## Numerically Efficient Fully Orthogonalized Single-step SNP-BLUP

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## Challenge

Computations of large animal breeding evaluations become numerially challenging when number of genotyped animals increases.

## Background: Single-step Genomic BLUP (ssGBLUP)

- A mixed model equation (MME).
- Combines pedigree (A) and genomic marker relationship information $\left(\mathbf{G}_{g}\right)$ through Single-step relationship matrix H.
- Requires inversion of full genomic relationship matrix $\mathbf{G}_{g}$
- Inversion $\mathbf{G}_{g}^{-1}$ becomes a bottleneck when number of genotyped increases.


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Single-step relationship matrix $\mathbf{H}$

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- Requires inversion of full genomic relationship matrix G
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$$
\mathbf{H}^{-1}=\mathbf{A}^{-1}+\left[\begin{array}{cc}
\mathbf{0} & \mathbf{0} \\
\mathbf{0} & \mathbf{G}_{w}^{-1}-\left(\mathbf{A}_{22}\right)^{-1}
\end{array}\right]
$$

Pedigree relationship matrix:

$$
\mathbf{A}^{-1}=\left[\begin{array}{ll}
\mathbf{A}^{11} & \mathbf{A}^{12} \\
\mathbf{A}^{21} & \mathbf{A}^{22}
\end{array}\right] \begin{aligned}
& (1=\text { non-genotyped }) \\
& (2= \\
& \text { genotyped })
\end{aligned}
$$

Adjusted genomic relationship matrix:

$$
\mathbf{G}_{w}=(1-w) \mathbf{G}_{g}+w \mathbf{A}_{22}
$$

Genomic relationship matrix:

$$
\mathbf{G}_{g}=\mathbf{Z}_{m} \mathbf{Z}_{m}^{\prime}
$$

$\mathbf{Z}_{m}$ is centered and scaled marker matrix

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> - Linearly equivalent MMEs [1]: decomposition of relationship matrix $\mathbf{G}$ of random effects $\widehat{\mathbf{u}}$ as

$$
\mathbf{G}=\mathbf{M} \widetilde{\mathbf{G}} \mathbf{M}^{\prime}
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- Alternative MME: parts of decomposition attached
to new model matrix $\mathbf{Z}=\mathbf{Z} \mathbf{M}$ and new set of
random effects $\widetilde{\mathrm{u}}$ related through $\widetilde{\mathbf{G}}$.
- Original MME solved from $\widehat{\mathbf{u}}=\mathbf{M u}$
- Inversion $\mathrm{G}^{-1}$ avoided if $\widetilde{G}=\mathbb{I}$, i.e. $\mathbb{u}$ "orthogonal".
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\end{array}\right]^{-1}\left[\begin{array}{l}
\mathbf{X}^{\prime} \mathbf{R}^{-1} \mathbf{y} \\
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\end{array}\right]
$$

