# Package 'MM4LMM'

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<b>Description</b> The main function MMEst() performs (Restricted) Maximum Likelihood in a variance component mixed models using a Min-Max (MM) algorithm (Laporte, F., Charcosset, A. & Mary-Huard, T. (2022) <doi:10.1371 journal.pcbi.1009659="">).</doi:10.1371>	
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MM4LMM-package

Min-Max algorithms for Variance Component Mixed Model Inference.

#### **Description**

This package provides a function to perform either ML or ReML estimation in a Variance Component Mixed Model. The optimization of the (possibly Restricted) log-likelihood is performed using the Min-Max algorithm described in Hunter et al. (2004). Depending on the number of variance components, different computational tricks are used to speed up inference. Additionally, the AnovaTest function provides Type I, Type III and Type III Kenward Roger approximation test series for fixed effects. The nullity of a specific linear combination can also be tested.

#### **Details**

Variance Component Mixed Models are mixed models of the form

$$Y = X\beta + \sum_{k=1}^{K} Z_k u_k$$

where Y is the response vector, X and  $\beta$  are respectively the incidence matrix and the coefficient vector associated with the fixed effects,  $u_k$  is the kth vector of random effects and corresponds to its associated incidence matrix. All random effect are assumed to be Gaussian with mean 0 and covariance  $\sigma_k^2 R_k$ , where  $R_k$  is a known correlation matrix and  $\sigma_k^2$  is an unknown variance parameter. All random effects are assumed to be independent. In many applications the last random component corresponds to the error and therefore both  $Z_k$  and  $R_k$  correspond to the identity matrix.

In such models the inference of both the unknown variance components  $\sigma_k^2,...,\sigma_K^2$  and the fixed effect  $\beta$  can be achieved through Maximum Likelihood (ML) or Restricted Maximum Likelihood (ReML) estimation. Neither ML nor ReML yield close form expressions of the estimates, consequently the maximization of the (restricted) log-likelihood has to be performed numerically. This package provides the user with Min-Max algorithms for the optimization. Efficient tricks such as profiling, MME trick and MM acceleration are implemented for computational efficiency (see Johnson et al. (1995), Varadhan et al. (2008) for details). The main function MMEst handles parallel inference of multiple models sharing the same set of correlation matrices - but possibly different fixed effects, an usual situation in GWAS analysis for instance.

## Author(s)

Fabien Laporte and Tristan Mary-Huard

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#### References

Laporte, F., Charcosset, A., & Mary-Huard, T. (2022). Efficient ReML inference in variance component mixed models using a Min-Max algorithm. PLOS Computational Biology, 18(1), e1009659. Johnson, D. L., & Thompson, R. (1995). Restricted maximum likelihood estimation of variance components for univariate animal models using sparse matrix techniques and average information.

Journal of dairy science, 78(2), 449-456.

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Hunter, D. R., & Lange, K. (2004). A tutorial on MM algorithms. The American Statistician, 58(1), 30-37.

Kenward, M. G., & Roger, J. H. (1997). Small sample inference for fixed effects from restricted maximum likelihood. Biometrics, 983-997.

Varadhan, R., & Roland, C. (2008). Simple and globally convergent methods for accelerating the convergence of any EM algorithm. Scandinavian Journal of Statistics, 35(2), 335-353.

Zhou, H., Hu, L., Zhou, J., & Lange, K. (2015). MM algorithms for variance components models. arXiv preprint arXiv:1509.07426.

AnovaTest

Type I and Type III Tests for mixed models.

## **Description**

This function computes Type I and Type III tests for each fixed effect of a model, as displayed by the MMEst function. Alternatively, a specific linear combination of the fixed parameters may be tested (at 0).

### Usage

```
AnovaTest(ResMMEst , TestedCombination=NULL , Type = "TypeIII" ,
    Cofactor = NULL , X = NULL , formula = NULL , VarList = NULL ,
    NbCores=1)
```

#### **Arguments**

ResMMEst A list as displayed by the MMEst function.

TestedCombination

A contrast matrix or a list of contrast matrices  $C_m$ . Each matrix corresponds to

a (set of) linear combination to be (jointly) tested at 0.

