Outline

- **Use of genotyping by sequencing**

- **Disentangling genomic covariance**

- **Biases in breeding values**

Sequencing

- **Whole genome sequencing at high depth**

  - Limited samples (10's – 100's)
  - Other samples are imputed to full sequence
  - Rare allele imputation poor

- **Whole genome sequencing at low depth (<1x)**

  - Larger samples
  - Good haplotype reference needed (HapMap in humans)

- **Reduced genome sequencing**

  - Larger samples
  - Rare alleles available
  - Low depth causes missing data and bias

Determining a heterozygote genotype (AB) by sequencing

Reading	1 Read	2 reads	3 reads	4 reads	5 reads
AA	50%	25%	12.5%	6.3%	3.1%
AB	0%	50%	75%	87.5%	93.8%
BB	50%	25%	12.5%	6.3%	3.1%

Using low-depth genotyping by sequencing

- **Using as low as 1 read works!**

  - On average the genotype is OK

  - A phenotype has more value than the 'damage' of inaccurate genotype

  - Multiplexing and doing more samples at lower depth attractive

- **But**

  - Inaccuracy in genotype causes bias

    - Under-estimated allele effects in GWAS;

    - over-estimation of genomic variances
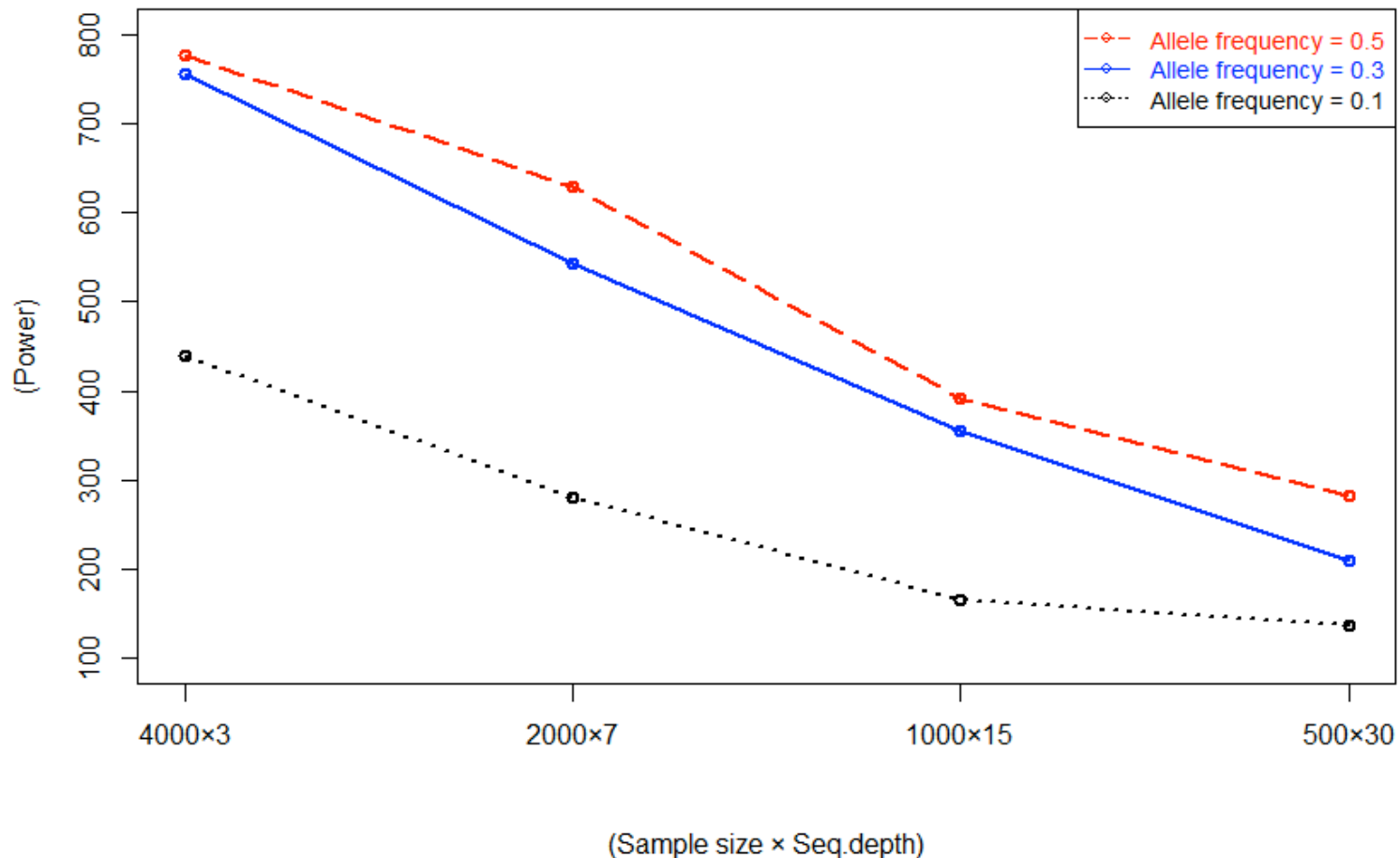
  - No reads = missing genotype!