## - Alternative MME: parts of decomposition attached

to new model matrix $\mathbf{Z}=\mathbf{Z M}$ and new set of random effects $\widetilde{u}$ related through $\widetilde{\mathbf{G}}$.

```
Original MME solved from }\widehat{\mathbf{u}}=\mathbf{M\widetilde{u}
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Linearly equivalent MME

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\end{array}\right]
$$

Calculation of original effects:

$$
\left[\begin{array}{c}
\hat{\mathbf{b}} \\
\widehat{\mathbf{u}}
\end{array}\right]=\left[\begin{array}{cc}
\mathbf{I}_{b} & \mathbf{0} \\
\mathbf{0} & \mathbf{M}
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- Unlimited number of possible decompositions of ssGBLUP relationship matrix [1]:

$$
\mathbf{I}=\mathbb{M}_{i} \widetilde{\mathrm{G}}_{i} \mathrm{M}_{i}
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- Linearly equivalent ssGBLUP formulations with varying number of random effects ũ without or with marker effects (ssSNP-BLUP).
- Numerical solutions same as with original ssGBLUP.
- Fully orthogonalized ( $\widetilde{\mathbf{G}}=\mathbf{I})$ ssSNP-BLUPs:
- almost identical number of iterations when solved using iterative methods, e.g. PCG.


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Linearly equivalent ssGBLUP formulations with varying number of random effects ü without or with marker effects (ssSNP-BLUP)
$\mathbf{M}_{1}=\left[\begin{array}{cc}\mathbf{I}_{1} & \mathbf{A}_{\text {impp }} \\ \mathbf{0} & \mathbf{I}_{2}\end{array}\right]$
$\mathbf{M}_{2}=\left[\begin{array}{ccc}\mathbf{I}_{1} & \sqrt{w} \mathbf{A}_{\text {imp }} & \sqrt{1-w} \mathbf{Z}_{\text {imp }} \\ \mathbf{0} & \sqrt{w} \mathbf{I}_{2} & \sqrt{1-w} \mathbf{Z}_{m}\end{array}\right]$
$\mathbf{M}_{3}=\left[\begin{array}{ccc}\sqrt{1-w} \mathbf{I}_{1} & \sqrt{w} \mathbf{J}_{1} & \sqrt{1-w} \mathbf{Z}_{m m p} \\ \mathbf{0} & \sqrt{w} \mathbf{J}_{2} & \sqrt{1-w} \mathbf{Z}_{m}\end{array}\right]$
$\mathbf{M}_{4}=\left[\begin{array}{ccc}\left(\mathrm{L}_{1}{ }^{\prime}\right)^{-1} & \sqrt{w} \mathbf{A}_{\mathbf{A}_{m p}} \mathbf{J}_{\mathbf{2}}\left(\mathrm{L}^{\prime}\right)^{-1} & \sqrt{1-w} \mathbf{Z}_{\text {mpp }} \\ \mathbf{0} & \sqrt{w} \mathbf{J}_{2}\left(\mathrm{~L}^{\prime}\right)^{-1} & \sqrt{1-w} \mathbf{Z}_{m}\end{array}\right]$
$\mathbf{M}_{\mathrm{S}}=\left[\begin{array}{ccc}\sqrt{1-w}\left(\mathrm{~L}_{1}^{\prime}\right)^{-1} & \sqrt{w} \mathbf{J}_{1}\left(\mathrm{~L}^{\prime}\right)^{-1} & \sqrt{1-w} \mathbf{Z}_{i m p} \\ \mathbf{0} & \sqrt{w} \mathbf{J}_{2}\left(\mathrm{~L}^{\prime}\right)^{-1} & \sqrt{1-w} \mathbf{Z}_{m}\end{array}\right]$
$\mathbf{M}_{6}=\left[\begin{array}{ccc}\left(\mathrm{L}_{1}{ }^{\prime}\right)^{-1} & \sqrt{w} \mathbf{A}_{\text {imp }} \hat{\mathbf{J}}_{2}\left(\hat{\mathrm{~L}}^{\prime}\right)^{-1} & \sqrt{1-w} \mathbf{Z}_{i m p} \\ 0 & \sqrt{w} \mathbf{J}_{2}\left(\mathrm{~L}^{\prime}\right)^{-1} & \sqrt{1-w} \mathbf{Z}_{m}\end{array}\right]$


## Numerical solutions same as with original ssGBLUP

- Fully orthogonalized $(\widetilde{\mathbf{G}}=\mathbf{I})$ ssSNP-BLUPs:
> - almost identical number of iterations when solved using iterative methods, e.g. PCG.


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$\mathbf{G}_{w}=w \mathbf{A}_{22}+(1-w) \mathbf{Z}_{m} \mathbf{Z}_{m}^{\prime}$
$\mathbf{z}_{i m p}=\mathbf{A}_{i m p} \mathbf{Z}_{m}$
Alternative RPG form
$J_{1}$ picks non-genot.
$\mathrm{J}_{2}$ picks genotyped
Orthogonalized $\widetilde{\mathbf{G}}_{2}$
$\mathrm{L}_{1}=\operatorname{chol}\left(\mathrm{A}^{11}\right)$
Orthogonalized $\tilde{\mathbf{G}}_{3}$
Reduced $\mathbf{A}_{22}$ of $\widetilde{\mathbf{G}}_{4}$
$\widehat{\mathrm{L}}=\operatorname{chol}\left(\widehat{\mathbf{A}}^{-1}\right)$, where
$\widehat{\mathbf{A}}$ : genot. and ancestors


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\end{array}\right] \\
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\mathbf{0} & \sqrt{w} \mathbf{I}_{2} & \sqrt{1-w} \mathbf{Z}_{m}
\end{array}\right] \\
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\mathbf{M}_{4}
\end{array}=\left[\begin{array}{ccc}
\left(\mathrm{L}_{1}^{\prime}\right)^{-1} & \sqrt{w} \mathbf{A}_{i m p} \mathbf{J}_{2}\left(\mathrm{~L}^{\prime}\right)^{-1} & \sqrt{1-w} \mathbf{Z}_{i m p} \\
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## Result 1: Iteration Convergence of Fully Orthogonalized $\left(\mathbf{G}=\mathbf{M}_{i} \mathbf{M}_{i}^{\prime}\right)$ MMEs

- Assuming singular values of (here usually) wide matrices $\mathbf{M}_{i}$ to be known: diagonal $\mathbf{D}_{i}$.


## Since $\mathrm{M}_{i}$ are all decompositions of same G,

 singular values are shared: $\mathbf{D}_{i}=\mathbf{D}$```
If fixed effects are neglected, Z = I, and single trait
case is assumed, all fully orthogonalized MME
share eigenvalues (D2}+\lambda\mathbf{I}\mathrm{ and rest equal }
```

