

# *Continuing the transformation*



## Predicting genetic changes with genomic information

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# Overview

- Genomic information can be used to achieve more genetic change  
– can we predict effect on genetic change, and inbreeding
- Combine genotype information with information from phenotypes and pedigree – ‘additional value’ of a genomic test
- How to best use genetic testing in breeding programs

# Genomic breeding values

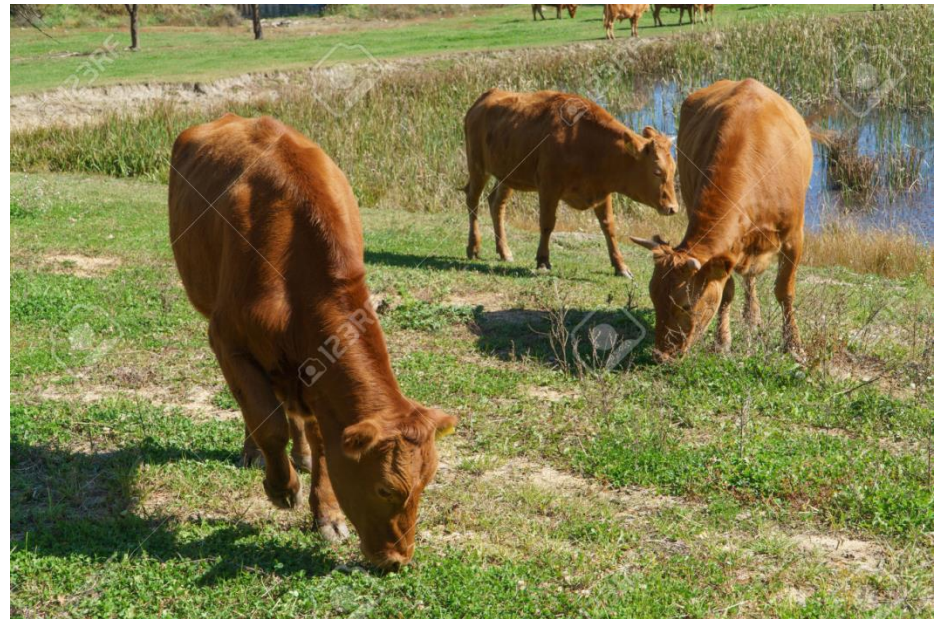
Good for:

Hard to measure, late in life traits      HTML

- Lean meat yield, meat quality
- Reproductive Rate
- Adult Weight

But how does it change  
selection response?

- Overall
- For each trait



# Potential benefits of GS - some principles

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

Trait Measurability	$h^2 = 0.1 = r^2$		$h^2 = 0.3 = r^2$	
	% $\Delta$ Acc	% $\Delta$ Gain	% $\Delta$ Acc	% $\Delta$ 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

→ More increase in accuracy when there is limited information

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Note that: gain = (accuracy \* selection intensity/generation interval) \* genetic SD

These effects even higher in selected populations (need to account for Bulmer effect)

# How is additional response calculated?

## ■ Selection index theory

- Index weights for various information sources
- Accuracies of EBV and GEBV -overall index and per trait-
- Response -overall index and per trait-
  - Some traits benefit more than other from GS

# Some definitions

		Information used to estimate it
■ TBV	True breeding value	
■ EBV	Estimated breeding value	Phenotype + Pedigree
■ GBV	Genomically estimated BV	DNA test
■ GEBV	Genomically enhanced estimated BV	Phenotype + Pedigree + DNA test

# Selection Index Approach

Genomic Selection: Predict TBV with Accuracy =  $x$

$$\rightarrow \text{GS explains } x^2 \% \text{ of } V_A \rightarrow V_{\text{qtl}} = x^2 V_A$$

*Lande and Thompson, 1990 Genetics:*

$$V_{\text{polygenic}} = (1-x^2)V_A$$

## Models

Without GS:  $V_{\text{Pheno}} = V_{\text{AddGen}} + V_{\text{error}}$

With GS:  $V_{\text{Pheno}} = V_{\text{polygenic}} + V_{\text{qtl}} + V_{\text{error}}$

Predict from  
phenotypes pedigree

Predict from DNA markers



# Selection Index *some formal stuff*

*for reference only*

First summarize some definitions

$$I = b_1x_1 + b_2x_2 + \dots + b_nx_n$$

$$\text{var}(X) = P$$

Index:

$X_i$  = selection criteria

$b_i$  = index weight

Single trait breeding objective:

$$H = A \text{ (breeding value)}$$

Breeding Objective

$\text{Cov}(X, A) = G$  (a vector with ST objective)

Optimal weights are  $b = P^{-1}G$

Calculate optimal index weights given the breeding objective

$$\text{Var}(I) = \text{var}(b'X) = b'\text{var}(X)b = b'Pb = \sigma_I^2$$

$$\text{Var}(H) = \sigma_a^2$$

Variance of index = covariance between index and breeding objective

$$\text{Cov}(I, H) = \text{cov}(b'X, A) = b'\text{cov}(X, A) = b'G = b'Pb = \sigma_I^2$$

$$b = P^{-1}G \rightarrow Pb = G$$

# Accuracy of selection index (single trait)

$r_{IA}$  = correlation between Index (=EBV) and A

$$= \frac{\text{cov}(I,A)}{\sigma_I \sigma_A} = \frac{\sigma_I^2}{\sigma_I \sigma_A} = \frac{\sigma_I}{\sigma_A} = \sqrt{(b'Pb / \sigma_a^2)}$$

Because  $\text{cov}(I,A) = \text{var}(I)$

Selection Index = Best Linear Prediction BLP

Index (I) is best estimate of breeding value:  $I = E(A | X) = \text{cov}(X,A) / \text{var}(X)$

Same as BLUP, but without fixed effects.

$$\text{Var}(I) = \text{var}(EBV) = r_{IA}^2 \cdot \text{var}(BV) = r_{IA}^2 \sigma_a^2 \quad r_{IA}^2 \text{ also known as reliability}$$

# Selection Index Approach with genomic info

use info on various information sources: *below for one trait only*

Variance-covariance of information sources  
P-matrix

covariance with TBV  
G-matrix

Without  
GS:

Own perf.	$V_p$					
Sire		$V_p$				
Dam			$V_p$			
FullSibs	etc			$\{t-(1-t)/n\}V_p$		
HalfSibs					$\{t-(1-t)/n\}V_p$	
Progeny						$\{t-(1-t)/n\}V_p$

$V_a$
$V_a/2$
$V_a/2$
$V_a/2$
$V_a/4$
$V_a/2$

With  
GS:

Own perf.	$V_p - V_q$						0
Sire		$V_p - V_q$					0
Dam			$V_p - V_q$				0
FullSibs	etc			$\{t-(1-t)/n\}(V_p - V_q)$			0
HalfSibs					$\{t-(1-t)/n\}(V_p - V_q)$		0
Progeny						$\{t-(1-t)/n\}(V_p - V_q)$	0
QTL	0	0	0	0	0	0	$V_q$

