Accuracy of imputation in a sparsely-genotyped pig pedigree

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Background

 Implementation of genomic selection requires large numbers of individuals to be densely genotyped

 Candidates for selection can be genotyped to increase accuracy of selection and decrease generation interval

 Currently not cost effective to genotype all candidates in some livestock species

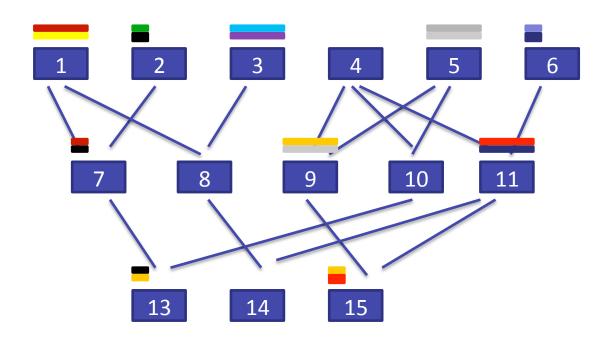
Background

- Strategy for implementing genomic selection in pigs
 - Genotype sires and some dams at 60k
 - Genotype selected candidates at low density
 - No commercial low-density chip
 - Impute from low to high density
 - Use imputed genotypes across all traits
- Impute
 - Use pedigree information
 - Close relatives are not always genotyped at high density

Imputation approach - Alphalmpute

- Long range phasing, segregation analysis and haplotype library imputation
- Uses information from close and distant relatives/ SNPs to impute alleles
- Imputes individuals with varying densities of genotypes and levels of relatives genotyped
- Where alleles can not be imputed allele probabilities are calculated and summed to give a probable genotype score

Imputation approach - Alphalmpute



Genotyping strategy in terms of high density, low density and not genotyped



Haplotype library for population

- Individual's are densely, sparsely, or not genotyped
- Pedigree information available
- Single locus segregation analysis for each SNP
- Long-range phasing and haplotype libraries
- Match each pair of haplotypes with low density genotypes and allele probabilities

Testing data

- Samples
 - Single line
 - 60k genotyped: N=3,534
 - Pedigree: N=6,473
 - Test samples
 - Most recently-born animals, no progeny: N=509

SNPs

- Known position on SSC01
- Filtered for MAF, quality and call rate
- M=4,221

N	Genotype Category
51	Both Parents
62	Sire and MGS
46	Dam and PGS
45	Sire
14	Dam
291	Other

Imputation

- Mask genotypes to simulate low-density genotyping
- Low-density genotyped SNPs selected based on position and MAF
- Genotyping scenarios:

No. SNPs	%60k (whole- genome basis)	Whole genome density
725	18	6k
379	9	3k
184	5	1500
93	2	768
46	1	384

Results – 6k density

		Paternal Maternal					
		Alleles	Alleles		Correct Incorrect		
	N	(%)	(%)	Corr.a	(%)	(%)	
Both Parents	51	99.9	99.9	1.00	0.2	99.4	
Sire and MGS	62	99.4	99.9	0.99	1.4	97.7	
Dam and PGS	46	99.7	99.7	0.99	0.8	98.7	
Sire	45	99.9	99.9	0.99	1.7	97.2	
Dam	14	98.8	98.8	0.98	2.0	96.3	
Other	291	99.4	99.4	0.97	2.9	95.1	

^aCorrelation between sum of allele probabilities and called genotypes

Results – 3k density

		Paternal Maternal						
		Alleles	Alleles Alleles			Correct Incorrect		
	N	(%)	(%)	Corr.a	(%)	(%)		
Both Parents	51	99.8	99.8	1.00	0.4	99.2		
Sire and MGS	62	99.1	88.9	0.98	2.4	96.6		
Dam and PGS	46	99.8	99.8	0.99	1.3	98.2		
Sire	45	98.7	99.9	0.98	2.4	96.1		
Dam	14	99.9	98.0	0.96	4.3	93.4		
Other	291	98.1	99.2	0.96	4.5	92.8		