Numbers of distinct (approximate) eigenvalues are thus same $\Rightarrow$ this explains same iteration counts.

Singular value decomposition of

$$
\mathbf{M}_{i}=\mathbf{U}_{i}\left[\begin{array}{ll}
\mathbf{D}_{i} & 0
\end{array}\right] \mathbf{V}_{i}^{\prime}
$$

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If fixed effects are neglected, $\mathbf{Z}=\mathbf{I}$, and single trait case is assumed, all fully orthogonalized MME share eigenvalues $\mathbf{D}^{2}+\lambda I$ and rest equal $\lambda$.

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Eigendecomposition of G

$$
\mathbf{G}=\mathbf{M}_{i} \mathbf{M}_{i}^{\prime}=\mathbf{U}_{i} \mathbf{D}_{i}^{2} \mathbf{U}_{i}^{\prime}=\mathbf{U D}^{2} \mathbf{U}^{\prime}
$$

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Eigendecomposition of $\mathbf{G}$

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\mathbf{G}=\mathbf{M}_{i} \mathbf{M}_{i}^{\prime}=\mathbf{U}_{i} \mathbf{D}_{i}^{2} \mathbf{U}_{i}^{\prime}=\mathbf{U D}^{2} \mathbf{U}^{\prime}
$$

Eigendecomposition of "MME"

$$
\widetilde{\mathbf{Z}}_{i}^{\prime} \widetilde{\mathbf{Z}}_{i}+\lambda \mathbf{I} \approx \mathbf{V}_{i}\left[\begin{array}{cc}
\mathbf{D}^{2}+\lambda \mathbf{I} & \mathbf{0} \\
\mathbf{0} & \lambda \mathbf{I}
\end{array}\right] \mathbf{V}_{i}^{\prime}
$$