$V_a - V_q$
$(V_a - V_q)/2$
$(V_a - V_q)/2$
$(V_a - V_q)/2$
$(V_a - V_q)/4$
$(V_a - V_q)/2$
$V_q$

# Selection Index Approach with genomic info

Pseudo BLUP: Genomic Breeding value is an additional trait with  $h^2 = 1$

Without  
GS:

Own perf.	$V_p$							$V_a$
Sire		$V_p$						$V_a/2$
Dam			$V_p$					$V_a/2$
FullSibs	etc			$\{t+(1-t)/n\}V_p$				$V_a/2$
HalfSibs					$\{t+(1-t)/n\}V_p$			$V_a/4$
Progeny						$\{t+(1-t)/n\}V_p$		$V_a/2$

With  
GS:

Own perf.	$V_p$						$V_q$	$V_a$
Sire		$V_p$					$V_q/2$	$(V_a)/2$
Dam			$V_p$				$V_q/2$	$(V_a)/2$
FullSibs	etc			$\{t+(1-t)/n\}(V_p)$			$V_q/2$	$(V_a)/2$
HalfSibs					$\{t+(1-t)/n\}(V_p)$		$V_q/4$	$(V_a)/4$
Progeny						$\{t+(1-t)/n\}(V_p)$	$V_q/2$	$(V_a)/2$
QTL	$V_q$	$V_q/2$	$V_q/2$	$V_q/2$	$V_q/4$	$V_q/2$	$V_q$	$V_q$

P-matrix

G-matrix

# Selection index: example of 2 approaches own phenotype + GBV

$$h^2 = 0.5$$

GBV reliability “ $x^2$ ” = 0.5

	P		G (,BV)		b	varIndex	acc														
Phenotype	1		0.5	a	0.5000	0.2500	0.7071														
Phenotype	<table style="width: 100%; border-collapse: collapse;"> <tr><td style="padding: 5px;">P</td><td style="padding: 5px;">1</td><td style="padding: 5px;">0.25</td></tr> <tr><td style="padding: 5px;">mBV</td><td style="padding: 5px;">0.25</td><td style="padding: 5px;">0.25</td></tr> </table>		P	1	0.25	mBV	0.25	0.25	<table style="width: 100%; border-collapse: collapse;"> <tr><td style="padding: 5px;">G</td><td style="padding: 5px;">0.5</td><td style="padding: 5px;">a</td></tr> <tr><td style="padding: 5px;">q</td><td style="padding: 5px;">0.25</td><td style="padding: 5px;">q</td></tr> </table>		G	0.5	a	q	0.25	q	<table style="width: 100%; border-collapse: collapse;"> <tr><td style="padding: 5px;">0.3333</td></tr> <tr><td style="padding: 5px;">0.6667</td></tr> </table>	0.3333	0.6667	0.3333	0.8165
P	1	0.25																			
mBV	0.25	0.25																			
G	0.5	a																			
q	0.25	q																			
0.3333																					
0.6667																					
Corrected Phenotype	<table style="width: 100%; border-collapse: collapse;"> <tr><td style="padding: 5px;">P</td><td style="padding: 5px;">0.75</td><td style="padding: 5px;">0</td></tr> <tr><td style="padding: 5px;">mBV</td><td style="padding: 5px;">0</td><td style="padding: 5px;">0.25</td></tr> </table>		P	0.75	0	mBV	0	0.25	<table style="width: 100%; border-collapse: collapse;"> <tr><td style="padding: 5px;">G</td><td style="padding: 5px;">0.25</td><td style="padding: 5px;">u</td></tr> <tr><td style="padding: 5px;">q</td><td style="padding: 5px;">0.25</td><td style="padding: 5px;">q</td></tr> </table>		G	0.25	u	q	0.25	q	<table style="width: 100%; border-collapse: collapse;"> <tr><td style="padding: 5px;">0.3333</td></tr> <tr><td style="padding: 5px;">1.0000</td></tr> </table>	0.3333	1.0000	0.3333	0.8165
P	0.75	0																			
mBV	0	0.25																			
G	0.25	u																			
q	0.25	q																			
0.3333																					
1.0000																					

Note weights on QTL info

MBV = GBV = “QTL”

# Selection index: example of 2 approaches

- information from relatives

	P		G		b	VarIndex	accuracy
ownPoly	0.75	0	0.125	0.25	0.3143	0.3429	0.8281
Own GBV	0	0.25	0	0.25	1.0000		
sirepoly	0.125	0	0.75	0.125	0.1143		

# Selection index: example of 2 approaches

- information from relatives

	P				G	b	VarIndex	accuracy
ownPoly	0.75	0	0.125	0	0.25	0.3143	0.3429	0.8281
ownGBV	0	0.25	0	0.125	0.25	1		
sirepoly	0.125	0	0.75	0	0.125	0.1143		
sireMBV	0	0.125	0	0.25	0.125	0		

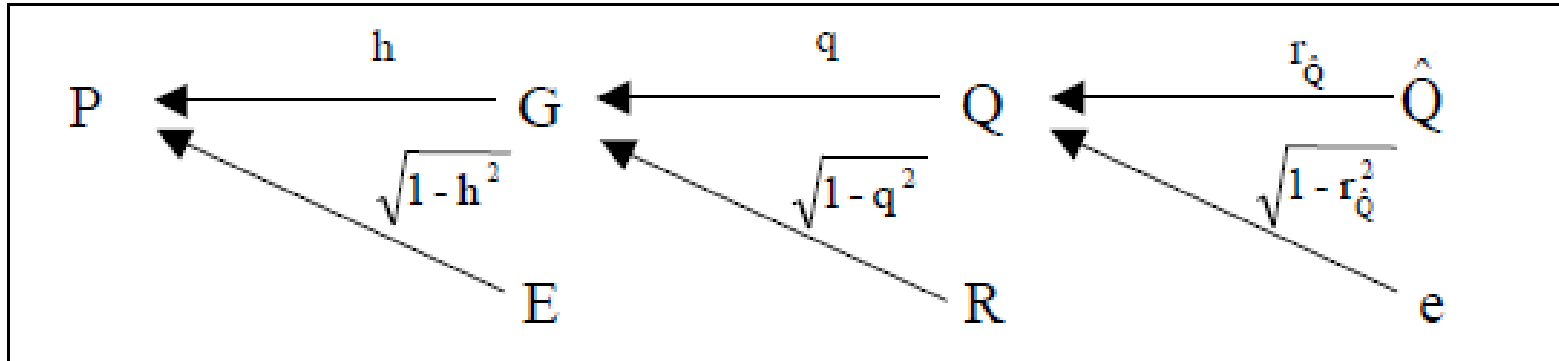
	P				G	b	varIndex	acc
ownPoly	0.75	0	0.125		0.25	0.3182	0.3409	0.8257
ownGBV	0	0.25	0.125		0.25	0.9545		
Sirepheno	0.125	0.125	1		0.25	0.0909		

