

meBLUP: a single step gBLUP combining pedigree and SNP information without genomic inverse

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From Mixed Models (MME) to Model Equations (ME)

HENDERSON'S **mixed model equations**

$$\begin{pmatrix} X^T D^{-1} X & X^T D^{-1} Z \\ Z^T D^{-1} X & Z^T D^{-1} Z + G^{-1} \end{pmatrix} \begin{pmatrix} \beta \\ u \end{pmatrix} = \begin{pmatrix} X^T D^{-1} y \\ Z^T D^{-1} y \end{pmatrix}$$

for estimating β and u , which only involve the inverse G^{-1} .
The vector \hat{u} (containing the estimated breeding values) is usually called the **best linear unbiased predictor (BLUP)** of the random effects.

The model equation formulation of a mixed model

FELLNER observed, that Henderson's mixed model equations are just the normal equations of a simple form giving the **ME formulation**:

$$Ax = b,$$

where

$$A = \begin{pmatrix} X & Z \\ 0 & -I \end{pmatrix}, x = \begin{pmatrix} \beta \\ u \end{pmatrix}, b = \begin{pmatrix} y \\ 0 \end{pmatrix},$$

Thus computing this \hat{u} of the ME formulation will be the same as that of the mixed model.

Note here is no G^{-1} .

Model equations animal model BLUP $Ax = b$

- ▶ setting up and solving ME $Ax = b$ equivalent to using MME

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<http://www.wcgalp.org/system/files/proceedings/2018/blup-without-inverse-relationship-matrix.pdf>

Setting up model equations in $Ax = b$

```
snp1-pedi.txt      snp1-data.txt      snp4.txt
1 6 7              bf  wt  sex Animal  SNPS
2 1 7              10.5 90.1 1 1      1 0112
3 1 4              13.2 87.3 1 2      2 2101
4 2 1              12.6 93.5 2 3      3 1022
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```
ANIMAL 8 13 14    1.4 -0.7 -0.7    2
ANIMAL 9 8 14     1.4 -0.7 -0.7    2
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ANIMAL 12 10 11   1.4 -0.7 -0.7    2
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```

```
SNP 16      1.0      3
SNP 17      1.0      3
SNP 18      1.0      3
SNP 19      1.0      3
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```
SNP 16           1.0           3
SNP 17           1.0           3
SNP 18           1.0           3
SNP 19           1.0           3
```

```
GENOMIC 16 17 18 19 8 20 0.0 1.0 1.0 2.0 -1.0 1.0 4
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solve $Ax = b$ by PCG

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- ▶ inverses have max. dimension $n_{trait} * n_{trait}$

Implementation, Scope and Performance

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- ▶ preconditioning: diagonal blocks ($n_{trait} * n_{trait}$)
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33mio	1	20mio	41K@116000	4	18.8sec

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33mio	2	53mio	41K@116000	1	52.7sec

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33mio	2	53mio	41K@116000	1	52.7sec
33mio	2	53mio	41K@135000	1	60.1sec

Validation me_BLUP vs ss_BLUP

Correlations true/estimated breeding value

for the last generation (without phenotypes)

h ²	number of SNPs used			method
	40000	50000	60000	
0.05	0.435+	0.446+	0.450-	meBLUP
0.05	0.426	0.445	0.455	ssBLUP
0.10	0.425+	0.429-	0.446-	meBLUP
0.10	0.418	0.436	0.459	ssBLUP
0.20	0.591+	0.593+	0.599+	meBLUP
0.20	0.581	0.586	0.597	ssBLUP
0.30	0.528-	0.537-	0.555-	meBLUP
0.30	0.530	0.551	0.584	ssBLUP
0.40	0.572-	0.591-	0.598-	meBLUP
0.40	0.577	0.615	0.635	ssBLUP
0.50	0.617+	0.640-	0.650-	meBLUP
0.50	0.616	0.655	0.679	ssBLUP
0.60	0.717+	0.735-	0.745-	meBLUP
0.60	0.704	0.745	0.767	ssBLUP

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- ▶ The resulting linear system is always positive definite and can be solved by conjugate gradients, without slow convergence.
- ▶ The ME approach leads to a different way of combining pedigree data and genomic data. It is not equivalent to ssBLUP.

Outlook

- ▶ estimate (co)variances SNP, Genomic, scaling vector
- ▶ nonlinear optimization as initial part of PCG
- ▶ weighing of sources of SNP vs pedigree
- ▶ finding ME that inherently do not double count

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Thank you for your attention!