Extension of single-step GBLUP to many genotyped individuals

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## Genomic selection and single-step

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





Aguilar et al., 2010 Christensen and Lund, 2010

- Simplicity
  - No DYD or DP
  - No index
  - No complexity
- Accuracy
  - Avoids double counting
  - Avoids fixed index
  - Accounts for preselection bias

## **Current implementation of SS**

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

- G and A<sub>22</sub> created explicitly
- Quadratic memory and cubic computations
- Cost per 100k genotypes 1.5 hr (Aguilar et al.,2014)



## Number of genotypes and impending problem

- > 2 M for Holsteins
- > 400k for Angus

Genomic pre-selection issue (Patry and Ducrocq, 2011; VanRaden et al., 2013)

- BLUP increasingly biased
- Need all data on preselection included

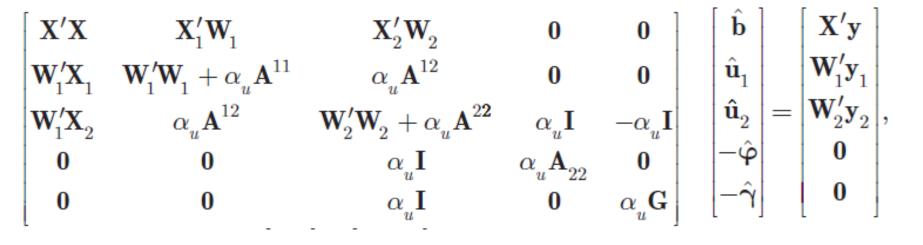
### **Unsymmetric equations**

# $\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z} \\ \mathbf{H}\mathbf{Z}'\mathbf{X} & \mathbf{H}\mathbf{Z}'\mathbf{Z} + \alpha\mathbf{I} \end{bmatrix} \begin{bmatrix} \hat{\mathbf{b}} \\ \hat{\mathbf{u}} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{y} \\ \mathbf{H}\mathbf{Z}'\mathbf{y} \end{bmatrix}$

Misztal et al., 2009

No convergence without good preconditioner No convergence with large H or A

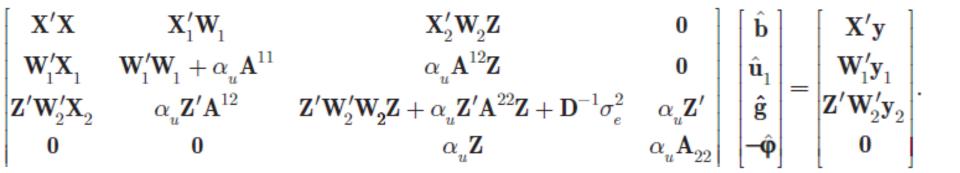
## No G or A<sub>22</sub> inverse model



Legarra and Ducrocq (2011)

Slow convergence with few genotypes Divergence with many genotypes

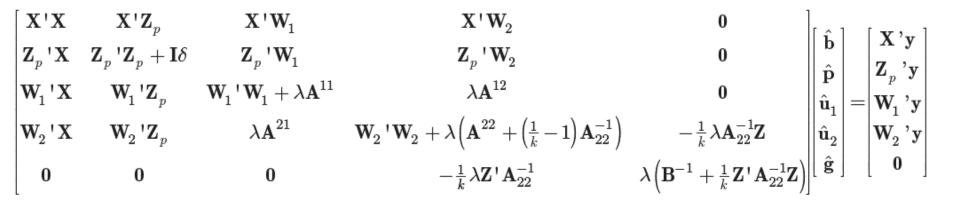
## SNP model for genotyped animals



Legarra and Ducrocq, 2011

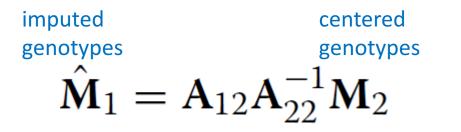
No successful programming

## SNP model for genotyped animals



Liu et al, 2014

## SNP effects for all animals (Fernando et al., 2014)



$$\begin{bmatrix} \mathbf{y}_1 \\ \mathbf{y}_2 \end{bmatrix} = \begin{bmatrix} \mathbf{X}_1^* \\ \mathbf{X}_2^* \end{bmatrix} \boldsymbol{\beta}^* + \begin{bmatrix} \mathbf{Z}_1 & \mathbf{0} \\ \mathbf{0} & \mathbf{Z}_2 \end{bmatrix} \begin{bmatrix} \hat{\mathbf{M}}_1 \boldsymbol{\alpha} + \boldsymbol{\epsilon} \\ \mathbf{M}_2 \boldsymbol{\alpha} \end{bmatrix} + \mathbf{e}$$