## Result 1: Iteration Convergence of Fully Orthogonalized ( $\mathbf{G}=\mathbf{M}_{i} \mathbf{M}_{i}^{\prime}$ ) MMEs

- Assuming singular values of (here usually) wide matrices $\mathbf{M}_{i}$ to be known: diagonal $\mathbf{D}_{i}$.
- Since $\mathbf{M}_{i}$ are all decompositions of same $\mathbf{G}$, singular values are shared: $\mathbf{D}_{i}=\mathbf{D}$.
- If fixed effects are neglected, $\mathbf{Z}=\mathbf{I}$, and single trait case is assumed, all fully orthogonalized MME share eigenvalues $\mathbf{D}^{2}+\lambda \mathbf{I}$ and rest equal $\lambda$.
- Numbers of distinct (approximate) eigenvalues are thus same $\Rightarrow$ this explains same iteration counts.

Singular value decomposition of

$$
\mathbf{M}_{i}=\mathbf{U}_{i}\left[\begin{array}{ll}
\mathbf{D}_{i} & 0
\end{array}\right] \mathbf{V}_{i}^{\prime}
$$

## Eigendecomposition of $\mathbf{G}$

$$
\mathbf{G}=\mathbf{M}_{i} \mathbf{M}_{i}^{\prime}=\mathbf{U}_{i} \mathbf{D}_{i}^{2} \mathbf{U}_{i}^{\prime}=\mathbf{U D}^{2} \mathbf{U}^{\prime}
$$

Eigendecomposition of "MME"

$$
\widetilde{\mathbf{Z}}_{i}^{\prime} \widetilde{\mathbf{Z}}_{i}+\lambda \mathbf{I} \approx \mathbf{V}_{i}\left[\begin{array}{cc}
\mathbf{D}^{2}+\lambda \mathbf{I} & \mathbf{0} \\
\mathbf{0} & \lambda \mathbf{I}
\end{array}\right] \mathbf{V}_{i}^{\prime}
$$

## Numerical efficiency:

- Linearly equivalent ssSNP-BLUPs: same solution and convergence, which formulation to choose?
- Select numerically efficient formulation
- How to get a more efficient formulation?

Group of non-genotyped (1) split to non-ancestors (n) and ancestors (a) of genotyped (2).

Smaller pedigree of non-genotyped ancestors (a) and genotyped (2) individuals can be formed with
$\widehat{\mathrm{A}}^{-1}=\widehat{\mathrm{LL}}^{\prime}$ as its pedigree relationship matrix.

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## Smaller pedigree of non-genotyped ancestors (a) and genotyped (2) individuals can be formed with <br> $\widehat{\mathbf{A}}^{-1}=\widehat{L L^{\prime}}$ as its pedigree relationship matrix.

$$
\begin{gathered}
\text { "Cholesky" of }= \\
\mathbf{L}=\left[\begin{array}{ccc}
\mathbf{L}_{n n} & \mathbf{0} & \mathbf{0} \\
\mathbf{L}_{a n} & \mathbf{L}_{a a} & \mathbf{L}_{a 2} \\
\mathbf{L}_{2 n} & \mathbf{L}_{2 a} & \mathbf{L}_{22}
\end{array}\right]=\left[\begin{array}{ccc}
\mathbf{L}_{n n} & \mathbf{0} & \mathbf{0} \\
\mathbf{L}_{a n} & {[\widehat{\mathbf{L}}]} \\
\mathbf{L}_{2 n} &
\end{array}\right]
\end{gathered}
$$

## Numerical efficiency: Smaller Pedigree of Ancestors of Genotyped: $\widehat{\mathbf{A}}^{-1}$

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- Smaller pedigree of non-genotyped ancestors (a) and genotyped (2) individuals can be formed with $\widehat{\mathbf{A}}^{-1}=\widehat{\mathbf{L}} \widehat{\mathbf{L}}^{\prime}$ as its pedigree relationship matrix.