Type "TypeII" or "KR" (default is "TypeIII"). AnovaTest will compute tests

of the given type for each fixed effect in the model. The option is ignored if a TestedCombination is provided. If Type is "KR" then AnovaTest will compute Type III test using Kenward Roger approximation, see Kenward and Roger

(1997) for details.

Cofactor The incidence matrix corresponding to fixed effects common to all models used

in MMEst. If NULL, a single intercept is used as cofactor. This entry is needed

when Type is "KR".

X The incidence matrix or a list of incidence matrices corresponding to fixed ef-

fects specific to each model used in MMEst (default is NULL). This entry is needed

when Type is "KR".

formula The formula object specifying the fixed effect part of all models separated by

+ operators used in MMEst (default is NULL). This entry is needed when Type is

"KR".

VarList The list of correlation matrices associated with random and residual effects used

in MMEst (default is NULL). This entry is needed when Type is "KR".

NbCores The number of cores to be used.

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#### **Details**

If no TestedCombination is provided, the function performs either Type I or Type III tests for each fixed effect in the model (default is Type III). If TestedCombination is provided, each linear combination in TestedCombination is tested at 0 using a Wald test. No check is performed regarding the estimability of the linear combination to be tested. If the dimension of the combination does not match with the dimension of ResMMEst, a NA is returned.

## Value

The output of the function is a list with as many items as in the original input list ResMMEst. For each item of ResMMEst, a table is provided that contains the Wald test statistics, p-values and degrees of freedom for all tested hypotheses.

## Author(s)

F. Laporte and T. Mary-Huard

#### References

Kenward, M. G., & Roger, J. H. (1997). Small sample inference for fixed effects from restricted maximum likelihood. Biometrics, 983-997.

```
require('MM4LMM')
data(QTLDetectionExample)
Pheno <- QTLDetectionExample$Phenotype
Geno <- QTLDetectionExample$Genotype
Kinship <- QTLDetectionExample$Kinship

##Build the VarList object
VL <- list(Additive = Kinship , Error = diag(1,length(Pheno)))

##Perform inference
Result <- MMEst(Y=Pheno , X = Geno , VarList = VL)

##Compute tests
AOV <- AnovaTest(Result,Type="TypeI")

##Test specific contrast matrix
TC = matrix(c(0,1),nrow=1)
AOV2 <- AnovaTest(Result, TestedCombination = TC , Type="KR" , X = Geno , VarList = VL)</pre>
AOV3 <- AnovaTest(Result, TestedCombination = TC , Type="KR" , X = Geno , VarList = VL)
```

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MMBlup	BLUP from MM4LMM results

## **Description**

This function computes the BLUP for each random vector included in the MMEst output. Note that this function can be called only if the argument X of MMEst was set to NULL

## Usage

```
MMBlup(Y,Cofactor = NULL, X = NULL, fmla = NULL,ZList=NULL,VarList,ResMM)
```

## **Arguments**

Υ	The vector of response values used in the function MMEst.
Cofactor	The incidence matrix corresponding to fixed effects common to all models to be adjusted used in the function MMEst. If NULL, a vector full of 1 is used.
Χ	Must be NULL.
fmla	The formula object specifying the fixed effect part of all models separated by + operators used in the function MMEst (default is NULL).
ZList	The list of incidence matrices associated with random and residual effects used in the function MMEst (default is NULL).
VarList	The list of covariance matrices associated with random and residual effects used in the function MMEst.
ResMM	A list as displayed by the MMEst function for a Variance Component Analysis.

### Value

The function returns a list where each element corresponds to the Best Linear Unbiased Prediction of a random component of the model.

## Author(s)

**GQMS** CoreFunctions Team

```
require('MM4LMM')
data(VarianceComponentExample)
DataHybrid <- VarianceComponentExample$Data
KinF <- VarianceComponentExample$KinshipF
KinD <- VarianceComponentExample$KinshipD

##Build incidence matrix for each random effect
Zf <- t(sapply(as.character(DataHybrid$CodeFlint), function(x)
    as.numeric(rownames(KinF)==x)))</pre>
```

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MMEst

MM inference method for variance component mixed models

## **Description**

This is the main function of the *MM4LMM* package. It performs inference in a variance component mixed model using a Min-Max algorithm. Inference in multiple models (e.g. for GWAS analysis) can also be performed.

