Accuracy of genomic prediction by singular value decomposition of the genotype matrix

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Aim

 \hookrightarrow Evaluate the effect of the number of components in singular value decomposition (SVD) of the genotype matrix on the accuracy of genomic prediction.

Definition: Orthogonal matrix

A matrix $\mathbf{U} \in \mathbb{R}^{n \times m}$ is said to be *orthogonal* iff

$$\mathbf{U}^{\mathsf{T}}\mathbf{U} = \mathbf{I}_m$$

Definition: Eigenvector and eigenvalue

A non-null vector v is called an *eigenvector* of a matrix $\mathbf{A} \in \mathbb{R}^{n \times n}$ when there is a scalar λ such that

$$\mathbf{A}\mathbf{v} = \lambda \mathbf{v}$$

 λ is said to be the *eigenvalue* of **A** associated to the eigenvector v.

Theorem: SVD

For every matrix $\mathbf{X} \in \mathbb{R}^{n \times m}$, there exist two orthogonal matrices $\mathbf{U} \in \mathbb{R}^{n \times n}$ and $\mathbf{V} \in \mathbb{R}^{m \times m}$ such that

$$\mathbf{X} = \mathbf{U}\mathbf{S}\mathbf{V}^{\mathsf{T}}$$

where $\mathbf{S} \in \mathbb{R}^{n \times m}$ is a diagonal matrix whose non-zero values are the square roots of the eigenvalues of $\mathbf{A} = \mathbf{X}^{\mathsf{T}}\mathbf{X}$.

We call the diagonal values of **S** the singular values of **A**.

Principal Component Ridge Regression

$$\begin{aligned} \mathbf{y} &= \mathbf{X}\mathbf{b} + \mathbf{e} \\ &[\mathbf{X}^{\mathsf{T}}\mathbf{X} + \lambda \mathbf{I}] \, \hat{\mathbf{b}} = \mathbf{X}^{\mathsf{T}}\mathbf{y} \\ &\mathbf{X} = \mathbf{U}\mathbf{S}\mathbf{V}^{\mathsf{T}} \\ &\mathbf{T} = \mathbf{U}^{\mathsf{T}}\mathbf{S} \\ &[\mathbf{S}^{\mathsf{T}}\mathbf{S} + \lambda \mathbf{I}] \, \hat{\mathbf{s}} = \mathbf{T}^{\mathsf{T}}\mathbf{y} \\ &\hat{\mathbf{b}} = \mathbf{V}\hat{\mathbf{s}} \end{aligned}$$
 Truncated-SVD version of SNP-BLUP
$$\mathbf{X} \approx \mathbf{U}_k \mathbf{S}_k \mathbf{V}_k^{\mathsf{T}} \\ &\mathbf{T} = \mathbf{U}_k^{\mathsf{T}}\mathbf{S}_k \\ &[\mathbf{S}_k^{\mathsf{T}}\mathbf{S}_k + \lambda \mathbf{I}] \, \hat{\mathbf{s}}_k = \mathbf{T}^{\mathsf{T}}\mathbf{y} \\ &\hat{\mathbf{b}} = \mathbf{V}_k \, \hat{\mathbf{s}}_k \end{aligned}$$

Materials and methods

- Genotypes* from 1,927 Atlantic Salmon (Salmo salar) fish
 - 16,454 SNP markers (all on chromosome 1)

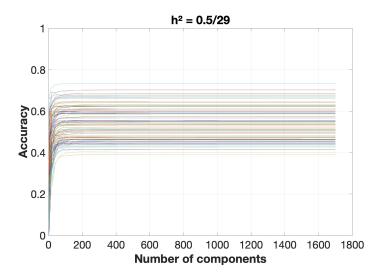
^{*}Data provided by AquaGen AS.

