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## AGHmatrix Tutorial

R package to compute and analyze relationship matrices for diploid and autotetraploid species

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## 1 Overview

AGHmatrix software is an R-package under development mainly to build relationship matrices using pedigree (A matrix) and/or molecular markers (G matrix) with the possibility to build a combined matrix of Pedigree corrected by Molecular (H matrix). The package also works with Diploid and Autotetraploid Data.

For the pedigree diploid data, it uses the method proposed by [Henderson \(1976\)](#) and described in [Mrode \(2014\)](#).

For the pedigree autotetraploid data, it uses the method proposed [Kerr \*et al.\* \(2012\)](#) and described in [Slater \*et al.\* \(2014\)](#).

For the molecular diploid data, it can use 2 methods: [Powell \*et al.\* \(2010\)](#) and [VanRaden \(2008\)](#).

For the molecular autotetraploid data, it uses a variation of diploid methods which are under development.

The combined matrix H is under development.

### 1.1 Citation

How to cite this software:

Available soon...

## 2 About R

R (R Core Team 2012) is a free programming language widely used in statistical computing. To download R, please visit the Comprehensive R Archive Network (<http://cran.r-project.org>). An alternative is to install the RStudio software, it is a more intuitive/graphical way to use R. To download it, please go to (<https://www.rstudio.com/products/RStudio/>).

For a quick start, we recommend to follow:

- Our R Introduction presentation available at <http://augusto-garcia.github.io/R-Introduction/>.
- “Introduction to R” section in “OneMap Tutorial” available at <http://cran.r-project.org/web/packages/onemap/index.html> for a quick introduction.
- “Verzani’s simpleR — Using R for Introductory Statistics” available at <http://cran.r-project.org/doc/contrib/Verzani-SimpleR.pdf> for a deeper introduction.

## 3 Installing the Package

After you have R installed in your machine, you can install the AGHmatrix package.

Within R, you need to install and load the package devtools:

```
install.packages("devtools")
library(devtools)
```

This will allow you to automatically build and install packages from github. If you use Windows, first install Rtools package [Rtools](#). On a Mac, you will need Xcode (available on the App Store). On Linux, you are good to go.

Then, to install AGHmatrix from github:

```
install_github("prmunoz/AGHmatrix")
```

### 3.1 Loading AGHmatrix package

After, open R (or RStudio) and type:

```
library(AGHmatrix)
```

The package should be available in your R package active list.

## 4 Loading your pedigree file

After load the package you have to load your data file. For it, you can use the function `read.data()` or `read.csv()` for example.

Your data should be available in R as a dataframe where column 1 should be the individual names (id), column 2 and 3 should be the parent names. In the package there is a data example. To look it, type:

```
data(ped.mrode)
ped.mrode

##      Ind Par1 Par2
## 1 Anc1    0    0
## 2 Anc2    0    0
## 3 Var1 Anc1 Anc2
## 4 Var2 Anc1    0
## 5 Var3 Var2 Var1
## 6 Var4 Var3 Anc2
```

The example *ped.mrode* (above) has 3 columns, where column 1 is the individual names, column 2 is the parental 1 names, column 3 is the parental 2 names. There is no header and the unknown value default is 0. Your personal data has to be in the same format than *ped.mrode*.

## 5 Relationship Matrix with Pedigree Data

In this section is presented how to load the data and how to construct the relationship matrix for diploid and autotetraploid species. The pedigree-base relationship matrix calculation, matrix A, is performed according with a recursive method as presented in Mrode (2005) and described by Henderson (1976). This method is expanded for n-ploidy according with Kerr *et al.* (2012) described in Slater *et al.* (2014).

In the algorithm, first occurs the preprocessing of the data. To the preprocessing of the pedigree, first, the individuals are numerated 1 to , where is the total individuals of the pedigree data. Then, it is verified if they are chronological sorted (i.e., if the parents of a given individual n are located before it in the list). If not, the algorithm performs necessary permutes. After preprocessing, occurs the matrix computation as presented in Mrode (2014) for diploid and Slater *et al.* (2014) for autotetraploidy.

### 5.1 Building matrix A

To build the A matrix you need to type the function with the data, ploidy, double reduction and unknown values. For example, if ploidy equals to 2 and unknown value equals 0 is calculated as presented in Mrode (2014) with the following code:

```
Amatrix(data=ped.mrode,ploidy=2,unk=0)
```

If ploidy equals to 4 and double reduction equals to 10% is calculated as presented in Slater *et al.* (2014) with the following code:

```
Amatrix(data=ped.mrode,ploidy=4,w=0.1,unk=0)
```

If you want to save your matrix in an object, you can use the following code:

```
matrix.example <- Amatrix(data=ped.mrode,ploidy=4,w=0.1,unk=0)
```

More information about the Amatrix function you can have typing:

```
?Amatrix
```

## 5.2 Exporting your data as ASREML csv format

In this section, we present how to use the function `formatmatrix` in order to export a matrix to a compatible ASREML format (csv file with 3 columns). In order to do it, we need to build a matrix, its inverse, and export it using `formatmatrix` function. This function has as options: *round.by*, which set the number of decimals you desire, *exclude.0*, if TRUE, remove all the zeros from your data, *name* what is the desired name of your file.

