A short manual for LFMM (command-line version)

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Please, print this reference manual only if it is necessary.

This short manual aims to help users to run LFMM command-line engine on Mac and Linux.

1 Description

We proposed an integrated framework based on population genetics, ecological modeling and machine learning techniques for screening genomes for signatures of local adaptation. We implemented fast algorithms using a hierarchical Bayesian mixed model based on a variant of principal component analysis in which residual population structure is introduced via unobserved factors. These algorithms can detect correlations between environmental and genetic variation at the same time as they infer the background levels of population structure. A description of the method is available in our paper:

Eric Frichot, Sean Schoville, Guillaume Bouchard, Olivier François, 2013. Landscape genomic tests for associations between loci and environmental gradients Molecular Biology and Evolution, 30 (7), 1687-1699.

2 Installation

We provide a set of R and perl scripts and C programs to convert to LFMM format and to display manhattan plot. By consequence, R and perl are mandatory to display manhattan plot.

To install LFMM CL version, you just have to execute the install script (install.sh) in LFMM main directory. To execute it in a terminal shell, go to LFMM main directory and write "./install.sh". If the script is not executable, type "chmod +x install.sh" and then "./install.sh". A binary called LFMM should be created in LFMM main directory.

3 Data format

Input files are composed of two files: a genotype file in the lfmm format and a variable file in the env format.

The **lfmm** format for the **genotype file** has one row for each individual. Each row contains one value per SNP (separated by spaces or tabulations): the number of alleles. The number of alleles can be the number of reference alleles or the number of derived alleles as long as it is the same choice for an entire SNP. The missing genotypes are encoded with the value -9 or 9. Here is an example with 3 individuals and 4 SNPs:



The **env** format for the **variable file** has one row for each individual. Each row contains one value per environmental variable (separated by spaces or tabulations). Below, an example of variable file for n=3 individuals and D=1 covariable. Warning: If you set several covariables, the program can be launched for each covariable sequentially or all variables together (see command-line options).

0.252477		
0.216618		
-0.47509		

The **zscore** format for the **output** file has one row for each SNP. Each row contains three values: The first value is the zscore, the second value is the $-\log 10$ (pvalue), the third value is the p-value, the forth value is the $-\log 10$ (qvalue) (based on Benjaminy-Hochberg approach) and the fifth value is the qvalue (separated by spaces or tabulations). Below, an example of output file for L=4 loci.

```
0.698819 0.314558 0.484665 0.0676981 0.855661
1.35961 0.759568 0.173953 0.224253 0.596688
0.771135 0.355929 0.440627 0.0795791 0.83257
0.879092 0.420959 0.379351 0.100315 0.793752
```

4 Run the programs

The program is executed by a command line. The format is:

```
./bin/LFMM -x genotype_file -v variable_file -K latent_factors_number
```

All the previous options are mandatory. There is no order for the options in the command line. Here is a more precise description of the options:

- -x genotype_file is the path for the genotype file (in lfmm format).
- -v variable_file is the path for the variable file (in env format).
- -K latent_factor_number is the number of latent factors. Several methods to choose K are described in the tutorial. Check the references for more informations.

Other options are not mandatory:

- -d nd fit LFMM with the nd-th variable only from the variable file. By default (if NULL and all is FALSE), fit LFMM with each variable from the variable file sequentially and independently.
- -a is a boolean option. If true, fit LFMM with all variables from the variable file at the same time. This option is not compatible with d option (by default: FALSE).
- -o output_file is the base of the path for the zscore output files (in zscore format). There is one output file per environmental variable. By default, the base of the output file is the base of the name of the input file. Then it is completed with the number of the covariable (with a "a" for all and a "s" for sequentially), the number of latent factors and zscore extension. For example, for the 3rd variable sequentially and 3 latent factors, the output file is by default input_file_s3.3.zscore.
- -m is a boolean option. If true, the input file contains missing genotypes. By default the program assumes that there is no missing data. The program is a bit slower with missing data. Please, do not use this option if it is not necessary.
- -p p is the number of processes (CPU) that you choose to use if you run the algorithm in parrallel. Be careful, the number of processes has to be lower or equal than the number of physical processes available on your computer. By default, the number of processes is 1.
- -i iteration_number is the number of iterations in the Gibbs Sampling algorithm. This number should be large enough (by default: 1000).
- -b burnin_number is the number of burning iterations in the Gibbs Sampling algorithm (by default 100).
- -s seed is the seed to initialize the randomization. By default, the seed is randomly chosen.
- -C dev_file is the path for the file containing the DIC and the deviance criterion. By default, the name of the DIC file is the name of the input file with .dic extension.

If you need a summary of the options, you can use the -h option by typing the command line

```
./LFMM -h
```

A full example is available at the end of this note.

