

User Manual for

m r MLM

multi-locus random-SNP-effect Mixed Linear Model tools for
genome-wide association study

(version 5.0.1)

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Disclaimer: While extensive testing has been performed by Yuan-Ming Zhang's Lab at Crop Information Center of College of Plant Science and Technology, Huazhong Agricultural University, the results are, in general, reliable, correct or appropriate. However, results are not guaranteed for any specific datasets. We strongly recommend that users validate the mrMLM results with other software packages, i.e., GEMMA, EMMAX, GAPIT v2 & PLINK.

Download website:

<https://cran.r-project.org/web/packages/mrMLM/index.html> or
<https://bigd.big.ac.cn/biocode/tools/BT007077>

Citation:

Method or software **References**

mrMLM	Wang et al. <i>Scientific Reports</i> 2016, 6:19444.
ISIS EM-BLASSO	Tamba et al. <i>PLoS Computational Biology</i> 2017, 13(1): e1005357.
pLARmEB	Zhang et al. <i>Heredity</i> 2017, 118: 517–524.
FASTmrEMMA	Wen et al. <i>Briefings in Bioinformatics</i> 2018, 19(4): 700–712.
pKWmEB	Ren et al. <i>Heredity</i> 2018, 120(3): 418–428.
FASTmrMLM	Tamba & Zhang, <i>bioRxiv</i> , 2018. doi: 10.1101/341784. Zhang et al. <i>Genomics, Proteomics & Bioinformatics</i> 2020, 18: 481–487.
Software mrMLM	Zhang et al. <i>Genomics, Proteomics & Bioinformatics</i> 2020, 18: 481–487.

Note: These references are listed in section of References.

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INTRODUCTION

1.1 Why mrMLM?

mrMLM (**m**ulti-locus **r**andom-SNP-effect **M**ixed **L**inear **M**odel) program is an R package for multi-locus genome-wide association studies (GWAS). At present this program (v5.0.1) includes six methods: 1) mrMLM, 2) FASTmrMLM (Fast multi-locus random-SNP-effect EMMA), 3) ISIS EM-BLASSO (Iterative Sure Independence Screening EM-Bayesian LASSO), 4) pLARmEB (polygenic-background-control- based least angle regression plus empirical Bayes), 5) pKWmEB (polygenic- background-control-based Kruskal-Wallis test plus empirical Bayes); and 6) fast mrMLM (FASTmrMLM).

Different from the previous versions, the software program of mrMLM v5.0.1 may run on personal computer such as desktop and laptop, which have smaller RAM than server. If there is (almost) homozygous association mapping population available, these methods are very effective. If there are many heterozygous marker genotypes in association mapping populations, the HIVmrMLM software is available.

mrMLM 5.0.1 works well on Windows, Linux (desktop) and MacOS.

1.2 Getting started

The software package mrMLM runs only in the R software environment and can be freely downloaded from the R website (<https://cran.r-project.org/web/packages/mrMLM.GUI/index.html>), or the BioCode (<https://bigd.big.ac.cn/biocode/tools/7077>), or requested from the maintainer, Dr Yuan-Ming Zhang at College of Plant Science and Technology, Huazhong Agri Univ (soy Zhang@mail.hzau.edu.cn).

Note: Users may need to install Rtools <https://cran.r-project.org/bin/windows/Rtools/> and add it into the system of PATH (Fig 1). The purpose is to ensure that the results can be written to the computer.

1.2.1 One-Click installation

Within R environment, the mrMLM software can be installed online using the below command:

```
install.packages("mrMLM")
```

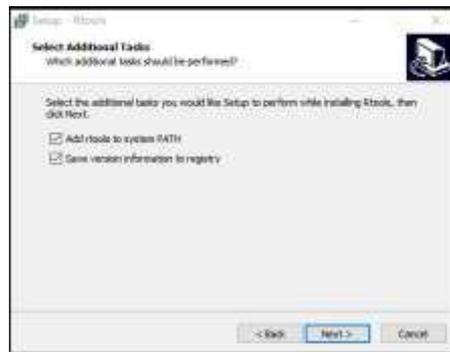


Figure 1. Install Rtools

1.2.2 Step-by-step installation

1.2.2.1 Install the add-on packages

Offline installation Users may download the below 26 packages from [CRAN](https://cran.r-project.org/) (<https://cran.r-project.org/>), [github](https://github.com/) (<https://github.com/>) and [google search](#).

BEDMatrix, coin, colorspace, data.table, doParallel, foreach, iterators, lars, later, libcoin, lpsolve, matrixStats, modeltools, multcomp, mvtnorm, ncvreg, pillar, Rcpp, RcppArmadillo, RcppEigen, sampling, sandwich, sbl, TH.data, tibble, zoo.

Under the R environment, then, users find “Packages”—“Install package(s) from local files...”, select all the above 26 packages, and install them offline.

1.2.2.2 Install mrMLM

Open R GUI, select "Packages"—"Install package(s) from local files..." and then find the mrMLM package which you have downloaded on your desktop.

