

Optimizing Breeding Program Design

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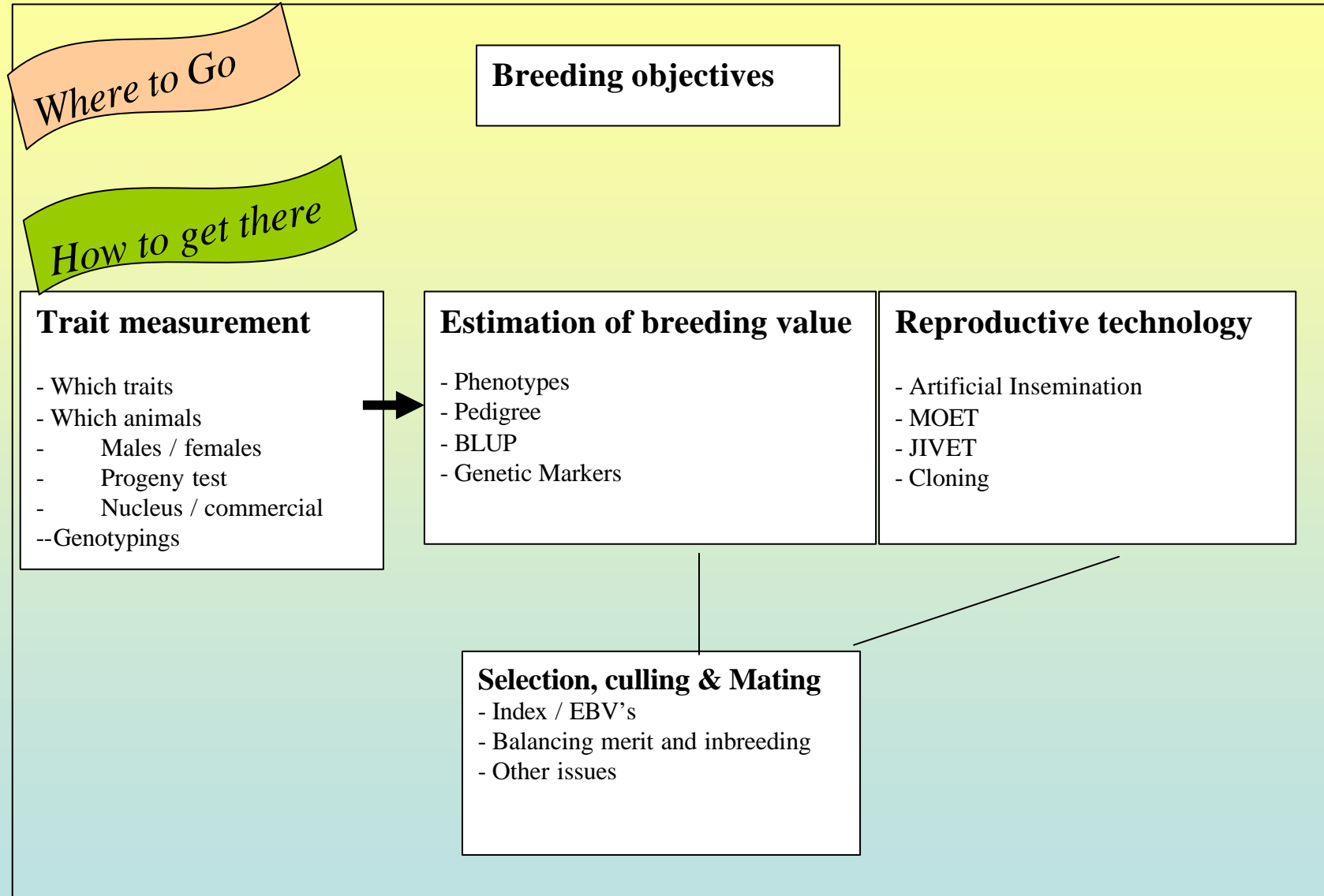
Outline / Introduction

- What is a *Good* Breeding Program
- What are design issues?
- What kind of decisions are involved?

