

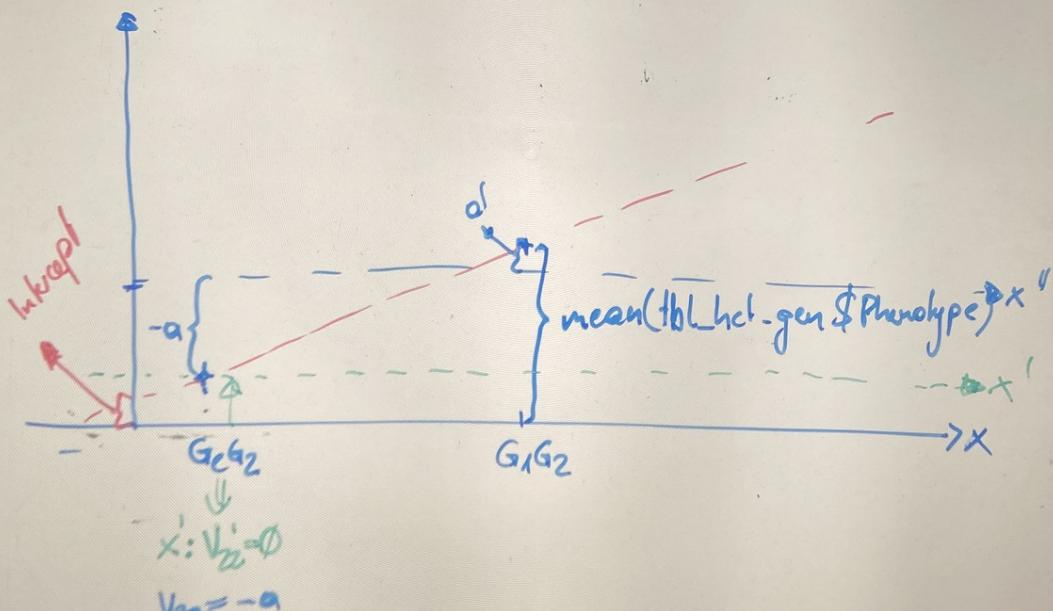
OHP Picture 1

Error in Notebook: "qq-estimate-bv.nb.html"

Problem in computation of genotypic value d:

Before: $n\text{-geno-}d \leftarrow \text{mean(tbl_het.gen\$Phenotype)}$
- "Intercept"

Correct: $n\text{-geno-}d \leftarrow \text{mean}(\dots) - \text{"Intercept"} - n\text{-geno-}a$



OHP Picture 2

□ One locus : G , 2 alleles G_1, G_2 with

Genotypes	V_{ij}	BV_{ij}	$f(G_1) = p$
G_1G_1	a	$2q\alpha$	$f(G_2) = q = 1-p$
G_1G_2	d	$(q-p)\alpha$	
G_2G_2	$-a$	$-2p\alpha$	

$$\text{with } \alpha = a + (q-p)d$$

□ Relationship between BV :

$$BV_{11} - BV_{12} = BV_{12} - BV_{22} = \alpha$$

□ Difference between genotypic values (V_{ij}) and the breeding values (BV_{ij})

$$\text{For } G_1G_1 : V_M - BV_M = a - 2q\alpha$$

$$\begin{aligned}
 &= a - [2q(a + (q-p)d)] \\
 &= a - 2qa - 2q^2d + 2pqd \\
 &= \cancel{q(1-2q)} + \cancel{(2pq - 2q^2)d} = \mu - 2q^2d
 \end{aligned}$$

OHP Picture 3

Summary:			
	V_{ij}	BV_{ij}	D_{ij}
G_1G_1	a	$2qx$	$-2q^2d$
G_1G_2	d	$(q-p)a$	$2pqd$
G_2G_2	$-a$	$-2px$	$-2p^2d$