User Manual file Users can decompress the mrMLM package and find the User Manual file (name: [Instruction.pdf](#)) in the folder of “../mrMLM/inst”.

1.2.3 Run mrMLM

Once the software mrMLM is installed, users may run it using two commands:

```
library("mrMLM")
```

```
mrMLM(***) (***: please see § 2.1.2 Example)
```

If users re-use the software mrMLM, users also use the above two commands.

2. Function

2.1 Function mrMLM()

2.1.1 Parameter settings

Parameter	Meaning	File format	Note
fileGen	File path & name in your computer, i.e., fileGen="D:/Users/Genotype_num.csv"	*.csv; *.txt (Genotypic values. Row : markers; Column : individuals). When PC=TRUE, the file format are PLINK binary files: *.bed; *.bim; *.fam.	Tables 1~3
filePhe	File path & name in your computer, i.e., filePhe="D:/Users/Phenotype.csv"	*.csv; *.txt (Phenotypic values. Row : individual; Column : traits)	Table 4
fileKin	File path & name in your computer, i.e., fileKin="D:/Users/Kinship.csv" or fileKin=NULL	*.csv; *.txt (Kinship matrix. Row & Column : individuals)	Table 5
filePS	File path & name in your computer, i.e., filePS="D:/Users/PopStr.csv" or filePS=NULL	*.csv; *.txt [Population structure. Row : individual; Column : sub-populations 1, 2, ..., k (No. of sub-populations)]	Table 6~8
PopStrType	Three types of population structures: Q (Q matrix), PCA (principal components), EvolPopStr (evolutionary population structure)		
fileCov	File path & name in your computer, i.e., fileCov="D:/Users/Covariate.csv" or fileCov=NULL	*.csv; *.txt (Covariate. Row : individual; Column : covariate 1, 2, ..., k (No. of covariate))	Table 9
Genformat	Format for genotypic codes: Num (number), Cha (character) & Hmp (Hapmap), i.e., Genformat="Num"		
method	Six multi-locus GWAS methods. Users may select one to six methods. For example, method=c("mrMLM", "FASTmrMLM", "FASTmrEMMA", "pLARMmEB", "pKWmEB", "ISIS EM-BLASSO")		
Likelihood	This parameter is only for FASTmrEMMA, including restricted maximum likelihood (REML) and maximum likelihood (ML). Likelihood="REML" or Likelihood="ML"		
trait	Traits analyzed from number 1 to number 2. For example, trait=1:3 indicates that users analyze the first to third traits.		
SearchRadius	This parameter is only for mrMLM and FASTmrMLM, indicating Search Radius in search of potentially associated QTN. SearchRadius=20 indicates that only one potentially associated QTN was selected within 20 kb.		
CriLOD	Critical LOD score for significant QTN. CriLOD=3 indicates that the critical LOD score for significant QTN is set at 3.0.		
SelectVariable	This parameter is only for pLARMmEB. SelectVariable=50 indicates that 50 potentially associated variables are selected from each chromosome. Users may change this number in real data analysis in order to obtain the best final results.		
Bootstrap	This parameter is only for pLARMmEB, including FALSE & TRUE. Bootstrap=FALSE indicates the analysis of only real dataset; Bootstrap=TRUE indicates the analysis of both real dataset and four resampling datasets.		
DrawPlot	This parameter is for all the six methods, including FALSE and TRUE. DrawPlot=FALSE indicates no figure output; DrawPlot=TRUE indicates the output of the Manhattan and QQ plots.		
Plotformat	This parameter is for all the figure files, including *.jpeg, *.png, *.tiff and *.pdf. Plotformat="jpeg" indicates the *.jpeg format of plot file.		
dir	Save path in your computer, i.e., "D:/Users"		
PC	This parameter is used to specify whether only small RAM device is available to run the mrMLM program, such as desktop or laptop. The default value is PC=FALSE. PC=TRUE indicates running the program on desktop or laptop with low RAM.		
RAM	This parameter is the RAM of your desktop or laptop. The default value is RAM=4. RAM=4 indicates the RAM of your device or maximum RAM you want to use is 4G.		

2.1.2 Example

The full codes

```
mrMLM(fileGen="D:/Users/Genotype_num.csv",filePhe="D:/Users/Phenotype.csv",fileKin=NULL,filePS=NULL
,PopStrType=NULL,fileCov=NULL,Genformat="Num",method=c("mrMLM","FASTmrMLM","FASTmrEMMA
","pLARmEB","pKWmEB","ISIS EM-BLASSO"),Likelihood="REML",trait=1:3,SearchRadius=20,CriLOD=3,
SelectVariable=50,Bootstrap=FALSE,DrawPlot=FALSE, Plotformat="jpeg", dir="D:/Users")
```

The reduced codes

```
mrMLM(fileGen="D:/Users/Genotype_num.csv", filePhe="D:/Users/Phenotype.csv", Genformat="Num",
method=c("mrMLM","FASTmrMLM","FASTmrEMMA","pLARmEB","pKWmEB","ISIS EM-BLASSO"),
trait=1:3, CriLOD=3, dir="D:/Users")
```

It should be noted that users must set "fileGen", "filePhe", "Genformat", "method", "trait", "CriLOD" and "dir", and the other eight parameters can be default in function, including PopStrType="Q"; Likelihood="REML" only for FASTmrEMMA; SearchRadius=20 only for mrMLM and FASTmrMLM; SelectVariable=50 & Bootstrap=FALSE only for pLARmEB; DrawPlot=TRUE; Plotformat= "jpeg".