Outline

- What is a *Good* Breeding Program
- What are design issues?
- What kind of decisions are involved? measurement / selection
- **New Technologies**
 - Genetic Markers
 - Reproductive technologies
- Their *joint* effect on breeding programs

Animal Breeding in a Nutshell



The framework of genetic improvement

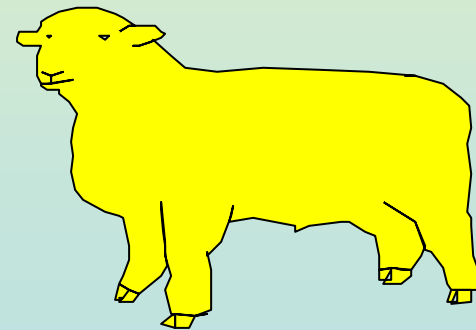
$$\text{Response} = \frac{i_m r_m + i_f r_f}{L_m + L_f} S_G$$

Genetic Variation in profit
+ restrict inbreeding

Selection intensity

Selection accuracy

Generation Interval



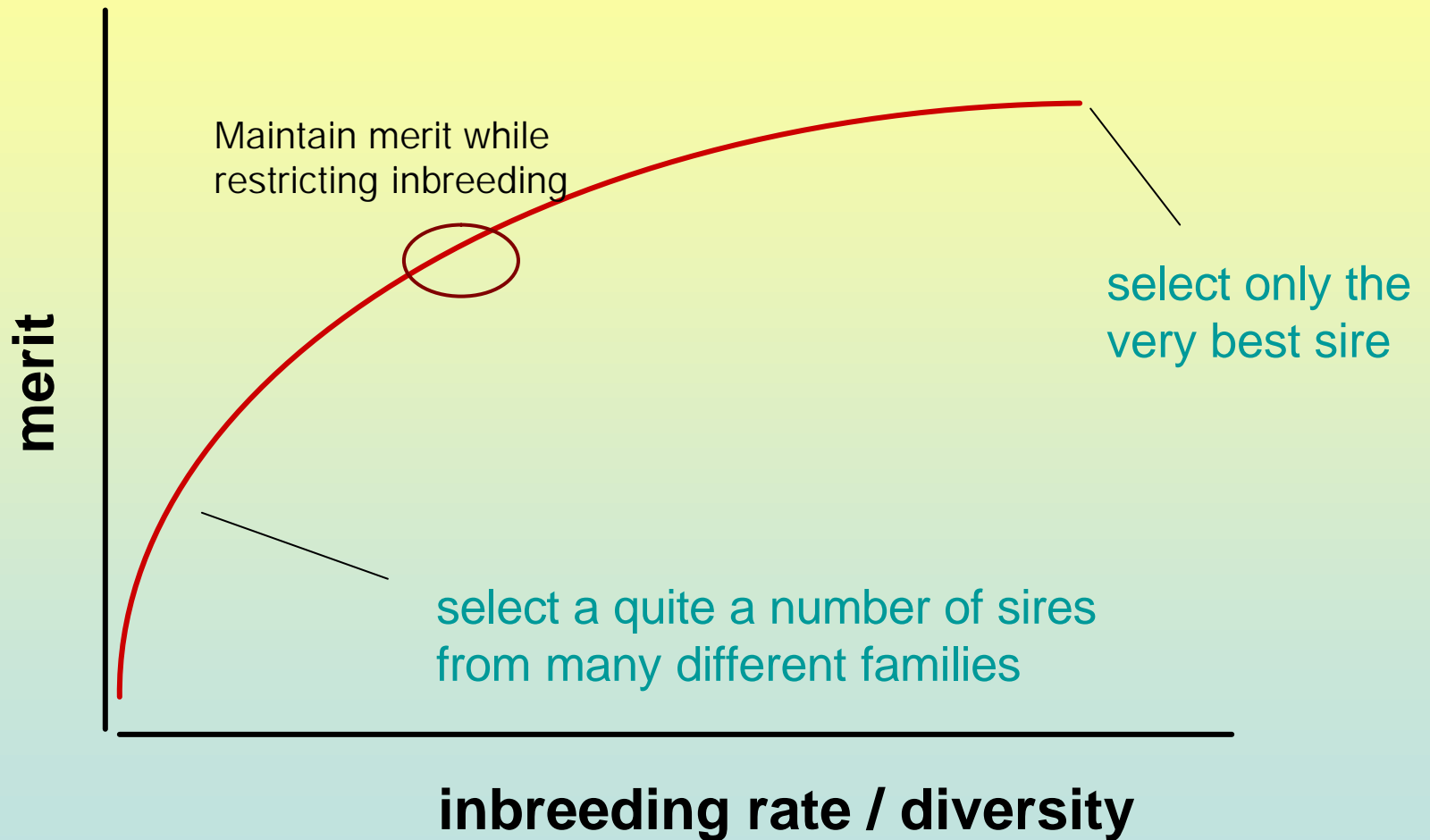
The *balancing act* of genetic improvement