Top: sire is genotyped, bottom: sire not genotyped, not same accuracy

Conclusion: Relatives info needs to be 'corrected for markers'

# Path coefficient method following

*Dekkers Dec 2007 JABG*



P = Phenotype

G = Breeding Value

Q = BV component associated with markers (=GBV)

$Q_{\text{hat}}$  = estimate of Q

Accuracy GBV = "x" =  $q \cdot r_{Q_{\text{hat}}}$

Phenotypic correlation:  $r_{P, Q_{\text{hat}}} = h \cdot x$

Genetic correlation  $r_{G, Q_{\text{hat}}} = x$



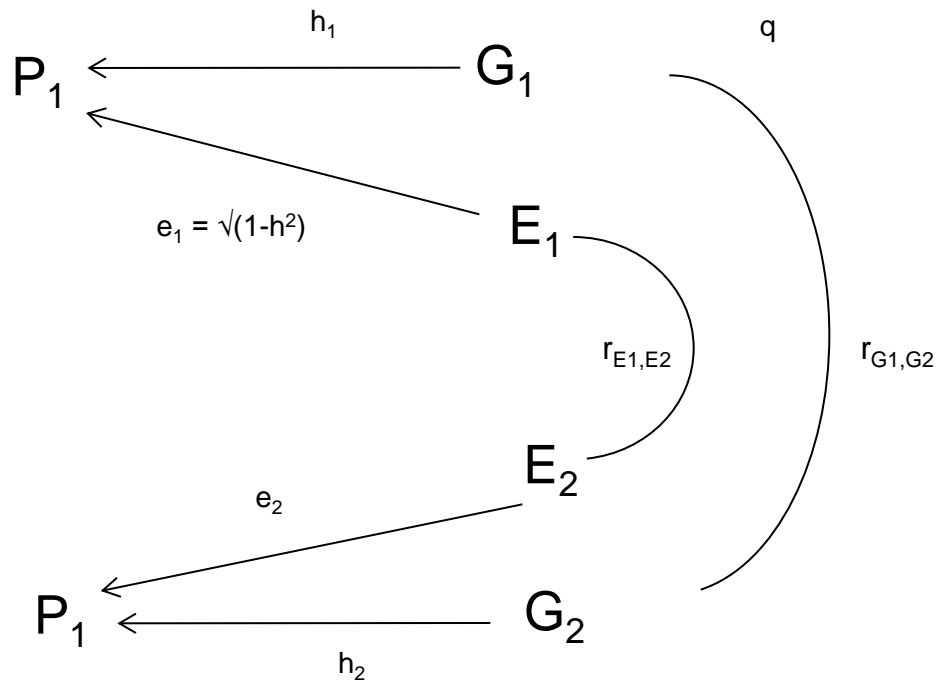
# Conclusion: single trait

- Can include GBV as a correlated trait
  - And use standard software for selection index
- $r_g =$  accuracy, same as 'x'
- $r_p = h.x$
- econ value for GBV = 0
- This is equivalent to treating it as an extra info source in a single trait multiple info sources approach:  
EBV = f(own perf, dam, sire, sibs, progeny, GBV)

## Extension to multiple traits

- Some traits may have GBV, others may not
- Need correlations....
  - between GBV and other trait phenotypes
  - between GBV and other trait genotypes
  - between different GBVs
- These can be predicted from genetic correlations between traits, only when assuming infinitesimal model

# Path coefficient method Dekkers Dec 2007 JABG



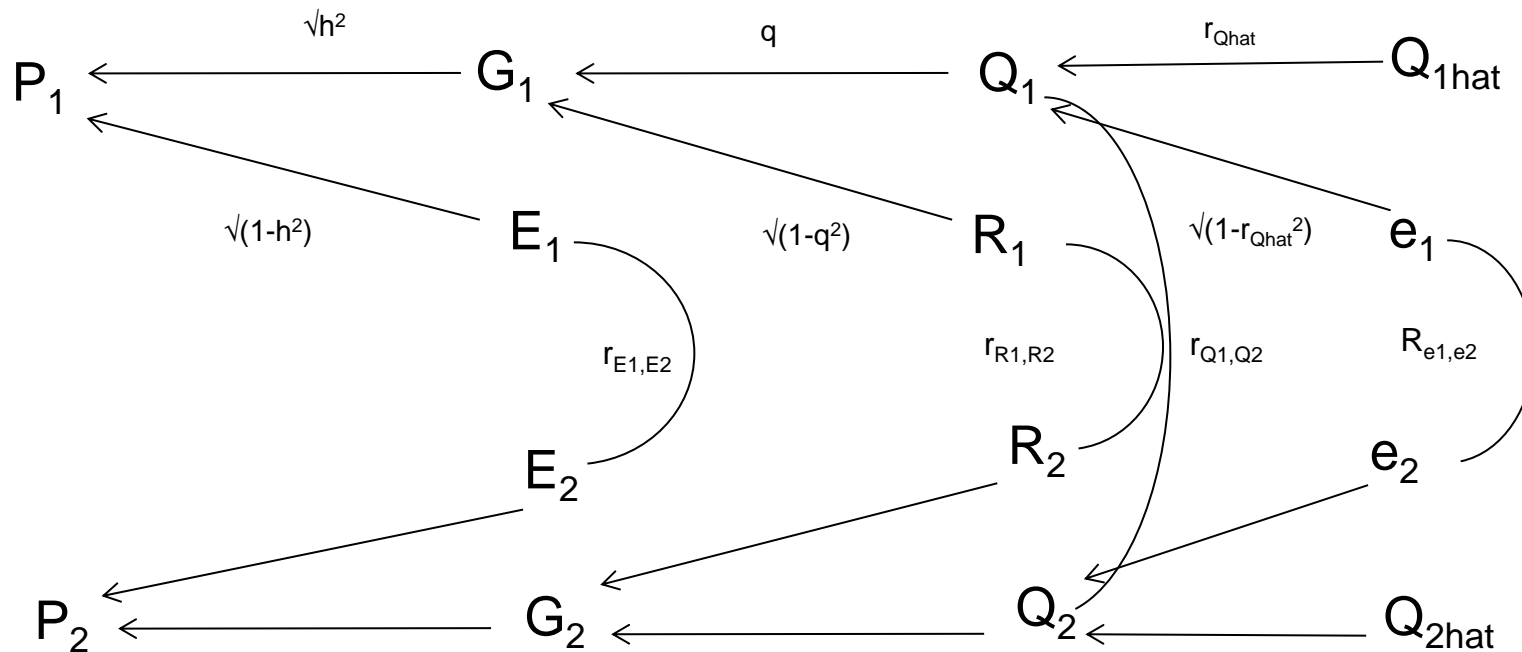
$$r_{P_1 G_2} = h_1 r_{G_1, G_2}$$

$$r_{p_1 p_1} = h_1 h_2 r_{G_1, G_2} + e_1 e_2 r_{E_1, E_2}$$

$$e_i = \text{sqrt}(1-h_i^2)$$

# Path coefficient method following

*Dekkers Dec 2007 JABG*



$$r_{G_i, Q_{hatj}} = r_{Q_{1hat}} \cdot r_{Q_1, Q_2}$$

$$r_{Q_{hati}, Q_{hatj}} = r_{Q_{1hat}} \cdot r_{Q_{2hat}} \cdot r_{Q_1, Q_2}$$

$$r_{P_i, Q_{hatj}} = h_i r_{Q_{1hat}} \cdot r_{Q_1, Q_2}$$

# Summary



To predict accuracy of GEBV we can use selection index approach

- Either: GBV + polygenic (no correlation)
- Or: GBV + P, correlation is  $r^2$
- The latter is easier: Genomic BV as a correlated trait.