The codes for lower RAM consumption

The full codes

```
mrMLM(fileGen="D:/Users/Genotype",filePhe="D:/Users/Phenotype.csv",fileKin=NULL,filePS=NULL,PopStrT
ype=NULL,fileCov=NULL,method=c("mrMLM","FASTmrMLM","FASTmrEMMA","pLARmEB","pKWmEB")
,Likelihood="REML",trait=1:3,SearchRadius=20,CriLOD=3,SelectVariable=50,Bootstrap=FALSE,DrawPlot=FA
LSE, Plotformat="jpeg", dir="D:/Users", PC=TRUE, RAM=4)
```

The reduced codes are as follows.

```
mrMLM(fileGen="D:/Users/Genotype",filePhe="D:/Users/Phenotype.csv",Genformat="Num",method=c("mrML
M","FASTmrMLM","FASTmrEMMA","pLARmEB","pKWmEB"), trait=1:3, CriLOD=3, dir="D:/Users",
PC=TRUE)
```

The codes for lower RAM consumption may run on personal computer such as desktop and laptop, which has smaller RAM than server. The parameter "PC" and "RAM" are added in the version 5.0.1 to realize this function. Please see 2.1.1 for details. Note that this technique of low RAM consumption is not suitable for "ISIS EM-BLASSO" algorithm.

2.2 Function MultiManhattan()

Users can use this function to draw or adjust the Manhattan plot according to their own needs.

2.2.1 Parameter settings

Parameter	Meaning
ResultIntermediate	Intermediate results obtained by the mrMLM software
ResultFinal	Final results obtained by the mrMLM software
mar	A numerical vector of the form c(bottom, left, top, right), which gives the number of lines of margin to be specified on the four sides of the plot, and the default is c(2.9, 2.8, 0.7, 2.8)
LabDistance	Distance between label and axis; the default is 1.5
ScaleDistance	Distance between scale values and axis; the default is 0.4
LabelSize	Size of all the three labels; the default is 0.8
ScaleSize	Size of scale values; the default is 0.7
AxisLwd	The width of axis, a positive number; the default is 5
TckLength	The length of tick marks; the default is -0.03
LogTimes	Magnification of $\{-\log_{10}(\text{P-value})\}$; the default is 2
LODTimes	Magnification of $\{\text{LOD score}\}$; the default is 1.2
lodline	The significant LOD score; the default is 3
dirplot	Path to save plot; the default is current working directory
PlotFormat	Format of the plot, <i>i.e.</i> , *.tiff, *.png, *.jpeg, *.pdf
width	Figure width; the default is 28000
height	Figure height; the default is 7000
pointsize	Word resolution, with the unit of 1/72 inch, being pixels per inch (ppi); the default is 60
res	Figure resolution, with the unit of pixels per inch (ppi); the default is 600
MarkGene	To mark genes in plot or not; if “TRUE” is selected, a file, namely “Reference information to mark gene.csv”, that contains the <i>x</i> and <i>y</i> axis information of all the significant QTNs will generate. The default is “FALSE”, indicating that no candidate or known gene names are marked in Manhattan plot.
Pos_x	The <i>x</i> axis positions of all the marked genes are input
Pos_y	The <i>y</i> axis positions of all the marked genes are input

GeneName	All the gene names are input
GeneNameColour	The color of gene names
...	Arguments passed to points, axis, text

2.2.2 Example

The full codes

```
MultiManhattan(ResultIntermediate="D:/Users/intermediate result.csv", ResultFinal="D:/Users/Final result.csv",
mar=c(2.9,2.8,0.7,2.8),LabDistance=1.5,ScaleDistance=0.4,LabelSize=0.8,ScaleSize=0.7,AxisLwd=5,TckLength=
-0.03,LogTimes=2,LODTimes=1.2,lodline=3,dirplot="D:/Users", PlotFormat=c("tiff","png","jpeg","pdf"), width=
28000, height=7000, pointsize = 60, res=600, MarkGene=TRUE, Pos_x=c(139,195), Pos_y=c(7.5,7),
GeneName=c("Gene1","Gene2"), GeneNameColour="blue")
```

The reduced codes

```
MultiManhattan(ResultIntermediate="D:/Users/intermediate result.csv", ResultFinal="D:/Users/Final result.csv")
```

It should be noted that users must set up two parameters "ResultIntermediate" and "ResultFinal", and the other twenty-one parameters may be default in this Function.

