



Genomic Selection Impacts

Julius van der Werf

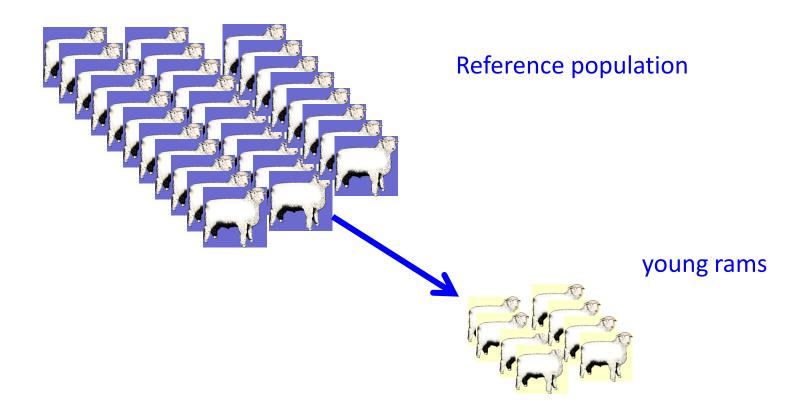
School of Environmental & Rural Science, UNE, Armidale

CRC for Sheep Industry Innovation, Armidale



ARMIDALE GENETICS

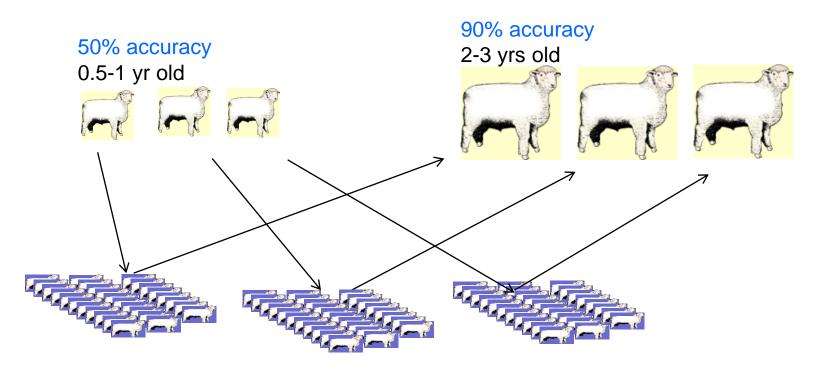
Genomic Prediction: basic idea



3) Computer centre can predict breeding value for young rams based on genomic relationship, combines it with other info

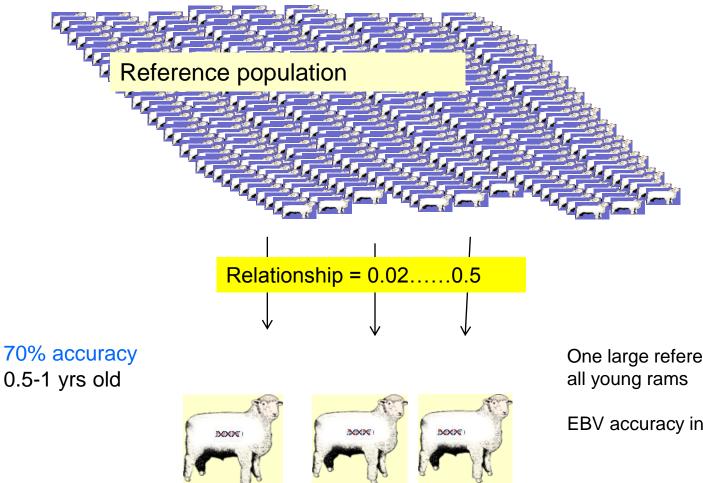
Can predict breeding value of young animals for 'any trait' measured in reference