Under-estimation of allele effect using GBS data for GWAS
From: Ashraf et al., Theor. Appl. Genet., 2014
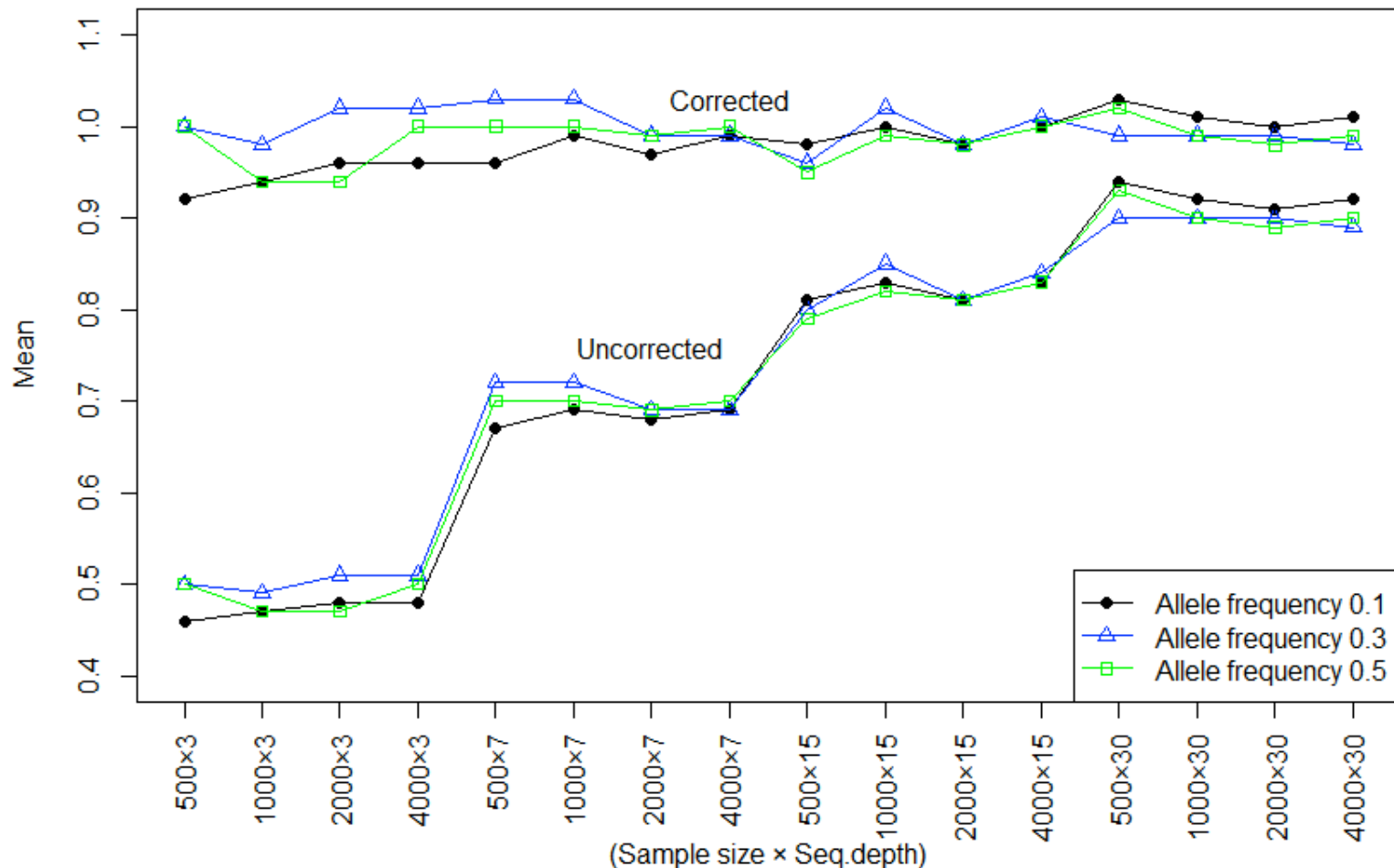
At constant sequencing, highest power