2.3 Dataset format

Numeric format for dataset “fileGen” (Table 1) The first column, named "**rs#**", stands for marker ID, i.e., “PZB00859.1”. The second column, named "**chrom**", stands for chromosome, i.e., numeric variable “1”. The third column, named "**pos**", stands for the position (bp) of SNP on the chromosome. The fourth column, named "**genotype for code 1**", indicates reference base for code variable $x = 1$. Among the remaining columns, each column lists all the genotypes for one individual, and the first row shows the individual names. For each marker, homozygous genotypes are expressed by 1 and -1, respectively, and the heterozygous and missing genotypes are indicated by zero. If the base for the first individual is missing, the base firstly observed in this row is what we list. Note that the genotype with code **1** will be also appeared in the **Result** files.

Table 1. The numeric format of the genotypic dataset

rs#	chrom	pos	genotype for code 1	33-16	Nov-38	A4226	A4722
PZB00859.1	1	157104	C	1	1	1	1

PZA01271.1	1	1947984	C	1	-1	1	-1
PZA03613.2	1	2914066	G	1	1	1	1
PZA03613.1	1	2914171	T	1	1	1	1
PZA03614.2	1	2915078	G	1	1	1	1
PZA03614.1	1	2915242	T	1	1	1	1
PZA02117.1	1	223466480	A	1	1	1	-1
PZA00403.5	1	223466873	T	1	1	1	0
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮

Character format for dataset “fileGen” (Table 2) The first three columns are same as those in Table 1. The differences are that the marker values are character, such as **A, T, C, G** and **N**, and the other notations are heterozygous genotypes. The “**N**” indicates the missing of genotypes. The first rows from the fourth to last columns are individual name.

Table 2. The character format of the genotypic dataset

rs#	chrom	pos	33-16	Nov-38	A4226	A4722
PZB00859.1	1	157104	C	C	C	C
PZA01271.1	1	1947984	C	G	C	G
PZA03613.2	1	2914066	G	G	G	G
PZA03613.1	1	2914171	T	T	T	T
⋮	⋮	⋮	⋮	⋮	⋮	⋮

Hapmap format for dataset “fileGen” (Table 3) Please see the TASSEL software in details. Here we introduce simply. The first eleven columns describe the specific information of markers and individuals, and these column names must be “rs#”, “alleles”, “chrom”, “pos”, “strand”, “assembly#”, “center”, “protLSID”, “assayLSID”, “panelLSID” and “QCcode”. In the “rs#” (1st), “chrom” (3rd) and “pos” (4th) columns, the information has been described as the above. The values of marker genotypes should be character, such as **AA, TT, CC, GG, NN, AC** and **AG**, where the “**NN**” indicates the missing or unknown of genotypes. In the 2nd and 5th to 11th columns, “**NA**” indicates **no information** available. All the individual genotypic information will be showed from the 12th to last columns. In each column, individual name is listed in the first row, i.e., “33-16”, and the others are the genotypes (character).

Table 3. The hapmap format of the genotypic dataset

rs#	alleles	chrom	pos	strand	assembly#	center	protLSID	assayLSID	panelSID	QCcode	33-16	...
PZB00859.1	A/C	1	157104	+	AGPv1	Panzea	NA	NA	maize282	NA	CC	...
PZA01271.1	C/G	1	1947984	+	AGPv1	Panzea	NA	NA	maize282	NA	CC	...
PZA03613.2	G/T	1	2914066	+	AGPv1	Panzea	NA	NA	maize282	NA	GG	...
PZA03613.1	A/T	1	2914171	+	AGPv1	Panzea	NA	NA	maize282	NA	TT	...
PZA03614.2	A/G	1	2915078	+	AGPv1	Panzea	NA	NA	maize282	NA	GG	...
PZA03614.1	A/T	1	2915242	+	AGPv1	Panzea	NA	NA	maize282	NA	TT	...
PZA02117.1	A/G	1	223466480	+	AGPv1	Panzea	NA	NA	maize282	NA	AA	...
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	...

Before implementing GWAS, the above character genotypes should be transferred into numeric information, here the homozygous genotype of each marker for the first individual is transferred into 1, another homozygous genotype for this marker is transferred into -1, and heterozygous and missing genotypes are transferred into zero. If the base for the first individual is missing, the base firstly observed in this row is what we list.

PLINK binary format for dataset “fileGen” Please see the PLINK software and the sample data in “mrMLM/inst/extdata” for details. Here we introduce simply. PLINK binary files consist of three files with a same name: 1) ***.bed** contains the genotypic information of individuals, 2) ***.bim** contains the information of genetic marker, and 3) ***.fam** contains the information of individuals. The following provides two ways to convert hapmap to *.bed/*.bim/*.fam.

Under linux system.

