initialized with the Q matrix from the first run with K=4
project.snmf = snmf("genotypes.geno", K = 4,
    Q.input.file = "./genotypes.snmf/K4/run1/genotypes_r1.4.Q")
# I: init the Q matrix of a run from a smaller run with 100 randomly chosen
# SNPs.
project.snmf = snmf("genotypes.geno", K = 4, I = 100)
# CPU: run snmf with 2 CPUs.
project.snmf = snmf("genotypes.geno", K = 4, CPU = 2)
# percentage: run snmf and calculate the cross-entropy criterion with 10% of
# masked genotypes, instead of 5% of masked genotypes.
project.snmf = snmf("genotypes.geno", K = 4, entropy = TRUE, percentage = 0.1)
# seed: choose the seed for the random generator.
project.snmf = snmf("genotypes.geno", K = 4, seed = 42)
```

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```
# alpha: choose the regularization parameter.
project.snmf = snmf("genotypes.geno", K = 4, alpha = 100)
# tolerance: choose the tolerance parameter.
project.snmf = snmf("genotypes.geno", K = 4, tolerance = 0.0001)
# Manage an snmf project #
# All the runs of snmf for a given file are
# automatically saved into an snmf project directory and a file.
# The name of the snmfProject file is the same name as
# the name of the input file with a .snmfProject extension
# ("genotypes.snmfProject").
# The name of the snmfProject directory is the same name as
# the name of the input file with a .snmf extension ("genotypes.snmf/")
# There is only one snmf Project for each input file including all the runs.
# An snmfProject can be load in a different session.
project.snmf = load.snmfProject("genotypes.snmfProject")
# An snmfProject can be exported to be imported in another directory
# or in another computer
export.snmfProject("genotypes.snmfProject")
dir.create("test", showWarnings = TRUE)
#import
newProject = import.snmfProject("genotypes_snmfProject.zip", "test")
# combine projects
combinedProject = combine.snmfProject("genotypes.snmfProject", "test/genotypes.snmfProject")
remove.snmfProject("test/genotypes.snmfProject")
# An snmfProject can be erased.
# Caution: All the files associated with the project will be removed.
remove.snmfProject("genotypes.snmfProject")
```

snmf.pvalues

P-values for snmf population differentiation tests

#### **Description**

Returns a vector of p-values computed from an snmf run.

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# Usage

snmf.pvalues (object, genomic.control, lambda, ploidy, entropy, fisher, K, run)

#### **Arguments**

object An snmfProject object.

genomic.control

A Boolean value. If TRUE, the p-values are automatically calibrated using genomic control. If FALSE, the p-values are calculated by rescaling the chi-squared

test statistics using the lambda parameter.

lambda A numeric value. The lambda value is used as an inflation factor to rescale the

chi-squared statistics in the computation of p-values. This option requires that genomic.control = FALSE. The default value of lambda is equal to 1.0 (no

rescaling).

ploidy An integer value among 1 or 2. Tests are implemented for haploids and diploids

(to be extended to polypoids).

entropy A Boolean value. If TRUE, the run of minimum entropy is used for computing

the p-values.

fisher A Boolean value. If TRUE, F-distributions are used to test the null-hypothesis,

Chi-squared otherwise.

K An integer value. The number of genetic clusters.

run An integer for the run number used the computation of p-values (by default, the

minimum entropy run).

#### Value

p. values A vector of p-values for each locus for the population differentiation test.

GIF The inflation factor value used in the test.

#### Author(s)

Olivier François

#### References

Martins, H., Caye, K., Luu, K., Blum, M. G. B., Francois, O. (2016). Identifying outlier loci in admixed and in continuous populations using ancestral population differentiation statistics. Molecular Ecology, 25(20), 5029-5042.

