

ENLARGING A TRAINING SET FOR GENOMIC SELECTION BY IMPUTATION OF UN-GENOTYPED ANIMALS IN POPULATIONS OF VARYING GENETIC ARCHITECTURE

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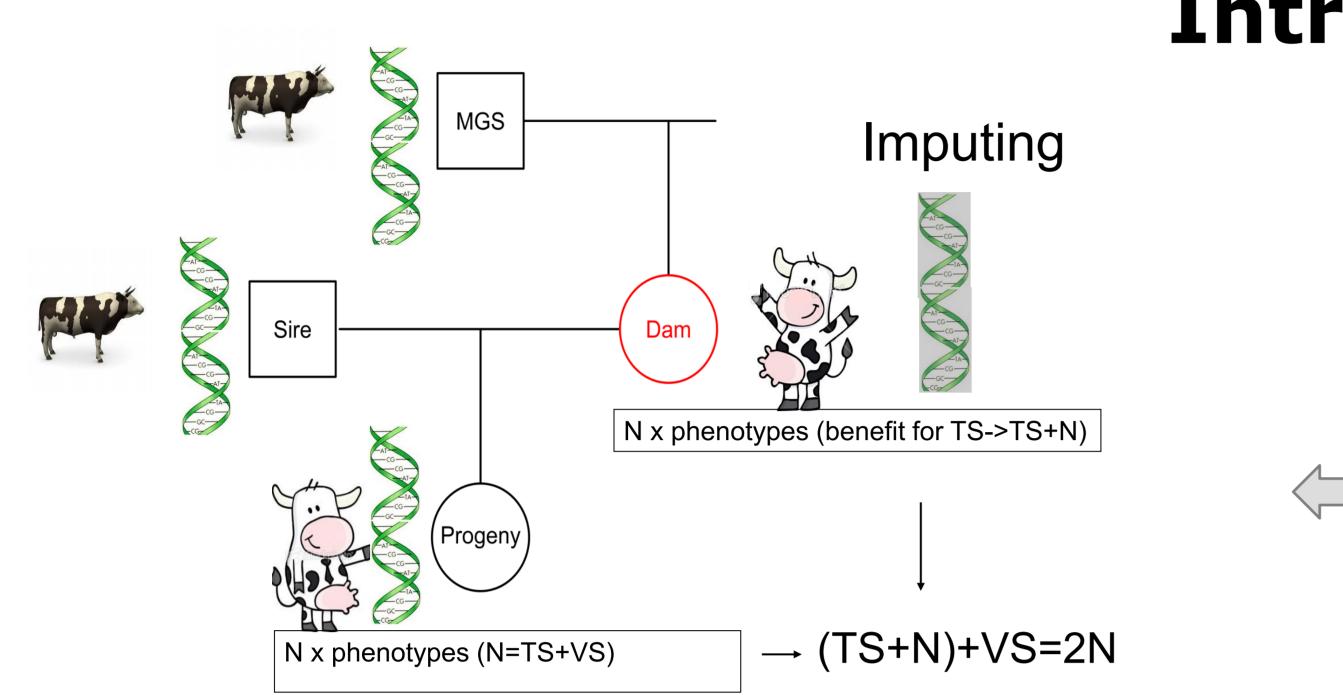


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Introduction

The most common application of imputation is to infer genotypes (GT) of a high-density panel of markers on animals that are genotyped for a low-density panel.

Assumed family members with available genotypic information (black) (MGS=M, Sire=S, Progeny=P) used for imputing an un-genotyped dam (Dam=D, red).

$$P(D = G_i \mid M = G_j \cap S = G_k \cap P = G_i) = \frac{P(P = G_i \mid S = G_k \cap D = G_i)P(D = G_i \mid M = G_j)}{\sum_{m=1}^{3} P(P = G_i \mid S = G_k \cap D = G_m)P(D = G_m \mid M = G_j)}$$

Another application of imputation is to increase the size of the training set with un-genotyped animals. This strategy can be particularly successful when a set of closely related individuals are genotyped.

The posterior probabilities of the dam's three possible GT can be calculated following Bayes' theorem and the allele frequency (use the given GT for MGS ($M=G_j$) and Sire ($S=G_k$) -> P($D=G_i|M=G_j \cap S=G_k \cap P=G_i$)=P($D=G_i|P=G_i$), P($P=G_i|S=G_k \cap D=G_i$)=P($P=G_i|D=G_i$), P($D=G_i|M=G_i$)=P($D=G_i$))

$$= P(D = G_i | P = G_i) = \frac{P(P = G_i | D = G_i)P(D = G_i)}{\sum_{m=1}^{3} P(P = G_i | D = G_m)P(D = G_m)}$$

Data

QMSim (Sargolzaei et al., 2009) simulated data were used with the following characteristics to analyze