5 How to choose K, the number of latent factors

The number of latent factors is the number of principal components (or latent factors) that is required to describe the neutral structure of the data. Several values should be tested. A too small value of K leads to liberal tests and may generate False Positive results. A too large value of K leads to conservative tests and may generate False Negative results. We advise you to be really careful in the choice of K and to test several values of K.

5.1 based on a population structure analysis

with a Principal Component Analysis approach In our paper, we used the number of significative principal components in the Tracy-Widom test of SmartPCA (http://helix.nih.gov/Applications/eigensoft.html) [3]. This heuristic may be a bit too conservative.

with on a clustering approach We also used the Bayesian clustering programs STRUCTURE

(http://pritch.bsd.uchicago.edu/software/structure2_1.html) [5], TESS

(http://membres-timc.imag.fr/Olivier.Francois/tess.html) [1,2] and sNMF (a fast program to estimate ancestry coefficients)

(http://membres-timc.imag.fr/Olivier.Francoissnmff.html) to find K the number of components which could better describe our simulated data.

5.2 based on p-values distribution

coming soon...

6 variance

coming soon an explanation of the percentage of variance explained by each variable.

7 Annex programs

We provide a set of R and perl scripts and C programs convert to LFMM format and to plot manhattan plot.

7.1 Data Format

Input files are composed of two mandatory files (a genotype file and an environmental variable file) and one optional file (the snp information file). The snp file is interesting to analyze zscore results and display results with manhattan plots. It is not necessary to provide information about individuals. All data formats are described with the same example. These files are available in examples/format_example/.

7.1.1 Genotype Data

• lfmm (example.lfmm)

The **lfmm** format has one row for each individual. Each row contains one value per SNP (separated by spaces or tabulations): the number of alleles. The number of alleles can be the number of reference alleles or the number of derived alleles as long as it is the same choice for an entire SNP. The missing genotypes are encoded with the value -9 or 9.

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

• ped (example.ped)

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: http://pngu.mgh.harvard.edu/ purcell/plink/data.shtml.

```
1 SAMPLEO 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
```

• ancestrymap (example.ancestrymap)

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. It is assumed that the genotypes for a given SNP name are written in consecutive lines. It is also assumed that the genotypes for a set of individuals are given in the same order as lines. The number of alleles can be the number of reference alleles or the number of derived alleles as long as it is the same choice for an entire SNP. It is assumed that the missing genotypes are encoded with the value 9.

```
rs0000
                     SAMPLEO
                     SAMPLE1
rs0000
rs0000
                     SAMPLE2
                                2
rs1111
                     SAMPLEO
                                0
rs1111
                     SAMPLE1
                                1
rs1111
                     SAMPLE2
                                0
rs2222
                     SAMPLEO
                                0
                     SAMPLE1
rs2222
                                9
rs2222
                     SAMPLE2
                                1
rs3333
                     SAMPLEO
                     SAMPLE1
rs3333
                                2
rs3333
                     SAMPLE2
```

• geno (example.geno)

The **geno** format has one row for each SNP. Each row contains 1 character per 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.

```
112
010
091
121
```

• vcf (example.vcf)

The \mathbf{vcf} The \mathbf{vcf} format is described here

http://www.1000genomes.org/wiki/Analysis/Variant%20Call%20Format/vcf-variant-call-format-version-41

```
##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">
##INFO=<ID=SM, Number=1, Type=Integer, Description="SampleVariant meta">
##CHROM POS ID REF ALT QUAL FILTER INFO FORMAT SAMPLEO 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 . 1/0:13

1 1005 rs3333 G A 999 . VM=3; SM=102 GT:GM 0/0:11 ./:12 0/1:13
```

Tips: As LFMM does not model allele frequencies, genotype file can be the number of copy of either the reference allele or the derived allele.

7.1.2 Snp Data

Warning: SNP data information has to be in the same order as in genotypic data file.

• pedsnp (example.pedsnp or example.map)

The snp file contains 1 line per SNP. There are 6 columns (last 2 optional): 1st column is chromosome.

Use X for X chromosome, 2nd column is SNP name, 3rd column is genetic position (in Morgans), 4th column is physical position (in bases), Optional 5th and 6th columns are reference and variant alleles.

11	rs0000	0.000000	OAC
11	rs1111	0.001000	100000 A G
11	rs2222	0.002000	200000 A T
11	rs3333	0.003000	300000 C A

• snp (example.snp)

The snp file contains 1 line per SNP. There are 6 columns (last 2 optional): 1st column is SNP name, 2nd column is chromosome. Use X for X chromosome, 2nd column is SNP name, 3rd column is genetic position (in Morgans) (If unknown, ok to set to 0.0), 4th column is physical position (in bases), Optional 5th and 6th columns are reference and variant alleles.

