Genomic prediction and GWAS with sequence information versus HD or 50k SNP chips

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- Whole genome sequence data expected to perform better in GWAS and GS, ore persistent across generations / breeds
  - Causal mutation (QTN) is included
  - No dependency on LD between SNP and QTL



# Identifying QTN with GS?

WAGENINGEN UR



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HD SNP:



The potential benefit of using sequence data for

• QTL detection

i.r.t.

- Explaining total genetic variation
- Prediction accuracy GS



# Method (1): Imputation to sequence

5556 Holstein Friesian bulls CRV

777K SNP genotypes (Illumina BovineHD BeadChip)

1147 animals (multiple breeds)

28M SNP (whole-genome sequence data)

1000 bull genomes project Run 4

# 5556 Holstein Friesian bulls with phenotype (PY) and imputed sequence

3469 bulls used for discovery and training & 2287 bulls used for validation



Aniek Bouwman

# Method (2): statistical models

GWAS using <u>single</u> SNP regression (GCTA)

- GWAS using single SNP regression
- Include GRM based on HD SNP set
- MAF >0.01

### <u>Conditional and joint GWAS (COJO)</u>

 Stepwise selection of SNP explaining additional variance
Conditional and joint multiple-SNP analysi



Conditional and joint multiple-SNP analysis of GWAS summary statistics identifies additional variants influencing complex traits

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# Method (3): Two validation methods

SNP set selection from GWAS:

- 1. Estimate heritability in validation animals using GRM based on selected sets of SNP
- 2. Train GRM (based on selected SNP sets) on discovery animals, back solve SNP and predict DGV for 2287 validation animals. Correlate DGV with phenotypes.



### Results GWAS: 50K



Chromosome



### Results GWAS : HD



Chromosome



### Results GWAS Sequence + cojo5



Chromosome



# Results GWAS: Cojo5 on Chr14 (DGAT)



Chromosome 14 position(Mb)



# Results: SNP set selection GWAS

#### Number of SNP selected in the different SNP sets:

	Sequence	HD	50k	СОЈО
All	13,789,029	656,044	49,580	
-log(p)>3	24,387	1,238	120	119
-log(p)>5	2,194	159	27	49



# **Results: Heritability**

#### h<sup>2</sup>, phenotypic variance explained by GRM using selected SNP sets

GRMs	Sequence	HD	50k	СОЈО
All	0.83	0.82	0.81	
-log(p)>3	0.53	0.40	0.22	0.24
-log(p)>5	<del>0.60</del> *	<del>0.43</del> *	<del>0.22</del> *	0.16

\* Scale problems with GRM when estimating variances



# **Results: Genomic prediction**

#### Correlation between genomic breeding value and phenotype

GRMs	Sequence	HD	50k	COJO
All	0.68	0.68	0.68	
-log(p)>3	0.58	0.56	0.42	0.38
-log(p)>5	0.39	0.30	0.28	0.31





- Simply using sequence within Holstein population, unlikely to improve GS, but helps QTL detection.
- Subsets of selected SNP always poorer h<sup>2</sup> and GS
  - Full seq. accuracy GS of 0.68 and  $h^2 = 0.83$
  - 51 SNPs accuracy GS of 0.31 and  $h^2 \approx 0.16$  (DGAT!)

Good way to get realistic expectations from QTL.



# Acknowledgements



1000 bull genomes consortium

www.1000bullgenomes.com