Selection intensity vs risk/ inbreeding / diversity

Selection accuracy vs generation interval

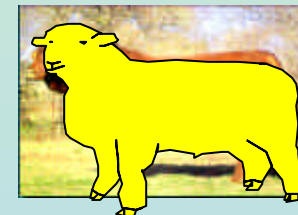
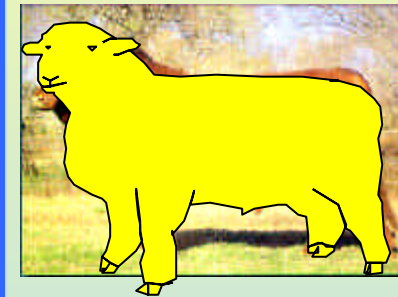
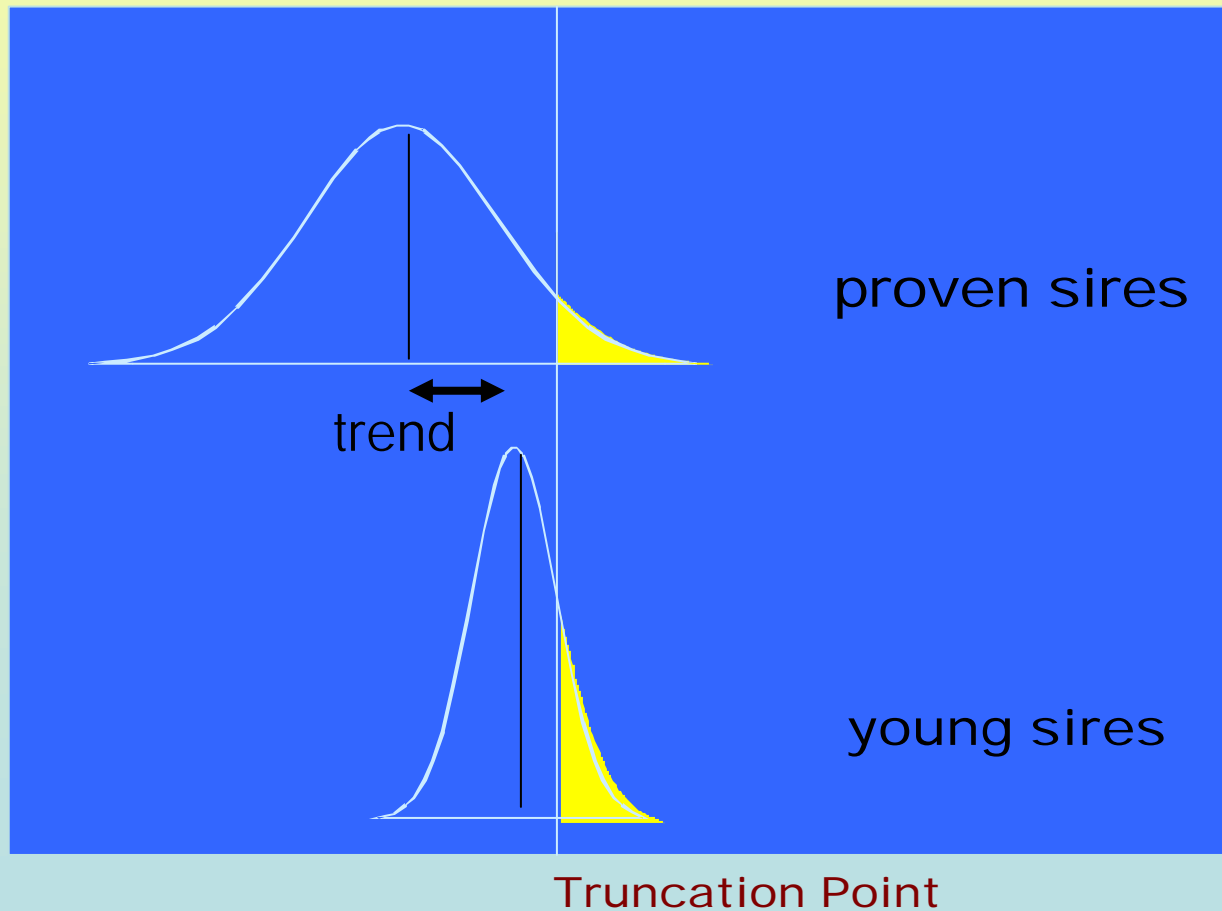
One trait versus another

