

Constructing A^{-1}

- Pedigree relationships:

$$u_i = 0.5(u_{s_i} + u_{d_i}) + \varphi_i$$

$$\mathbf{A}^{-1} = (\mathbf{I} - \mathbf{P})' \mathbf{M}^{-1} (\mathbf{I} - \mathbf{P})$$

Henderson (1976)
Quaas (1988)

```
ImP=(/1., -.5, -.5/)
Minv=(/2., 4/3., 1., 0./)
ainv=0.0
```

```
do
  read(2,*,iostat=io) animal, sire, dam,par_stat
  if (io /= 0) exit
  p(1)=animal
  p(2)=sire
  p(3)=dam
  do i=0,nanimal
    do k=1,3
      do l=1,3
        ainv(k,l)=ainv(k,l)+ImP(k)*ImP(l)*Minv(par_stat)
      enddo
    enddo
  enddo
```

par_stat:
3 - # of known parents



UNIVERSITY OF
GEORGIA

College of Agricultural &
Environmental Sciences

*Animal Breeding and
Genetics Group*

Theory of GBLUP and single-step GBLUP

Daniela Lourenco

BLUPF90 TEAM – 08/2024



UNIVERSITÀ
DEGLI STUDI
FIRENZE

Statistical methods before genomics

- BLUP (Henderson, 1949 - 1976)
 - Best: minimizes MSE
 - Linear: linear function of the data
 - Unbiased: $E(u) = E(\hat{u})$
 - Prediction: for random effects

Statistical Science
1991, Vol. 6, No. 1, 15-51

**That BLUP Is a Good Thing: The Estimation of
Random Effects**

G. K. Robinson

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{W} \\ \mathbf{W}'\mathbf{X} & \mathbf{W}'\mathbf{W} + \mathbf{A}^{-1} \frac{\sigma_e^2}{\sigma_u^2} \end{bmatrix} \begin{bmatrix} \hat{\boldsymbol{\beta}} \\ \hat{\mathbf{u}} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{y} \\ \mathbf{W}'\mathbf{y} \end{bmatrix}$$

Henderson's MME

- Model

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{W}\mathbf{u} + \mathbf{e}$$

- Joint probability of phenotypes and EBV

$$p(\mathbf{y}, \mathbf{u}) = p(\mathbf{u}|\mathbf{y}) p(\mathbf{y}) = p(\mathbf{y}|\mathbf{u}) p(\mathbf{u})$$

- Joint probability density function of phenotypes and EBV

$$p(\mathbf{y}, \mathbf{u}) = p(\mathbf{y}|\mathbf{u}) p(\mathbf{u}) = \frac{1}{\sqrt{2\pi|\mathbf{R}|}} e^{-\frac{1}{2}(\mathbf{y}-\mathbf{X}\boldsymbol{\beta}-\mathbf{W}\mathbf{u})'\mathbf{R}^{-1}(\mathbf{y}-\mathbf{X}\boldsymbol{\beta}-\mathbf{W}\mathbf{u})} \frac{1}{\sqrt{2\pi|\mathbf{G}|}} e^{-\frac{1}{2}(\mathbf{u}-\mathbf{0})'\mathbf{G}^{-1}(\mathbf{u}-\mathbf{0})}$$

$$\begin{cases} \mathbf{X}'\mathbf{R}^{-1}\mathbf{X}\boldsymbol{\beta} + \mathbf{X}'\mathbf{R}^{-1}\mathbf{W}\mathbf{u} = \mathbf{X}'\mathbf{R}^{-1}\mathbf{y} \\ \mathbf{W}'\mathbf{R}^{-1}\mathbf{X}\boldsymbol{\beta} + (\mathbf{W}'\mathbf{R}^{-1}\mathbf{W} + \mathbf{G}^{-1})\mathbf{u} = \mathbf{W}'\mathbf{R}^{-1}\mathbf{y} \end{cases} \quad \begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{W} \\ \mathbf{W}'\mathbf{X} & \mathbf{W}'\mathbf{W} + \mathbf{A}^{-1} \frac{\sigma_e^2}{\sigma_u^2} \end{bmatrix} \begin{bmatrix} \hat{\boldsymbol{\beta}} \\ \hat{\mathbf{u}} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{y} \\ \mathbf{W}'\mathbf{y} \end{bmatrix}$$

Henderson's MME for dairy in 1989

- BLUP (Henderson, 1949 - 1976)
- Implementation for dairy in 1989

National genetic improvement programs for dairy cattle in the United States

G. R. Wiggans

J Anim Sci 1991. 69:3853-3860.

Challenges

Genetic improvement programs are in a period of rapid change. Advances in computer capability enable adoption of sophisticated computational procedures. Advances in repro-



ELSEVIER

Journal of Dairy Science
Volume 71, Supplement 2, June 1988, Pages 54-69



Implementation of an Animal Model for Genetic Evaluation of Dairy Cattle in the United States

G.R. Wiggans, I. Misztal, L.D. Van Vleck

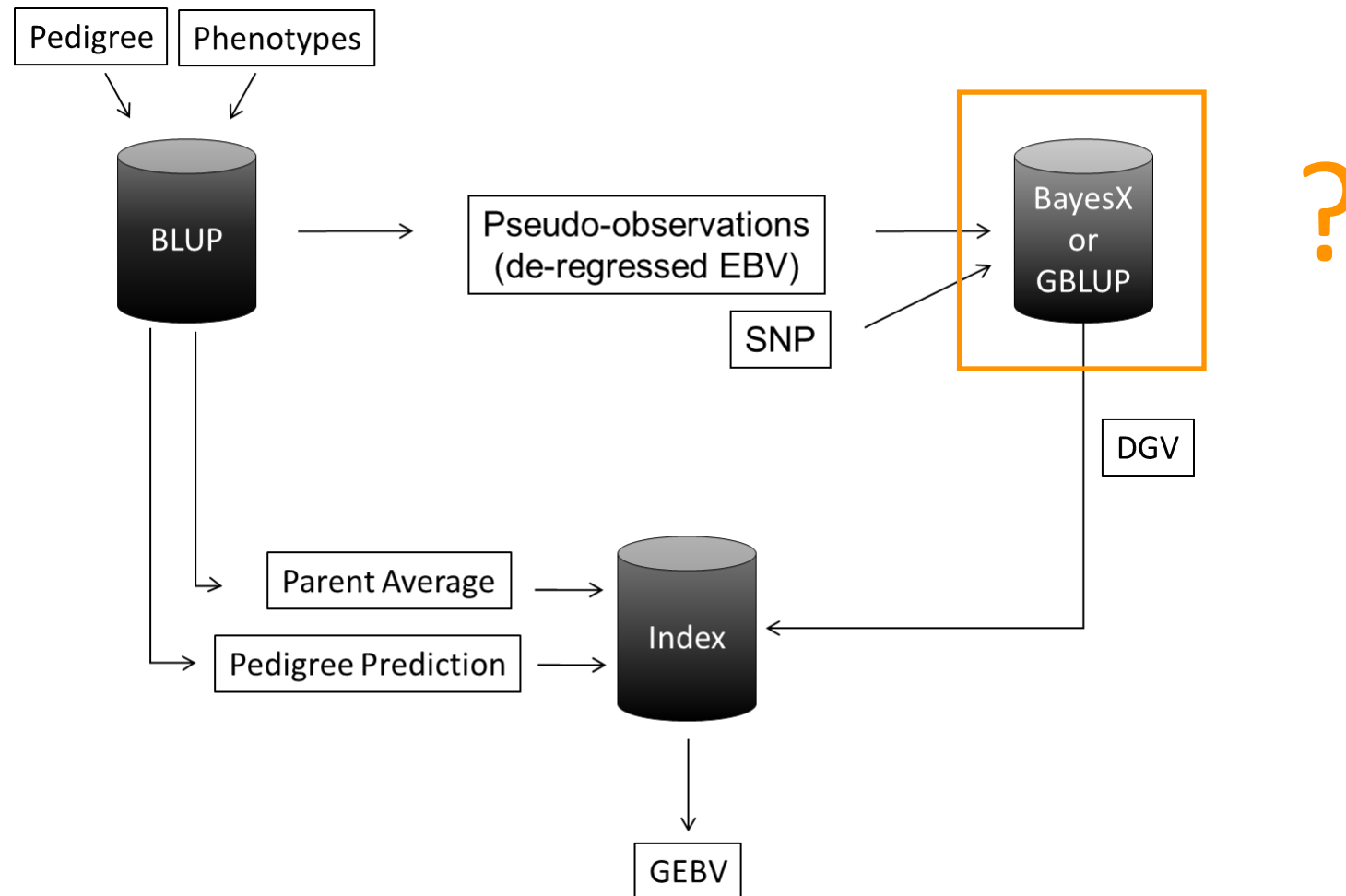
- 9.5 M animals
- 11 M lactations
- 23.5 M equations to solve
- 7.5 hours

ACKNOWLEDGMENTS

This research was conducted using the Cornell National Supercomputer Facility, a resource of the

Moving from 1989 to 2009

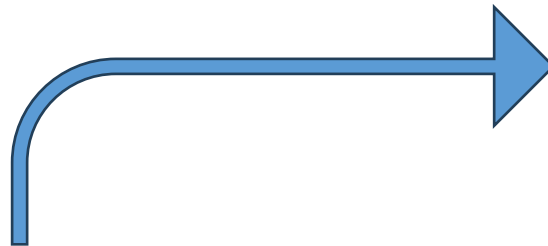
- How to add genomic information to the evaluation system in 2009?