$$
\left.\mathbf{L}=\left[\begin{array}{ccc}
\mathbf{L}_{n n} & \mathbf{0} & \mathbf{0} \\
\mathbf{L}_{a n} & \mathbf{L}_{a a} & \mathbf{L}_{a 2} \\
\mathbf{L}_{2 n} & \mathbf{L}_{2 a} & \mathbf{L}_{22}
\end{array}\right]=\left[\begin{array}{ccc}
\mathbf{L}_{n n} & \mathbf{0} & \mathbf{0} \\
\mathbf{L}_{a n} & {[\widehat{\mathbf{L}}}
\end{array}\right]\right]
$$

"Cholesky" of = '
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Smaller pedigree of non-genotyped ancestors (a) and genotyped (2):

$$
\widehat{\mathbf{L}}=\left[\begin{array}{l}
\widehat{\mathbf{L}}_{a} \\
\widehat{\mathbf{L}}_{2}
\end{array}\right]=\left[\begin{array}{ll}
\mathbf{L}_{a a} & \mathbf{L}_{a 2} \\
\mathbf{L}_{2 a} & \mathbf{L}_{22}
\end{array}\right],
$$

## Partial Orthogonalization of ssGBLUP Relationship Matrix H

- "Cholesky" matrix L (of $\mathrm{A}^{-1}=\mathrm{LL}^{\prime}$ ) naturally orthogonalizes non-genotyped non-ancestors (n)

Orthogonalizing group (n) in H from ssGBLUP relationship matrix $\mathbf{H}$.

- Smaller pedigree individuals are related through:

$$
\xrightarrow[\mathbf{L}^{\prime} \mathbf{H L}]{ }=\left[\begin{array}{cc}
\mathbf{I}_{n} & 0 \\
0 & \widehat{\mathbf{L}}^{\prime} \hat{\mathrm{H}} \hat{\mathrm{~L}}
\end{array}\right]
$$

- Former "on-the-fly" imputation operations of genomic information are now part of orthogonal projection $\mathrm{P}_{\stackrel{-}{\perp}}^{\perp}$


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\end{array}\right]
$$

$$
\widehat{\mathbf{L}}^{\prime} \hat{\mathbf{H}}^{\widehat{\mathbf{L}}}=\widehat{\mathbf{I}}+(1-w) \mathbf{P}_{\widehat{\mathbf{L}}_{a}^{\prime}}^{\perp}\left(\widehat{\mathbf{L}}_{2}^{\prime} \mathbf{Z}_{m} \mathbf{Z}_{m}^{\prime} \widehat{\mathbf{L}}_{2}-\widehat{\mathbf{I}}\right) \mathbf{P}_{\hat{\mathbf{L}}_{a}^{\prime}}^{\perp}
$$

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Orthogonalizing group (n) in H

$$
\left.\widehat{\mathbf{L}}^{\prime} \hat{\mathbf{H}}^{\mathbf{L}}=\widehat{\mathbf{I}}+(1-w) \mathbf{P}_{\widehat{\mathbf{L}}_{a}^{\prime}}^{\perp} \widehat{\mathbf{L}}_{2}^{\prime} \mathbf{Z}_{m} \mathbf{Z}_{m}^{\prime} \widehat{\mathbf{L}}_{2}-\widehat{\mathbf{I}}\right) \mathbf{P}_{\widehat{\mathbf{L}}_{a}^{\prime}}^{\perp}
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$$
\mathbf{L}^{\prime} \mathbf{H L}=\left[\begin{array}{cc}
\mathbf{I}_{n} & 0 \\
0 & \hat{\mathbf{L}}^{\prime} \hat{\mathbf{H}} \hat{\mathrm{L}}
\end{array}\right]
$$