Cost of imputation Requires new type of programming Extension to complex models unclear

## Can regular ssGBLUP be made more efficient?

## Scaling up A<sub>22</sub><sup>-1</sup>

$$A_{22}^{-1} = A^{22} - A^{21} (A^{22})^{-1} A^{12}$$

- $A_{22}^{-1}$  dense (Faux et al., 2014)
- For PCG iteration (Stranden et al., 2014)

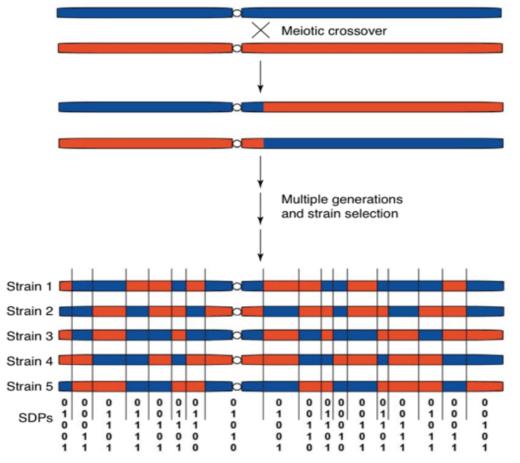
$$A_{22}^{-1}q = A^{22}q - \left\{A^{21}\left[\left(A^{22}\right)^{-1}\left(A^{12}q\right)\right]\right\}$$

• Seconds for 500k animals with good programming (Masuda et al., 2017)

Is dimensionality of genomic information limited?

- Regular G not positive definite past ~5k
   Blending with A (VanRaden, 2008)
- Dimensionality of SNP BLUP small (Maciotta et al., 2013)
- Success of imputation
- Manhattan plots noisy until averaged by 300k-10Mb (depending on species)

## Origin of Haplotype blocks



Cuppen, 2005

Heterogenetic and homogenic tracts in genome (Stam, 1980)

.....

E(#tracts)=4NeL (Stam, 1980) Ne – effective population size L –length of genome in Morgans

> Holsteins: Ne ≈100 L=30 Me=12,000

Inversion via SVD/eigenvalue decomposition

Assume 1 million animals genotyped with 60k chip

 $\mathbf{G} = \mathbf{Z}\mathbf{Z}' = \mathbf{U}\mathbf{D}\mathbf{U}'$  Eigenvalue decomposition (1M x 1M)

 $G^- = UD^-U'$  Generalized inverse (1M x 1M)

 $Z = USV = UD^{0.5}V - SVD decomposition (1M x 60k)$ 10h for 720k animals (Masuda, 2017)

**t** - index for non-negligible eigenvalues, say 10k  $\mathbf{G}^- = \mathbf{U}_t \ \mathbf{D}_t^{-1} \mathbf{U}_t' = \mathbf{U}_t \ \mathbf{S}_t^{-1} \mathbf{S}_t^{-1} \mathbf{U}_t' = \mathbf{U}_* \ \mathbf{U}_*$ 

For PCG iteration  $\mathbf{G}^{-1}\mathbf{q} = \mathbf{U}_* (\mathbf{U}_* \mathbf{q})$  - only 1 M x 10k elements

## Inverse by Woodbury formula

$$G = ZZ' + I\varepsilon,$$
  

$$G^{-1} = \frac{1}{\varepsilon}I - \frac{1}{\varepsilon}Z(\frac{1}{\varepsilon}Z'Z + I)^{-1}Z'\frac{1}{\varepsilon}$$
  
For PCG iteration:

Woodbury formula **Z'Z** 60k x 60k

$$\mathbf{G}^{-1}\mathbf{q} = \frac{1}{\varepsilon} \{\mathbf{I} - \mathbf{Z}(\mathbf{U}\mathbf{D}\mathbf{U}')^{-1}\mathbf{Z}'\}\mathbf{q} = \frac{1}{\varepsilon} \{\mathbf{I} - \mathbf{S}\mathbf{S}'\}\mathbf{q}$$
$$\mathbf{S} = \mathbf{Z}\mathbf{U}'\mathbf{D}^{-1/2}$$
With reduced rank  $\mathbf{S} = \mathbf{Z}\mathbf{U}_t'(\mathbf{D}_t)^{-\frac{1}{2}}$  (1M x 10k)

Ostersen et al., 2017

## If G has limited dimensionality, can G<sup>-1</sup> be sparse like A<sup>-1</sup>?

## Use of a la Henderson's rules?



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## A recursive algorithm for decomposition and creation of the inverse of the genomic relationship matrix

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Use of relatives for **G**<sup>-1</sup> Accuracies not good enough Theory not clear



## Assumption of limited dimensionality

S – n x 1 vector containing additive information of population (haplotypes, chromosome segments, LD blocks)?

Breeding value Very small error  $\mathbf{u} = \mathbf{Ts} + \mathbf{e}$ 

If  $\mathbf{U}_{c}$  contains n animals:

$$\mathbf{s} \approx \mathbf{T}_c^{-1} \mathbf{u}_c$$

Breeding values of any n animals contains all additive information

Choose core "c" and noncore "n" animals

$$\mathbf{u}_{n} = \mathbf{P}_{nc}\mathbf{u}_{c} + \varepsilon_{n}$$
$$\mathbf{u}_{c} = \mathbf{u}_{c}$$
$$\begin{bmatrix} \mathbf{u}_{c} \\ \mathbf{u}_{n} \end{bmatrix} = \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \mathbf{P}_{nc} & \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{u}_{c} \\ \mathbf{\varepsilon}_{n} \end{bmatrix}$$

 $\operatorname{var}(\boldsymbol{\varepsilon}_n) = \mathbf{M}_{\mathbf{nn}}$ 

$$\mathbf{G} = \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \mathbf{P}_{nc} & \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{G}_{cc} & \mathbf{0} \\ \mathbf{0} & \mathbf{M}_{nn} \end{bmatrix} \begin{bmatrix} \mathbf{I} & \mathbf{P}_{cn} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$
$$\mathbf{G}^{-1} = \begin{bmatrix} \mathbf{I} & -\mathbf{P}_{cn} \\ \mathbf{0} & \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{G}_{cc}^{-1} & \mathbf{0} \\ \mathbf{0} & \mathbf{M}_{nn}^{-1} \end{bmatrix} \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ -\mathbf{P}_{nc} & \mathbf{I} \end{bmatrix}$$

## How to estimate **P** and inv(**G**)?

$$\operatorname{var}\left(\begin{bmatrix}\mathbf{u}_{c}\\\mathbf{u}_{n}\end{bmatrix}\right) = \begin{bmatrix}\mathbf{G}_{cc} & \mathbf{G}_{cn}\\\mathbf{G}_{nc} & \mathbf{G}_{nn}\end{bmatrix}\sigma_{u}^{2}$$

**G** is "true" relationship matrix

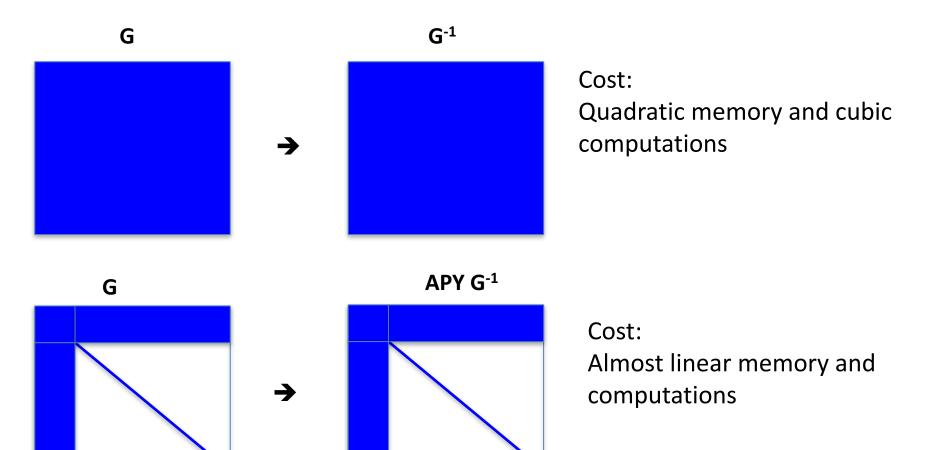
$$\mathbf{u}_n \mid \mathbf{u}_c = \mathbf{G}_{nc} \mathbf{G}_{cc}^{-1} \mathbf{u}_c, \quad \mathbf{P} = \mathbf{G}_{nc} \mathbf{G}_{cc}^{-1}$$