$r_g$  = accuracy of GBV = 'x'

$r_p$  = h.x

Econ value of GBV = 0

# How much genetic change?

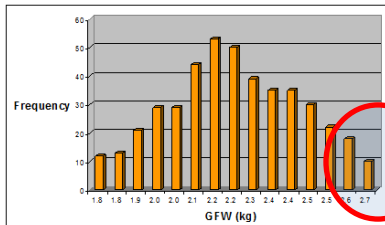
Breeder's Equation

Selection Response

$$= i r_{IA} \sigma_A$$



Selection accuracy



Selection intensity



Genetic variation

# How much genetic change?

Selection Differential

Mean of selected parents

$$S = i r_{IA} \sigma_A$$

# Response per year

$$S = i r_{IA} \sigma_A$$

- Response per generation

$$\frac{1}{2} S_{\text{sires}} + \frac{1}{2} S_{\text{dams}}$$

Superiority of parents  
averaged over males and females

- Response per year

$$\frac{\frac{1}{2} S_{\text{sires}} + \frac{1}{2} S_{\text{dams}}}{\frac{1}{2} L_{\text{sires}} + \frac{1}{2} L_{\text{dams}}} = \frac{S_{\text{sires}} + S_{\text{dams}}}{L_{\text{sires}} + L_{\text{dams}}}$$

Generation interval (in years)  
averaged over males and females

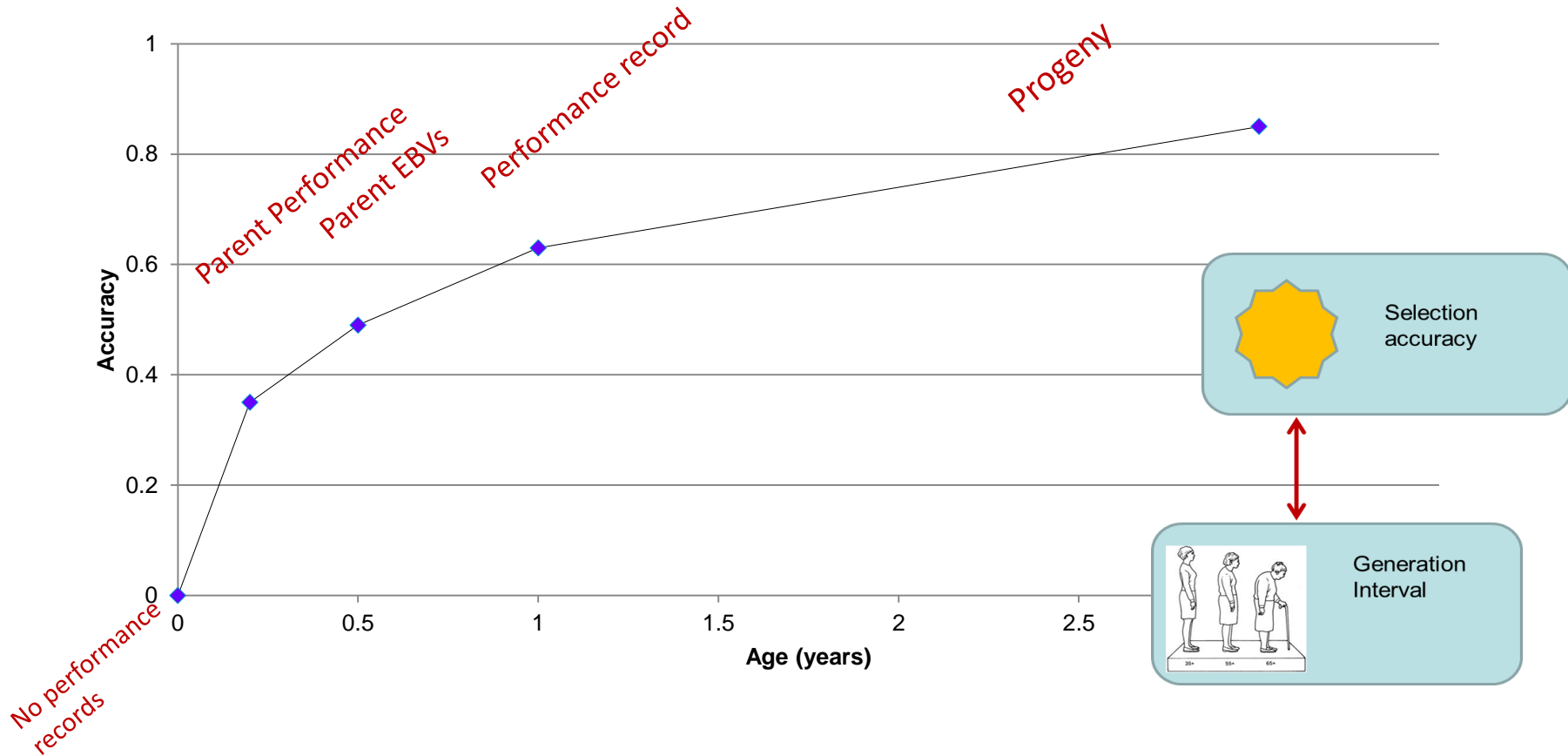
Generation interval is  
average age of sires (dams)  
when their progeny are born