Compare: Progeny Testing



Each progeny group only informs one sire

Genomic Testing



One large reference population informs all young rams

EBV accuracy increased at young age

Genomic Selection: Benefit

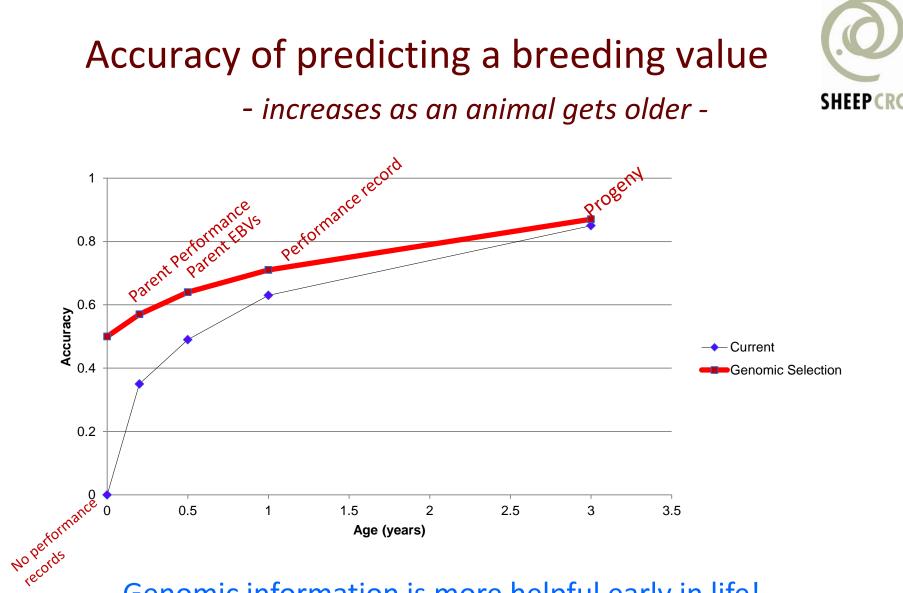
Overall:

More accurate prediction of genetic merit for breeding objective

Specific:

Traits that are usually difficult to improve difficult or expensive to measure can not be measured early low heritability

e.g. Carcass traits Lifetime time wool production Reproductive rate Parasite resistance



Genomic information is more helpful early in life!

Potential benefits of GS - some principles

% increase in EBV accuracy (male 1yo) and genetic gain

	$h^2 = 0$	$.1 = r^2$	$h^2 = 0.3 = r^2$		
Trait Measurability	%∆ Acc	%∆ Gain	%∆ Acc	%∆ Gain	
< 1 year, both sexes	15	7	7	7	
> 1 year, both sexes	68	19	59	37	
>1 year, females only	119	27	112	52	
on Corr. Trait, r _g = 0.9	20	12	20	26	
on Corr. Trait, r _g = 0.5	67	50	76	86	

Potential benefits of GS - some principles

% increase in EBV accuracy (male 1yo) and genetic gain

	$h^2 = 0.$	$1 = r^2$	$h^2 = 0$	$3=r^{2}$
Trait Measurability	%∆ Асс	%∆ Gain	%∆ Acc	%∆ Gain
< 1 year, both sexes	15	7	7	7
> 1 year, both sexes	68	19	59	37
>1 year, females only	119	27	112	52
on Corr. Trait, r _g = 0.9	20	12	20	26
on Corr. Trait, r _g = 0.5	67	50	76	86

These effects underestimated due to not accounting for Bulmer effect

Shifting the *trait* balance with genomic selection

Current				
Selection				
	Accuracy	Response		
Weight kg	0.71	0.79		
Dressing % Saleable meat	0.26	0.23		
yield %	0.33	0.29		
Overall Merit				
\$Index	0.58	2.03		





<u>% extra gain impact</u>

- Early trait small small small accuracy/ gen int
 Late Trait moderate gen int/acc
- Sex limited trait
 - females only, late
 - Males only early

very large small to modest

gen int acc/gen int

Benefits - Dairy



- Extra gain ~100%
- Breeding objective dominated by sex-limited trait
- No more progeny testing (save money)
- Very much shorter generation intervals
- More use of reproductive technologies
- Potential to select on hard to measure traits
 but only if these are being measured!
- Commercial males have more chance to be selected
- AI companies can easily afford testing
- Widely used in the industry