• lfmmsnp (example.lfmmsnp)

The snp file contains 1 line per SNP. There are 3 columns: 1st column is SNP name, 2nd column is chromosome, 3th column is physical position (in bases).

```
rs0000 11 0
rs1111 11 100000
rs2222 11 200000
rs3333 11 300000
```

Tips: SNP data information is not mandatory. But if you have it, we advise you to provide it. It is useful for post-treatment of LFMM analysis.

7.2 Data conversion

The LFMM command-line engine allows data in lfmm format. You can convert from ped, eigenstratgeno, ancestrymap to lfmm format using C progams.

The format of the command-line is (replacing iformat; by ped, ancestrymap, or geno):

```
./bin/<format>21fmm input_file [output_file]
```

where

- input_file is the path for the input file (in ¡format; format).
- output_file is the path for the output_file (in lfmm format). By default, the name of the output file is the name of the input_file with .lfmm extension.

For examples,

example.ped

```
./bin/ped21fmm examples/format_example/example.ped
```

• example.ancestrymap

```
./bin/ancestrymap21fmm examples/format_example/example.ancestrymap
```

• example.geno

```
./bin/geno2lfmm examples/format_example/example.geno
```

• example.vcf

```
./bin/vcf2geno examples/format_example/example.vcf
./bin/geno2lfmm examples/format_example/example.geno
```

7.3 Transform LFMM results

The LFMM command line engine outputs results without taking into account snp data informations. You can add these informations and display results in a table similar as the one in the GUI by using perl scripts. The output format will be called .res. The format is (L the number of loci):

• without snp data

```
perl ./scripts/nothing2lfmm.pl zscore.txt zscore.res L
```

• for example.pedsnp

```
perl ./scripts/pedsnp2lfmm.pl zscore.txt example.pedsnp zscore.res L
```

• for example.snp

```
perl ./scripts/snp2lfmm.pl zscore.txt example.snp zscore.res L
```

• for example.lfmmsnp

```
perl ./scripts/lfmmsnp2lfmm.pl zscore.txt example.lfmmsnp zscore.res L
```

7.4 Manhattan plot

You can create a manhattan plot in the pdf format using a provided R script. If you want, you can highlight a list of specific snps in green. The list of SNPs you want to display should be written in file manhattan_table.txt. The list of SNPs you want to highlight in green should be written in file toHighlight_table.txt.

```
Rscript scripts/manhattan.R manhattan_table.txt toHighlight_table.txt manhattan_plot.pdf
```

The format of these files is the same as the export format of a zscore table (zscore.res). Here is an example of format:

Name	Chr	Position	Zscore	-log10(p-val	ıe) p-value
rs0000	11	0	0.070834	0.0252502	0.943517
rs1111	11	100000	0.0534096	0.0189187	0.957373
rs2222	11	200000	0.0126014	0.00440559	0.989907
rs3333	11	300000	0.0261071	0.00915623	0.979138
rs4444	11	400000	0.0181521	0.00635249	0.985479
rs5555	11	500000	0.00728521	0.0025369	0.994175
rs6666	11	600000	0.00500635	0.0017534	1 0.995971

8 Tutorial

8.1 Data set

The data set that we analyze in this tutorial is an Asian human data set of SNPs data. This data is a worldwide sample of genomic DNA (10757 SNPs) from 388 individuals, taken from the Harvard Human Genome Diversity Project - Centre Etude Polymorphism Humain (Harvard HGDP-CEPH)2. In those data, each marker has been ascertained in samples of Mongolian ancestry (referenced population HGDP01224) [4]. We selected all samples from Asia. Using Tracy-Widom tests implemented in SmartPCA [3], we found that the number of principal components with P-values smaller than 0.01 was around KTW = 10. Using the Bayesian clustering programs STRUCTURE [5] and TESS [1,2], we found that K = 7 components could better describe our simulated data. We extracted climatic data population samples using the WorldClim data set at 30 arcsecond (1km2) resolution (Hijmans, Cameron, Parra, Jones, and Jarvis (2005)). We summarized the climatic variables by using the first axis of a principal component analysis for temperature variables and for precipitation variables. The data set is in directory examples/human_example/. The genotypic information are in panel11_Asia.1fmm. the SNPs information is in panel11.pedsnp. The environmental file is cov_panel11_Asia.env. There are 2 variables, one proxy for temperature and one for precipitation.