Orthogonal projection of
$a$

$$
\begin{aligned}
\mathbf{P}_{\widehat{\mathbf{L}}_{a}^{\prime}}^{\perp} & =\widehat{\mathbf{I}}-\widehat{\mathbf{L}}_{a}^{\prime}\left(\hat{\mathbf{L}}_{a} \hat{\mathbf{L}}_{a}^{\prime}\right)^{-1} \widehat{\mathbf{L}}_{a} \\
& =\widehat{\mathbf{I}}-\widehat{\mathbf{L}}_{a}^{\prime}\left(\widehat{\mathbf{A}}^{a a}\right)^{-1} \hat{\mathbf{L}}_{a}
\end{aligned}
$$

## Result 2: New ssSNP-BLUP Formulation using the Smaller Pedigree

- Let $\widetilde{\mathrm{L}}_{a}$ be sparsity preserving Cholesky factorization of $\widehat{\mathbf{A}}^{a a}$.

New fully orthogonalized ssSNP-BLUP: $\mathbf{H}=\mathbf{M} \widetilde{\mathbf{G}} \mathbf{M}^{\prime}$, where


## Sparse Cholesky L ${ }_{a}$ of

$$
\widehat{\mathbf{A}}^{a a}=\widehat{\mathbf{L}}_{a} \widehat{\mathbf{L}}_{a}^{\prime}=\widetilde{\mathbf{L}}_{a} \widetilde{\mathbf{L}}_{a}^{\prime}
$$

Main advantage of new formulation: size of numerical sparse Cholesky factorization matrix $L_{a}$ is number of non-genotyped ancestors ( $\mathbf{A}^{a a}$ ) instead of all non-genotyped ( $\mathbf{A}^{11}$ )

Four groups of random effects ü: non-genotyped ancestors (a) have two sets of animals effects.

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$$
\mathbf{M}=\left(\mathbf{L}^{\prime}\right)^{-1}\left[\begin{array}{cccc}
\mathbf{I}_{n} & \mathbf{0} & \mathbf{0} & \mathbf{0} \\
\mathbf{0} & \sqrt{1-w} \widehat{\mathbf{M}}_{a} & \sqrt{w \mathbf{\mathbf { I }}} & \sqrt{1-w}\left(\widehat{\mathbf{I}}-\widehat{\mathbf{M}}_{a} \widehat{\mathbf{M}}_{a}^{\prime}\right) \widehat{\mathbf{L}}_{2}^{\prime} \mathbf{Z}_{m}
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## New random effects ũ

$$
\widetilde{\mathbf{u}}=\left[\begin{array}{c}
\widetilde{\mathbf{u}}_{n} \\
\widetilde{\mathbf{u}}_{a} \\
\widetilde{\mathbf{u}}_{a+2} \\
\widetilde{\mathbf{u}}_{m}
\end{array}\right]
$$



## Result 3: Numerical Feasibility with Large Number of Genotyped

- Sparse Cholesky $\widetilde{\mathrm{L}}_{a}$ (of $\widehat{\mathbf{A}}^{a a}=\widetilde{\mathrm{L}}_{a} \widetilde{\mathrm{~L}}_{a}^{\prime}$ ) has fill-ins depending on pedigree and proportion of genotyped.

```
Memory reguirements of }\mp@subsup{L}{a}{}\mathrm{ must remain
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Investigated by simulating different proportions of
genotyped animals in Nordic dairy cattle population.
Pedigree of }9.2\mathrm{ million was "genotyped" with four
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Memory usage: }13\mathrm{ times size of pedigree when 2.8
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## Conclusions

1 All fully orthogonalized ssSNP-BLUPs were shown to share same convergence properties with respect to iterative solution methods.

2 New ssSNP-BLUP formulation was presented in which Single-step relationship matrix $\mathbf{H}$ was expressed using smaller pedigree of genotyped and their ancestors.

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## Lulke

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[^0]:    Fully orthogonalized ( $\widetilde{G}=\mathbf{I})$ ssSNP-BLUPs:

    - almost identical number of iterations when solved using iterative methods, e.g. PCG.