$$R = \frac{i_m r_m + i_f r_f}{L_m + L_f} \sigma_A$$

# Accuracy of predicting a breeding value (r)

- increases as an animal gets older

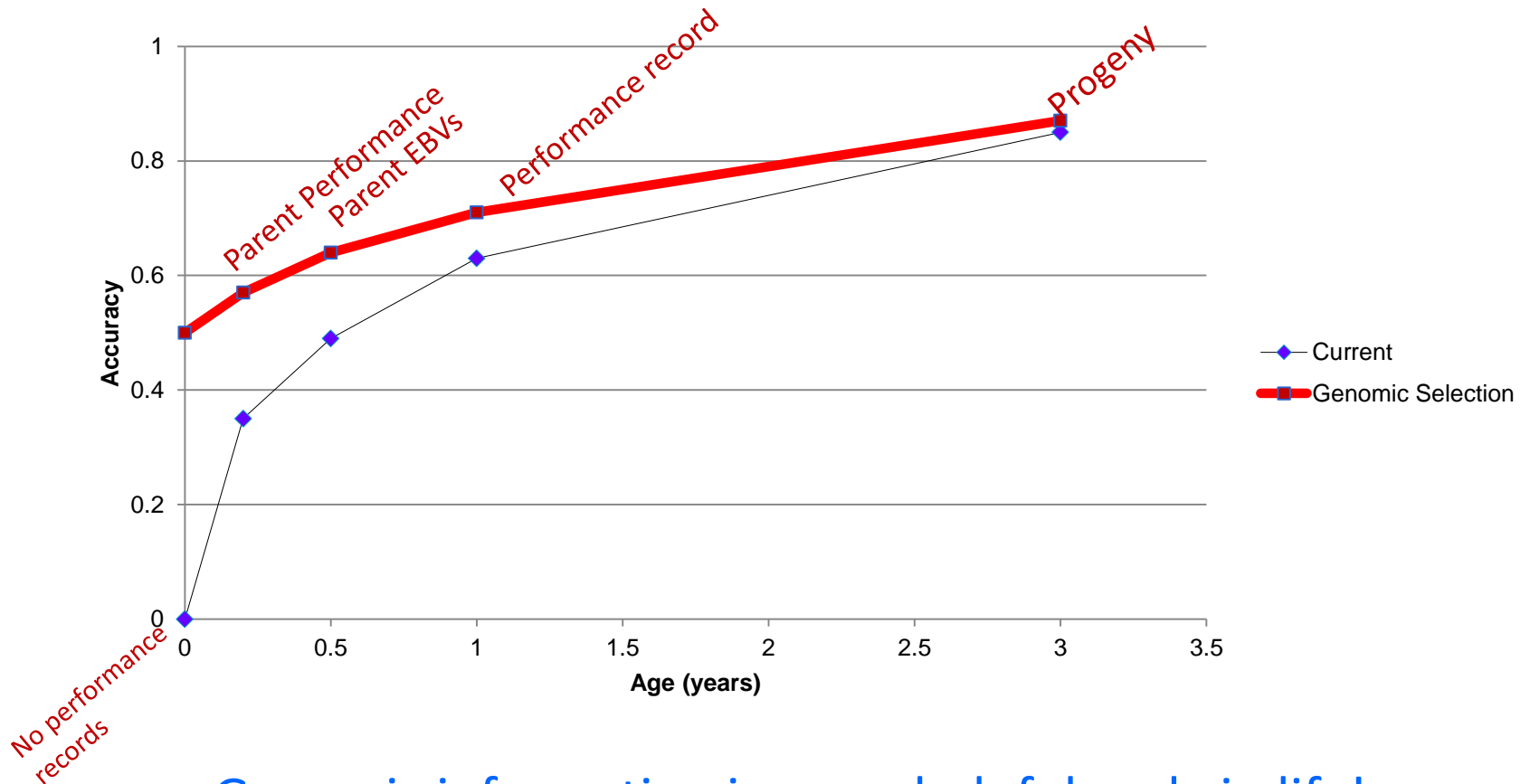


Assumed heritability = 25%

*Need to balance accuracy and generation interval!*

# Accuracy of predicting a breeding value

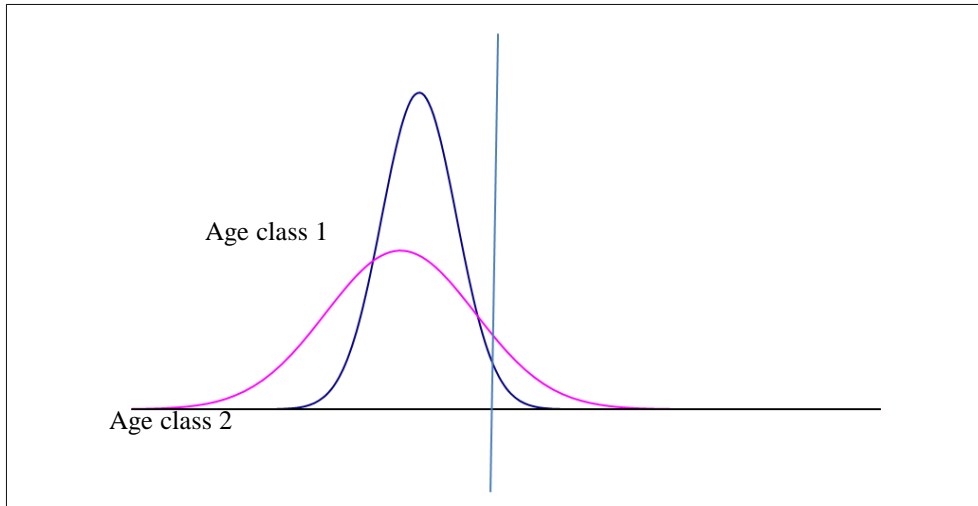
- increases as an animal gets older -



Genomic information is more helpful early in life!

# Optimizing age structure

Accuracy changes with age class !



Without genomic selection

ageclass	N in group	mean	SD	Nr Selected
1	50	10.20	0.4	2.7
2	50	10.00	0.8	7.3

Accuracy

With genomic selection

ageclass	N in group	mean	SD	Nr Selected
1	50	10.20	0.7	5.4
2	50	10.00	0.8	4.6

# Potential benefits of GS - some cases

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

Trait Measurability	$h^2 = 0.1 = r^2$		$h^2 = 0.3 = r^2$	
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on Corr. Trait, $r_g = 0.5$	67	50	76	86

These effects underestimated due to not accounting for Bulmer effect

# Effect of GS on genetic change

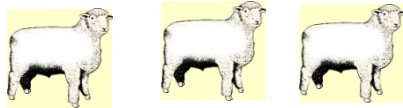
- More accuracy of GEBV,
  - esp younger animals
  - Depends on trait measurability (early/late trait, sex limited)
- More response due to higher accuracy
- .....Or lower generation interval
- E.g. dairy: accuracy of GBV is lower than of progeny test, but generation interval can be much reduced