$$\mathbf{G}^{-1} = \begin{bmatrix} \mathbf{G}_{cc}^{-1} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{bmatrix} + \begin{bmatrix} \mathbf{G}_{cc}^{-1} \mathbf{G}_{cn} \\ \mathbf{I} \end{bmatrix} \mathbf{M}^{-1} \begin{bmatrix} \mathbf{G}_{nc}^{-1} \mathbf{G}_{cc}^{-1} & \mathbf{I} \end{bmatrix}$$

## APY algorithm

(Algorithm for Proven and Young)

## **Properties of APY algorithm**





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#### Using recursion to compute the inverse of the genomic relationship matrix

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#### *Hot topic:* Use of genomic recursions in single-step genomic best linear unbiased predictor (BLUP) with a large number of genotypes

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#### Inexpensive Computation of the Inverse of the Genomic Relationship Matrix in Populations with Small Effective Population Size

Ignacy Misztal<sup>1</sup>

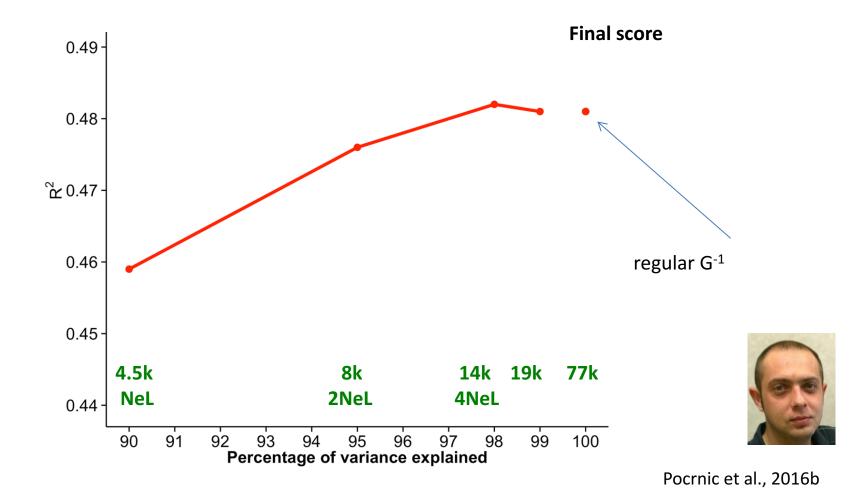
Animal and Dairy Science, University of Georgia, Athens, Georgia 30602

#### The Dimensionality of Genomic Information and Its Effect on Genomic Prediction

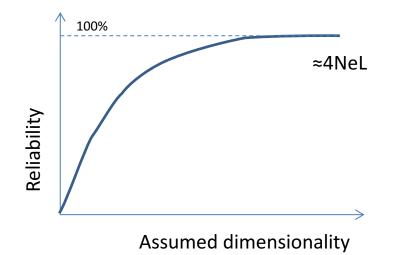
Ivan Pocrnic,\*.<sup>1</sup> Daniela A. L. Lourenco,\* Yutaka Masuda,\* Andres Legarra,<sup>†</sup> and Ignacy Misztal\* \*Department of Animal and Dairy Science, University of Georgia, Athens, Georgia 30602, and <sup>†</sup>Institut National de la Recherche Agronomique, GenPhySE, F-31326 Castanet-Tolosan, France