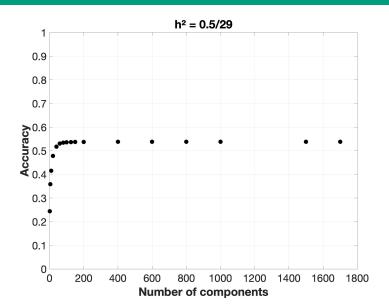
Simulation

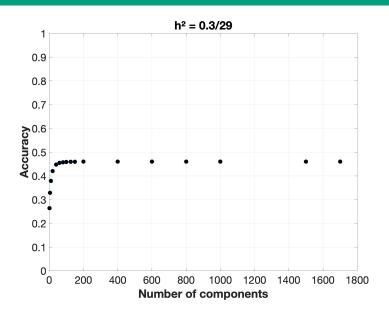
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$$b_1, \ldots, b_{1000}$$
 i.i.d. $b_i \sim N(0, V_m)$, where $V_m = \frac{h^2}{2 \sum_{i=1}^{1000} p_i (1-p_i)}$

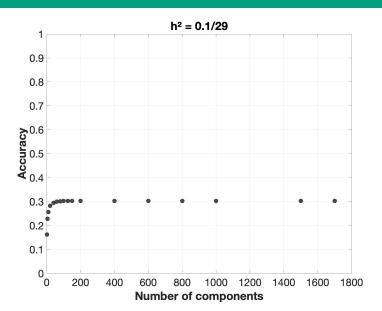
- $h^2 = \{0.1/29, 0.3/29, 0.5/29\}$
- 1,000 QTL randomly positioned along the 16,454 SNP loci
- mask the QTL from the genotype matrix
- calculate true breeding values g = Xb
- lacktriangle predict breeding values with PCRR $\hat{f g}={f X}\hat{f b}={f X}{f V}_k\hat{f s}_k$
 - (10-fold cross-validation)
- lacksquare estimate correlation coefficient $\hat{r}=rac{Cov(g,\hat{g})}{\hat{\sigma}_g\hat{\sigma}_{\hat{g}}}$
- lacksquare run 100 replicates and obtain average accuracy $ar{r} = \sum_{j=1}^{100} \hat{r}_j/100$

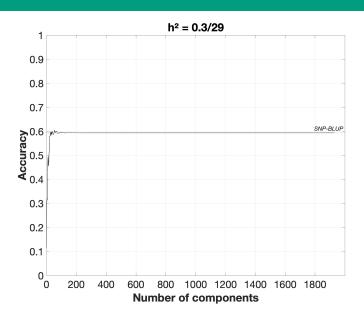
 $k = \{2, 5, 10, 20, 40, 60, 80, 100, 125, 150, 200, 400, 600, 800, 1000, 1500, 1700\}$

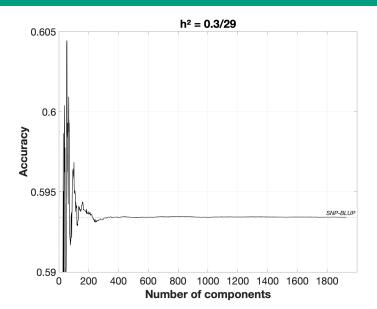












Discussion

- reduction of statistical noise
- accuracy not always an increasing function of the number of SVD components
- higher accuracy using APY (e.g., 0.5% gain)
- higher accuracy using PCRR/PCIG (e.g., 1.8% gain)

Conclusions

- SVD is useful for data reduction of the genotype matrix
- PCRR can be used for genomic prediction
 - good accuracies obtained with few components
- PCRR can provide higher accuracies than SNP-BLUP with certain numbers components
 - within replicates, maximum accuracy at 50–250 components
 - across replicates, maximum mean accuracy at 400–600 components

References



Ødegård, J., Indahl, U., Strandén, I., & Meuwissen, T. 2018. Large-scale genomic prediction using singular value decomposition of the genotype matrix. *Genetics Selection Evolution*, 50, 6.



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