## Usage

```
MMEst(Y, Cofactor = NULL, X = NULL, formula=NULL, VarList, ZList = NULL, Method = "Reml",
Henderson=NULL, Init = NULL, CritVar = 0.001, CritLogLik = 0.001,
MaxIter = 100, NbCores = 1, Verbose = TRUE)
```

## Arguments

Υ	A vector of response values.
Cofactor	An incidence matrix corresponding to fixed effects common to all models to be adjusted. If NULL, a single intercept is used as cofactor.
X	An incidence matrix or a list of incidence matrices corresponding to fixed effects specific to each model. If X is a matrix, one model per column will be fitted. If X is a list, one model per element of the list will be fitted (default is NULL).
formula	A formula object specifying the fixed effect part of all models separated by + operators. To specify an interaction between Cofactor and X use the colnames of X when it is a list or use "Xeffect" when X is a matrix.
VarList	A list of covariance matrices associated with random and residual effects.
ZList	A list of incidence matrices associated with random and residual effects (default is NULL).
Method	The method used for inference. Available methods are "Reml" (Restricted Maximum Likelihood) and "ML" (Maximum Likelihood).

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Henderson If TRUE the Henderson trick is applied. If FALSE the Henderson trick is not

applied. If NULL the algorithm chooses wether to use the trick or not.

Init A vector of initial values for variance parameters (default is NULL)

CritVar Value of the criterion for the variance components to stop iteration. (see Details)

CritLogLik Value of the criterion for the log-likelihood to stop iteration. (see Details)

MaxIter Maximum number of iterations per model.

NbCores Number of cores to be used.

Verbose A boolean describing if messages have to be printed (TRUE) or not (FALSE).

Default is TRUE.

#### **Details**

If X is NULL, the following model is fitted:

$$Y = X_C \beta_C + \sum_{k=1}^K Z_k u_k$$

with  $X_C$  the matrix provided in Cofactor,  $\beta_C$  the unknown fixed effects,  $Z_k$  the incidence matrix provided for the kth component of ZList and  $u_k$  the kth vector of random effects. If ZList is unspecified, all incidence matrices are assumed to be the Identity matrix. Random effects are assumed to follow a Gaussian distribution with mean 0 and covariance matrix  $R_k \sigma_k^2$ , where  $R_k$  is the kth correlation matrix provided in VarList.

If X is not NULL, the following model is fitted for each i:

$$Y = X_C \beta_C + X_{[i]} \beta_{[i]} + \sum_{k=1}^{K} Z_k u_k$$

where  $X_{[i]}$  is the incidence matrix corresponding to the ith component (i.e. column if X is a matrix, element otherwise) of X, and  $\beta_{[i]}$  is the (unknow) fixed effect associated to  $X_{[i]}$ .

All models are fitted using the MM algorithm. If Henderson=TRUE, at each step the quantities required for updating the variance components are computed using the Mixed Model Equation (MME) trick. See Johnson et al. (1995) for details.

## Value

The result is a list where each element corresponds to a fitted model. Each element displays the following:

Beta Estimated values of  $\beta_C$  and  $\beta_i$  Sigma2 Estimated values of  $\sigma_1^2, ..., \sigma_K^2$  Variance matrix of Beta

LogLik (Method)

The value of the (restricted, if Method is "Reml") log-likelihood

NbIt The number of iterations required to reach the optimum

Method The method used for the inference

attr An integer vector with an entry for each element of Beta giving the term in

Factors which gave rise to this element (for internal use in the function AnovaTest)

Factors Names of each term in the formula

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#### Author(s)

F. Laporte and T. Mary-Huard

#### References

Laporte, F., Charcosset, A., & Mary-Huard, T. (2022). Efficient ReML inference in variance component mixed models using a Min-Max algorithm. PLOS Computational Biology, 18(1), e1009659.