EAAP meeting 20

## Reliabilities – Holsteins (77k)

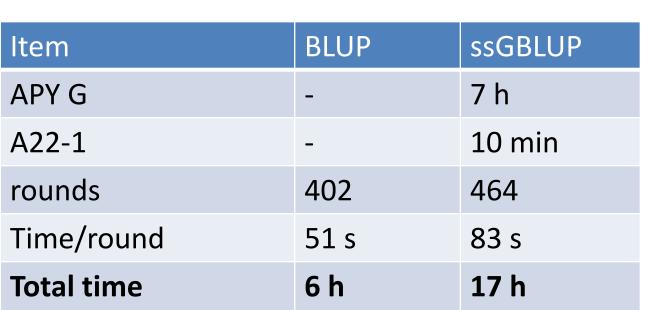


## Distribution of segments/haplotypes/..



## Costs with 720k genotyped animals

- 30 M Holsteins
- 50 M records
- 764k 60k genotypes





## Which core animals in APY?

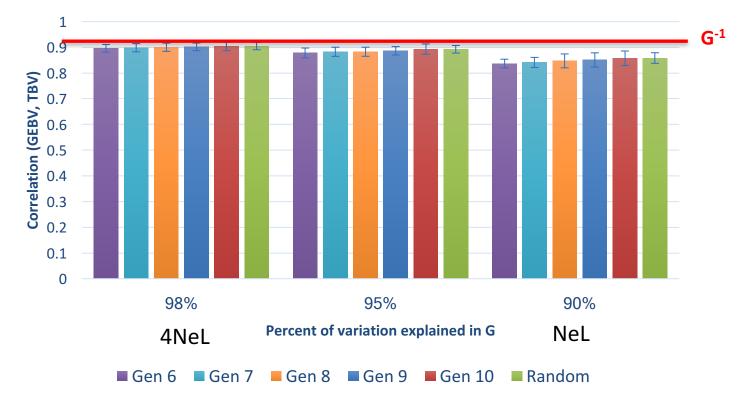
#### Bradford et al. (2017)



- Simulated populations (QMSim; Sargolzaei and Schenkel, 2009)
- Ne = 40
- #genotyped animals = 50,000
- Core animals:
  - Random gen 6 || gen 7 || gen 8 || gen 9 || gen 10 (y)
  - Random all generations

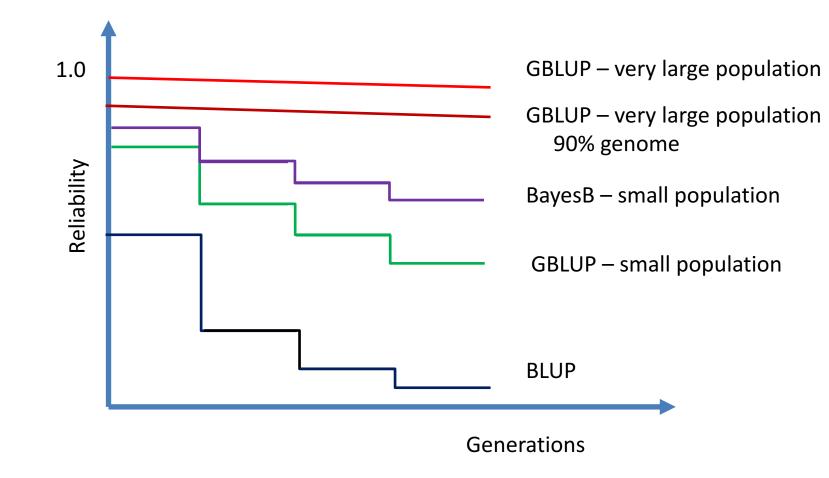
### Which core animals in APY?

Accuracy



Bradford et al. (2016)

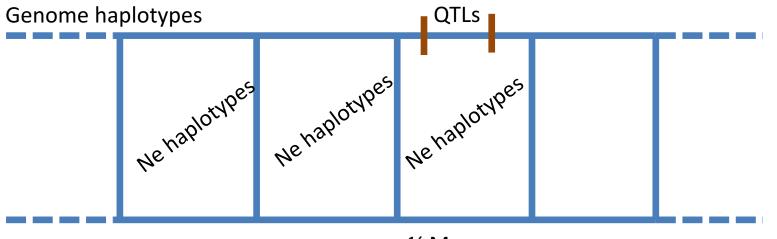
## Persistence over generations



Very large – equivalent to 4NeL animals with 99% accuracy Are SNP effects from Holstein national populations converging