Multistep

SNP-BLUP (ridge regression)

- SNP effect model = outputs SNP effects
- $a \sim N(0, \sigma_a^2)$



$$y = X\beta + Za + e$$

$$\begin{bmatrix} X'X & X'Z \\ Z'X & Z'Z + I \frac{\sigma_e^2}{\sigma_a^2} \end{bmatrix} \begin{bmatrix} \hat{\beta} \\ \hat{a} \end{bmatrix} = \begin{bmatrix} X'y \\ Z'y \end{bmatrix}$$

$$\text{GEBV} = Z\hat{a}$$

```

110101111511101111001000122115120512212502251110250122010201021000221121025000122010:
2110110102201212222012101222010120222111112021222111112102020101101020111112011012110:
1210100211120211112000212122210002112212212211000002022000021102212221212202000112020:
12000120022012121110012100222211021122110201121222120022002121212111202112022002022100:
2000020202210212211200220012222111012202021102022202022000122212101201021022010011010:
1011021202201211221110210011110010221121202211111020221001201222012111021021021012000:
121002120220012211000111122201001012011212111212012211022020102002021211222022010022110:
111001020221220210201011012220200121221112212211211122200220111201121211022000022012:
21101202021111210121102110222010010221212221102220201221020212112010211122022112011010:
2200011102210122101021102520201112120222122212220110121011102220050210121022010022125:
210102200121221211212021012222002012210212110201121021221002211011020211021112021012010:
1220111201222021021001000212100112012020200121002002121001120102202121211022010101100:
22111221012112022221022102110201021121211122000001112200022112202022211212001212110:
2002010012212101010210122211011122202020221100101112100112010220122202110210011020:
1200010202211122001010210022110002022212122222001011022111021201201121221111102112010:
1110002122112120121212100222110120222101022112222110220011202110020201102022100021020:
1100001202022002212120220012102000111221101102222120022002012001010212121022102010110:
112100210210010101111022002221200022211120202222211022210120201211122211112011011020:
21100202152100122120201100220020112512121502252222250221011201121051202222112111012110:
110011120220111211101020012221000112221212021211121200220012202220022212212112001112011:
2100021202211202211210210122210110122212221212111202012210122011211121112022000012101:
21000202022002022222001200222000122022220021102252200122001202111151001012022001012025:
212102121521002201200012101121201215110215122521121150220011102111050202221122011022010:
11110212152001221221102000122020122522211502152222115022011020212005020202202211112110:
12110102112220210101022002221201201121221012111101112210202001010112212121002021021:
2210001202212222102020211022211010121102212022222002210022112121021202011022010111010:
1100012202201212201100220111211000110211221212122002011222200222211021111212022011022010:
12101001112001211110021112220111112122221210201110202210021122210012121112101211110:
210100110220122121102110212101212022121212110111110221001202121110211011021100022020:
1210010202211212122100100021202011112211121220011111022100220102201212121021000012020:

```

- All SNP explain the same proportion of variance on the trait

SNP-BLUP (ridge regression)

- SNP effect model = outputs SNP effects
- All SNP explain the same proportion of variance on the trait

$$\mathbf{GEBV} = \mathbf{Z}\hat{\mathbf{a}}$$

$$\mathbf{u} = \mathbf{Z}\hat{\mathbf{a}}$$

$$\text{Var}(\mathbf{u}) = \text{Var}(\mathbf{Z}\mathbf{a})$$

$$\text{Var}(\mathbf{u}) = \mathbf{Z} \text{Var}(\mathbf{a}) \mathbf{Z}'$$

$$\text{Var}(\mathbf{u}) = \mathbf{Z}\mathbf{Z}'\sigma_a^2$$

$$\sigma_a^2 = \frac{\sigma_u^2}{2 \sum_{i=1}^{SNP} p_i(1 - p_i)}$$

$$\text{Var}(\mathbf{u}) = \mathbf{Z}\mathbf{Z}' \frac{\sigma_u^2}{2 \sum_{i=1}^{SNP} p_i(1 - p_i)}$$

$$\text{Var}(\mathbf{u}) = \frac{\mathbf{Z}\mathbf{Z}'}{2 \sum_{i=1}^{SNP} p_i(1 - p_i)} \sigma_u^2$$

Genomic
relationship matrix
VanRaden (2008)

$$\mathbf{G} = \frac{\mathbf{Z}\mathbf{Z}'}{2 \sum_{i=1}^{SNP} p_i(1 - p_i)}$$

$$\text{Var}(\mathbf{u}) = \mathbf{G}\sigma_u^2$$



GBLUP assumption!!!

GBLUP: Genomic BLUP

- GEBV-based model = outputs genomic breeding values (GEBV)
- $\mathbf{u} \sim N(0, \mathbf{G}\sigma_u^2)$

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{W}\mathbf{u} + \mathbf{e}$$

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{W} \\ \mathbf{W}'\mathbf{X} & \mathbf{W}'\mathbf{W} + \mathbf{G}^{-1} \frac{\sigma_e^2}{\sigma_u^2} \end{bmatrix} \begin{bmatrix} \hat{\boldsymbol{\beta}} \\ \hat{\mathbf{u}} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{y} \\ \mathbf{W}'\mathbf{y} \end{bmatrix}$$

Fernando & Grossman (1989)

Bernardo (1994)

Nejati-Javaremi et al. (1997)

$$\mathbf{G} = \frac{\mathbf{Z}\mathbf{Z}'}{2 \sum p_i(1 - p_i)}$$

VanRaden (2008)

GBLUP: Genomic BLUP

$$G = \frac{ZZ'}{2 \sum p_i(1 - p_i)}$$

```

025 110101111511110111110010001221151205122125022511110250122010201021000221121025000122010:
036 21101101022012122222012101222010120222111112021222111112102020101101020111112011012110:
050 121010021112021111200021212222100021122122122110000020220000211022122212122020001112020:
054 120001200220121211100121002222110211221102011212221200220021212121111202112022002022100:
066 20000202022102122112002200122221101220202110202222020220001222121011201021022010011010:
097 101102120220121122111021001111100102211212022111111020221001201222012111021021021012000:
101 121002120220011221100011112220100101120112121211121201221002102002021211222022010022110:
151 111001020221220210201011012220200121221111221221121111222002201112011212111022000022012:
172 211012020211112101211021102220101001221212221102220201221020212112010211122022112011010:
224 220001110221012210101021102520201112120222122212220110121011102220050210121022010022125:
277 210102200121221211212021012222002012210212110201121021221002211011020211021112021012010:
314 122011120122220210210010002121001120120202001210020021210011201022021212111022010101100:
419 221112210121120222221022102110201021121211122000000111220002211122020222112120012121110:
439 200202100122121210101021012221101112220202022110010111210011201022012220211021010011020:
456 12000102022111220010102100221100020222121222220010110221110212012011212211111102112010:
501 111000021221121201212121002221101202222101022112222110220011202110020201102022100021020:
571 110000120202200221212022001210200011122110110222221200220020212001010212121022102010110:
579 11210021021001010111102200222120002221111202022222110222101202012111222111112011011020:
581 21100202152100122120201100220020112512121502252222250221011201121051202222112111012110:
657 11001112022011121110102001222100011222121202121112120022001220222002221212112001112011:
660 210002120221120221121021012221011012221222121211120201221012201121111211112022000012101:
730 210002020220020222220012002220001220222220021102252200122001202111151001012022001012025:
732 21210212152100220120001210112120121511021512252112115022001102111050202221122011022010:
764 111102121520012212211020001220201225222115021522221150220110202120050202022022111112110:
780 121101021122220210101022002221201201121221012111110111221020202001010112212121002021021:
800 22100012022122221020202110222110101211202212022222200221002211121021202011022010111010:
816 11000122022012122011002201112110001102112212122002011222200222211021111212022011022010:
832 121010011120011211110021112220111112122221210201111020221002112221001212111121012111110:
900 21010011022012212121102110212101212022121212110111110221001202121110211011021100022020:
901 12100102022112121221001000212020111122111212200111110221002201022012212121021000012020:

```

Genomic relationship matrix

$$\mathbf{G} = \frac{\mathbf{Z}\mathbf{Z}'}{2 \sum p_i(1 - p_i)} = \frac{(\mathbf{M} - 2\mathbf{P})(\mathbf{M} - 2\mathbf{P})'}{2 \sum p_i(1 - p_i)}$$

Genotypes {0,1,2}

Shifted to refer to the average of a population with allele frequencies p

Scaled to refer to the genetic variance of a population with allele frequencies p

Each SNP contributes $2pq\alpha^2$ to the genetic variance

Genomic relationship matrix

$$\mathbf{G} = \frac{\mathbf{Z}\mathbf{Z}'}{2 \sum p_i(1 - p_i)} = \frac{(\mathbf{M} - 2\mathbf{P})(\mathbf{M} - 2\mathbf{P})'}{2 \sum p_i(1 - p_i)}$$

Genotypes {0,1,2}

Shifted to refer to the average of a population with allele frequencies p

Scaled to refer to the genetic variance of a population with allele frequencies p

If base allelic frequencies are used, \mathbf{G} is an unbiased and efficient estimator of IBD realized relationships

Pedigree vs. Genomic relationships

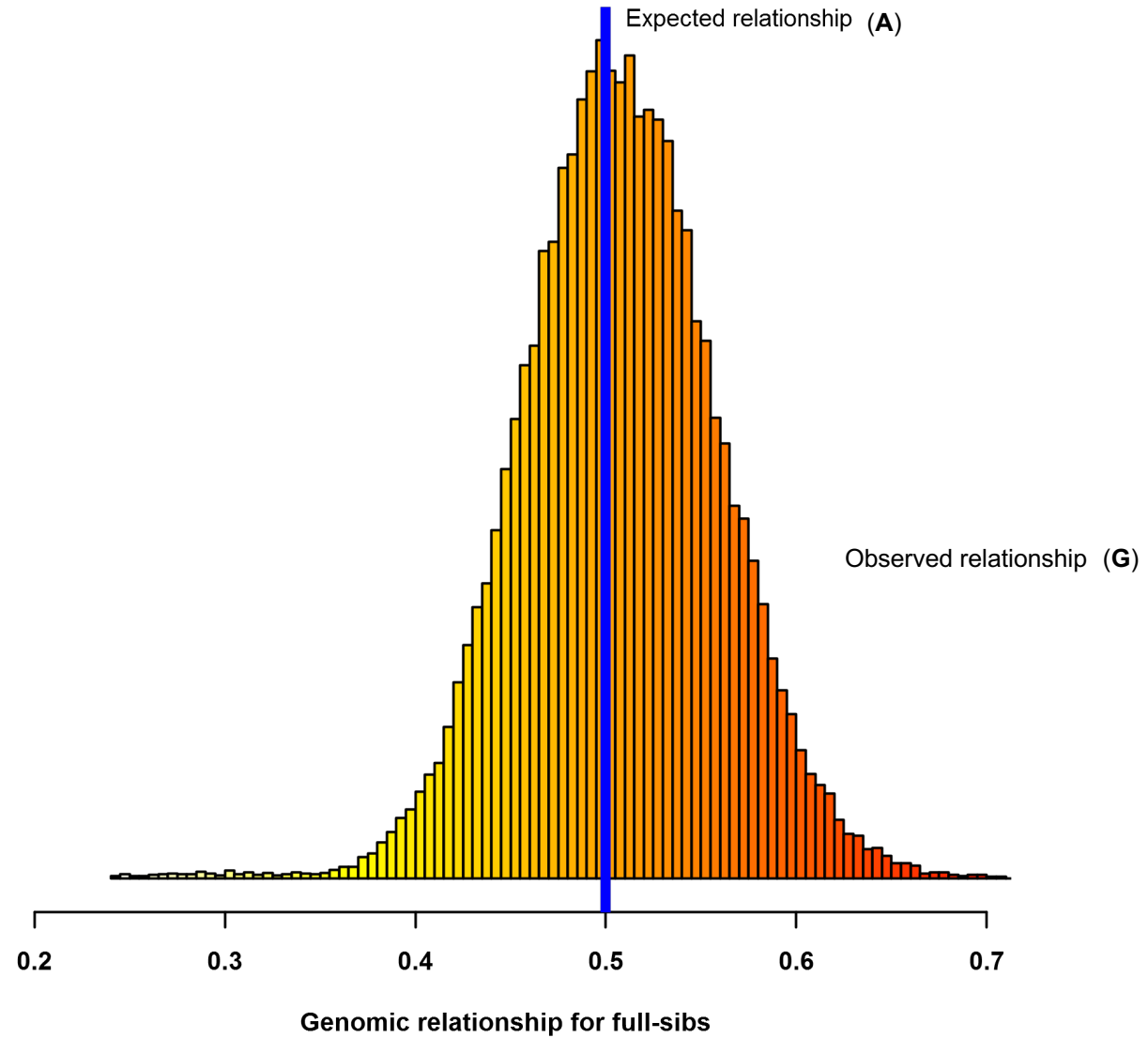
- **A** provides Identical By Descent relationships (IBD)
- **G** provides Identical by State relationships (IBS)

- **A** is the expectation of realized or observed relationships
- **G** contains realized or observed relationships

- SNPs more informative than **A**
 - Two full sibs might have a correlation of 0.4 or 0.6

- Many markers needed to better estimate relationships
 - IBS as an estimator of IBD

Pedigree vs. Genomic relationships



Some “interesting” properties of **G**

- VanRaden (2008)
 - **G** can be singular if few SNP or identical genotypes (twins)
 - **G** must be singular if number of individuals > number of SNP

$$\mathbf{G} = 0.95 \frac{\mathbf{ZZ}'}{2 \sum p_i(1 - p_i)} + 0.05\mathbf{I} \quad \text{OR} \quad \mathbf{G} = 0.95 \frac{\mathbf{ZZ}'}{2 \sum p_i(1 - p_i)} + 0.05\mathbf{A} \quad \rightarrow \quad \mathbf{G} = \alpha\mathbf{G}_0 + \beta\mathbf{A}$$

- Blending \approx Adding a residual polygenic effect

Some “interesting” properties of **G**

- For all matrices of the kind

$$\mathbf{G} = \frac{\mathbf{Z}\mathbf{Z}'}{2 \sum p_i(1 - p_i)} = \frac{(\mathbf{M} - 2\mathbf{P})(\mathbf{M} - 2\mathbf{P})'}{2 \sum p_i(1 - p_i)}$$
 - We don't need to put the same p 's in the upper and and in the lower part
- Changing allele frequencies in \mathbf{P} shifts EBV's by a constant
 - Irrelevant if there is an overall mean or fixed effect in the model (Stranden and Christensen, 2011)
- Changing allele frequencies in $\frac{1}{2\sum p_i q_i}$ “scales”

GBLUP

- GEBV-based model = outputs genomic breeding values (GEBV)
- $\mathbf{u} \sim N(0, \mathbf{G}\sigma_u^2)$

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{W}\mathbf{u} + \mathbf{e}$$

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{W} \\ \mathbf{W}'\mathbf{X} & \mathbf{W}'\mathbf{W} + \mathbf{G}^{-1} \frac{\sigma_e^2}{\sigma_u^2} \end{bmatrix} \begin{bmatrix} \hat{\boldsymbol{\beta}} \\ \hat{\mathbf{u}} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{y} \\ \mathbf{W}'\mathbf{y} \end{bmatrix}$$

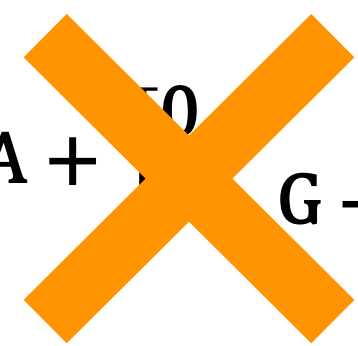
Only for
genotyped individuals!!!

$$\mathbf{G} = \frac{\mathbf{Z}\mathbf{Z}'}{2 \sum p_i(1 - p_i)}$$

VanRaden (2008)

Not all individuals are genotyped

- Genomic evaluation would be simpler if all individuals were genotyped
- What to do when there are genotyped and non-genotyped individuals?
 - SNPs are capturing relationships
 - Pedigrees give information about relationships
 - Genomic and pedigree relationships can be combined in a single matrix!