Balancing inbreeding and merit



Optimizing Generation Interval

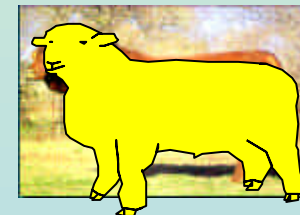
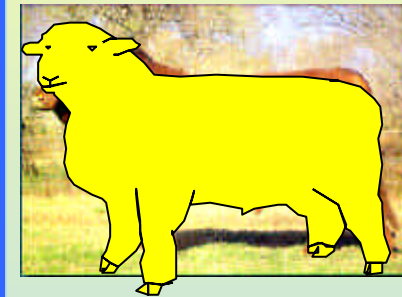
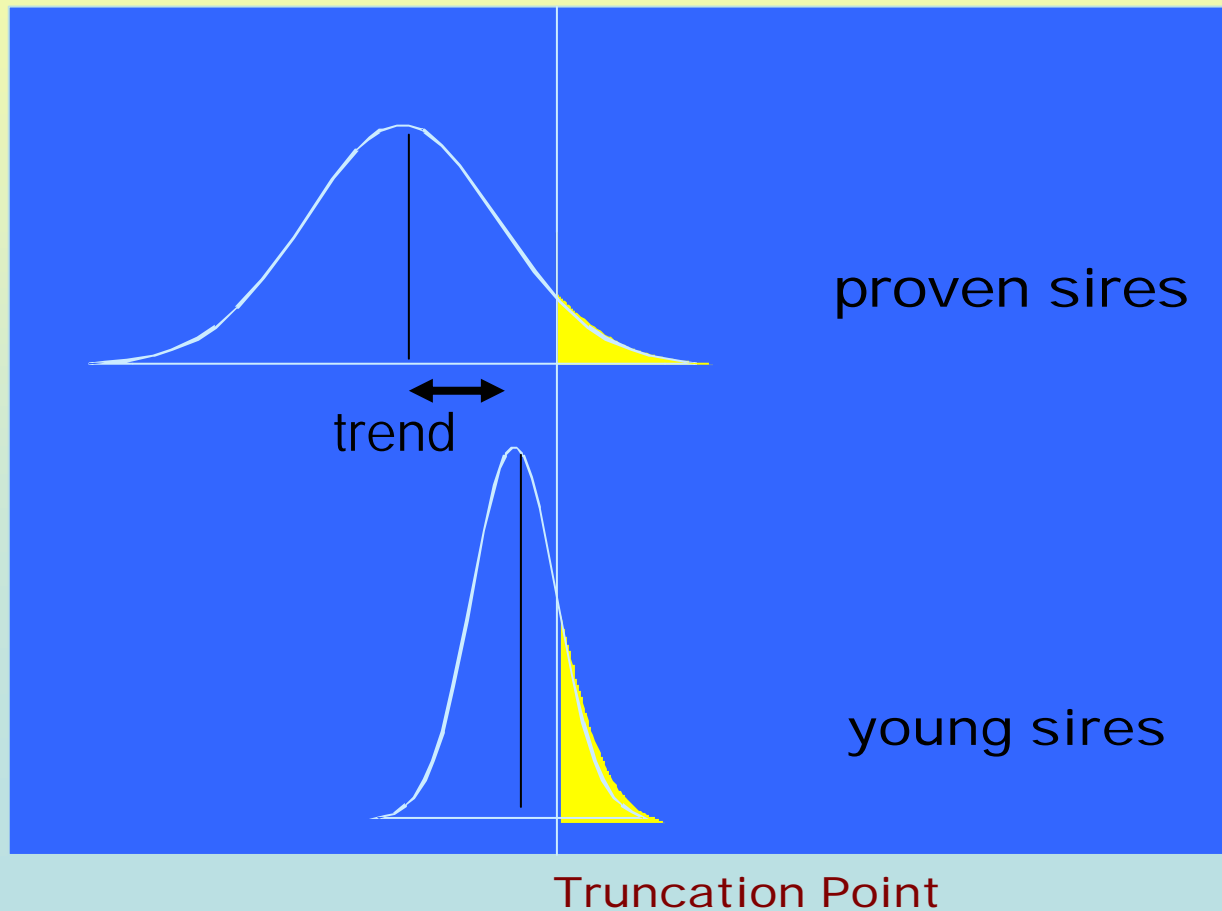
- Dilemma between young and old sires