to detect a significant association at low depth
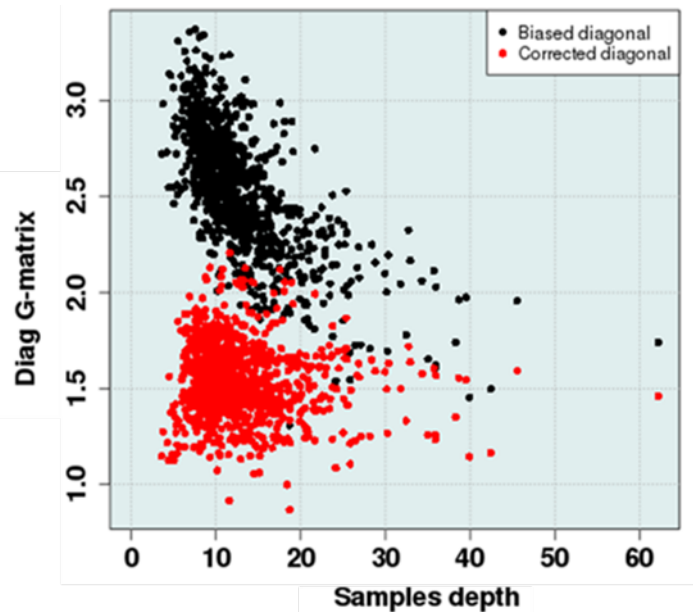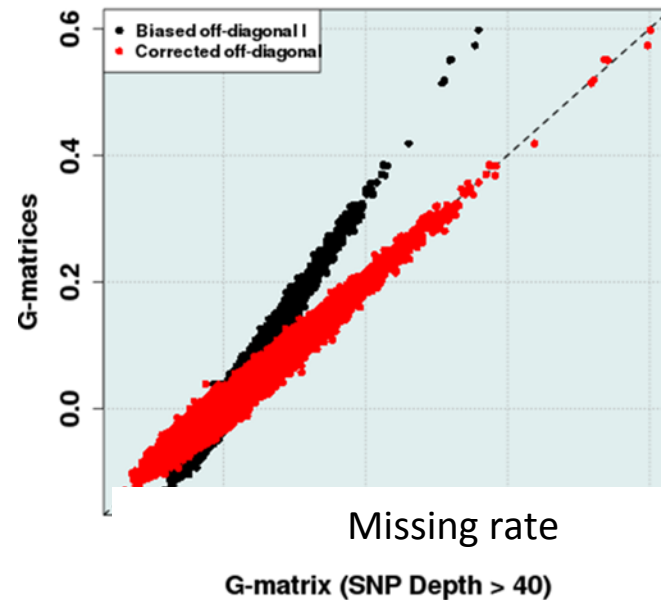## From: Ashraf et al., Theor. Appl. Genet., 2014