$$\begin{array}{c}
 \text{Non-genotyped} \rightarrow \\
 \mathbf{A} = \begin{bmatrix} \mathbf{A}_{11} & \mathbf{A}_{12} \\ \mathbf{A}_{21} & \mathbf{A}_{22} \end{bmatrix} \\
 \leftarrow \text{Genotyped}
 \end{array}
 \quad
 \mathbf{A} = \begin{bmatrix} \mathbf{A}_{11} & \mathbf{A}_{12} \\ \mathbf{A}_{21} & \mathbf{G} \end{bmatrix}
 \quad
 \mathbf{H} = \mathbf{A} + \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{G} & -\mathbf{A}_{22} \end{bmatrix}$$


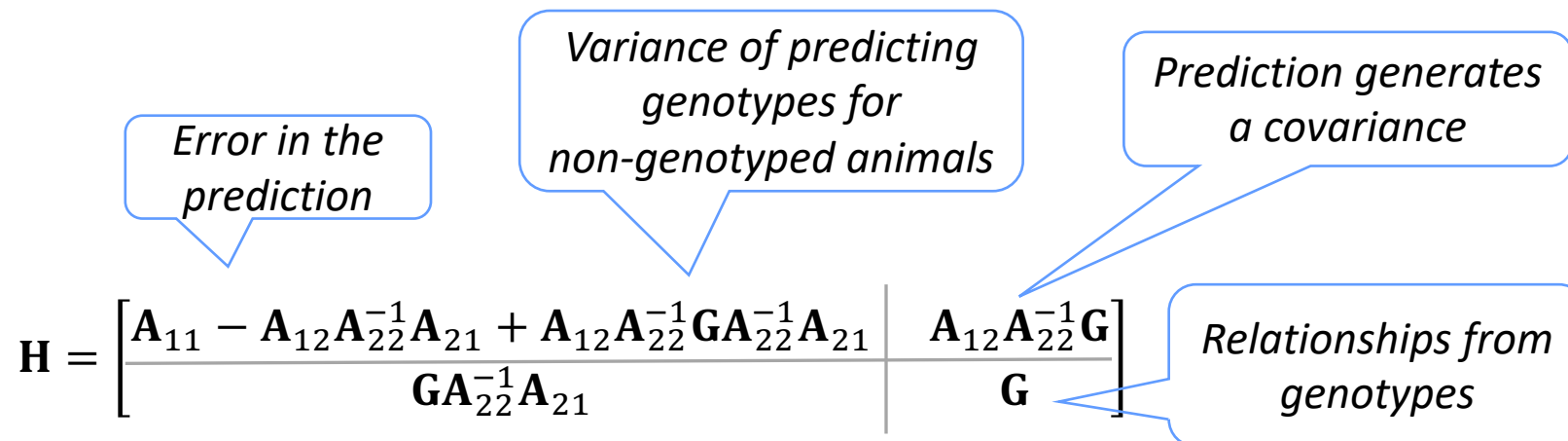
Not all animals are genotyped

- Genomic info can be extended to non-genotyped animals
 - joint distribution of EBV for non-genotyped (u_1) and genotyped (u_2)

$$p(u_1, u_2) = p(u_2)p(u_1|u_2)$$

Legarra et al., 2009

$$\mathbf{H} = \begin{pmatrix} \text{var}(u_1) & \text{cov}(u_1, u_2) \\ \text{cov}(u_2, u_1) & \text{var}(u_2) \end{pmatrix} = \begin{pmatrix} \mathbf{A}_{11} + \mathbf{A}_{12}\mathbf{A}_{22}^{-1}(\mathbf{G} - \mathbf{A}_{22})\mathbf{A}_{22}^{-1}\mathbf{A}_{21} & \mathbf{A}_{12}\mathbf{A}_{22}^{-1}\mathbf{G} \\ \mathbf{G}\mathbf{A}_{22}^{-1}\mathbf{A}_{21} & \mathbf{G} \end{pmatrix}$$



The diagram shows the H matrix partitioned into two columns. The first column contains the variance-covariance matrix for non-genotyped animals, and the second column contains the relationships from genotypes. Callouts explain the components:

- Error in the prediction:** Points to the top-left element of the first column, $\mathbf{A}_{11} - \mathbf{A}_{12}\mathbf{A}_{22}^{-1}\mathbf{A}_{21}$.
- Variance of predicting genotypes for non-genotyped animals:** Points to the top-right element of the first column, $\mathbf{A}_{12}\mathbf{A}_{22}^{-1}\mathbf{G}$.
- Prediction generates a covariance:** Points to the bottom-right element of the first column, $\mathbf{G}\mathbf{A}_{22}^{-1}\mathbf{A}_{21}$.
- Relationships from genotypes:** Points to the bottom-right element of the second column, \mathbf{G} .

$$\mathbf{H} = \left[\begin{array}{c|c} \mathbf{A}_{11} - \mathbf{A}_{12}\mathbf{A}_{22}^{-1}\mathbf{A}_{21} + \mathbf{A}_{12}\mathbf{A}_{22}^{-1}\mathbf{G}\mathbf{A}_{22}^{-1}\mathbf{A}_{21} & \mathbf{A}_{12}\mathbf{A}_{22}^{-1}\mathbf{G} \\ \hline \mathbf{G}\mathbf{A}_{22}^{-1}\mathbf{A}_{21} & \mathbf{G} \end{array} \right]$$

Some properties of \mathbf{H}

- Always semi-positive definite
 - eigenvalues are always positive or zero
- Positive definite & invertible if \mathbf{G} is invertible
- In practice, if \mathbf{G} is too different from \mathbf{A}_{22} (wrong pedigree or genotyping)
 - Numerical problems
- If no one is genotyped, Single-step is BLUP
- If everyone is genotyped, Single-step is GBLUP

Realized relationship matrix (**H**)

Animal	Sire	Dam
1	0	0
2	0	0
3	1	2
4	1	2

Pedigree
Relationship
Matrix (**A**)

$$\begin{bmatrix} 1.0 & 0.0 & 0.5 & 0.5 \\ . & 1.0 & 0.5 & 0.5 \\ . & . & 1.0 & 0.5 \\ . & . & . & 1.0 \end{bmatrix}$$

Genomic
Relationship
Matrix (**G**)
for animals 3 and 4

$$\begin{bmatrix} 1.0 & 0.52 \\ . & 1.0 \end{bmatrix}$$

Realized
Relationship
Matrix (**H**)

$$\begin{bmatrix} 1.004 & 0.0 & 0.507 & 0.507 \\ . & 1.004 & 0.507 & 0.507 \\ . & . & 1.0 & 0.52 \\ . & . & . & 1.0 \end{bmatrix}$$



Understanding H

- It is a projection of **G** on the rest of the individuals so that **G** makes sense
 - e.g., parents of two animals related in **G** should be related in **A**
- It is a Bayesian update of the pedigree matrix based on new information from genotypes
- Typically
 - **A** in the millions
 - **G** and **A**₂₂ in the thousands
 - Leads to a very efficient method of genomic evaluation:
 - Single Step GBLUP

Single-step Genomic BLUP (ssGBLUP)

- Because not all animals are genotyped
 - 5% to 15% in large populations

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z} \\ \mathbf{Z}'\mathbf{X} & \mathbf{Z}'\mathbf{Z} + \mathbf{H}^{-1} \frac{\sigma_e^2}{\sigma_u^2} \end{bmatrix} \begin{bmatrix} \hat{\boldsymbol{\beta}} \\ \hat{\mathbf{u}} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{y} \\ \mathbf{Z}'\mathbf{y} \end{bmatrix}$$

$$\mathbf{H}^{-1} = \mathbf{A}^{-1} + \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{G}^{-1} - \mathbf{A}_{22}^{-1} \end{bmatrix}$$

Aguilar et al., 2010
Christensen and Lund, 2010



Single-step Genomic BLUP (ssGBLUP)

Single-step H matrix inverse - 2009

- May 4-6 2009 meeting at UW - Madison:
 - Symposium on Statistical Genetics of Livestock for the Post-Genomic Era
- May 11 email from Dave Johnson, LIC, NZL to Ignacy Misztal:
 - Good to meet you again last week. I have been thinking about your whole-population H matrix. It occurred to me, as I looked for something to do as I waited in LA airport, that H has a simple relatively sparse inverse. Namely $H^{-1} = A^{-1} + \{(0,0), (0, G^{-1} - A_{22}^{-1})\}$. Have you looked at this at all?