Johnson, D. L., & Thompson, R. (1995). Restricted maximum likelihood estimation of variance components for univariate animal models using sparse matrix techniques and average information. Journal of dairy science, 78(2), 449-456.

Hunter, D. R., & Lange, K. (2004). A tutorial on MM algorithms. The American Statistician, 58(1), 30-37.

Zhou, H., Hu, L., Zhou, J., & Lange, K. (2015). MM algorithms for variance components models. arXiv preprint arXiv:1509.07426.

```
require('MM4LMM')
#### Example 1: variance component analysis, 1 model
data(VarianceComponentExample)
DataHybrid <- VarianceComponentExample$Data</pre>
KinF <- VarianceComponentExample$KinshipF</pre>
KinD <- VarianceComponentExample$KinshipD</pre>
##Build incidence matrix for each random effect
Zf <- t(sapply(as.character(DataHybrid$CodeFlint), function(x)</pre>
  as.numeric(rownames(KinF)==x)))
Zd <- t(sapply(as.character(DataHybrid$CodeDent), function(x)</pre>
  as.numeric(rownames(KinD)==x)))
##Build the VarList and ZList objects
VL = list(Flint=KinF , Dent=KinD , Error = diag(1,nrow(DataHybrid)))
ZL <- list(Flint=Zf , Dent=Zd , Error = diag(1,nrow(DataHybrid)))</pre>
##Perform inference
#A first way to call MMEst
ResultVA <- MMEst(Y=DataHybrid$Trait , Cofactor = matrix(DataHybrid$Trial)</pre>
                   , ZList = ZL , VarList = VL)
length(ResultVA)
print(ResultVA)
#A second way to call MMEst (same result)
Formula <- as.formula('~ Trial')</pre>
ResultVA2 <- MMEst(Y=DataHybrid$Trait , Cofactor = DataHybrid,</pre>
                    formula = Formula
                   , ZList = ZL , VarList = VL)
length(ResultVA2)
print(ResultVA2)
```

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```
#### Example 2: Marker Selection with interaction between Cofactor and X matrix
Formula <- as.formula('~ Trial+Xeffect+Xeffect:Trial')</pre>
ResultVA3 <- MMEst(Y=DataHybrid$Trait , Cofactor = DataHybrid,</pre>
                  X = VarianceComponentExample$Markers,
                   formula = Formula
                   , ZList = ZL , VarList = VL)
length(ResultVA3)
print(ResultVA3[[1]])
#### Example 3: QTL detection with two variance components
data(QTLDetectionExample)
Pheno <- QTLDetectionExample$Phenotype
Geno <- QTLDetectionExample$Genotype</pre>
Kinship <- QTLDetectionExample$Kinship</pre>
##Build the VarList object
VLgd <- list(Additive=Kinship , Error=diag(1,length(Pheno)))</pre>
##Perform inference
ResultGD <- MMEst(Y=Pheno , X=Geno</pre>
                   , VarList=VLgd , CritVar = 10e-5)
length(ResultGD)
print(ResultGD[[1]])
```

MMVcov

Covariance Matrix for variance estimators.

## Description

This function computes the covariance matrix of variance estimators using either the inverse of the Expected Hessian Matrix or the inverse of the Average Information Matrix.

## Usage

```
MMVcov(ResMM , Y , Cofactor = NULL , formula = NULL,
    ZList = NULL , VarList , information="Expected")
```

#### **Arguments**

ResMM A list as displayed by the MMEst function for a Variance Component Analysis

(only the first element of the list will be analyzed).

Y The vector of response values used in the function MMEst.

Cofactor The incidence matrix corresponding to fixed effects common to all models to be

adjusted used in the function MMEst. If NULL, a vector full of 1 is used.

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formula The formula object specifying the fixed effect part of all models separated by +

operators used in the function MMEst (default is NULL).

ZList The list of incidence matrices associated with random and residual effects used

in the function MMEst (default is NULL).

VarList The list of covariance matrices associated with random and residual effects used

in the function MMEst.

information A string specifying the method used to approximate the covariance matrix. It

can be either "Expected" (default) to use the Expected Hessian Matrix or "AI" to use the Average Information Matrix. The AI matrix is always computed using Reml estimates whereas the expected hessian matrix could also be used for ML

estimates.

## **Details**

If information is not specified then the algorithm computes the covariance matrix using the Expected matrix based on the inference method (Reml or ML) used in the first item of ResMM. If information is equal to "AI" then it computes the AI matrix to calculate the covariance matrix. Only the first item of ResMM is analyzed.

#### Value

The output of the function is a list:

vcov The covariance matrix of the variance estimators

SE The standard errors of the variance estimators (the square root of the covariance

matrix diagonal)

## Author(s)

F. Laporte and T. Mary-Huard