Bias from low-sequencing depth in computed genomic relationships
Cericola et al., PAG conference 2015

Diagonals are biased with sample sequencing depth

Off-diagonals are biased with sample missing rate when using standard VanRaden adjustment

Missing rate

Lessons learnt from genotyping by sequencing

- Great tool for genotyping
  - Cost effective, competitive with arrays
  - Directly applicable to tetraploids and pools
    - Use allele-frequency estimates instead of genotypes

- Biases due to genotype inaccuracy, but we learn to deal with it

- Implications (in animals)
  - For species without genome sequence and arrays
  - Similar biases for imputed genotyped?
  - Better missing rate adjustment

Disentangling covariance between traits: mapping SNPs to different correlation structures
Janss, WCGALP Conference 2014

- 748 F2 Mice

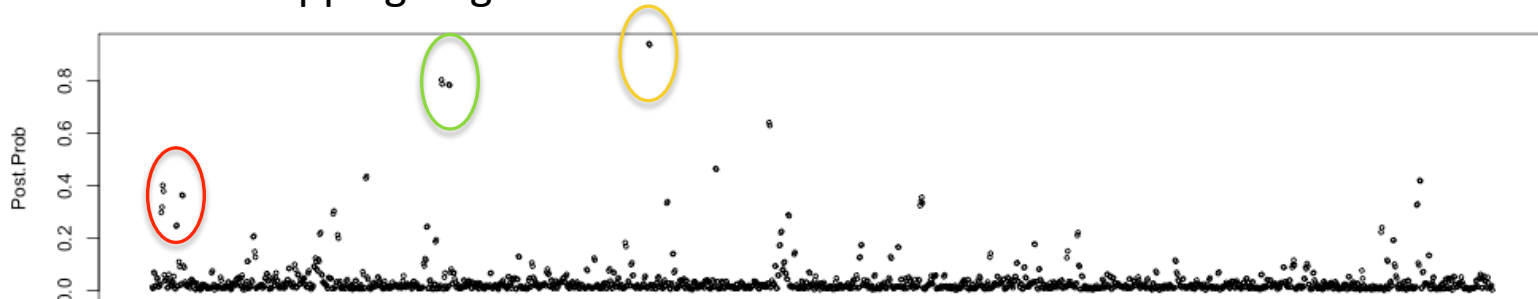- Body Weight, Feed Intake and Feed Efficiency

$$\begin{cases} BW = Xb_1 + \overset{>0}{r_1}g_1 + \overset{>0}{r_4}g_2 + e_1 \\ FI = Xb_2 + \overset{>0}{r_2}g_1 + \overset{<0}{r_5}g_2 + e_2 \\ FE = Xb_3 + r_3 g_1 + r_6 g_2 + e_3 \end{cases}$$
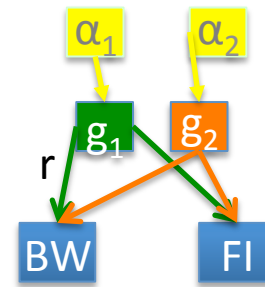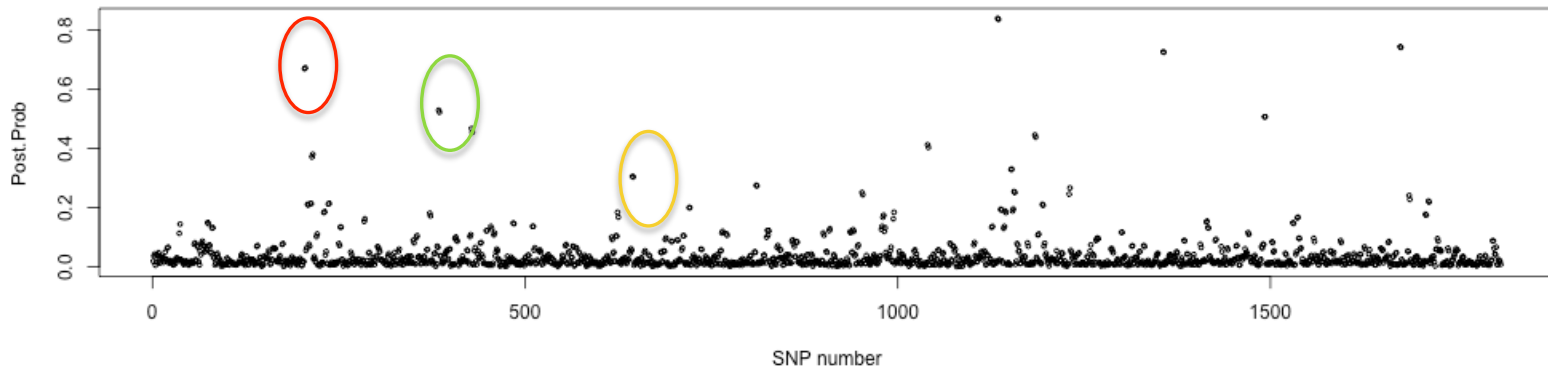

- Map SNPs to g1 (BW+, FI+) and g2 (BW+, FI-)

  using SNPs and mixture model in g1, g2
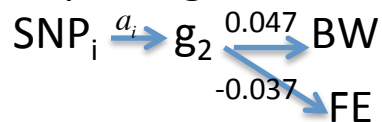
Two genomic values explaining BW, FI, FE

SNPs mapping to g1

SNPs mapping to g2

Interpreting SNP effects

$$SNP_i \xrightarrow{a_i} g_2 \xrightarrow{0.047} BW$$
$$\xrightarrow{-0.037} FE$$