Chapman memorial lecture, Madison, WI, April 2024 (16)

USDA
VanRaden



Journal of Dairy Science
Volume 93, Issue 2, February 2010, Pages 743-752



*Hot topic: A unified approach to utilize phenotypic, full pedigree, and genomic information for genetic evaluation of Holstein final score*¹

I. Aguilar^{*†}, I. Misztal^{*}, D.L. Johnson[‡], A. Legarra[§], S. Tsuruta^{*}, T.J. Lawlor[#]

Combining two sources of relationships

$$\mathbf{H} = \mathbf{A} + \begin{bmatrix} \mathbf{A}_{12}\mathbf{A}_{22}^{-1}(\mathbf{G} - \mathbf{A}_{22})\mathbf{A}_{22}^{-1}\mathbf{A}_{21} & \mathbf{A}_{12}\mathbf{A}_{22}^{-1}(\mathbf{G} - \mathbf{A}_{22}) \\ (\mathbf{G} - \mathbf{A}_{22})\mathbf{A}_{22}^{-1}\mathbf{A}_{21} & \mathbf{G} - \mathbf{A}_{22} \end{bmatrix}$$

- **A**
 - Contains expected relationships
 - Is limited by the pedigree depth and completeness
 - Depends on accuracy of recording pedigrees

- **G**
 - Contains number of alleles shared between animals weighted by heterozygosity
 - No limitations regarding to the number of past generations
 - Depends on allele frequency and quality of genomic data

Combining two sources of relationships

$$\mathbf{H}^{-1} = \mathbf{A}^{-1} + \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{G}^{-1} - \mathbf{A}_{22}^{-1} \end{bmatrix}$$

Computed using Henderson-Quaas' algorithm with inbreeding

Computed using VanRaden's formula, which considers inbreeding

Computed using Colleau's algorithm, which considers inbreeding

- Tuning

- Base of \mathbf{G} is *genotyped* animals
- Base of \mathbf{A} is *founders of the pedigree*
- For SSGBLUP, Vitezica et al. 2011 modeled a mean in genotyped animals:

$$p(\mathbf{u}_2) = N(\mathbf{1}\mu, \mathbf{G})$$

$$\text{Integrate } \mu : \mathbf{G}^* = a + b\mathbf{G}$$

$$\mu = (\text{Pedigree base}) - (\text{Genomic base})$$

Tries to put \mathbf{G} and \mathbf{A} in the same scale

Single-step

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z} \\ \mathbf{Z}'\mathbf{X} & \mathbf{Z}'\mathbf{Z} + \mathbf{H}^{-1} \frac{\sigma_e^2}{\sigma_u^2} \end{bmatrix} \begin{bmatrix} \hat{\boldsymbol{\beta}} \\ \hat{\mathbf{u}} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{y} \\ \mathbf{Z}'\mathbf{y} \end{bmatrix}$$

ssGBLUP

Misztal et al. (2009)
Legarra et al. (2009)
Aguilar et al. (2010)
Christensen & Lund (2010)

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z}\mathbf{M} & \mathbf{X}'_n\mathbf{Z}_n \\ \mathbf{M}'\mathbf{Z}'\mathbf{X} & \mathbf{M}'\mathbf{Z}'\mathbf{Z}\mathbf{M} + \mathbf{I} \frac{\sigma_e^2}{\sigma_\alpha^2} & \mathbf{M}'_n\mathbf{Z}'_n\mathbf{Z}_n \\ \mathbf{Z}'_n\mathbf{X}_n & \mathbf{Z}'_n\mathbf{Z}_n\mathbf{M}_n & \mathbf{Z}'_n\mathbf{Z}_n + \mathbf{A}^{nn} \frac{\sigma_e^2}{\sigma_g^2} \end{bmatrix} \begin{bmatrix} \hat{\boldsymbol{\beta}} \\ \hat{\boldsymbol{\alpha}} \\ \hat{\boldsymbol{\epsilon}} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{y} \\ \mathbf{M}'\mathbf{Z}'\mathbf{y} \\ \mathbf{Z}'_n\mathbf{y}_n \end{bmatrix}$$

ssSNPBLUP or ssBR

Fernando et al. (2014)
Liu et al. (2014)
Mantysaari & Strandén (2016)

Fernando et al. *Genetics Selection Evolution* 2014, **46**:50
<http://www.gsejournal.org/content/46/50>

equation (3) results in the usual non-genomic MME for the BVM.

Theory underlying SSBV-BLUP

Legarra et al. [11] proposed an ingenious strategy to combine information from genotyped and non-genotyped animals in a single BLUP analysis based on a BVM, which we refer to as SSBV-BLUP. Suppose \mathbf{g} is partitioned as:

$$\mathbf{g} = \begin{bmatrix} \mathbf{g}_1 \\ \mathbf{g}_2 \end{bmatrix} = \begin{bmatrix} \mathbf{g}_1 \\ \mathbf{T}_2\boldsymbol{\alpha} \end{bmatrix},$$

We confirmed that regular ssGBLUP and ssBR with an extra polygenic effect led to the same predictions.



J. Dairy Sci. 101:10082–10088
<https://doi.org/10.3168/jds.2018-14913>

© 2018, The Authors. Published by FASS Inc. and Elsevier Inc. on behalf of the American Dairy Science Association®.
This is an open access article under the CC BY-NC-ND license (<http://creativecommons.org/licenses/by-nc-nd/4.0/>).

Short communication: Genomic prediction using different single-step methods in the Finnish red dairy cattle population

H. Gao,*† M. Koivula,‡ J. Jensen,* I. Strandén,‡ P. Madsen,* T. Pitkänen,‡ G. P. Aamand,† and E. A. Mäntysaari‡

*Center for Quantitative Genetics and Genomics, Department of Molecular Biology and Genetics, Aarhus University, DK-8830 Tjele, Denmark

†Nordic Cattle Genetic Evaluation, DK-8200 Aarhus, Denmark

‡Natural Resources Institute Finland (Luke), FIN-31600 Jokioinen, Finland

Bases for Genomic Predictions

Bases for Genomic Prediction

Andres Legarra Daniela A.L. Lourenco Zulma G. Vitezica

2022-05-11





UNIVERSITY OF
GEORGIA
College of Agricultural &
Environmental Sciences

Quality Control of SNP data and creation of genomic matrices with BLUPF90 software

SNP file

SNP

ANIMAL

025	1101011110511110111110010001221151205122125022511110250122010201021000221121025000122010
036	21101101022012122222012101222010120222111112021222111112102020101101020111112011012110:
050	121010021112021111200021212222100021122122122110000020220000211022122212122020001112020:
054	120001200220121211100121002222110211221102011212221200220021212121111202112022002022100:
066	200002020221021221120022001222211101220202110202222020220001222121011201021022010011010:
097	101102120220121122111021001111100102211212022111111020221001201222012111021021021012000:
101	121002120220011221100011112220100101120112121211121201221002102002021211222022010022110:
151	111001020221220210201011012220200121221111221221121111222002201112011212111022000022012:
172	211012020211112101211021102220101001221212221102220201221020212112010211122022112011010:
224	220001110221012210101021102520201112120222122212220110121011102220050210121022010022125:
277	210102200121221211212021012222002012210212110201121021221002211011020211021112021012010:
314	122011120122220210210010002121001120120202001210020021210011201022021212111022010101100:
419	221112210121120222221022102110201021121211122000000111220002211122020222112120012121110:
439	200202100122121210101021012221101112220202022110010111210011201022012220211021010011020:
456	1200010202211122001010210022110002022212122222001011022111021201201121221111102112010:
501	111000021221121201212121002221101202222101022112222110220011202110020201102022100021020:
571	110000120202200221212022001210200011122110110222221200220020212001010212121022102010110:
579	11210021021001010111102200222120002221111202022222110222101202012111222111112011011020:
581	2110020252100122120201100220020112512121502252222250221011201121051202222112111012110:
657	110011120220111211101020012221000112221212021211121200220012202220022212212112001112011:
660	210002120221120221121021012221011012221222121211120201221012201121111211112022000012101:
730	210002020220020222220012002220001220222220021102252200122001202111151001012022001012025:
732	21210212521002201200012101121201215110215122521211150220011102111050202221122011022010:
764	11110212520012212211020001220201225222115021522221150220110202120050202022022111112110:
780	121101021122220210101022002221201201121221012111110111221020202001010112212121002021021:
800	22100012022122221020202110222110101211202212022222200221002211121021202011022010111010:
816	11000122022012122011002201112110001102112212122002011222200222111021111212022011022010:
832	12101001112001121111002111222011111212222121020111102022100211222100121211112101211110:
900	210100110220122121211021102121012120221212121101111110221001202121110211011021100022020:
901	121001020221121212210010002120201111221112122001111110221002201022012212121021000012020:

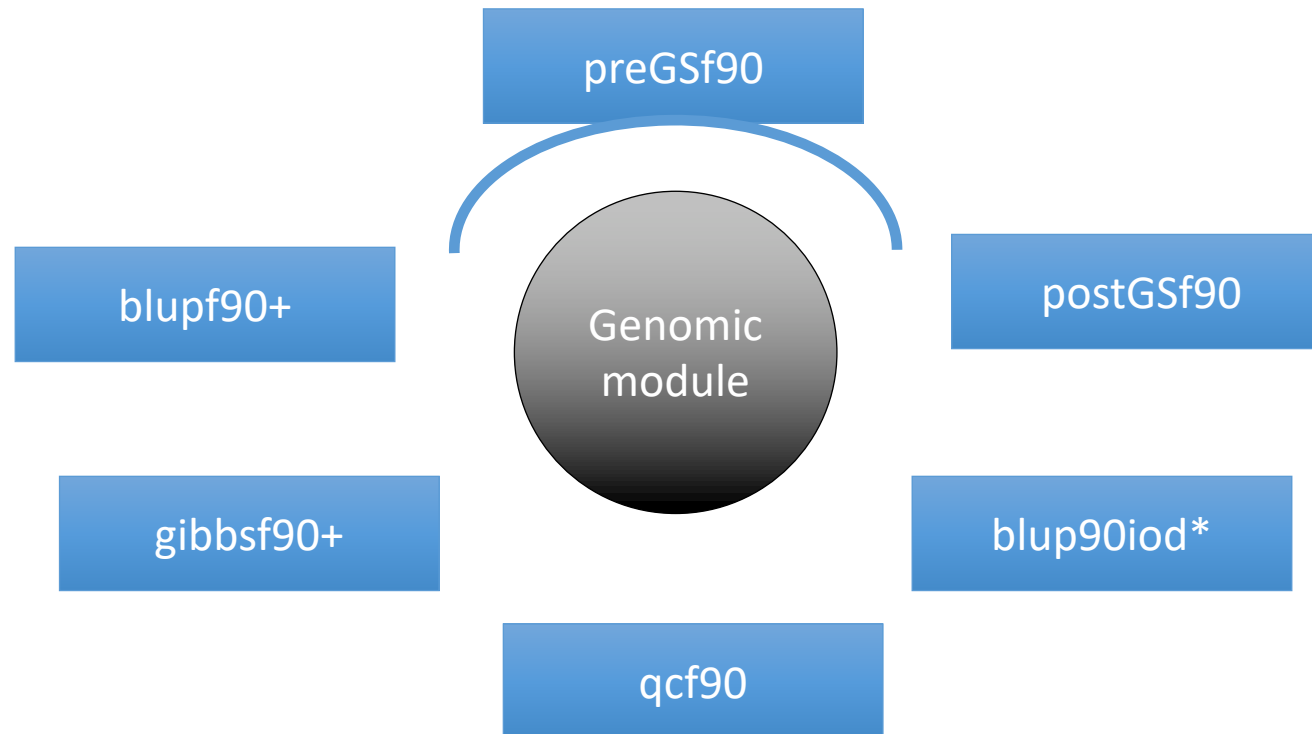
Quality control

- Call rate
 - Animals
 - SNP
- Minor Allele Frequency (MAF)
- Hardy-Weinberg Equilibrium (HWE)
- Non-mapped SNP
- Mendelian Conflicts
- Duplicate genotypes
- Linkage disequilibrium (LD)

**Which software in the
BLUPF90 family?**

preGSf90

- Interface program to the genomic module to process the genomic information in the BLUPF90 family of programs



preGSf90

- Performs Quality Control of SNP information
- Creates the genomic relationship matrix (**G**)
 - and relationships based on pedigree (**A₂₂**)
 - Inverse of relationship matrices



preGSf90

- Same parameter file as for all BLUPF90 programs
- Needs an extra OPTION in renf90.par
 - `OPTION SNP_file marker.geno`
- Reads 2 extra files (besides data and pedigree):
 - `marker.geno`
 - `marker.geno_XrefID` (created by renumf90)

`_XrefID` has 2 columns: Renumbered ID Original ID

Run renumf90 before preGSf90

- Use renumf90 for renumbering data and creating XrefID and files

```
EFFECT
1 cross alpha
RANDOM
animal
FILE
ped3.txt
FILE_POS
1 2 3 0 0
SNP_FILE
marker.geno
PED_DEPTH
0
(CO) VARIANCES
0.30
```

Parameter files

RENUMF90
renum.par

DATAFILE
phenotypes.txt
TRAITS
3
FIELDS_PASSED TO OUTPUT

WEIGHT(S)

RESIDUAL_VARIANCE
0.9038

EFFECT
1 cross alpha # mu
EFFECT
2 cross alpha # animal

RANDOM
animal

FILE
pedigree

SNP_FILE

marker.geno

(CO)VARIANCES
0.9951E-01

BLUPF90
renf90.par

DATAFILE
renf90.dat
NUMBER_OF_TRAITS
1
NUMBER_OF_EFFECTS
2
OBSERVATION(S)
1
WEIGHT(S)

EFFECTS: POSITIONS_IN_DATAFILE NUMBE
2 1 cross
3 15800 cross

RANDOM_RESIDUAL VALUES
0.90380
RANDOM_GROUP
2

RANDOM_TYPE
add_animal
FILE

renadd02.ped
(CO)VARIANCES
0.99510E-01

OPTION SNP_file marker.geno.

New pedigree file from RENUMF90

- 1 - **renumbered animal ID**
- 2 - parent 1 number or UPG
- 3 - parent 2 number or UPG
- 4 - 3 minus number of known parents
- 5 - known or estimated year of birth
- **6** - number of known parents
 if animal is genotyped 10 + number of known parents
- 7 - number of records
- 8 - number of progenies as parent 1
- 9 - number of progenies as parent 2
- **10** - **original animal ID**

SNP file, XrefID, and ped from renumf90

SNP File

First col: original ID

Second col: SNP genotypes {codes: 0,1,2, and 5 (missing)}

All SNP should start in the same column!!!

```
80 211010110020120110110101101111
8014 211101015111011202211101115111
516 211001012022520211202101211021
181 211101111122011205502000201010
```

No changes!!!

Renumbered ID

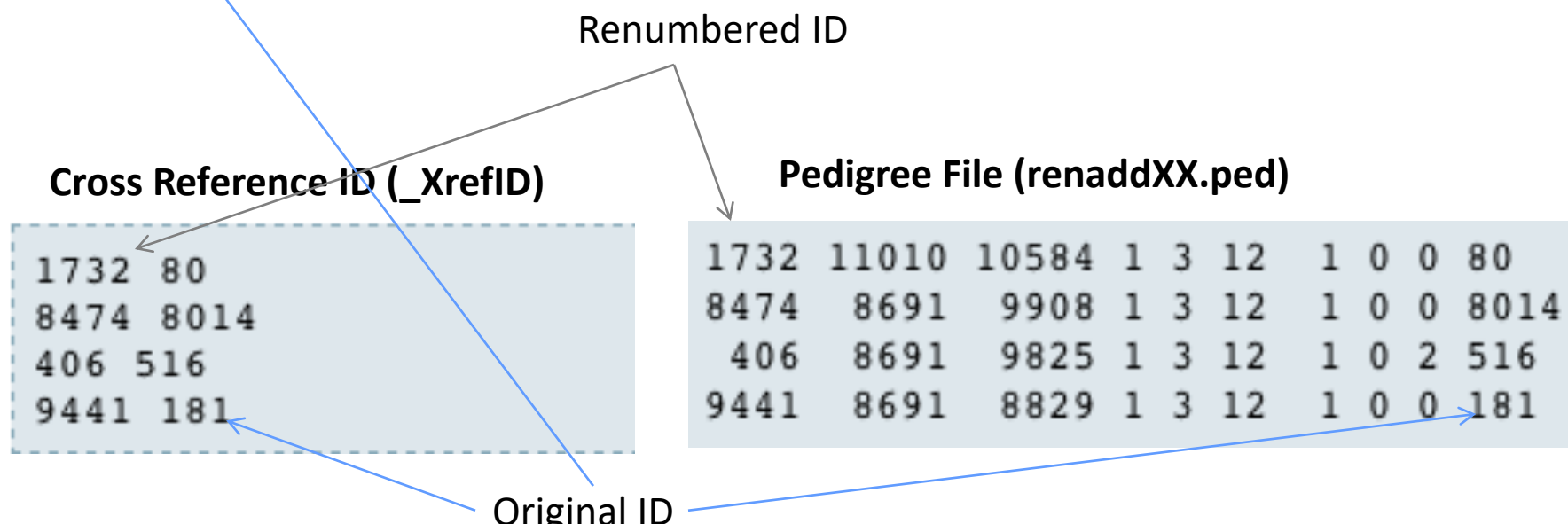
Cross Reference ID (_XrefID)

```
1732 80
8474 8014
406 516
9441 181
```