Genetic Evaluation helps

BLUP EBV Optimizes generation interval

- Dilemma between young and old sires



Breeding Objectives

$$\text{Index} = a_1 \text{EBV}_{\text{MFD}} + a_2 \text{EBV}_{\text{FW}} + \dots + a_m \text{EBV}_{\text{NLW}}$$

Economic values



Some useful terms

- Tactical vs strategic optimization
 - (e.g. BLUP vs Selection Index)

Modeling:

- | | | |
|------------------------------------|----|---------------------|
| • Stochastic | vs | Deterministic |
| – Easier | | More complex |
| – More accurate | | Need to approximate |
| – More CPU | | Less CPU |
| – Not so suitable for optimization | | Can optimize |
| – Variation in outcome | | One outcome |

(Stochastic) Modeling of BP

- Draw a base population $A_i = [0 ; V_A]$

- Give the phenotypes $P_i = A_i + E_i$

- Select $= f(\text{BLUP-EBV}, \text{EBV} + \text{QTL}, \text{Inbreeding})$

- Mate (random/ assortative) N generations

- Define offspring $A_i = .5A_s + .5A_d + MS$

$$\frac{1}{2} (1 - \frac{1}{2} (F_s + F_d)) V_A$$

Deterministic Models

- Predict within and between family variance
- Predict selection intensity
- Predict inbreeding
- Predict selection accuracy based on BLUP
“pseudo BLUP”

Components of dG

- Genetic Variance
- Selection Intensity
- Selection Accuracy
- Generation Interval