Benefits - Beef



- Extra gain ~25-50%
- Breeding objective has some hard to measure traits
- More emphasis on carcass and meat, less on growth
- More emphasis on females reproductive rate
- Somewhat shorter generation intervals
- More use of reproductive technologies
- Potential to select on hard to measure traits if these are being measured!

but only

- Genotyping cost can be high for breeders
- Who pay for the reference population?

Benefits - Sheep



- Extra gain ~25-50%
- Breeding objective has some hard to measure traits
- More emphasis on: carcass and meat, reproductive rate, 'lifetime wool', parasite resistance
- Somewhat shorter generation intervals
- Some more use of reproductive technologies
- Potential to select on hard to measure traits but only if these are being measured!
- Genotyping cost can be high for breeders
- Implemented in Australia, New Zealand
- Who pay for the reference population?

Benefits – Pigs & Poultry



- Extra gain ~50%?
- More emphasis on meat quality, Feed Efficiency?
- Sex limited traits
- shorter generation intervals in layers
- Potential to select on hard to measure traits if these are being measured!

but only

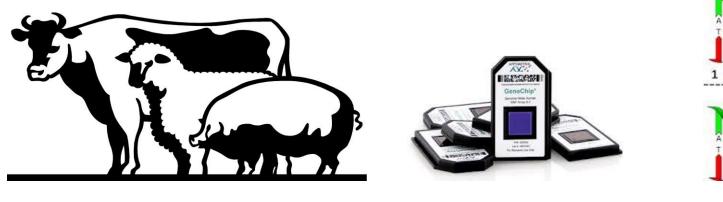
– Genotyping cost can be high?

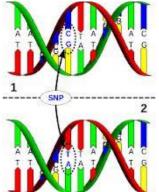
Conclusion



- Genomic selection can benefit breeding programs
- A challenge to implement:
 - cost to breeders
 - Need for phenotypes \rightarrow reference population / multi breed
- Reference population needs to contain (indirect) relatives of selection candidates *at this stage*
- Reference population needs to be continuously updated

Genomic information and inbreeding





Why use information from relatives?

- High degree of similarity between relatives
- More accurately predict breeding value