The **impact** of enlarging a training set (TS) with imputed dams (TSA) **on the accuracy of genomic**

imputing methods:

- One chromosome of 100cM, 2000 bi-allelic randomly allocated markers, $h^2=0.2$
- Two levels of LD x 2 levels of selection (LowLD_Sel, HighLD_Sel, LowLD_NoSel, HighLD_NoSel)
- 20 generations (for HighLD with bottleneck)
- 2000 genotyped female progeny from the last generation with known genotype for sire and mgs

predictions was evaluated for:

- different populations
- varying levels of heritability
- different sizes of genotyped females (TS)

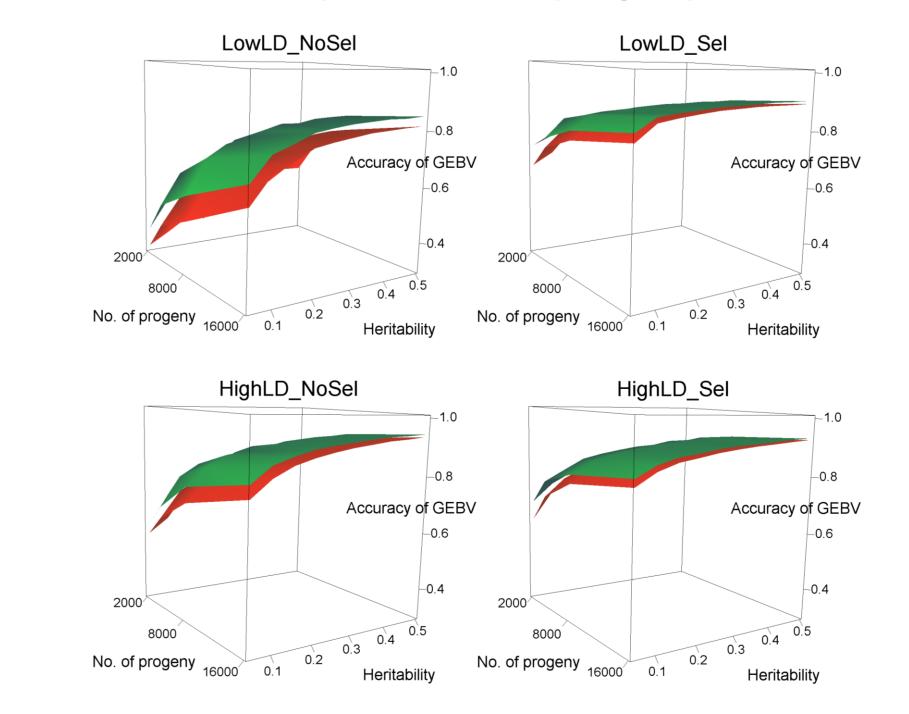
Variants of differing heritability were generated by adding a residual term to the simulated true breeding values ($h^2 = 0.05, 0.10, ..., 0.50$).



Correlation between true and imputed genotypes from different imputation methods and programs

Imputing	Scenario			
Method	LowLD_NoSel	LowLD_Sel	HighLD_NoSel	HighLD_Sel
Single_Step ^{ig}	0.76 ± 0.003	0.83 ± 0.038	0.88 ± 0.004	0.90 ± 0.013
Single_Step ^{gc}	0.81 ± 0.003	0.86 ± 0.028	0.90 ± 0.003	0.93 ± 0.009
Two_Steps	0.57 ± 0.008	0.74 ± 0.066	0.80 ± 0.006	0.85 ± 0.021
findhap.f90	0.52 ± 0.006	0.69 ± 0.065	0.74 ± 0.006	0.82 ± 0.030
Alphalmpute	0.83 ± 0.003	0.87 ± 0.024	0.86 ± 0.004	0.89 ± 0.010

Accuracies of genomic prediction for different values of h², number of female progeny in the last generation, and population structure, red/green surfaces: with TS/TSA (90% of the progeny in TS/~+imputed dams^{Single_Step^{gc}})



(allele substitution effects of every locus on the simulated phenotypes were fitted in a multiple random regression model similar to the GBLUP method of Meuwissen et al. (2001), accuracies were evaluated as correlation between simulated TBV and estimated GEBV)

Single_Step=Bayes based methods (^{ig} with integer genotypes, ^{gc} with "gene content"=non integer genotypes);Two_Steps: unambiguously inferred genotypes were used to build low-density panels and then imputed with fastPHASE; findhap.f90, AlphaImpute: Imputing-Software

Conclusion

With the underlying family structure (typical for Holsteins) imputation can be used to achieve an extra increase in accuracy of genomic predictions by enlarging the training set with completely ungenotyped dams. This strategy was shown to be

particularly useful for populations with lower levels of linkage disequilibrium, for genomic selection on traits with low heritability, and for species or breeds for which the size of the reference population is limited.