#### See Also

snmf

48 struct2geno

#### **Examples**

```
### Example of analyses using snmf ###
data("tutorial")
# creation of a genotype file, "genotypes.lfmm".
# The data contain 400 SNPs for 50 individuals.
write.geno(tutorial.R, "genotypes.geno")
###################
# snmf runs
################
# main options, K: the number of ancestral populations,
         entropy: cross-entropy criterion,
         CPU: the number of CPUs.
project.snmf = snmf("genotypes.geno",
                    K = 4
                    entropy = TRUE,
                    ploidy = 2,
                    repetitions = 10,
                    project = "new")
# genome scan using adjusted p-values (genomic control method)
p = snmf.pvalues(project.snmf, entropy = TRUE, ploidy = 2, K = 4)
par(mfrow = c(2,1))
hist(p$pvalues, col = "orange")
plot(-log10(p$pvalues), pch = 19, col = "blue", cex = .7)
```

struct2geno

Conversion from the STRUCTURE format to the geno format.

#### **Description**

The function converts a multiallelic genotype file in the STRUCTURE format into a file in the 'geno' for snmf and the 'lfmm' format for 1fmm.

#### Usage

```
struct2geno (input.file, ploidy, FORMAT, extra.row, extra.column)
```

struct2geno 49

# **Arguments**

input.file	A character string. A path to a STRUCTURE or a TESS input file of multiallelic markers (eg, microsatellites) for haploid or diploid individuals. Missing data must be encoded as "-9" or as any negative value. Individual genotypes are encoded using either one or two rows of data.
ploidy	An integer value (1 or 2). Value 2 for diploids and 1 for haploids.
FORMAT	An integer value equal to 1 for markers encoded using one row of data for each individual, and 2 for markers encoded using two rows of data for each individual.
extra.row	An integer value indicating the number of extra rows in the header of the input file (eg, marker ids).
extra.column	an integer value indicating the number of extra columns in the input file. Extra columns can include individual ids, pop ids, geographic coordinates, etc.

#### Value

NULL. Output files in the 'geno' and the 'lfmm' format record individual genotypes for each allele at each marker.

# Author(s)

Olivier Francois

#### See Also

lfmm.data geno lfmm snmf

# **Examples**

```
### Example of conversion from a STRUCTURE format ###
### Artificial data with 10 diploid individuals and 10 STR markers
### FORMAT = 1
### Input file: 'dat.str'
dat.str <- matrix(sample(c(101:105,-9),
                  200, prob = c(rep(1,5), 0.1),
                  replace = TRUE),
                  nrow = 10, ncol = 20)
write.table(dat.str,
           file = "dat.str",
            col.names = FALSE,
            row.names = FALSE,
            quote = FALSE)
### Conversion
struct2geno("dat.str", ploidy = 2, FORMAT = 1)
### snmf run and barplot
s <- snmf("dat.str.geno", K = 2, project = "new")</pre>
barchart(s, K = 2, run = 1, xlab = "Individuals")
```

50 tracy.widom

tracy.widom	Tracy-Widom test for eigenvalues

# **Description**

Perform tracy-widom tests on a set of eigenvalues to determine the number of significative eigenvalues and calculate the percentage of variance explained by each principal component. For an example, see pca.

#### Usage

```
tracy.widom (object)
```

# **Arguments**

object a pcaProject object.

#### Value

tracy.widom returns a list containing the following components:

eigenvalues The sorted input vector of eigenvalues (by descreasing order).

twstats The vector of tracy-widom statistics.

pvalues The vector of p-values associated with each eigenvalue.

effecn The vector of effective sizes.

percentage The vector containing the percentage of variance explained by each principal

component.

# Author(s)

Eric Frichot

#### References

Tracy CA and Widom H. (1994). Level spacing distributions and the bessel kernel. Commun Math Phys. 161:289–309. Patterson N, Price AL and Reich D. (2006). Population structure and eigenanalysis. PLoS Genet. 2:20.