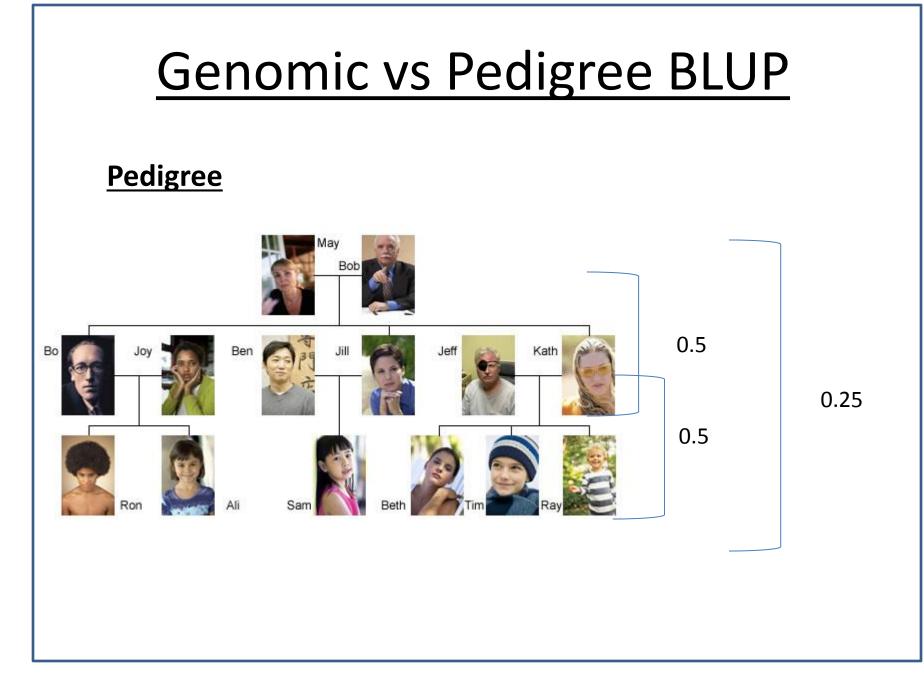


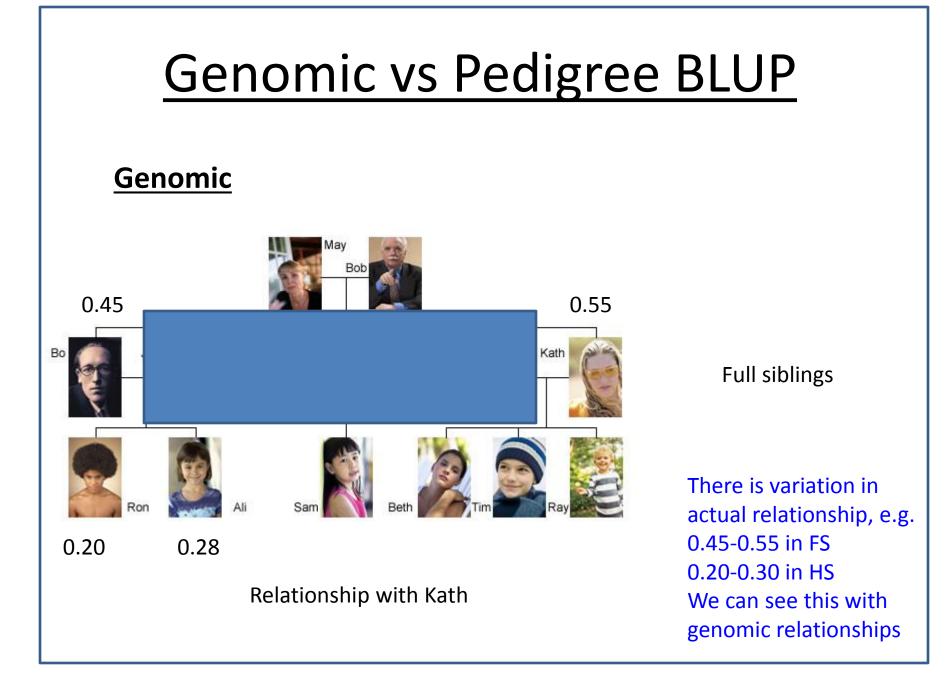


Relationships between Individuals

• Estimated using:

- Expected probabilities from PEDIGREE
- Estimated proportion of genome shared
 - Thousands of genetic markers (SNPs)





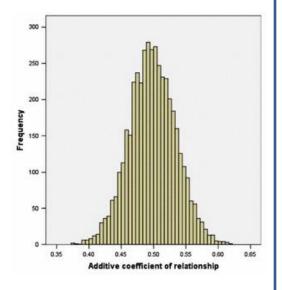
Estimates of relationship using genotypes:

- The expectations A
- Replaced by the estimated G

 Genomic relationship matrix



- But which half?
- Variation around the expectation?



There is variation in actual relationship, e.g. 0.40-0.50 in FS

We can see this with genomic relationships

Genomic Prediction: GBLUP

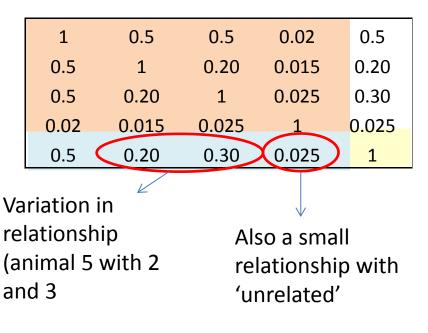
Example: Data on sire 1, his sons (2 and 3) and an unrelated individual (4)

want to predict 5 (also a son of 1) \leftarrow no data

A-matrix (pedigree-based)

1	0.5	0.5	0	0.5
0.5	1	0.25	0	0.25
0.5	0.25	1	0	0.25
0	0	0	1	0
0.5	0.25	0.25	0	1

G-matrix (DNA-based)



What information is used in BVs?