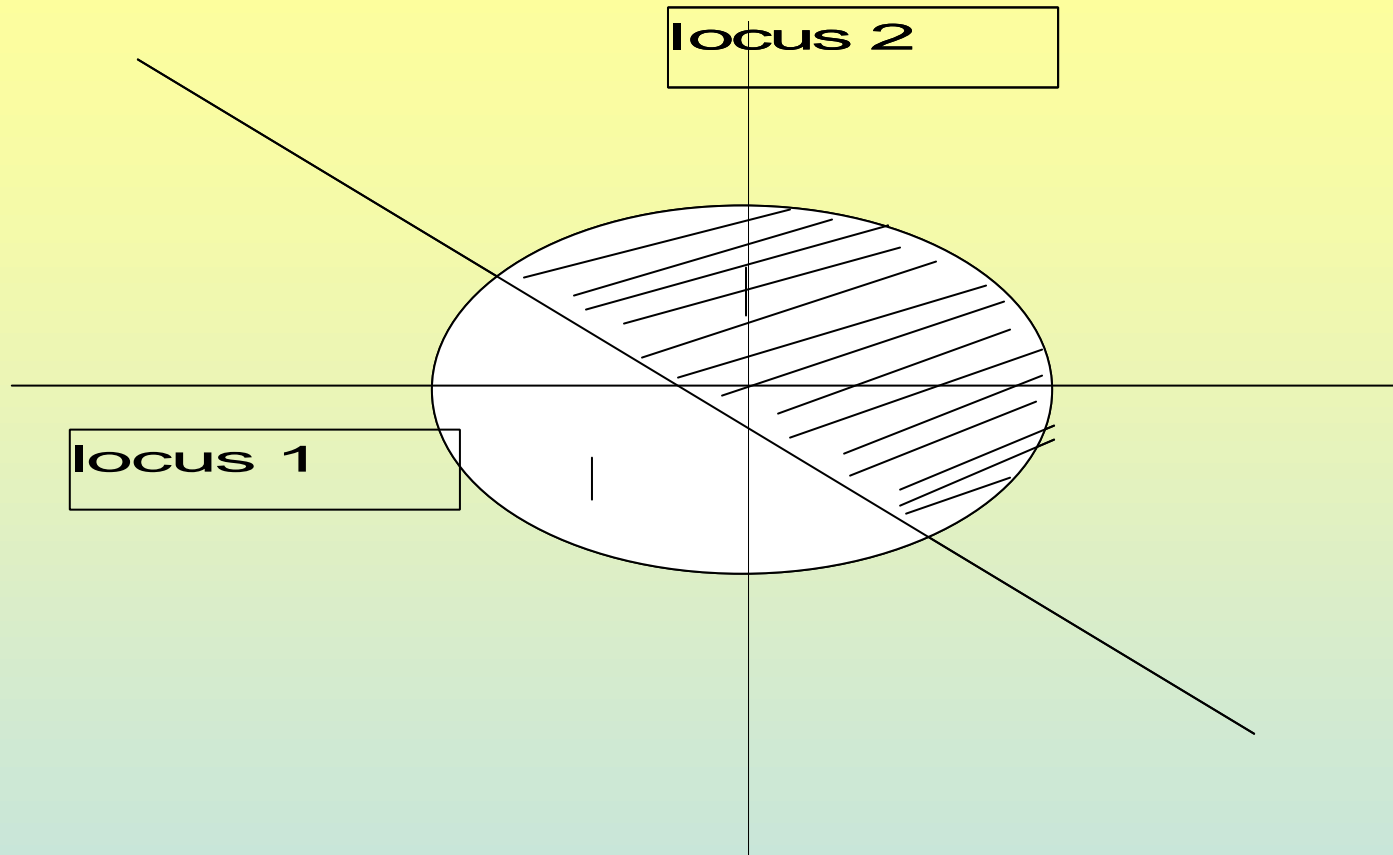
Genetic Variance:

Loss of variance due to selection

- Variance among selected individuals is lower
- Selected parents have a reduced variance
 - $V_{P_S} = (1-k) V_p$ $k = i (i-x)$
- “Bulmer Effect”: Also their genetic variance is lower