#### See Also

pcalfmm.datalfmm

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#### **Examples**

```
# Creation of the genotype file "genotypes.lfmm"
# with 1000 SNPs for 165 individuals.
data("tutorial")
write.lfmm(tutorial.R, "genotypes.lfmm")
#################
# Perform a PCA #
#################
# run of PCA
# Available
               options, K (the number of PCs calculated),
               center and scale.
# Creation of genotypes.pcaProject - the pcaProject object.
               a directory genotypes.pca containing:
# Create files: genotypes.eigenvalues - eigenvalues,
               genotypes.eigenvectors - eigenvectors,
#
               genotypes.sdev - standard deviations,
               genotypes.projections - projections,
# Create a pcaProject object: pc.
pc = pca("genotypes.lfmm", scale = TRUE)
# Perform Tracy-Widom tests #
# Perfom Tracy-Widom tests on all eigenvalues.
# Create file: genotypes.tracyWidom - tracy-widom test information,
               in the directory genotypes.pca/.
tw = tracy.widom(pc)
# Plot the percentage of variance explained by each component.
plot(tw$percentage)
# Display the p-values for the Tracy-Widom tests.
tw$pvalues
# remove pca Project
remove.pcaProject("genotypes.pcaProject")
```

tutorial

Example tutorial data sets

#### **Description**

This dataset is composed of a genotypic matrix stored tutorial.R with 50 individuals genotyped for 400 SNPs. The last 50 SNPs are correlated with an environmental variable (tutorial.C). This dataset is a subset of the data shown in the computer note associated with the package (Frichot and Francois 2015).

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#### Usage

tutorial

#### Value

tutorial.R A genotypic matrix for 50 individuals genotyped at 400 SNPs. The last 50 SNPs are correlated with an environmental variable stored in tutorial.C.

tutorial. C An environmental variable for 50 invdividuals.

vcf

vcf format description

# **Description**

Description of the vcf format. The vcf format can be used as an input format for genotypic matrices in the functions snmf, 1fmm, and pca.

#### **Details**

The vcf format is described here.

Here is an example of a genotypic matrix using the vcf format with 3 individuals and 4 loci:

```
##fileformat=VCFv4.1
##FORMAT=<ID=GM,Number=1,Type=Integer,Description="Genotype meta">
##INFO=<ID=VM,Number=1,Type=Integer,Description="Variant meta">
##INFO=<ID=SM,Number=1,Type=Integer,Description="SampleVariant meta">
#CHROM POS ID REF ALT QUAL FILTER INFO FORMAT SAMPLE0 SAMPLE1 SAMPLE2
1 1001 rs0000 T C 999 . VM=1;SM=100 GT:GM 1/0:1 0/1:2 1/1:3
1 1002 rs1111 G A 999 . VM=2;SM=101 GT:GM 0/0:6 0/1:7 0/0:8
1 1003 notres G AA 999 . VM=3;SM=102 GT:GM 0/0:11 . /.:12 0/1:13
1 1004 rs2222 G A 999 . VM=3;SM=102 GT:GM 0/0:11 . /.:12 0/1:13
1 1005 rs3333 G A 999 . VM=3;SM=102 GT:GM 1/0:11 1/1:12 0/1:13
```

#### Author(s)

Eric Frichot

# See Also

```
vcf2geno vcf2lfmm geno lfmm ped ancestrymap
```

vcf2geno 53

vcf2geno	Convert from vcf to geno format	
----------	---------------------------------	--

# **Description**

A function that converts from the vcf format to the geno format. Note: This function may be obsolete. Conversion in accepted format such as ped can be obtained with the program vcftools.

# Usage

```
vcf2geno(input.file, output.file = NULL, force = TRUE)
```

# **Arguments**

input.file	A character string containing a path to the input file, a genotypic matrix in the vcf format.
output.file	A character string containing a path to the output file, a genotypic matrix in the geno format. By default, the name of the output file is the same name as the input file with a .geno extension.
force	A boolean option. If FALSE, the input file is converted only if the output file does not exist. If TRUE, convert the file anyway.

# Value

output.file A character string containing a path to the output file, a genotypic matrix in the geno format.