Dominance deviation

In general for genotype G_iG_j , the difference between genotypic value and breeding value can be expressed as:

$$V_{ij} - BV_{ij} = \mu + D_{ij}$$

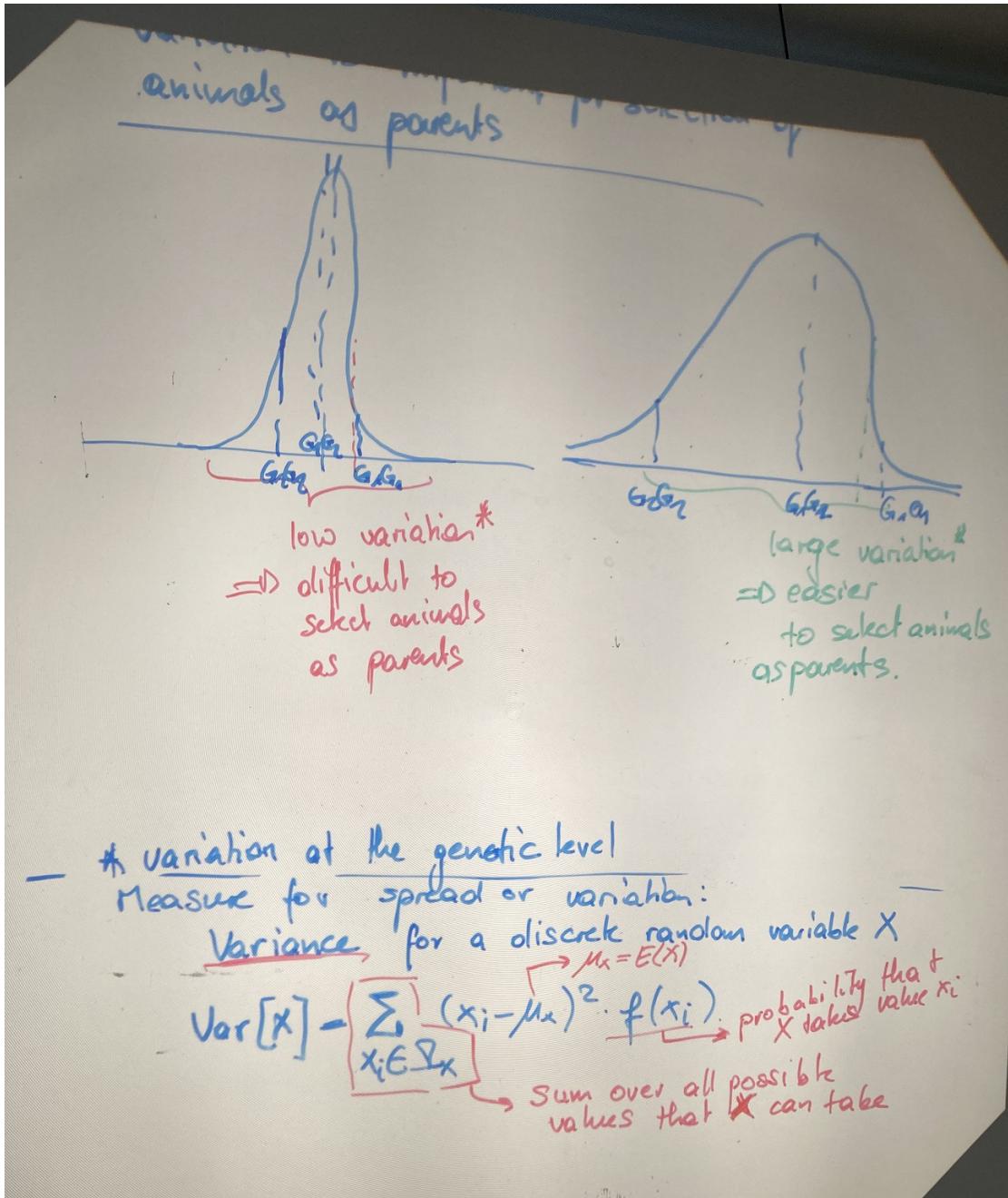
\Rightarrow Solve for genotypic value V_{ij}

$$\Rightarrow V_{ij} = \mu + BV_{ij} + D_{ij}$$

population mean
 additive part, stands for the value due to the number (" G_i ") of positive alleles in a given genotype
 ! selection of parents

Deviation caused by potential dominance effects at a given locus.