V_A in selected group is reduced to $(1 - r^2 k)$.

r = selection accuracy



Note:

Negative Covariance after selection

Effect of selection on V_A

$$\mathbf{s}_{A_{t_sel}}^2 = (1 - r^2 k) \mathbf{s}_{A_t}^2$$

Effect of selection in next generation

full residual variancebut genetic variance is still reduced

Only part coming from parents is reduced

New variance is generated due to *Mendelian Sampling*

This is NOT affected by selection “Bulmer Effect”:

$$\sigma_{A_{t+1}}^2 = \frac{1}{4} \sigma_{As_t}^2 + \frac{1}{4} \sigma_{Ad_t}^2 + \frac{1}{2} \sigma_{A_0}^2$$

Variance over time

$$\sigma_{A_{t+1}}^2 = \frac{1}{4} \sigma_{As_t}^2 + \frac{1}{4} \sigma_{Ad_t}^2 + \frac{1}{2} \sigma_{A_0}^2$$

(P-males=10%, P-females = 50%)

Gen	VA	V(sire)	V(dam)	Herit	Mean	R
1	1.00	0.58	0.68	0.50	0.00	0.90
2	0.82	0.51	0.58	0.45	0.90	0.77
3	0.77	0.49	0.56	0.44	1.68	0.74
4	0.76	0.49	0.55	0.43	2.42	0.73
5	0.76	0.49	0.55	0.43	3.15	0.73
6	0.76	0.49	0.55	0.43	3.88	0.73
7	0.76	0.49	0.55	0.43	4.61	0.73
8	0.76	0.49	0.55	0.43	5.34	0.73
9	0.76	0.49	0.55	0.43	6.08	0.73
10	0.76	0.49	0.55	0.43	6.81	0.73

Note

Only need to worry about this in deterministic simulation

Selection intensity (i)

- i is lower in finite populations
 - Order statistics effect
 - Not a large effect
- i is lower when EBVs are correlated
 - More important, esp. when correlation are high
 - And note that they can be high!!!

Order statistics effects

- mainly in very small groups/ populations -

Population size	Selected fraction		
	P=2%	P = 10%	P= 50%
2	-	-	0.564
4	-	-	0.663
10	-	1.539	0.739
20	-	1.638	0.767
50	2.249	1.705	0.785
100	2.328	1.73	0.791
400	2.396	1.75	0.796
infinite	2.421	1.755	0.798

For a good approximation,
use Burrow's formula

$$i_{\infty} = \frac{1 - p}{2 * i_{\infty} * p * (N + 1)}$$

Effect of correlated EBVs - family structure -

**How much do correlations reduce the
selection intensity**

$$i_r(t) = \sqrt{(1-t)} \cdot i(0)$$

Rawlings (1976)

For f families each with n_f individuals

$$i_r(t) = \sqrt{1 - \frac{t(n_f - 1)}{n_f f - 1}} i(0)$$

or

$$i_r(t) = \sqrt{(1 - t_{av})} i(0)$$

**t_{av} is the average correlation
between EBV's in the population**

Overestimates si when t_{av} is high!