Clark et al, 2013 GSE

Va= ¼ sire + ¼ dam + ½ MS

Across family

Within Family

 Table 2- The proportion of variation in breeding value explained by between family (Sire and Dam) and within family (MS) information.

		N	NZ dairy bulls			Australian dairy bulls				
	BV	Sire	Dam	MS+e	Prop. of PT	BV	Sire	Dam	MS+e	Prop. of PT
Parent Ave	erage	0.56	0.44	0.001	0.001	PA EBV	0.44	0.52	0.04	0.05
Genomic E	BV	0.43	0.26	0.31	0.56	GEBV	0.33	0.37	0.30	0.36
Progeny Te	est	0.21	0.31	0.48	1.0	РТ	0.16	0.32	0.52	1.0

Correlation of breeding values and co-selection of relatives

Breeding value type	Half sib correlation	Full Sib correlation	Accuracy
Parent Average	0.55	1.0	0.45
Genomic BV	0.50	0.85	0.57
Progeny Test	0.26	0.53	1.0

Full Sibs - share the same Parent average BV (½ sire ½ dam)
- no longer the case with genomics

Half Sibs

- Share different PA breeding values
- Small advantage of using G to restrict inbreeding

Truncation selection on breeding values estimated using TBLUP or GBLUP

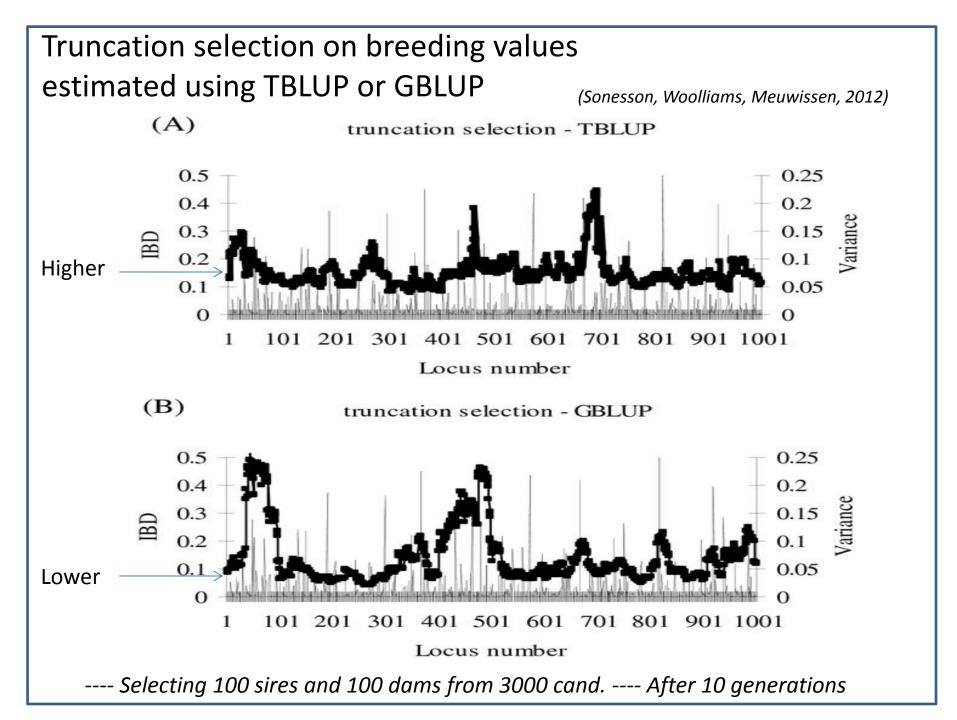
(Sonesson, Woolliams, Meuwissen, 2012)