OHP Picture 4



OHP Picture 5

Variance with V_{ij} as random variable:

$$\begin{aligned} \bar{v}_G^2 = \text{Var}[V] &= (V_{11} - \mu)^2 \cdot f(G_1 G_1) \\ &\quad + (V_{12} - \mu)^2 \cdot f(G_1 G_2) \\ &\quad + (V_{22} - \mu)^2 \cdot f(G_2 G_2) \\ &\quad \downarrow \text{population mean} \\ &= (\alpha - [(p-q)\alpha + 2pq\delta])^2 \cdot p^2 \\ &\quad + (\alpha - [(p-q)\alpha + 2pq\delta])^2 \cdot 2pq \\ &\quad + ((-\alpha) - [(p-q)\alpha + 2pq\delta])^2 \cdot q^2 \end{aligned}$$

Use: $V_{ij} - \mu = \beta V_{ij} + D_{ij}$

$$\Rightarrow \text{Var}[V] = (\beta V_{11} + D_{11})^2 \cdot p^2 + (\beta V_{12} + D_{12})^2 \cdot 2pq + (\beta V_{22} + D_{22})^2 \cdot q^2$$

... $\bar{v}_G^2 = \text{Var}[V] = 2pq\alpha^2 + (2pq\delta)^2$

OHP Picture 6

Variance:

$$\begin{aligned} \sigma_G^2 &= 2pq\alpha^2 + (2pqd)^2 \\ &= \sigma_A^2 + \sigma_D^2 \rightarrow \text{dominance variance} \end{aligned}$$

genetic-additive
Variance

□ BV_{ij} as random variables:

$$\begin{aligned} \text{Var}[BV_{ij}] &= (BV_{11} - E[BV])^2 \cdot p^2 \\ &\quad + (BV_{12} - E[BV])^2 \cdot 2pq \\ &\quad + (BV_{22} - E[BV])^2 \cdot q^2 \end{aligned} \quad \left. \begin{array}{l} = BV_{11}^2 \cdot p^2 + BV_{12}^2 \cdot 2pq \\ + BV_{22}^2 \cdot q^2 \end{array} \right\} - 2px$$

$$\begin{aligned} E[BV] &= 2pq\alpha \cdot p^2 + (q-p)\alpha \cdot 2pq + (\cancel{\alpha}) \cdot q^2 \\ &= 2p^2qa + qa \cdot 2pq - px \cdot 2pq - q^2px \\ &= 2pq\alpha + 2pq\alpha - 2pq\alpha - 2pq\alpha \\ &= \emptyset \end{aligned}$$

OHP Picture 7

$$\begin{aligned}
 \text{Var}[B\bar{r}] &= BV_{11}^2 \cdot p^2 + BV_{12}^2 \cdot 2pq + BV_{22}^2 \cdot q^2 \\
 &= (2q\alpha)^2 \frac{p^2}{p^2} + ((q-p)\alpha)^2 2pq + (-2\alpha)^2 q^2 \\
 &= 4p^2 q^2 \alpha^2 + (q^2 - 2q\alpha p + p^2 \alpha^2) \cdot 2pq \\
 &\quad + 4p^2 q^2 \alpha^2 \\
 &= 4p^2 q^2 \alpha^2 + 2pq\alpha^2 - 4p^2 q^2 \alpha^2 + 2pq\alpha^2 \\
 &\quad + 4pq\alpha^2 \\
 &= 2pq\alpha^2 + 2pq\alpha^2 + 4pq\alpha^2 \xrightarrow{(p+q)^2} \\
 &= 2pq\alpha^2 \underbrace{(q^2 + p^2 + 2pq)}_{=1} \xrightarrow{(p+q)^2} \\
 &- 2pq\alpha^2 = \overline{\sigma_A}^2 \\
 \text{Var}[D] &= \dots = \overline{\sigma_D}^2
 \end{aligned}$$

OHP Picture 8

Two and more loci

Population

Interaction between Loci G and H

□ Two loci G and H with two alleles each

Genotype	Genotypic Values	BV	D (I_{GH})
G_1G_1	H_1H_1		
G_1G_1	H_1H_2		
G_1G_2	H_2H_2		
G_1G_2	H_1H_2		

V_{ijkl}, corresponding to genotypic value of genotype $G_iG_jH_kH_l$, is not the sum of the genotypic values for the single loci V_G and V_H

OHP Picture 9

$$V = V_G + V_H + I_{GH}$$

$$V_G = \mu_G + BV_G + D_G$$

$$V_H = \mu_H + BV_H + D_H$$

$$V = \mu_G + BV_G + D_G + \mu_H + BV_H + D_H + I_{GH}$$

□ Re-group:

$$V = \underbrace{\mu_G + \mu_H}_{\mu} + \underbrace{BV_G + BV_H}_{U} + \underbrace{D_G + D_H}_{D} + I_{GH}$$

□ More than 2 loci : A, B, C, ...