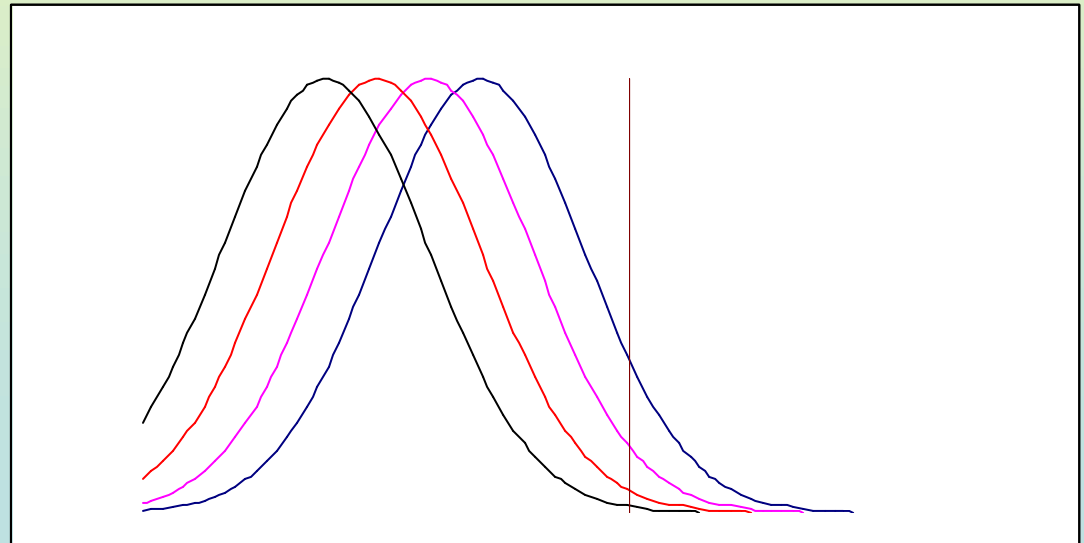
Meuwissen 1991

Selection Accuracy

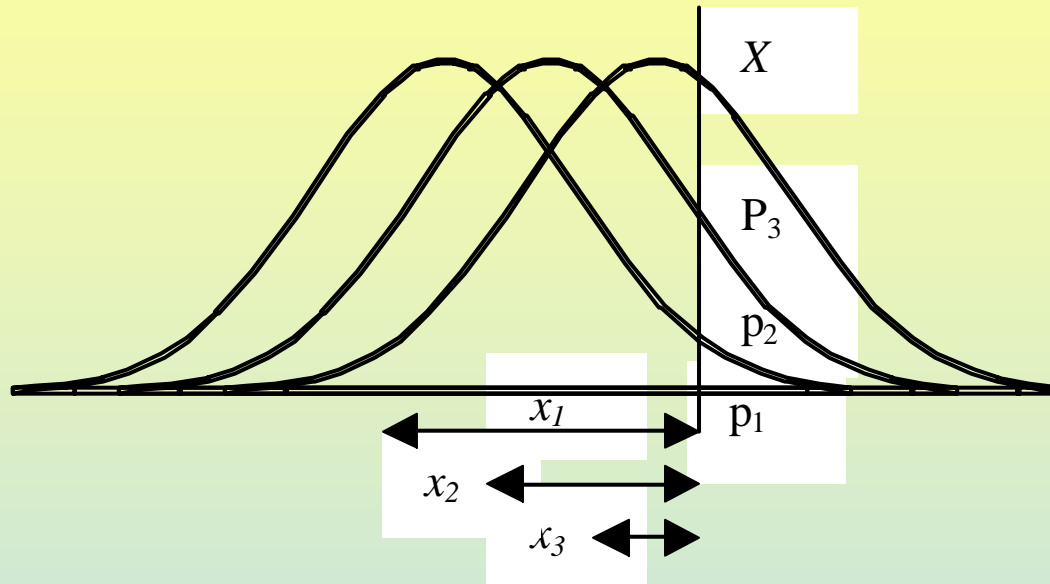
- Pseudo BLUP
 - Selection Index
 - Iterated to accommodate EBV parents
- Note that different age classes have different r

Overlapping Generations

- Selection across age classes
 - Different means (how much?)
 - Different SD?



An algorithm for finding common truncation point



ageclass	N in group	mean	SD	Proportion Selected	Nr Selected	Mean of selected	
1	50	10	1	0.28	14.17	11.18	
2	35	9.5	1	0.14	4.96	11.03	
3	15	9	1	0.06	0.87	10.74	
					20.0	11.12	mean of selected

Summary

- Need a deterministic model that predicts and optimizes genetic gain
 - Selection intensity
 - Selection accuracy
 - Generation intervals
 - Genetic Variation after selection
 - Inbreeding