Two genomic values explaining BW, FI, FE

Trait	Correlations g1	g2
BW	0.34	0.34
FI	0.40	0.08
FE	-0.07	0.49

Real (agricultural) applications

- Pooling data and prediction across breed
  - Consider traits in different breeds as different but correlated traits
  - Disentangle SNPs that contribute to correlation and those that are breed specific

- Multi-environment data
  - Consider traits multi-trait
  - Map common and environment specific SNPs
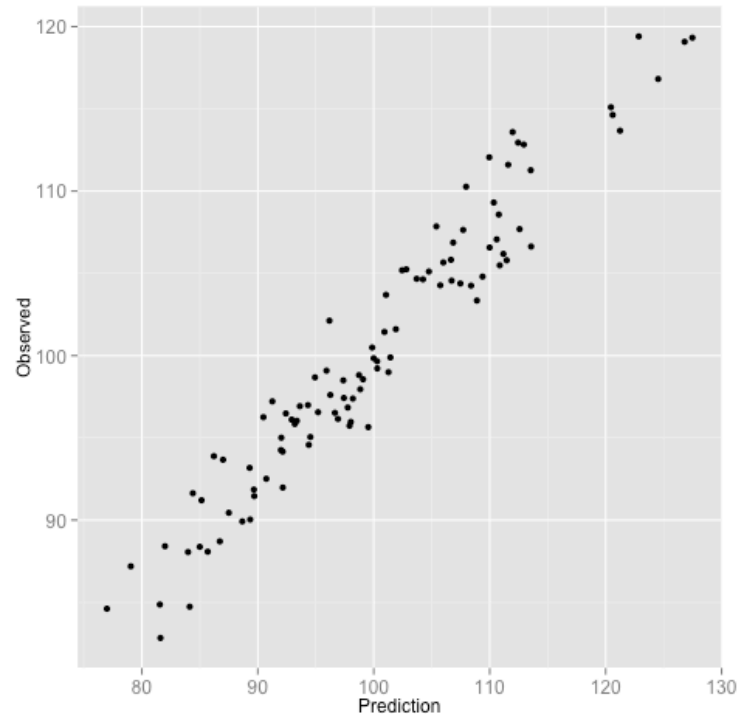 - Breeding values for all environments

Learning about biased breeding values

Unbiased predictions

Biased predictions
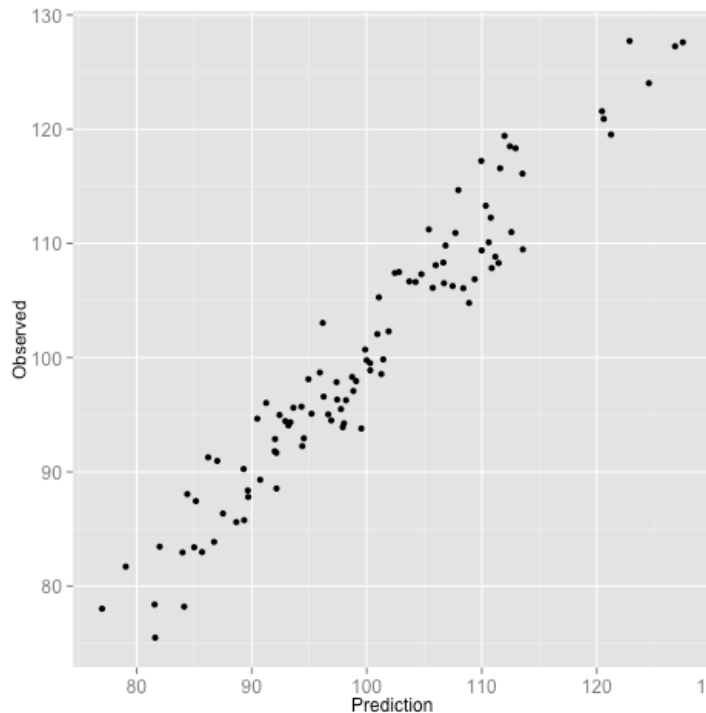
Can compare between old/new, genotyped/non-genotyped, different accuracies, etc.

Can not compare to anything outside the group

Data design and bias

Fixed

Trait = block + genotype + residual   Biased

Random

Trait = block + genotype + residual   Unbiased

Trait = block + genotype + residual   Biased

Trait = block + genotype + Parent x Parent + res   Unbiased

Summary

- Useful cross-fertilization between animals and plants
- Use of sequencing for genotyping
- Refinements in field-data analysis
- Multi-trait models for across breed & GxE

Thank you