# Benefits across Species

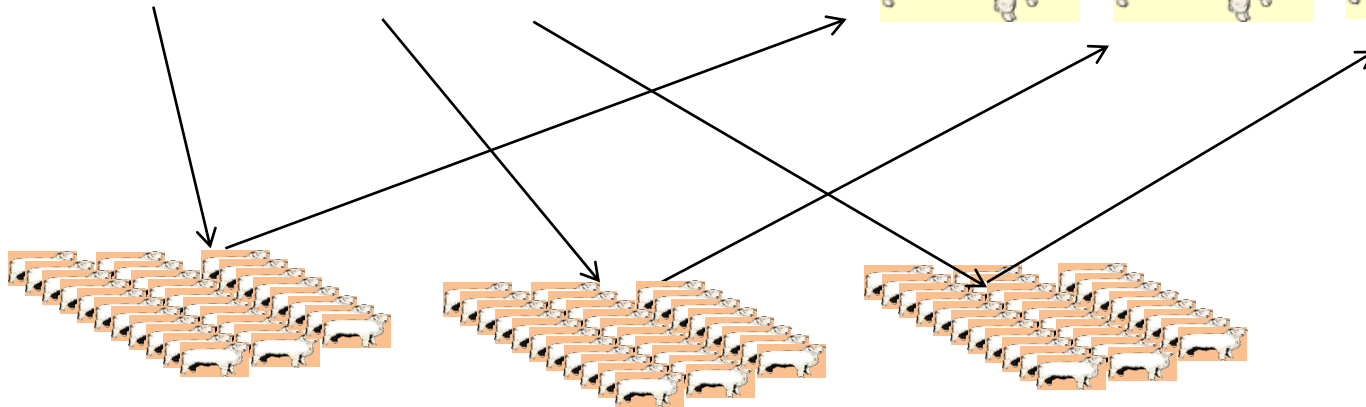
	<u>% extra gain</u>	<u>impact</u>
• Early trait	small	small accuracy/ gen int
• Late Trait	moderate	gen int/acc
• Sex limited trait		
– females only, late	very large	gen int
– Males only early	small to modest	acc/gen int

# Compare: Progeny Testing

50% accuracy  
0.5-1 yr old

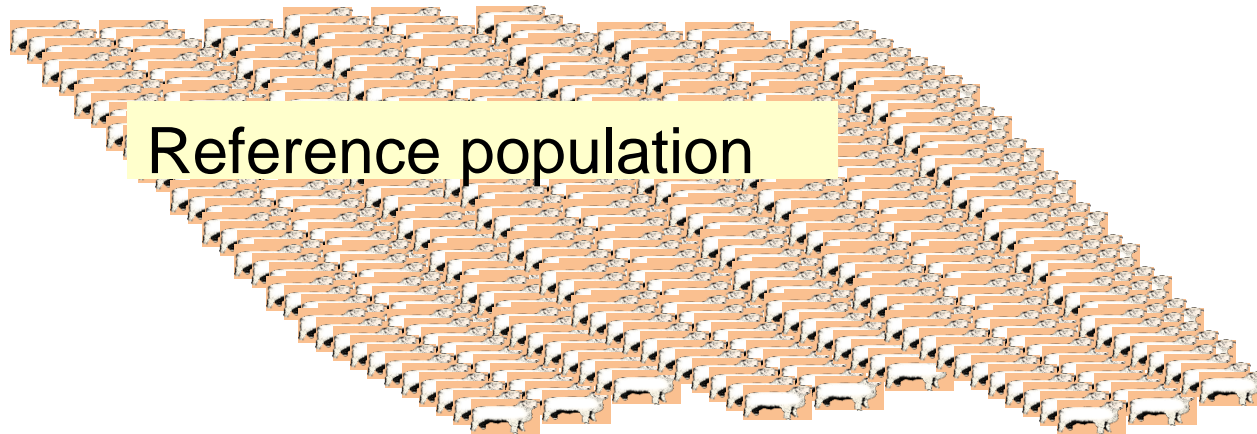


90% accuracy  
2-3 yrs old



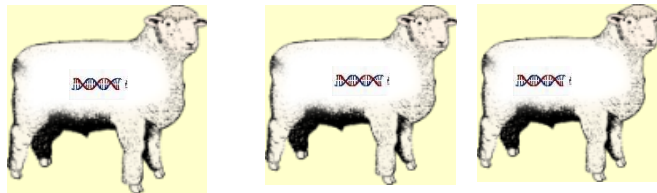
Each progeny group only informs one sire

# Genomic Testing



Relationship = 0.02.....0.5

70% accuracy  
0.5-1 yrs old



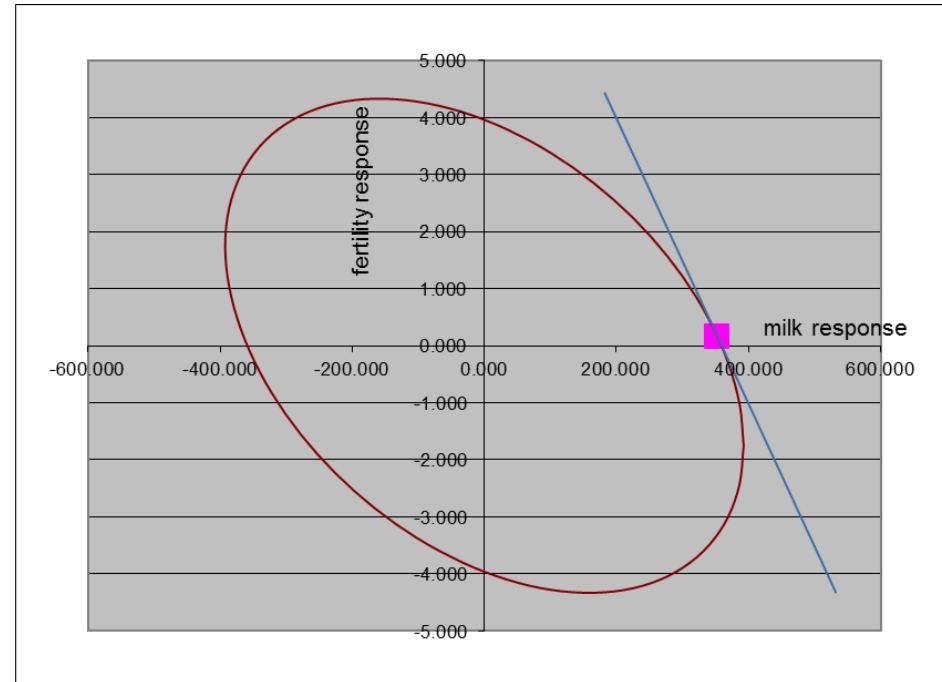
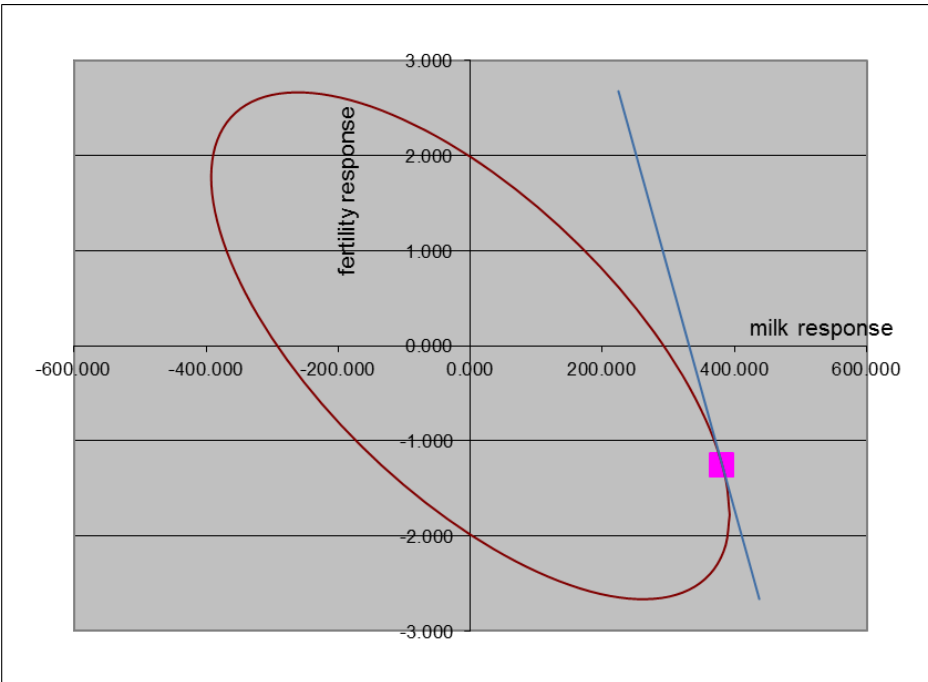
One large reference population informs all young rams