8.2 Run LFMM

Here is an example of command line to analyse our dataset

```
./bin/LFMM -x examples/human_example/panel11_Asia.lfmm -v examples/human_example/cov_panel11_Asia.env -K 7 -d 1
```

Output for LFMM

```
LFMM Copyright (C) 2012 Eric Frichot
This program comes with ABSOLUTELY NO WARRANTY; for details type './LFMM -1'.
This is free software, and you are welcome to redistribute it
under certain conditions; type './LFMM -1' for details.
                            LFMM Version 2.0
***
              E. Frichot, S. Schoville, G. Bouchard, O. Francois
****
****
                            Please cite our paper !
      Information at http://membres-timc.imag.fr/Olivier.Francois/lfmm/
./bin/LFMM -x examples/human_example/panel11_Asia.lfmm -v examples/human_example/cov_panel11_Asia.env
Summary of the options:
       -n (number of individuals)
                                      388
       -L (number of loci)
                                      10757
       -K (number of latent factors)
                                      7
       -o (output file)
                                      examples/human_example/panel11_Asia
       -i (number of iterations)
                                      1000
       -b (burnin)
                                      100
       -s (seed random init)
                                      11071964616852227840
       -p (number of processes (CPU))
       -x (genotype file)
                                      examples/human_example/panel11_Asia.lfmm
       -v (variable file)
                                       examples/human_example/cov_panel11_Asia.env
       -D (number of covariables)
       -d (the dth covariable)
                                       1
       -C (DIC file)
                                       examples/human_example/panel11_Asia.dic
Read variable file:
                                                                  OK.
        examples/human_example/cov_panel11_Asia.env
Read genotype file:
                                                               OK.
        examples/human_example/panel11_Asia.lfmm
<<<<
        Analyse for covariable 1
               Start of the Gibbs Sampler algorithm.
End of the Gibbs Sampler algorithm.
       ED:4173727.18
                             DIC: 4173727.955
       The zscores for variable 1 were registered in:
                examples/human_example/panel11_Asia_s1.7.zscore.
       The columns are: zscore, -log10(p-value), p-value, -log10(q-value), q-value.
       The execution for covariable 1 worked without error.
>>>>
```

Here is an example of command line to create zscore.res for the variable.

perl ./scripts/pedsnp2zscore.pl examples/human_example/panel11_Asia_s1.7.zscore examples/human_example/panel11_Asia_s1.7.res 10757

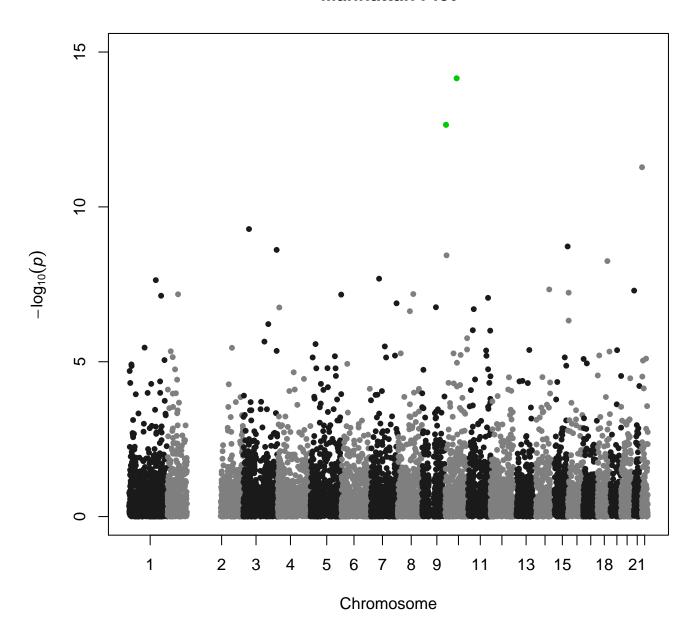
Then, we created a file examples/human_example/toHiglight_table.txt as follows:

```
Name Chr Position Zscore -log10(p-value) p-value -log10(q-value) q-value
Affx-3561055 10 67583134 7.55076 13.3638 4.32731e-14 9.33209 4.65489e-10
Affx-3235806 10 4518973 7.25392 12.3927 4.04887e-13 8.66201 2.17768e-09
```

Here is an example of command line to create a manhattan plot (in examples/human_example/manhattan_plot_2.pdf) with all SNPs and the two previous SNPs highlighted in green:

Rscript scripts/manhattan.R examples/human_example/panel11_Asia_s1.7.res examples/human_example/toHighlight_table.txt examples/human_example/panel11_Asia_s1.7.pdf

Manhattan Plot



9 Contact

If you need assistance, do not hesitate to send me an email (efrichot@gmail.com). A FAQ (Frequently Asked Questions) section is available on our webpage (ttp://membres-timc.imag.fr/Olivier.Francois/lfmm.html). LFMM software is still under development. All your comments and feedbacks are more than welcome.

References

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