#### Author(s)

Eric Frichot

#### See Also

vcf geno ancestrymap2lfmm ancestrymap2geno ped2lfmm ped2geno lfmm2geno geno2lfmm

# **Examples**

54 vcf2lfmm

```
as the input file with a .geno extension.
# Create files: "example.geno",
#
                "example.vcfsnp" - SNP informations,
#
                "example.removed" - removed lines.
output = vcf2geno("example.vcf")
                from the vcf format ("example.vcf")
# Conversion
                to the geno format with the output file called "plop.geno".
# Create files: "plop.geno",
                "plop.vcfsnp" - SNP informations,
                "plop.removed" - removed lines.
output = vcf2geno("example.vcf", "plop.geno")
# As force = false and the file "example.geno" already exists,
# nothing happens.
output = vcf2geno("example.vcf", force = FALSE)
```

vcf21fmm

Convert from vcf to 1fmm format

# **Description**

A function that converts from the vcf format to the lfmm format. Note: This function may be obsolete. Conversion in accepted format such as ped can be obtained with the program vcftools.

# Usage

```
vcf2lfmm(input.file, output.file = NULL, force = TRUE)
```

#### **Arguments**

input.file A character string containing a path to the input file, a genotypic matrix in the

vcf format.

output.file A character string containing a path to the output file, a genotypic matrix in the

1fmm format. By default, the name of the output file is the same name as the

input file with a .lfmm extension.

force A boolean option. If FALSE, the input file is converted only if the output file

does not exist. If TRUE, convert the file anyway.

#### Value

output.file A character string containing a path to the output file, a genotypic matrix in the

1fmm format.

#### Author(s)

Eric Frichot

write.env 55

#### See Also

vcf lfmm.data ancestrymap2lfmm ancestrymap2geno ped2lfmm ped2geno vcf2geno

#### **Examples**

```
# Creation of a file called "example.vcf"
# with 4 SNPs for 3 individuals.
data("example_vcf")
write.table(example_vcf,"example.vcf",col.names =
   c("#CHROM", "POS", "ID", "REF", "ALT", "QUAL", "FILTER", "INFO",
    "FORMAT", "SAMPLE0", "SAMPLE1", "SAMPLE2"),
    row.names = FALSE, quote = FALSE)
                from the vcf format ("example.vcf")
# Conversion
                to the lfmm format ("example.lfmm").
#
                the name of the output file is the same name
# By default,
                as the input file with a .lfmm extension.
# Create files: "example.lfmm",
                "example.vcfsnp" - SNP informations,
#
                "example.removed" - removed lines.
#
output = vcf2lfmm("example.vcf")
# Conversion
                from the vcf format ("example.vcf")
#
                to the lfmm format with the output file called "plop.lfmm".
# Create files: "plop.lfmm",
                "plop.vcfsnp" - SNP informations,
                "plop.removed" - removed lines.
output = vcf2lfmm("example.vcf", "plop.lfmm")
# As force = false and the file "example.lfmm" already exists,
# nothing happens.
output = vcf2lfmm("example.vcf", force = FALSE)
```

write.env

Write files in the env format

# Description

Write a file in the env format.

# Usage

```
write.env(R, output.file)
```

#### **Arguments**

R

A matrix containing the environmental variables with one line for each individual and one column for each environmental variable. The missing genotypes have to be encoded with the value 9.

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output.file A character string containing a path to the output file, an environmental data

matrix in the env formt.

Value

output.file A character string containing a path to the output file, an environmental data

matrix in the env formt.

# Author(s)

Eric Frichot

#### See Also

```
read.env env 1fmm
```

# **Examples**

```
# Creation of an environmental matrix C
# containing 2 environmental variables for 3 individuals.
# C contains one line for each individual and one column for each variable.
C = matrix(runif(6), ncol=2, nrow=3)

# Write C in a file called "tuto.env".
# Create file: "tuto.env".
write.env(C, "tuto.env")

# Read the file "tuto.env".
C = read.env("tuto.env")
```

write.geno

Write files in the geno format

# **Description**

Write a file in the geno format.