EBV accuracy increased at young age



# Genomic Selection also changes the balance between traits

## *Selection for milk Yield and Fertility*



	economic weights		progeny measured		response (4 yrs)	
	milk	fertility	milk	fertility	milk	feed
left	0.2	<b>8</b>	50	10	381	-1.25
right	0.2	<b>8</b>	50	50	352	0.17

# Shifting the *trait* balance with genomic selection

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

# Shifting the *trait* balance with genomic selection

	Current Selection		Genomic Selection		Difference
	Accuracy	Response	Accuracy	Response	
Weight kg	0.71	0.79	0.75	0.76	-4%
Dressing %	0.26	0.23	0.59	0.42	83%
Saleable meat yield %	0.33	0.29	0.60	0.46	59%
Overall Merit \$Index	0.58	2.03	0.69	2.43	20%

*Note: not only more gain overall, but shift to HTML traits*

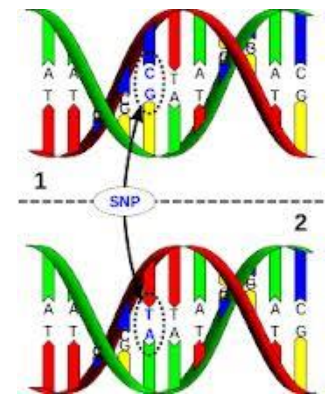
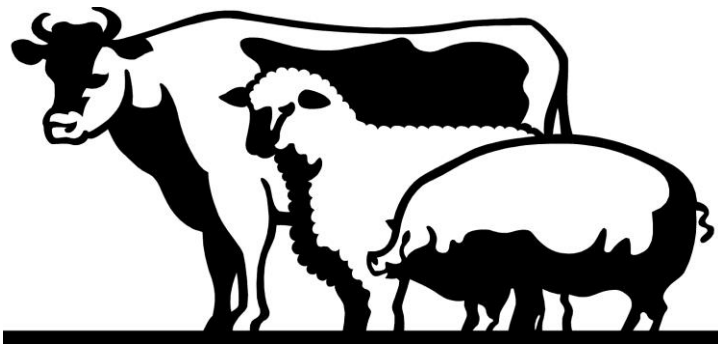
# Effect of GS on genetic change

With multiple trait selection:

The accuracy of hard to measure trait will improve more, and as a result, these traits get more pushed,

possibly at the expense of the easy to measure traits (but these were 'overemphasized' response before GS)

# Genomic information and inbreeding

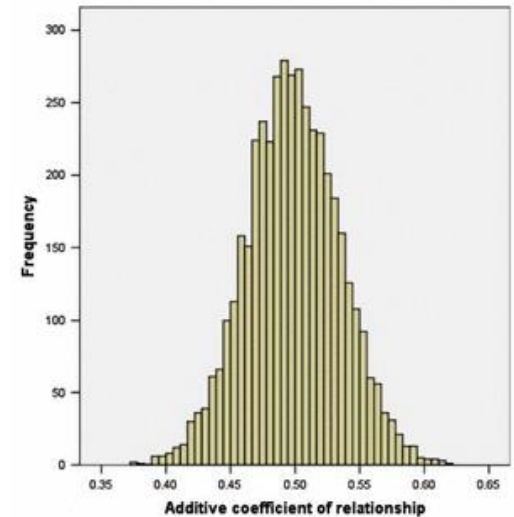


# Relationships between Individuals

- Estimated using:
  - Expected relatedness from PEDIGREE
  - ‘Observed’ relatedness from 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
  
- Still half mum and half dad
  - 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

# What information is used in BVs?

Clark et al, 2013 GSE

- $$V_a = \frac{1}{4} \text{ sire} + \frac{1}{4} \text{ dam} + \frac{1}{2} \text{ MS}$$

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.

	NZ dairy bulls					Australian dairy bulls				
	BV	Sire	Dam	MS+e	Prop. of PT	BV	Sire	Dam	MS+e	Prop. of PT
Parent Average		0.56	0.44	0.001	0.001	<b>PA EBV</b>	0.44	0.52	0.04	0.05
Genomic BV		0.43	0.26	0.31	0.56	<b>GEBV</b>	0.33	0.37	0.30	0.36
Progeny Test		0.21	0.31	0.48	1.0	<b>PT</b>	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