□ Decompose $V = V_A + V_B + V_C + \dots + I_{ABC\dots}$

$$= \mu_A + BV_A + D_A + \mu_B + BV_B + D_B + \mu_C + BV_C + D_C + \dots + I_{ABC\dots}$$

$$= \mu_A + \mu_B + \mu_C + \dots + BV_A + BV_B + BV_C + \dots + D_A + D_B + D_C + \dots + I_{ABC\dots}$$

$$= \mu + U + D + I_{ABC\dots}$$

OHP Picture 10

Genetic Model:

◻ $P = G + E$ for phenotypic observation
 ↓
 genotypic value V

◻ With observation y (instead of P)

$$y = V + E$$

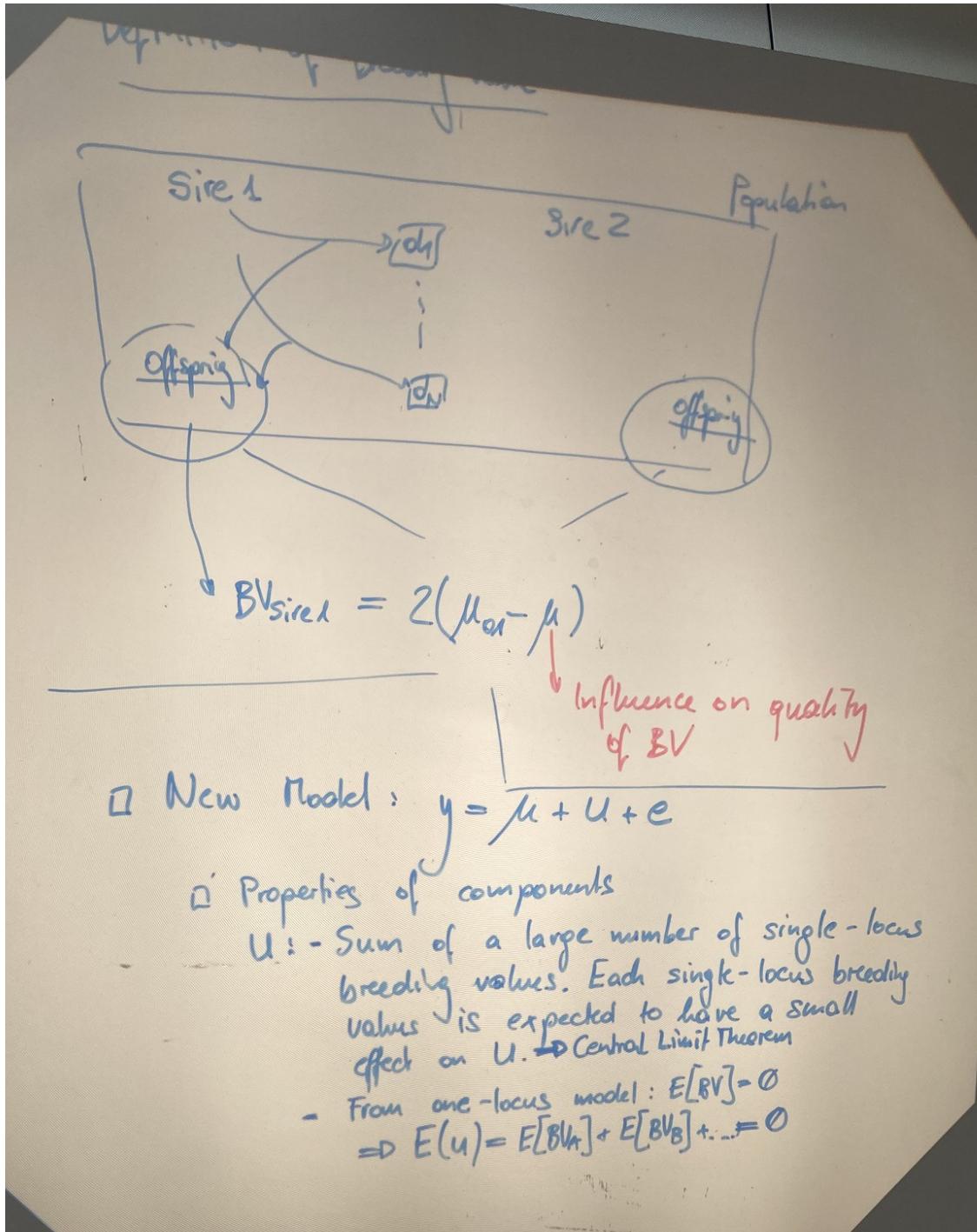
◻ Insert decomposition of V :