# Usage

```
write.geno(R, output.file)
```

#### **Arguments**

R A matrix containing the genotypes with one line for each individual and one

column for each SNP. The missing genotypes have to be encoded with the value

9.

output.file A character string containing a path to the output file, a genotypic matrix in the

geno format.

write.lfmm 57

# Value

output.file A character string containing a path to the output file, a genotypic matrix in the geno format.

# Author(s)

Eric Frichot

#### See Also

read.geno geno snmf geno21fmm lfmm2geno ancestrymap2geno ped2geno vcf2geno

# **Examples**

```
# Creation of a file called "genotypes.geno" in the working directory,
# with 1000 SNPs for 165 individuals.
data("tutorial")

# Write R in a file called "genotypes.geno".
# Create file: "genotypes.geno".
write.geno(tutorial.R, "genotypes.geno")

# Read the file "genotypes.geno".
R = read.geno("genotypes.geno")
```

write.lfmm

Write files in the 1fmm format

#### **Description**

Write a file in the 1fmm format.

#### Usage

```
write.lfmm(R, output.file)
```

# **Arguments**

R A matrix containing the genotypes with one line for each individual and one

column for each SNP. The missing genotypes have to be encoded with the value

9.

output.file A character string containing a path to the output file, a genotypic matrix in the

1fmm format.

#### Value

output.file A character string containing a path to the output file, a genotypic matrix in the

geno format.

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#### Author(s)

Eric Frichot

#### See Also

```
read.lfmm lfmm.data lfmm geno2lfmm lfmm2geno ancestrymap2lfmm ped2lfmm
```

#### **Examples**

```
# Creation of a file called "genotypes.geno" in the working directory,
# with 1000 SNPs for 165 individuals.
data("tutorial")

# write R in a file called "genotypes.lfmm"

# Create file: "genotypes.lfmm".
write.lfmm(tutorial.R, "genotypes.lfmm")

# read the file "genotypes.lfmm".
R = read.lfmm("genotypes.lfmm")
```

z.scores

z-scores from a lfmm run

# **Description**

Return the 1fmm output matrix of zscores for the chosen runs with K latent factors, the d-th variable and the all option. For an example, see 1fmm.

# Usage

```
z.scores (object, K, d, all, run)
```

# **Arguments**

object	A lfmmProject object.
K	The number of latent factors.
d	The d-th variable.
all	A Boolean option. If true, the run with all variables at the same time. If false, the runs with each variable separately.
run	A list of chosen runs.

#### Value

res A matrix containing a vector of z-scores for the chosen runs per column.

#### Author(s)

Eric Frichot

zscore.format 59

#### See Also

```
1fmm 1fmm.data
```

#### **Examples**

```
### Example of analyses using 1fmm ###
data("tutorial")
# creation of the genotype file, genotypes.lfmm.
# It contains 400 SNPs for 50 individuals.
write.lfmm(tutorial.R, "genotypes.lfmm")
# creation of the environment file, gradient.env.
# It contains 1 environmental variable for 40 individuals.
write.env(tutorial.C, "gradients.env")
#################
# runs of 1fmm #
#################
# main options, K: (the number of latent factors),
            CPU: the number of CPUs.
# Toy runs with K = 3 and 2 repetitions.
# around 15 seconds per run.
project = NULL
project = 1fmm("genotypes.1fmm", "gradients.env", K = 3, repetitions = 2,
    iterations = 6000, burnin = 3000, project = "new")
# get the z-scores for all runs for K = 3
z = z.scores(project, K = 3)
# get the z-scores for the 2nd run for K =3
z = z.scores(project, K = 3, run = 2)
# remove
remove.lfmmProject("genotypes_gradients.lfmmProject")
```

zscore.format

Output file format for 1fmm

#### **Description**

Description of the zscore output format of 1fmm.

#### **Details**

The zscore format has one row for each SNP. Each row contains three values: The first value is the zscore, the second value is the -log10(pvalue), the third value is the p-value (separated by spaces or tabulations).

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# Author(s)

Eric Frichot

# See Also

1fmm lfmm.data env

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