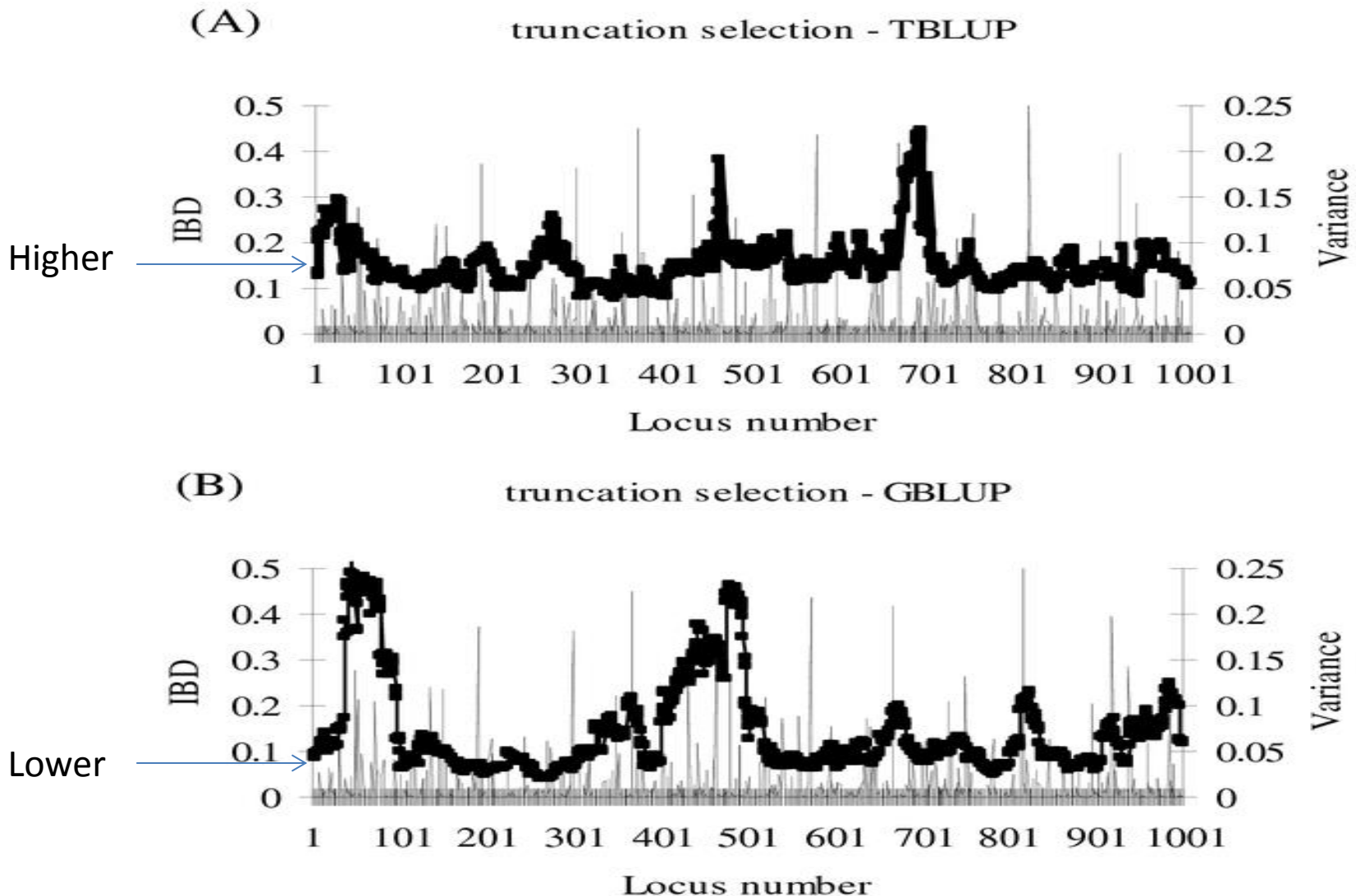
- share the same Parent average BV ( $\frac{1}{2}$  sire  $\frac{1}{2}$  dam)
- no longer the case with genomics

## 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

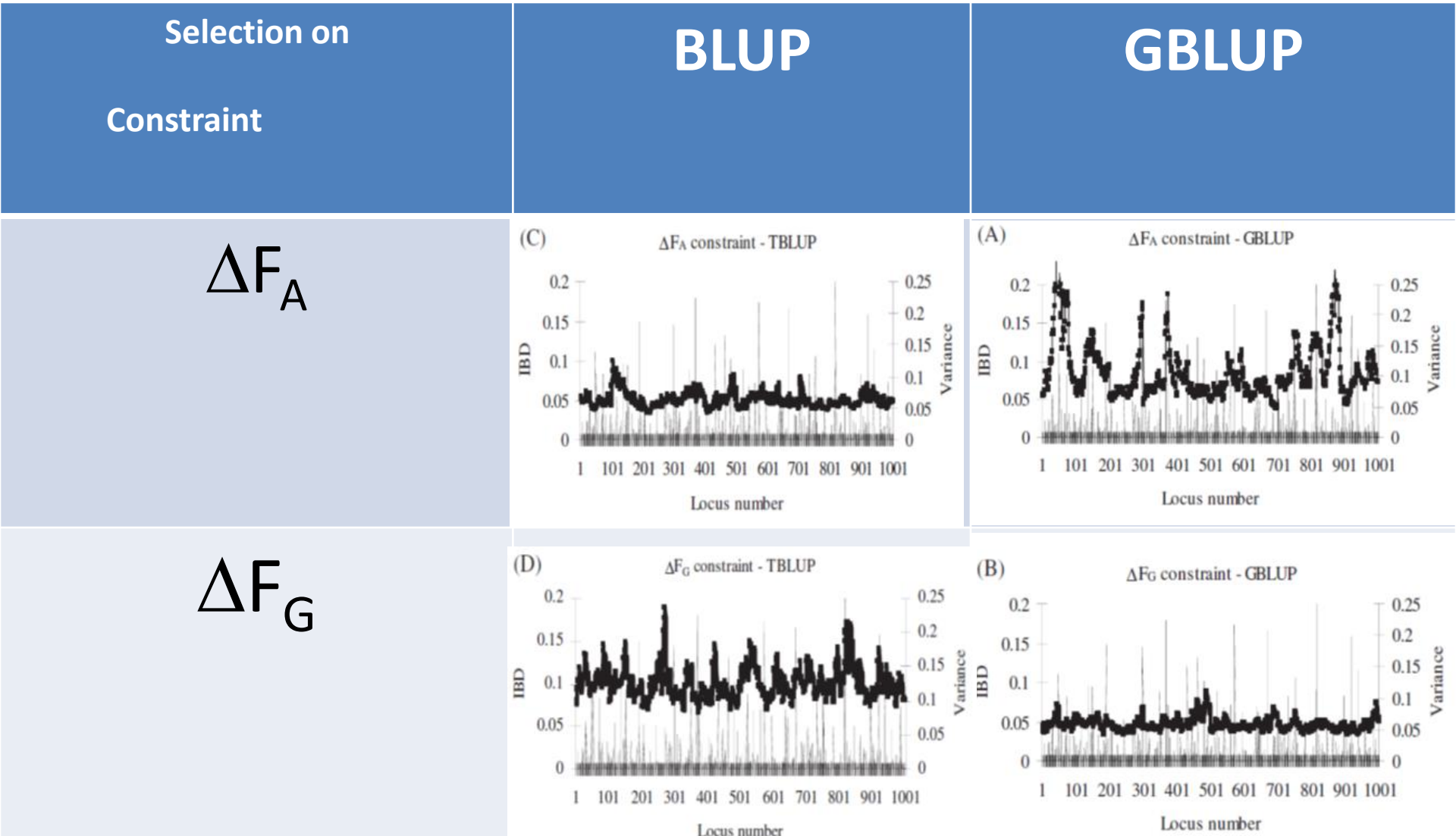
(Sonesson, Woolliams, Meuwissen, 2012)



---- Selecting 100 sires and 100 dams from 3000 cand. ---- After 10 generations

# Constraining Inbreeding:

Pedigree or Genomics, Optimal contributions



# Genomic selection and inbreeding

- Effect on IBD and variance at loci is different between GBLUP and BLUP
- GBLUP could give more 'local inbreeding'
- Can constrain inbreeding: better use G when applying GBLUP and use A when applying BLUP
- When constraining with G, can exploit some more variation within family (useful for large FS families)