$$y = \underbrace{\mu + u + d + I_{ABC-}}_{=V} + E$$

◻ Perspective of Livestock Breeding, breeding values (u) are of primary interest, hence

$$y = \mu + u + \underbrace{d + I_{ABC-}}_{=E^*} + E = \mu + u + E^*$$

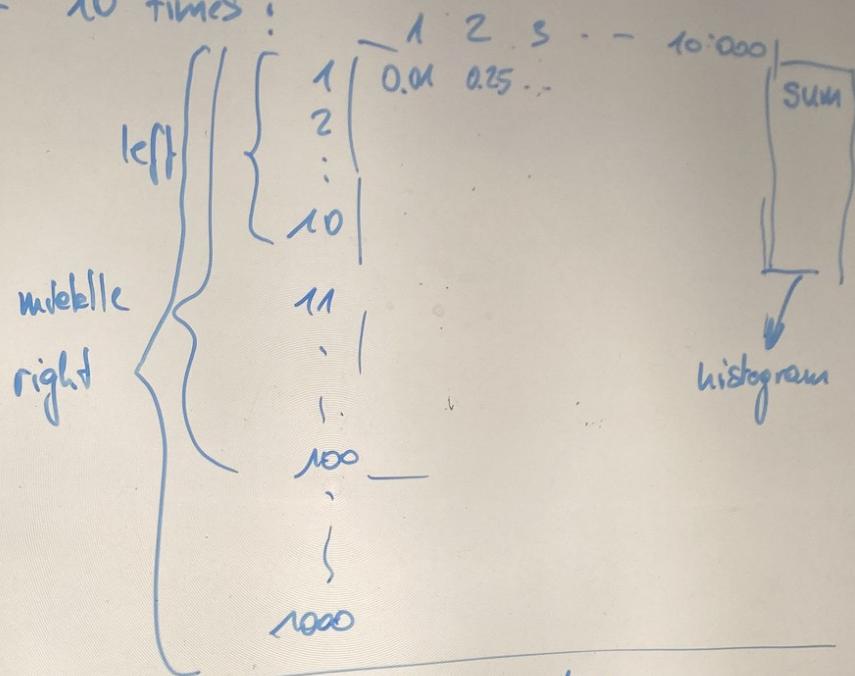
OHP Picture 11



OHP Picture 12

Central Limit Theorem (CLT)

- Take 10'000 random samples
- 10 times:



- CLT is useful for u and e
- From distribution theory, any sum of normal distributions, is normal $\Rightarrow y \sim \text{Normal}$

OHP Picture 13

New Model:

- no longer compute breeding values as with the single locus model, because we do not know the locations in the genome that are relevant for a given trait of interest.
- all that can be done is to predict breeding values based on phenotypic information, but using the developed genetic model.
- Principle of prediction:
predicted breeding value (\hat{u})
$$\hat{u} = b * (y - \bar{u})$$

Step 2
Observation

Step 1
appropriate mean performance

OHP Picture 14

- With every prediction, there is an error associated, error is quantified by accuracy
- Accuracy: Correlation between information source (y) and true breeding value (u)
$$r(u,y) = \frac{\text{Cov}(u,y)}{\sqrt{\text{Var}(u) \cdot \text{Var}(y)}}$$
$$= \frac{\text{Var}(u)}{\sqrt{\text{Var}(u) \cdot \text{Var}(y)}} = \sqrt{\frac{\text{Var}(u)}{\text{Var}(y)}} \text{ corr}$$