## Theory of limited dimensionality

Number of haplotypes: 4 Ne L Ne within each ¼ Morgan segment



¼ Morgan

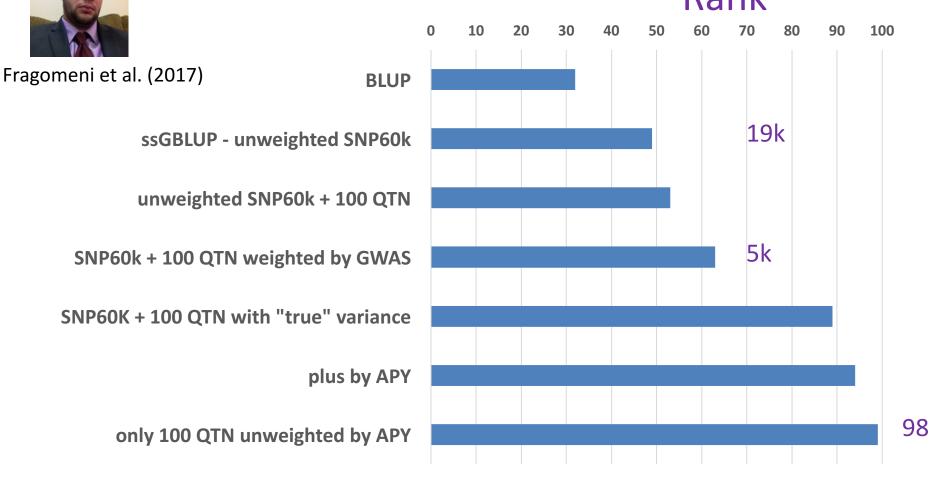
Dimensionality of ¼ Morgan case: Ne

or number of identified QTLs

→ Reduced dimensionality with weighted GRM

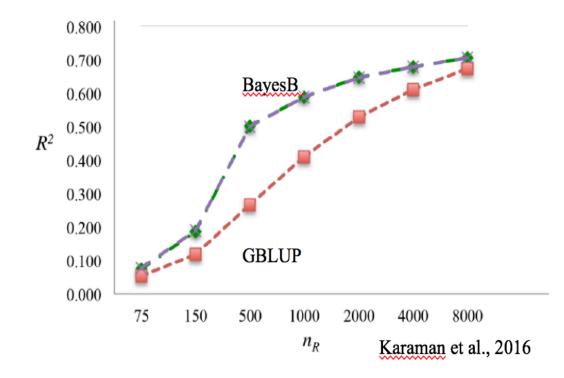
Fragomeni et al., 2018

### ssGBLUP accuracies using SNP60K and 100 QTNs – simulation study Rank



### Multitrait ssGBLUP or SNP selection?

- SNP selection/weighting (BayesB, etc.)
  - Large impact with few genotypes
  - Little or no impact with many



## Variance components

- Based on SNP
  - limitations
- REML based on relationships
  - Equations no longer sparse
  - YAMS sparse matrix package –up to 100 times speedup (Masuda et al., 2017)
  - APY for REML
- Method R (Legarra and Reverter, 2017)

## Extra topics

- Matching pedigrees and genomic relationships
- Missing pedigrees
- Crossbreeding
- Causative SNP
- Haplotypes for crossbreds (Christensen et al., 2016)
- Metafounders (Legarra et al., 2016)
- Approximation of reliabilities

## Conclusions

- Limited dimensionality of genomic information due to limited effective population size
- ssGBLUP suitable for any data set and model
- With large data sets for Holsteins:
  - Good persistence of predictions
  - Convergence of predictions from different countries



**United States** Department of Agriculture

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zoetis











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## Theory for APY

- Breeding values of core animals linear functions of:
  - Independent chromosome segments (Me)
  - Independent effective SNP
- E(Me)=4 Ne L (Stam, 1980; VanRaden, 2008)

Ne –effective population size

L – length of genome in Morgans

Me = 4 (Ne=100) (L=30) =12,000

## Accuracy and distance from markers to QTL

Fragomeni et al. (2017)