Pedigree File (renaddXX.ped)

```
1732 11010 10584 1 3 12 1 0 0 80
8474 8691 9908 1 3 12 1 0 0 8014
406 8691 9825 1 3 12 1 0 2 516
9441 8691 8829 1 3 12 1 0 0 181
```

Original ID



SNP map file

- OPTION map_file <file>
 - For GWAS and QC
- Format:
 - A header must be provided
 - Names for SNP, chromosome, and physical position are mandatory
 - SNPID for SNP
 - CHR for chromosome
 - POS for position

```
NUM CHR   POS      SNPID      NUM2
31428 14  7928189  ARS-BFGL-BAC-1020 2
32005 14  31819743 ARS-BFGL-BAC-10245 3
31371 14  6133529  ARS-BFGL-BAC-10345 4
31679 14  17544926 ARS-BFGL-BAC-10591 7
32053 14  34639444 ARS-BFGL-BAC-10867 8
31993 14  31267746 ARS-BFGL-BAC-10919 9
23506 10  18882288 ARS-BFGL-BAC-10952 10
23550 10  20609250 ARS-BFGL-BAC-10960 11
23566 10  21225382 ARS-BFGL-BAC-10975 12
23612 10  26527257 ARS-BFGL-BAC-10986 13
24705 10  78512500 ARS-BFGL-BAC-10993 14
24712 10  79252023 ARS-BFGL-BAC-11000 15
24732 10  80410977 ARS-BFGL-BAC-11003 16
24741 10  80783719 ARS-BFGL-BAC-11007 17
24827 10  84516867 ARS-BFGL-BAC-11025 18
25865 11  21276136 ARS-BFGL-BAC-11039 21
```

Saving 'clean' files

- OPTION saveCleanSNPs
- Save clean genotype data without excluded SNP and individuals
 - For example, for a SNP_file named *marker.geno*
- Clean files will be:
 - *marker.geno_clean*
 - *marker.geno_clean_XrefID*
- Removed SNP/animals will be output in files:
 - *marker.geno_SNPs_removed*
 - *marker.geno_Animals_removed*

Only QC in preGSf90

- preGSf90 does:
 - Quality control
 - Genomic relationship matrices and inverses
 - Inverse is costly
- How to do only QC avoiding the inverses:
 - `OPTION SNP_file marker.geno`
 - `OPTION saveCleanSNPs`
 - `OPTION stop_after_quality_control`

No QC in application programs

- ONLY use:
 - If QC was performed in a previous run
 - and “clean” genotype file is used
- OPTION SNP_file *marker.geno_clean*
- OPTION no_quality_control

Use in application programs


- Use `renumf90` to renumber and create XrefID and files

```
SNP_FILE  
marker.geno
```

```
EFFECT  
1 cross alpha  
RANDOM  
animal  
FILE  
ped3.txt  
FILE_POS  
1 2 3 0 0  
SNP_FILE  
marker.geno  
PED_DEPTH  
0  
(CO)VARIANCES  
0.30
```

- Run `preGSf90` with quality control, saving clean files
- Run further programs with clean files as needed
 - `blupf90+`, `gibbs2f90+`, ...

PreGSf90 wiki

 BLUPF90 [Log In](#)

[Media Manager](#) [Sitemap](#)

Trace: [start](#) · [application_programs](#) · [readme.pregsf90](#)

[readme.pregsf90](#)

PreGSF90 / PostGSF90

PreGSF90 is an interface program to the `genomic` module to process the genomic information for the BLUPF90 family of programs

This page also describes some options for PostGSF90 which is designed for genome-wide association study (GWAS).

Ignacio Aguilar and Ignacy Misztal, University of Georgia
email: iaguilar@inia.org.uy; ignacy@uga.edu
01/29/09 - 07/30/14

Summary

Program PreGSF90 helps to implement the genomic selection following the single-step methodology as presented by [Aguilar et al. 2010 JDS](#). In this methodology the relationship matrix **A** based on the pedigree information is replaced by matrix **H**, which combines the pedigree and genomic information.

The main difference between \mathbf{A}^{-1} and \mathbf{H}^{-1} is matrix of structure
$$\text{GimA22} = \text{inv}(\mathbf{G}) - \text{inv}(\mathbf{A}_{22}),$$
where **G** is a genomic relationship matrix and **A₂₂** is a relationship matrix for genotyped animals.

Efficient methods for the creation of the genomic relationship matrix, relationship based on pedigree and their inverses are described in [Aguilar et al., 2011 JABG](#). Program PreGSF90 could be run after `RENUMF90`. It is also run automatically by application programs like `BLUPF90`, `REMLF90`, `GIBBSxF90` or `BLUP90IOD` when their parameter file contains `OPTION SNP_file filename`.

Table of Contents

- ◊ [PreGSF90 / PostGSF90](#)
- ◊ [Summary](#)
- ◊ [Input files](#)
- ◊ [Output files](#)
- ◊ [Options for creation of genomic relationship Matrix \(G\)](#)
- ◊ [Quality Control \(QC\) for G](#)
- ◊ [Quality Control for Off-diagonal of A22 and G](#)
- ◊ [Options for H](#)
- ◊ [GWAS options \(PostGSF90\)](#)
- ◊ [Output files for GWAS \(postGSF90\)](#)
- ◊ [Misc options](#)
- ◊ [Save and Read options](#)
- ◊ [Save and Read intermediate files](#)
- ◊ [DEPRECATED OPTIONS](#)

preGSf90

- Performs Quality Control of SNP information
- Creates the genomic relationship matrix (**G**)
 - and relationships based on pedigree (**A₂₂**)
 - Inverse of relationship matrices



PreGSf90

- Created to construct the matrices using in ssGBLUP

$$\mathbf{H}^{-1} = \mathbf{A}^{-1} + \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{G}^{-1} - \mathbf{A}_{22}^{-1} \end{bmatrix}$$

$$\mathbf{G} \quad \mathbf{G}^{-1}$$

$$\mathbf{A}_{22} \quad \mathbf{A}_{22}^{-1}$$

$$\mathbf{G}^{-1} - \mathbf{A}_{22}^{-1}$$

Genomic Relationship Matrix - G


- $\mathbf{G} = \frac{\mathbf{ZZ}'}{2 \sum p_i(1-p_i)}$ (VanRaden, 2008)

- Z = matrix for SNP marker
- Dimension of Z = $n \times i$
- n animals
- i markers

Genotype Codes

- 0 – Homozygous
- 1 – Heterozygous
- 2 – Homozygous
- 5 – No Call (Missing)

SNP file



```
80 21101011002012011011010110111111211111210100
8014 21110101511101120221110111511112101112210100
516 21100101202252021120210121102111202212111101
181 21110111112201120550200020101022212211111100
```

PreGSf90

- Efficient methods
 - create the genomic relationship matrix and the relationship matrix based on pedigree
 - Invert the relationship matrices
- Computes statistics for the matrices
 - Means, Var, Min, Max
 - Correlations between diagonals
 - Correlations for off-diagonals
 - Correlations for the full matrices
 - Regression coefficients

Genomic Matrix default options

$$\mathbf{G}_0 = \frac{\mathbf{Z}\mathbf{Z}'}{2 \sum p_i(1 - p_i)} = \frac{(\mathbf{M} - 2\mathbf{P})(\mathbf{M} - 2\mathbf{P})'}{2 \sum p_i(1 - p_i)}$$

(VanRaden, 2008)

- With:
 - \mathbf{Z} centered using current allele frequencies:
 - Current genotyped animals

Genomic Matrix Options

- OPTION whichfreq *x*
 - 0: read from file *freqdata* or other specified name (needs OPTION FreqFile)
 - 1: 0.5
 - 2: current calculated from genotypes (default)
- OPTION FreqFile *file*
 - Reads allele frequencies from a file

Genomic Matrix default options

- **Blending** - to avoid singularity problems

$$\mathbf{G} = 0.95 * \mathbf{G}_0 + 0.05 * \mathbf{A}_{22}$$

- OPTION AlphaBeta 0.95 0.05 #(default)
- Beta may vary from 0.2 to 0.01

Genomic Matrix default options

- **Tuning**

- Adjust \mathbf{G} to have mean of diagonals and off-diagonals equal to \mathbf{A}_{22}
- OPTION tunedG 2 #(default) Chen et al. (2011)