	Genetic gain		Inbreeding
Breeding value estimation	ΔG (se)	ΔF _{ped} (se)	ΔF _{IBD} (se)
TBLUP	2.49 (0.035)	0.0156 (0.0001)	0.0235 (0.0009)
GBLUP	2.77 (0.026)	0.0053 (0.0002)	0.0209 (0.0005)

 $\Delta F_{PED} < \Delta F_{IBD}$

 $\Delta {\rm F}_{\rm PED}$ lower with GBLUP

 ΔF_{IBD} only slightly lower with GBLUP (and even less so if using BayesB)



Constraining Inbreeding:

Pedigree or Genomics, Optimal contributions

- Measures of genetic merit (ḡ)

 Pedigree vs genomic
 Pedigree based BLUP --- Genomic BLUP
- Measures of inbreeding
 - Pedigree vs genomic (A or G) NRM (Pedigree) --- GRM (genomic)

$$Max = \mathbf{c}_{t}' \, \bar{\mathbf{g}}_{t} - \lambda \mathbf{c}_{t}' \mathbf{A}_{t} \mathbf{c}_{t}$$

or: = $\mathbf{c}_{t}' \, \bar{\mathbf{g}}_{t} - \lambda \mathbf{c}_{t}' \mathbf{G}_{t} \mathbf{c}_{t}$

Measuring inbreeding

• Pedigree

- The probability that animals share alleles IBD.

- Genomics
 - GRM (IBS) or what is actually shared.
 - others

Genomic Inbreeding estimates

- Estimates of the number of homozygotes
 - Sharing of markers (IBS)
 - Long runs of homozygotes (more IBD)

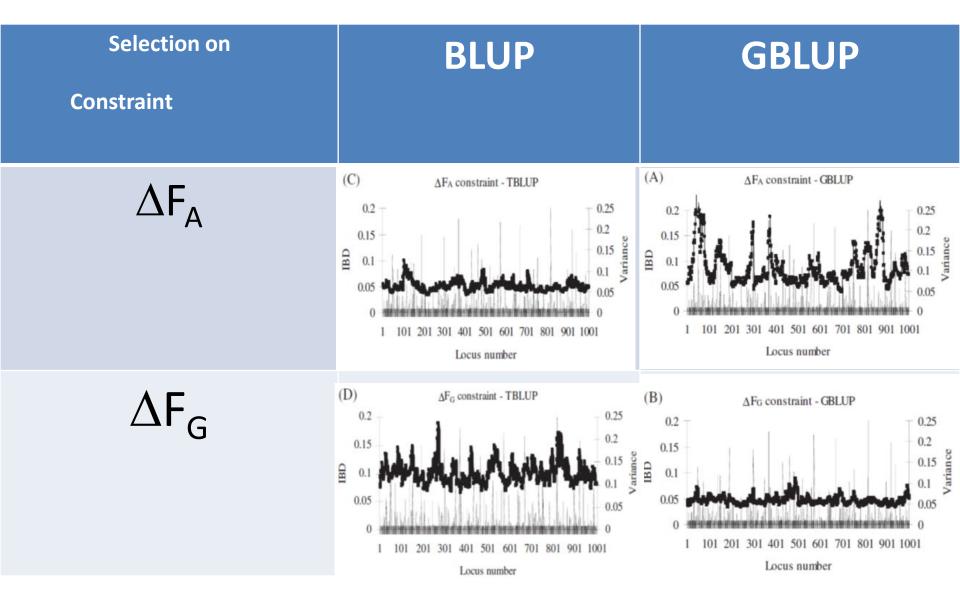
- Genomic relationships (IBS)
 - Various methods
 - Choice of allele frequencies