Firstly, if the suffix of your hapmap file is not “.hmp.txt” or it cannot be handled by TASSEL, you should transform it to “*.hmp.txt” by using the following codes on R environment.

```
Genotype <- read.csv("/home/data/Genotype_Hmp.csv",header=FALSE)
write.table(Genotype,"/home/data/Genotype.hmp.txt",quot=FALSE,row.name=F,col.name=F,sep="\t")
```

Secondly, quit R environment and set a path to the location where original and converted datasets are stored, e.g.,

```
cd /home/data
```

Then, use TASSEL to sort the hapmap file by running:

```
/home/tassel-5-standalone/run_pipeline.pl -SortGenotypeFilePlugin -inputFile  
Genotype.hmp.txt -outputFile Genotype.sort.hmp.txt -fileType Hapmap
```

Next, use TASSEL to transform *.hmp.txt to *.vcf by running:

```
/home/tassel-5-standalone/run_pipeline.pl -fork1 -h Genotype.sort.hmp.txt -Xmx50g  
-export -exportType VCF
```

Finally, use PLINK to transform *.vcf to plink binary files (*.bed/*.bim/*.fam) by running:

```
/home/plink/plink --vcf Genotype.vcf --make-bed --out Genotype
```

Under windows system

Firstly, please install the Java, TASSEL and PLINK software packages with Windows versions as Figure 2. Note that TASSEL and Java should be installed, while PLINK should be downloaded, decompressed, and used (it is unnecessary to install). These software packages can be downloaded from

Java: <https://www.oracle.com/java/technologies/downloads/#jdk17-windows>

TASSEL: <https://www.maizegenetics.net/tassel/>

PLINK: <https://www.cog-genomics.org/plink/1.9/>

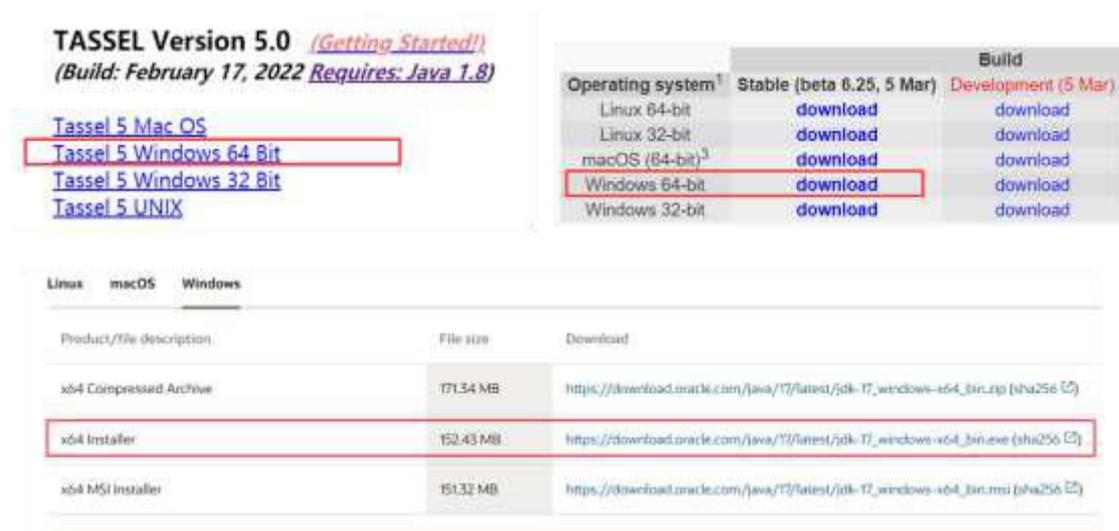


Figure 2. The installation package of TASSEL (top left), plink (top right) and Java(bottom).

Secondly, if the suffix of your hapmap file is not “.hmp.txt” or it cannot handle by TASSEL, you should transform it to “.hmp.txt” by using the following codes on R environment.

```
Genotype <- read.csv("E:/Genotype_Hmp.csv",header=FALSE)
write.table(Genotype,"E:/Genotype.hmp.txt",quot=FALSE,row.name=F,col.name=F,
sep="\t")
```

Thirdly, use TASSEL to sort the hapmap file. Open Windows PowerShell on your computer and run the following two code:

```
cd E:\TASSEL5
./run_pipeline -SortGenotypeFilePlugin -inputFile E:\Genotype.hmp.txt -outputFile
E:\Genotype.sort.hmp.txt -fileType Hapmap
```

Here “E:\TASSEL5” indicates the path to install TASSEL, while “E:\Genotype.hmp.txt” indicates the path and file name of your hapmap data which need to be sort.

Next, use TASSEL to transform *.hmp.txt to *.vcf. Open Windows PowerShell on your computer and run the following two codes:

```
cd E:\TASSEL5
./run_pipeline -fork1 -h E:\Genotype.sort.hmp.txt -Xmx10g -export -exportType VCF
```

Here “E:\TASSEL5” indicates the path to install TASSEL, while “E:\Genotype.sort.hmp.txt” indicates the path and file name of your hapmap data which need to be transformed. “-Xmx10g” is a parameter setup, indicating the maximum memory, 10 G, in the use of the TASSEL software. It should be less than the RAM of your device and larger than the memory in the process of data.