- Base of GBLUP is *genotyped* animals
- Base of pedigree is *founders of the pedigree*
- For SSGBLUP modelled as a mean for genotyped animals
 - $p(\mathbf{u}_2) = N(\mathbf{1}\mu, \mathbf{G})$
 - Integrate $\mu : \mathbf{G}^* = \mathbf{1}\mathbf{1}'\lambda + (1 - \lambda/2)\mathbf{G}$
 - $\mu = (\text{Genomic base}) - (\text{Pedigree base})$
 - Vitezica et al. 2011

Options for matching \mathbf{G} to \mathbf{A}_{22}

- OPTION tunedG x
 - 0: no adjustment
 - 1: $\text{mean}(\text{diag}(\mathbf{G}))=1$, $\text{mean}(\text{offdiag}(\mathbf{G}))=0$
 - 2: $\text{mean}(\text{diag}(\mathbf{G}))=\text{mean}(\text{diag}(\mathbf{A}_{22}))$, $\text{mean}(\text{offdiag}(\mathbf{G}))=\text{mean}(\text{offdiag}(\mathbf{A}_{22}))$ (default)
 - 3: $\text{mean}(\mathbf{G})=\text{mean}(\mathbf{A}_{22})$
 - 4: Use Fst adjustment. Powell et al. (2010) & Vitezica et al. (2011)

$$\lambda = \frac{1}{n^2} \left(\sum_i \sum_j \mathbf{A}_{22ij} - \sum_i \sum_j \mathbf{G}_{ij} \right) \quad \mathbf{G}^* = \mathbf{1}\mathbf{1}'\lambda + (1 - \lambda/2)\mathbf{G}$$

Storing and Reading Matrices

- preGSf90 saves $\mathbf{G}^{-1} - \mathbf{A}_{22}^{-1}$ by default (file: GimA22i)

To save 'raw' \mathbf{G} :

- OPTION saveG [all]
 - If the optional *all* is present all intermediate \mathbf{G} matrices will be saved!!!

To save \mathbf{G}^{-1}

- OPTION saveGInverse
 - Only the final \mathbf{G} , after blending, scaling, etc. is inverted !!!

To save \mathbf{A}_{22} and inverse

- OPTION saveA22 and OPTION saveA22Inverse

Saves in binary format!!!

Storing with Original IDs

- Some matrices could be stored in text files with the original IDs extracted from *renaddxx.ped* created by the RENUMF90 program (col #10)
- For example:
 - OPTION saveGOrig
 - OPTION saveDiagGOrig
 - OPTION saveHinvOrig
- Values
 - origID_i, origID_j, val

Genomic Matrix - Population structure

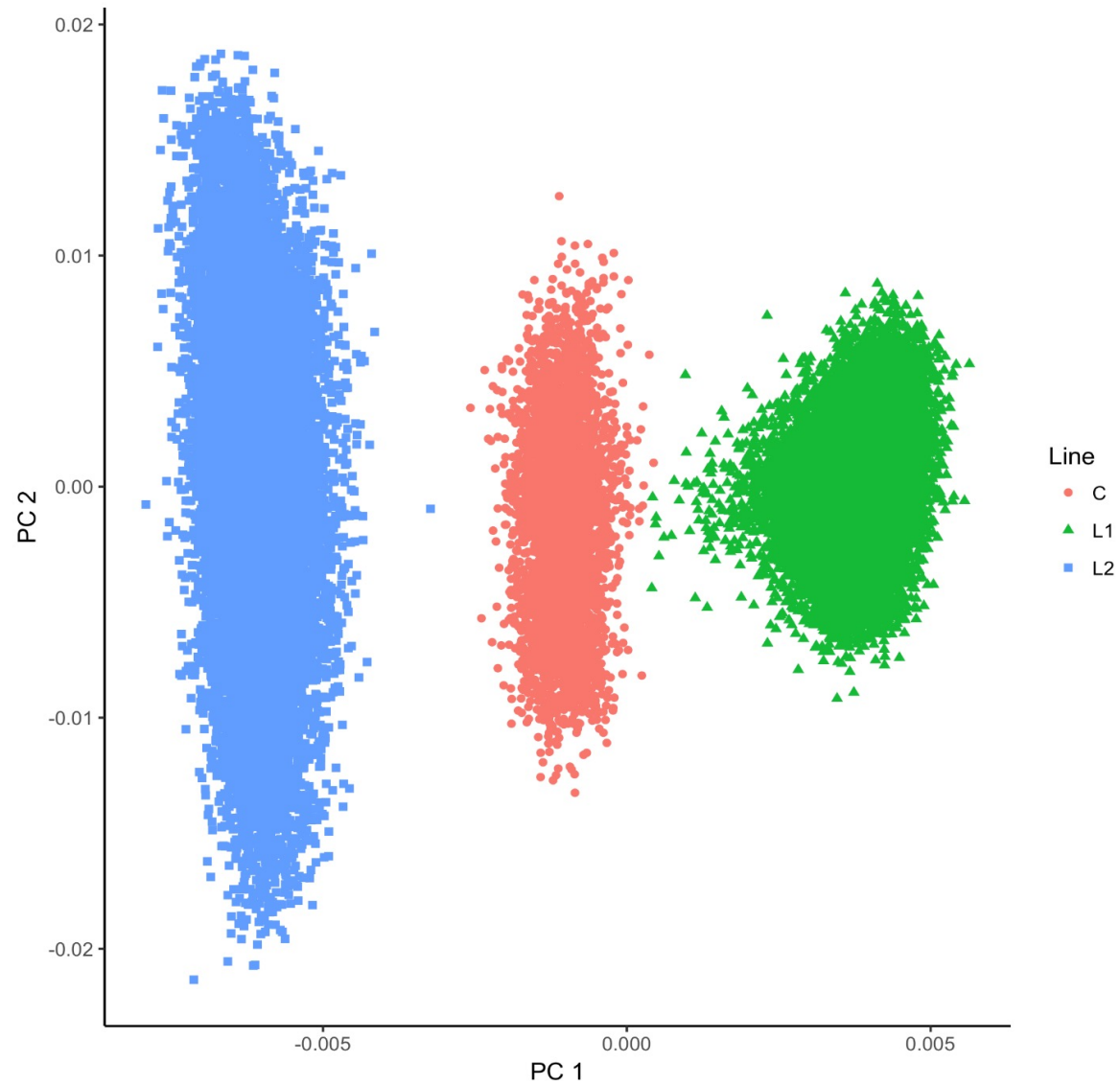
```
OPTION plotpca
```

Plot first two principal components to look for stratification in the population.

```
OPTION extra_info_pca file col
```

Reads from *file* the column *col* to plot with different colors for different classes.

Genomic Matrix - Population structure



Tricks to setup **G** for GBLUP #1

- Tricks are needed because preGSf90 is set up for ssGBLUP

1) Use a dummy pedigree

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

...

2) Use PED_DEPTH 1 in renumf90

3) Change blending parameters

- OPTION AlphaBeta 1.00 0.00 $\rightarrow G = 1.00*\mathbf{G} + 0.00*\mathbf{I}$
- OPTION AlphaBeta 0.95 0.05 $\rightarrow G = 0.95*\mathbf{G} + 0.05*\mathbf{I}$

4) No adjustment for compatibility with \mathbf{A}_{22}

- OPTION tunedG 0

Tricks to setup **G** for GBLUP #2

1) In renum.par, remove any information about the pedigree. Example:

```
FILE
pedigree.txt

FILE_POS
1 2 3 0 0

PED_DEPTH
3
```

3) Change blending parameters

- OPTION AlphaBeta 1.00 0.00 → $G = 1.00 * G + 0.00 * I$
- OPTION AlphaBeta 0.95 0.05 → $G = 0.95 * G + 0.05 * I$

4) No adjustment for compatibility with A_{22}

- OPTION tunedG 0

PreGSf90 inside BLUPF90 ??

- Almost all programs from BLUPF90 support creating genomic relationship matrices
- OPTION SNP_file xxxx

- Why preGSF90 ?
 - Same genomic relationship matrix for several models, traits, etc.
 - Just do it once and store GimA22i or Gi and A22i separate

Use in application programs

- Use renumf90 for renumbering and creating of XrefID and files
SNP_FILE
marker.geno
- Option 1:
run blupf90+ right after renumf90
- Option 2:
run preGSf90 with quality control, saving clean files
run blupf90+ with clean files