^aCorrelation between sum of allele probabilities and called genotypes

Results – 1500 density

		Paternal Maternal					
		Alleles	Alleles Alleles			Correct In	
	N	(%)	(%)	Corr.a	(%)	(%)	
Both Parents	51	99.8	99.7	0.99	0.6	98.9	
Sire and MGS	62	98.8	99.8	0.97	3.7	94.9	
Dam and PGS	46	99.7	99.6	0.98	2.3	97.0	
Sire	45	97.3	99.8	0.96	4.3	92.7	
Dam	14	99.7	97.2	0.96	4.6	92.3	
Other	291	97.0	98.7	0.94	6.8	89.1	

^aCorrelation between sum of allele probabilities and called genotypes

Results – 768 density

		Maternal	Paternal Maternal					
		Alleles	Alleles		Correct Incorrect			
	N	(%)	(%)	Corr.a	(%)	(%)		
Both Parents	51	99.7	99.7	0.99	1.3	98.1		
Sire and MGS	62	98.1	99.7	0.95	5.4	92.4		
Dam and PGS	46	99.6	99.7	0.97	3.5	95.8		
Sire	45	96.1	99.8	0.94	7.0	88.8		
Dam	14	99.9	94.8	0.92	8.3	86.4		
Other	291	94.4	97.3	0.90	9.6	82.2		

^aCorrelation between sum of allele probabilities and called genotypes

Results – 384 density

		Paternal Maternal					
		Alleles	Alleles		Correct Incorrect		
	N	(%)	(%)	Corr.a	(%)	(%)	
Both Parents	51	99.7	99.5	0.98	2.02	97.1	
Sire and MGS	62	96.7	99.6	0.93	8.1	88.1	
Dam and PGS	46	99.7	99.5	0.95	6.5	92.6	
Sire	45	92.3	99.8	0.89	11.8	80.3	
Dam	14	99.6	94.9	0.90	11.0	83.6	
Other	291	86.0	94.7	0.84	12.4	72.0	

^aCorrelation between sum of allele probabilities and called genotypes

Results – IMPUTE2 comparison

Correlation between imputed and called genotypes for test animals

38	4 density	3k	3k density		
Alpha Imput		Alpha E2 Imput		E2 N	
0.98	0.77	1.00	0.96	51	Both Parents
0.93	0.80	0.99	0.94	62	Sire and MGS
0.96	0.79	0.99	0.95	46	Dam and PGS
0.89	0.78	0.99	0.95	45	Sire
0.90	0.76	0.98	0.93	14	Dam
0.86	0.76	0.97	0.95	291	Other

^a**IMPUTE2**: B. N. Howie, P. Donnelly and J. Marchini (2009) A flexible and accurate genotype imputation method for the next generation of genome-wide association studies. PLoS Genetics 5 (6): e1000529

Results – Genomic breeding values using imputed genotypes

 Calculate gEBV using full SNP panel and using low-density SNP panel with imputation

Correlation between gEBV using lowdensity SNPs and gEBV using all SNPs

	3k density	384 density
Both Parents	0.99	0.90
Sire and MGS	0.99	0.87
Dam and PGS	1.00	0.94
Sire	0.98	0.86
Dam	0.98	0.84
Other	0.97	0.73

Discussion

- Improvements possible, especially for "other" category
 - In practice most animals have at least one parent genotyped
 - Recombination modeling
- Explore genotyping strategies
- Use to impute un-genotyped pedigree
 - Increase training set size
 - Simplify single step evaluation to GBLUP

Conclusions

- Imputation method is robust and useful with incomplete genotyping
- The correlation between imputed and called genotypes was near 1 at all densities when both parents were genotyped
- Even at very low densities and when not closely related to genotyped individuals, correlations were high
- Can calculate gEBV for selection candidates using small panel

Acknowledgements

- Lizhen Wang
- Selma Forni