Ntest	ΔF _d	ΔG (se)	ΔF_{ped} (se)	ΔF _{IBD} (se)
		ΔF _A constraint – GBL	JP	
3000	0.005	3.08 (0.035)	0.0050 (0.0001)	0.0211 (0.0004)
6000	0.005	3.10 (0.035)	0.0048 (0.0001)	0.0226 (0.0004)
6000	0.010	3.31 (0.037)	0.0098 (0.0003)	0.0422 (0.0008)
		ΔF _G constraint – GBL	JP	
3000	0.005	1.91 (0.026)	0.0041 (0.0001)	0.0051 (0.0001)
6000	0.005	1.95 (0.024)	0.0039 (0.0001)	0.0053 (0.0001)
6000	0.010	2.41 (0.028)	0.0071 (0.0002)	0.0102 (0.0002)
		ΔF _A constraint – TBLU	IP	
3000	0.005	2.26 (0.003)	0.0050 (0.0001)	0.0068 (0.0001)
6000	0.005	2.50 (0.003)	0.0049 (0.0001)	0.0074 (0.0001)
6000	0.010	2.63 (0.003)	0.0102 (0.0002)	0.0151 (0.0003)
		ΔF _G constraint – TBLU	JP	
3000	0.005	1.41 (0.041)	0.0193 (0.0004)	0.0121 (0.0002)
6000	0.005	1.44 (0.039)	0.0185 (0.0004)	0.0122 (0.0002)
6000	0.010	1.48 (0.046)	0.0300 (0.0008)	0.0183 (0.0003)

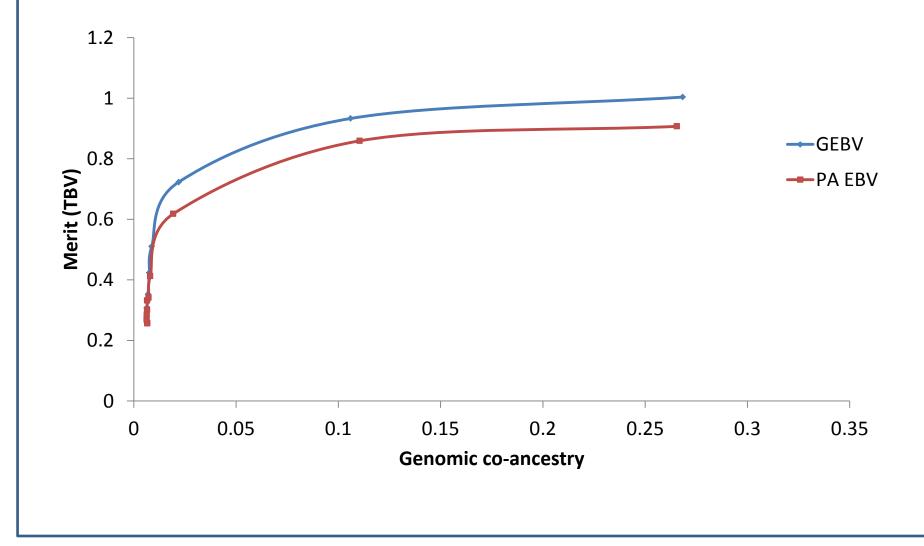
Constraining Inbreeding: Pedigree or Genomics, Optimal contributions

Selection on Constraint	BLUP	GBLUP
ΔF_A	$\Delta {\rm F_A}$ constrained $\Delta {\rm F_G}$ not well constrained	$\Delta {\rm F_A}$ constrained $\Delta {\rm F_G}$ badly constrained
ΔF_G	$\Delta { m F_A}$ not constrained $\Delta { m F_G}$ not constrained	$\Delta {\rm F_A}$ constrained $\Delta {\rm F_G}$ constrained