Finally, use PLINK to transform *.vcf to plink binary files (*.bed/*.bim/*.fam). Open Windows PowerShell on your computer and run the following two codes:

```
cd E:\plink_win64_20220305
./plink --vcf E:\TASSEL5\Genotype.vcf --make-bed --out E:\Genotype
```

Here “E:\TASSEL5\Genotype.vcf” indicates the path and file name of *.vcf file generated in previous step, while “E:\Genotype” indicates the path and file name of plink binary files which are the input data of mrMLM 5.0.1.

Format for the dataset “filePhe” (Table 4) The **Phenotypic** file should be a file with *.csv or *.txt format. The first column lists individual ID, i.e., “B46”, and “<Phenotype>” should be showed in the first row. Among the other columns, each column lists all the observations for the trait, its trait name is showed in the first row,

i.e., “trait1”, and phenotypic values are in the corresponding rows of their individuals. "NA" indicates the missing or unknown phenotypes.

Table 4. The format of Phenotypic dataset

<Phenotype>	trait1	trait2	trait3
B46	42	43.02	44.32
B52	72.5	71.88	72.8
B57	41	41.7	41.42
B64	74.5	74.43	74.5
⋮	⋮	⋮	⋮

The format for dataset “fileKin” (Table 5) The Kinship file should be a file with *.csv or *.txt format. In the first column in Table 5, “263” is sample size (n), and “33-16”, “Nov-38” and “A4226” are individual ID. Note that “ n ” is the number of common individuals between the phenotypic and genotypic datasets. All the kinship coefficients are listed as an $n \times n$ matrix.

`fileKin=NULL` indicates that the Kinship matrix is calculated by the software mrMLM. Here only the above n individuals are used to calculate the Kinship matrix. `fileKin="D:/Users/Kinship.csv"` means that the K matrix with name `Kinship.csv` is uploaded from the folder "D:/Users". If the number and order of individuals in `Kinship.csv` are not consistent with those of the above n individuals, our software may match the K matrix in order that the number and order of the transferred K matrix are consistent with those in the above n individuals.

Q matrix format for dataset “filePS” (Table 6) The Q matrix dataset in Table 6 consists of a $(n+2) \times (k+1)$ matrix, where n is the number of the above common individuals and k is the number of sub-populations. In the first column, “<PopStr>” and “<ID>” should present in the first and second rows, respectively; “33-16”, “Nov-38” and “A4226” are individual ID. In the 2nd to $(k+1)$ -th columns, “ Q_1 ” to “ Q_k ” indicate sub-populations. In the third row, “0.014”, “0.972” and “0.014” are the posterior probabilities of the “33-16” individual in the 1st, 2nd and 3rd subpopulations, respectively. When the Q matrix is uploaded to the software, [the software will automatically delete the column whose sum is the smallest.](#)

Table 5. The format of the Kinship dataset

263					
33-16	1.00809	0.45954	0.50677	0.42503	0.45591
Nov-38	0.45954	1.03352	0.43048	0.47044	0.39597
A4226	0.50677	0.43048	1.01717	0.45409	0.43775
A4722	0.42503	0.47044	0.45409	0.89002	0.34874
A188	0.45591	0.39597	0.43775	0.34874	1.0099
A214N	0.34693	0.33421	0.39779	0.29244	0.33058
A239	0.43593	0.46499	0.40323	0.36691	0.39597
A272	0.34874	0.40505	0.31423	0.3887	0.44138
A441-5	0.47952	0.44138	0.47226	0.47952	0.49224
A554	0.39779	0.45954	0.5431	0.48679	0.4214
⋮	⋮	⋮	⋮	⋮	⋮

Table 6. The format of the filePS dataset

<PopStr>			
<ID>	Q1	Q2	Q3
33-16	0.014	0.972	0.014
Nov-38	0.003	0.993	0.004
A4226	0.071	0.917	0.012
A4722	0.035	0.854	0.111
A188	0.013	0.982	0.005
A214N	0.762	0.017	0.221
A239	0.035	0.963	0.002
A272	0.019	0.122	0.859
⋮	⋮	⋮	⋮

Principal components format for dataset “filePS” (Table 7) The principal component dataset in Table 7 consists of a $(n+2) \times (k+1)$ matrix, where n is the number of the common individuals and k is the number of principal components. In the first column, “<PCA>” and “<ID>” should present in the first and second rows, respectively; “33-16”, “Nov-38” and “A4226” are individual ID. In the 2nd to $(k+1)$ -th columns, “PC₁” to “PC_k” indicate the first to k -th principal components. In the second column, “0.306”, ..., “0.216” are the scores of the first principal component for the 1st to 9-th individuals, respectively. [Note that the software doesn't delete any principle components.](#)

Table 7. The dataset format of principal components

<PCA>			
<ID>	PC1	PC2	PC3
33-16	0.306	0.029	0.226
Nov-38	-0.708	-2.071	1.413
A4226	-2.330	0.116	-0.824
A4722	1.059	0.470	-1.315
A188	-2.376	1.087	-0.135
A214N	-2.346	0.516	0.666
A239	-0.099	-0.318	-0.473
A272	-0.053	0.093	-0.275
A441-5	0.216	-0.535	-0.159
⋮	⋮	⋮	⋮

Evolutionary population structure format for dataset “filePS” (Table 8) The evolutionary population structure dataset in Table 8 consists of a $(n+2) \times 2$ matrix, where n is the number of the common individuals. In the first column, “<EvolPopStr>” and “<ID>” should present in the first and second rows, respectively; “33-16”, “Nov-38” and “A4226” are individual ID. In the second column, “EvolType” indicates the evolutionary type, i.e., the evolutionary types for individuals “33-16” and “A4722” are “A” and “B”, respectively, such as wild (A), landrace (B), and bred (C) soybeans.