```
require('MM4LMM')
data(VarianceComponentExample)
DataHybrid <- VarianceComponentExample$Data
KinF <- VarianceComponentExample$KinshipF
KinD <- VarianceComponentExample$KinshipD

##Build incidence matrix for each random effect
Zf <- t(sapply(as.character(DataHybrid$CodeFlint), function(x)
as.numeric(rownames(KinF)==x)))
Zd <- t(sapply(as.character(DataHybrid$CodeDent), function(x)
as.numeric(rownames(KinD)==x)))

##Build the VarList and ZList objects
VL = list(Flint=KinF , Dent=KinD , Error = diag(1,nrow(DataHybrid)))
ZL <- list(Flint=Zf , Dent=Zd , Error = diag(1,nrow(DataHybrid)))
##Perform inference</pre>
```

QTLDetectionExample

QTLDetectionExample

QTL Detection Example

## **Description**

This corresponds to (a sample of) the dataset presented in Rincent et al. (2014).

## Usage

```
data("QTLDetectionExample")
```

#### **Format**

```
The format is: List of 3

Phenotype Named num [1:259]

Genotype int [1:259,1:10]

Kinship num [1:259,1:259]
```

## **Details**

The list contains three elements:

- Phenotype: a numeric vector containing phenotypes of individuals
- Genotype: a matrix containing the genotypes of indviduals over 10 biallelic markers
- Kinship: a matrix of simple relatedness coefficients between individuals

## Source

```
https://link.springer.com/article/10.1007/s00122-014-2379-7
```

## References

Rincent, R., Nicolas, S., Bouchet, S., Altmann, T., Brunel, D., Revilla, P., ... & Schipprack, W. (2014). Dent and Flint maize diversity panels reveal important genetic potential for increasing biomass production. Theoretical and applied genetics, 127(11), 2313-2331.

## **Examples**

```
data(QTLDetectionExample)
names(QTLDetectionExample)
## maybe str(QTLDetectionExample) ; plot(QTLDetectionExample) ...
```

VarianceComponentExample

Variance Component Example

## **Description**

This corresponds to (a sample of) the dataset presented in Giraud et al. (2017).

## Usage

```
data("VarianceComponentExample")
```

#### **Format**

```
The format is: List of 3

Data 'data.frame': 432 obs. of 5 variables

Trial a factor with 2 levels

CodeHybrid a factor with 177 levels

CodeDent a factor with 116 levels

CodeFlint a factor with 122 levels

Trait a numeric vector

KinshipD num [1:116,1:116]

KinshipF num [1:122,1:122]
```

## **Details**

The list contains three elements:

- Data: a data frame containing the information about hybrids (trials, hybrid names, dent parental lines, flint parental lines and phenotypes)
- KinshipD: a matrix of simple relatedness coefficients between dent lines
- KinshipF: a matrix of simple relatedness coefficients between flint lines

#### Source

doi:10.1534/genetics.117.300305

## References

Giraud, H., Bauland, C., Falque, M., Madur, D., Combes, V., Jamin, P., ... & Blanchard, P. (2017). Reciprocal Genetics: Identifying QTL for General and Specific Combining Abilities in Hybrids Between Multiparental Populations from Two Maize (Zea mays L.) Heterotic Groups. Genetics, 207(3), 1167-1180.

```
data(VarianceComponentExample)
names(VarianceComponentExample)
## maybe str(VarianceComponentExample); plot(VarianceComponentExample) ...
```

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