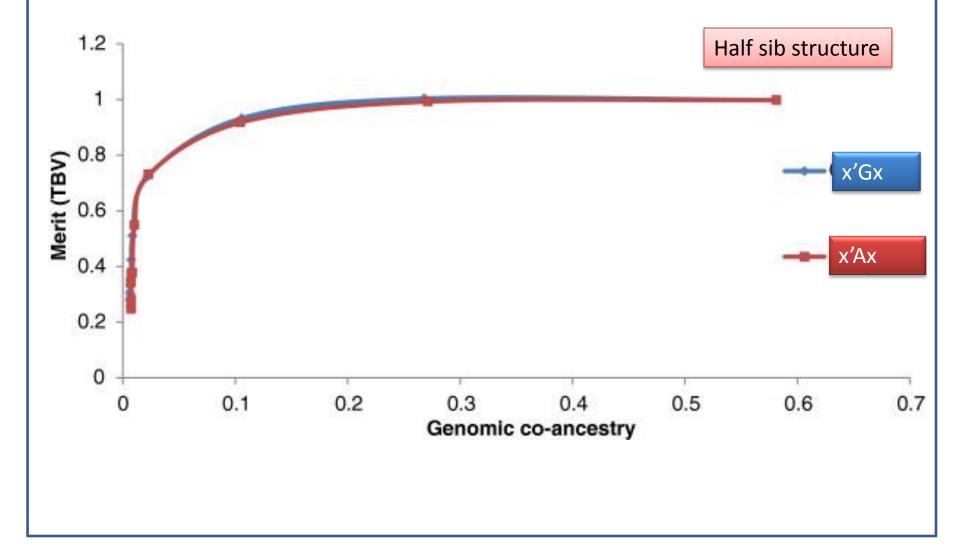
Constraining Inbreeding: Pedigree or Genomics, Optimal contributions



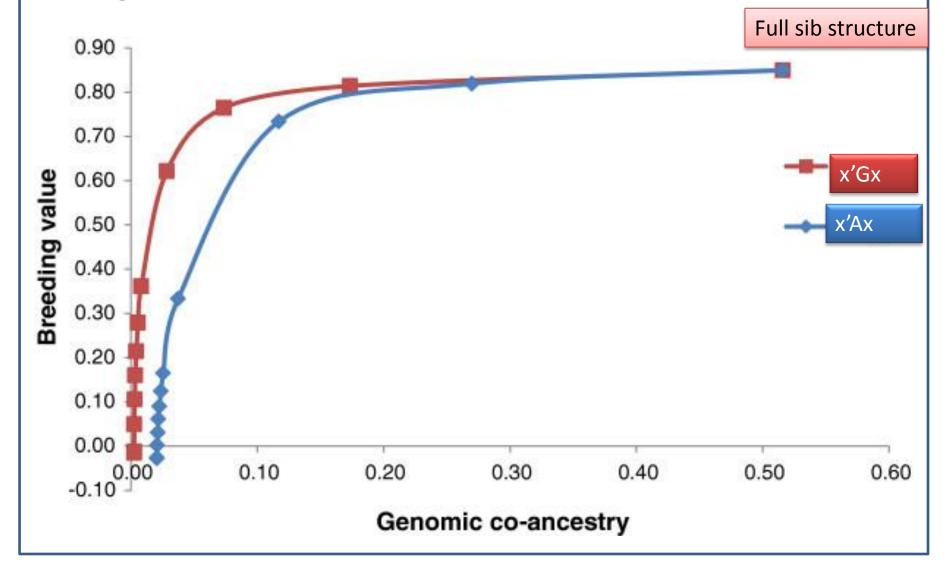
How much can we utilize additional diversity? selecting on GEBV vs EBV (parent average)



How much can we utilize additional diversity? using x'Ax vs x'Gx?



How much can we utilize additional diversity? using x'Ax vs x'Gx?



Genomic information helps to manage inbreeding

- 1. Uses genomic relationships helps to restrict genomic inbreeding.
- 2. Utilizes information about Mendelian sampling
 - 1. More accuracy: more ΔG for same ΔF
 - 2. More diversity \rightarrow more selection space giving raise to even more ΔG for same ΔF but mainly useful for large FS families