Table 8. The dataset format of evolutionary population structure

<EvolPopStr>	
<ID>	EvolType
33-16	A
A4722	B
A188	A
A239	B
⋮	⋮

filePS=NULL indicates no inclusion of population structure in the genetic model. filePS="D:/Users/PopStr.csv" means that population structure dataset with name PopStr.csv is uploaded from the folder “D:/Users”. If the number and order of individuals in PopStr.csv aren’t consistent with those of the above common individuals, our software may match the population structure matrix in order that the

number and order of new matrix are consistent with those in the above common individuals.

The format for dataset “fileCov” (Table 9) The “Covariate” dataset consists of the $(n+2) \times (k+1)$ matrix, where n is the number of the common individuals and k is the number of covariates. In the first column, “<Covariate>” and “<ID>” should present in the first and second rows, respectively. If covariate is categorical, it should be named as Cate_covariate*. If covariate is continuous, it should be named as Con_covariate* (Table 9).

fileCov=NULL indicates no inclusion of covariates in the genetic model. fileCov="D:/Users/covariate.csv" means that the covariates with name covariate.csv are uploaded from the folder “D:/Users”. If the number and order of individuals in the uploaded file are not consistent with those in the above common individuals, our software need to change the number and order of individuals in order to match the above datasets.

Table 9. The format of the fileCov dataset

<Covariate>				
<ID>	Cate_covariate1	Cate_covariate2	Con_covariate1	Con_covariate2
33-16	A	C	349.5	374
Nov-38	B	C	205	452
A4226	A	D	300	374
A4722	A	D	190	452
A188	B	C	213	374
⋮	⋮	⋮	⋮	⋮

2.4 Result

Once the running of the software mrMLM v5.0.1 is ended, the “results” files will appear on the Directory, which was set up by users before running the software. The results for each trait include “*_intermediate result.csv”, “*_Final result.csv”, and Manhattan and QQ plots.

In the *_intermediate result.csv file, there are thirteen columns, including Trait ID, Trait name, reference sequence number (rs#, marker name), chromosome, marker's

position (bp) on the chromosome, SNP effect (γ_k , Effect) (mrMLM, FASTmrMLM, and FASTmrEMMA), $-\log_{10}(P)$ (mrMLM, FASTmrMLM, FASTmrEMMA, and pKWmEB), and genotype for code 1.

In the **Final result** file, there are fourteen columns, including Trait ID, Trait name, method, reference sequence number (rs#, marker names), chromosome, marker's position (bp) in the chromosome, QTN effect, LOD score, $-\log_{10}(P)$, the proportion of phenotypic variance explained by significant QTN (r^2), minor allelic frequency, genotype for code 1, residual error variance, and total phenotypic variance.

In the Manhattan plot, each marker $-\log_{10}(P)$ median among the $-\log_{10}(P)$ values from the mrMLM, FASTmrMLM, FASTmrEMMA and pKWmEB approaches is used to draw the Manhattan plots. **If users do not select one of the above four approaches, the software program does not produce the Manhattan plot.** In the Manhattan plot, these dots are indicated by light colors. All the QTNs commonly identified by multiple approaches are indicated by the pink dots that are shown above dotted vertical lines, while all the QTNs identified by one single approach are indicated by the light color dots that are also shown above dotted vertical lines (Fig 3).

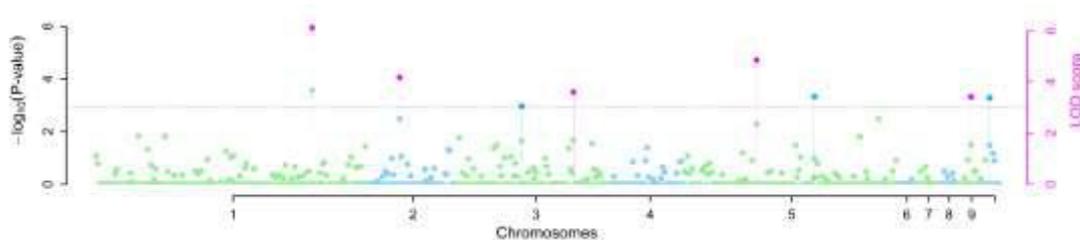


Figure 3. Manhattan plot

The setups for the resolution of the Manhattan plot are default. If users select the format of *.pdf, the **Figure width** is 16 [with the unit of inches], **Figure height** is 4 [with the unit of inches], and **Word resolution** is 20 [with the unit of 1/72 inch, ppi]. If users select the other three format, **Figure width** is 28000, **Figure height** is 7000 [with the unit of pixel (px)], **word resolution** is 60 [with the unit of 1/72 inch, being pixels per inch (ppi)], and **Figure resolution** is 600 [with the unit of pixels per inch (ppi)].