Below, an example of how to do it:

```
#setting the number of digits to display in R for 12
options(digits=12)

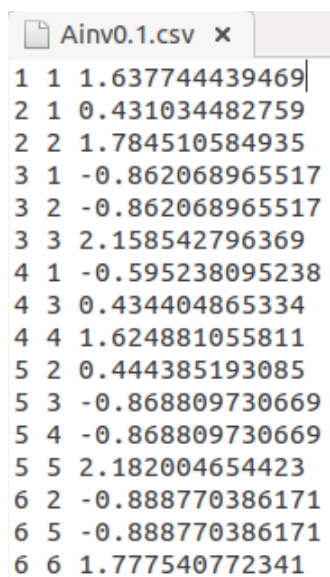
#loading the data example
data(ped.mrode)

#building the matrix
A<-Amatrix(data=ped.mrode, ploidy=4, w=0.1, unk=0)

#build the inverse
Ainv<-solve(A)

#exporting it. The function "formatmatrix" will convert the matrix in a 3-column table.
formatmatrix(Ainv, round.by=12, exclude.0=TRUE, name="Ainv0.1")
```

This script will create the following csv file presented in Figure 2.



Row	Col	Value
1	1	1.637744439469
2	1	0.431034482759
2	2	1.784510584935
3	1	-0.862068965517
3	2	-0.862068965517
3	3	2.158542796369
4	1	-0.595238095238
4	3	0.434404865334
4	4	1.624881055811
5	2	0.444385193085
5	3	-0.868809730669
5	4	-0.868809730669
5	5	2.182004654423
6	2	-0.888770386171
6	5	-0.888770386171
6	6	1.777540772341

Figure 1: csv file representing an inversed A matrix from ped.mrode data with  $w=0.1$ . The first 2 columns represent rows and columns of the matrix, the third column represents the value. All the rows with value equal to 0 it was excluded from the file.

### 5.3 Making a *loop* in order to get several matrices

In this section, we present a simple "for" function for the user be able to get in a practical way several matrices for different double reduction values to later be used in ASREML (for example). In R:

```
#setting the number of digits to display in R for 12
options(digits=12)

#loading the data example
data(ped.mrode)

#determining your double reduction range
double.red<-seq(0,0.2,0.05)

#extracting the length of double.red
n<-length(double.red)

#Looping it
for(i in 1:n){
  A<-Amatrix(data=ped.mrode,
```

```

        ploidy=4,
        w=double.red[i],
        unk=0)
#making the inverse
A.inv<-solve(A)
#exporting as csv
formatmatrix(data=A.inv,
             name=paste("Ainv_",double.red[i],sep=""),
             round.by=12,
             exclude.0=TRUE)
}

```

At the end, it will get 5 files represents 5 matrices (if double-reduction proportion of 0, 0.05, 0.1, 0.15, and 0.2). These matrices will be in a 3 column-way format as in Figure 2.

## 6 Relationship Matrices with Molecular Data - G Matrix

*This chapter is under construction!*

To build the relationship matrix based on markers, your data should be organized in a matrix (markers x individual) coded as 0,1,2 and missing data value. Your data can be easily loaded in R with the function `read.table()`. The function `Gmatrix` can construct the matrix proposed by [VanRaden \(2008\)](#) or the matrix proposed by [Powell et al. \(2010\)](#).

As example, here we build both matrices using fake data (`snp.table`, which is part of this R package).

```

#loading the data example
data(snp.table)

#looking the data, snp.table missing values is coded as -9
snp.table

```

##	Genotype2	Genotype3	Genotype8
## Marker1	2	2	2
## Marker2	2	2	2
## Marker3	2	2	2
## Marker4	2	2	2
## Marker5	1	-9	2
## Marker6	-9	2	-9
## Marker7	2	2	2
## Marker8	2	1	2
## Marker9	2	2	2
## Marker10	2	2	1

```

## Marker11      0      1      1
## Marker12      2      2      2
## Marker13      1      0      1
## Marker14      2      2      2
## Marker15      1      1      2
## Marker16      2      1      1
## Marker17      2      2      2
## Marker18      1      2      2
## Marker19      2      0      2
## Marker20      2      2      2
## Marker21      1      2      1
## Marker22      2      2      2
## Marker23      2      2      2
## Marker24      2     -9      1
## Marker25      0      2      2
## Marker26      0      1      0
## Marker27      1      1      1
## Marker28      1      2      2
## Marker29      2      0      0
## Marker30      2      2      2

#building the matrix based upon VanRaden
G<-Gmatrix(SNPdata=snp.table, missingValue=-9, method="VanRaden")

## Number of Markers: 30
## Number of Individuals: 3
##
## Completed! Time = 0.002 seconds

G

##          Genotype2  Genotype3  Genotype8
## Genotype2  1.55156951 -0.2511211 0.09865471
## Genotype3 -0.25112108  1.2556054 0.39461883
## Genotype8  0.09865471  0.3946188 1.22869955

#building the matrix based upon Powell
G<-Gmatrix(SNPdata=snp.table, missingValue=-9, method="Powell")

## Number of Markers: 30
## Number of Individuals: 3
##
## Completed! Time = 0 seconds

G

```



```
##          Genotype2  Genotype3  Genotype8
## Genotype2  1.14111111 -0.09166667  0.1388889
## Genotype3 -0.09166667  1.05333333  0.2233333
## Genotype8  0.13888889  0.22333333  1.0811111
```

More information about the Gmatrix function you can have typing:

```
?Gmatrix
```

To invert and export the matrix follow the steps already described in 5.2.

## Bibliography

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