Using the P-values in Figure 3, it is easy to draw the QQ plot (Fig 4). **If users do not select one of the above four approaches, the software program does not produce the QQ plot.** The setups for the resolution of the QQ plot are default. If users select the format of *.pdf, the **Figure width** is 7 [with the unit of inches], **Figure height** is 7 [with the unit of inches], and **Word resolution** is 25 [with the unit of 1/72 inch, ppi]. If users select the other three format, **Figure width** is 10000, **Figure height** is 10000 [with the unit of pixel (px)], **word resolution** is 60 [with the unit of 1/72 inch, being pixels per inch (ppi)], and **Figure resolution** is 600 [with the unit of pixels per inch (ppi)].

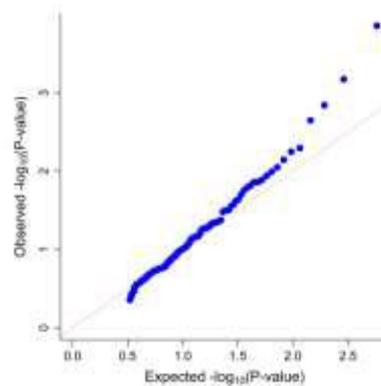


Figure 4. QQ plot

3. References

1. Zhang Yuan-Ming, Mao Yongcai, Xie Chongqing, Howie Smith, Luo Lang, Xu Shizhong. Mapping quantitative trait loci using naturally occurring genetic variance among commercial inbred lines of maize (*Zea mays* L.). *Genetics* 2005, **169**: 2267–2275.
2. Wang Shi-Bo, Feng Jian-Ying, Ren Wen-Long, Huang Bo, Zhou Ling, Wen Yang-Jun, Zhang Jin, Jim M. Dunwell, Xu Shizhong*, Zhang Yuan-Ming*. Improving power and accuracy of genome-wide association studies via a multi-locus mixed linear model methodology. *Scientific Reports* 2016, **6**: 19444.
3. Tamba Cox Lwaka, Ni Yuan-Li, Zhang Yuan-Ming*. Iterative sure independence screening EM-Bayesian LASSO algorithm for multi-locus genome-wide association studies. *PLoS Computational Biology* 2017, **13**(1): e1005357.
4. Zhang Jin#, Feng Jian-Ying#, Ni Yuan-Li, Wen Yang-Jun, Niu Yuan, Tamba Cox Lwaka, Yue Chao, Song Qijian, Zhang Yuan-Ming*. pLARmEB: Integration of least angle regression with empirical Bayes for multi-locus genome-wide association studies. *Heredity* 2017, **118**: 517–524.
5. Ren Wen-Long#, Wen Yang-Jun#, Jim M. Dunwell, Zhang Yuan-Ming*. pKWmEB: Integration of Kruskal-Wallis test with empirical Bayes under polygenic background control for multi-locus genome-wide

- association study. *Heredity* 2018, **120**: 208–218.
6. Wen Yang-Jun, Zhang Hanwen, Ni Yuan-Li, Huang Bo, Zhang Jin, Feng Jian-Ying, Wang Shi-Bo, Jim M. Dunwell, Zhang Yuan-Ming*, Wu Rongling*. Methodological implementation of mixed linear models in multi-locus genome-wide association studies. *Briefings in Bioinformatics* 2018, **19**(4): 700–712.
 7. Tamba CL, Zhang YM. A fast mrMLM algorithm for multi-locus genome-wide association studies. *bioRxiv*, 2018. doi: 10.1101/341784.
 8. Zhang Yuan-Ming*, Jia Zhenyu, Jim M. Dunwell. Editorial: The applications of new multi-locus GWAS methodologies in the genetic dissection of complex traits. *Frontiers in Plant Science* 2019, **10**: 100.
 9. Zhang Ya-Wen, Tamba Cox Lwaka, Wen Yang-Jun, Li Pei, Ren Wen-Long, Ni Yuan-Li, Gao Jun, Zhang Yuan-Ming*. mrMLM v4.0: An R platform for multi-locus genome-wide association studies. *Genomics, Proteomics & Bioinformatics* 2020, **18**: 481–487.
 10. Li M, Zhang YW, Zhang ZC, Xiang Y, Liu MH, Zhou YH, Zuo JF, Zhang HQ, Chen Y, Zhang YM. A compressed variance component mixed model for detecting QTNs, and QTN-by-environment and QTN-by-QTN interactions in genome-wide association studies. *Molecular Plant* 2022, online, S1674-2052(22)00060-0. doi: 10.1016/j.molp.2022.02.012.