

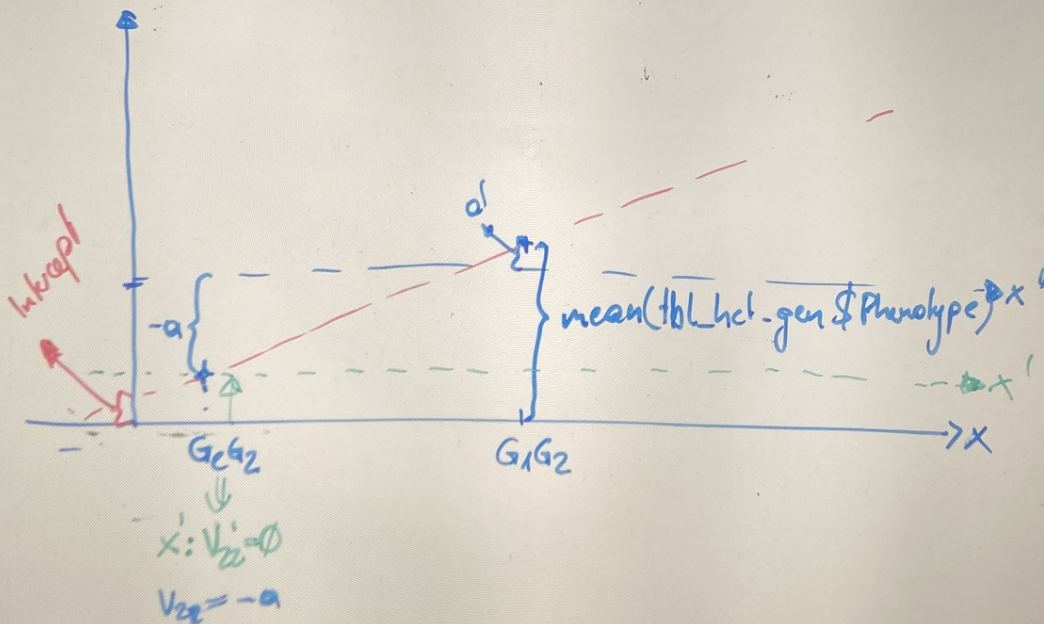
OHP Picture 1

Error in Notebook: "gg-estimate_by_nb.html"

□ Problem in computation of genotypic value d :

Before: $n_geno_d \leftarrow \text{mean}(\text{tbl_het_gen} \$ \text{Phenotype}) - \text{"Intercept"}$

Correct: $n_geno_d \leftarrow \text{mean}(\dots) - \text{"Intercept"} - n_geno_a$



□ One locus : G , 2 alleles G_1, G_2 with
 $f(G_1) = p$
 $f(G_2) = q = 1 - p$

Genotypes	U_{ij}	BV_{ij}
G_1G_1	a	$2qa$
G_1G_2	d	$(q-p)a$
G_2G_2	$-a$	$-2pa$

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

□ Relationship between BV:
 $BV_{11} - BV_{12} = BV_{12} - BV_{22} = \alpha$

□ Difference between genotypic values (U_{ij}) and the breeding values (BV_{ij})
 For G_1G_1 : $U_{11} - BV_{11} = a - 2qa$
 $= a - [2q(a + (q-p)d)]$
 $= a - 2qa - 2q^2d + 2pqa$
 $= \underbrace{q(1-2q)}_{\mu} + (2pq - 2q^2)d = \mu - 2q^2d$

Summary:

	V_{ij}	BV_{ij}	D_{ij} → Dominance deviation
G_1G_1	a	$2qa$	$-2q^2d$
G_1G_2	d	$(q-p)a$	$2pqd$
G_2G_2	$-a$	$-2pa$	$-2p^2d$

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

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

⇒ Solve for genotypic value V_{ij}

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

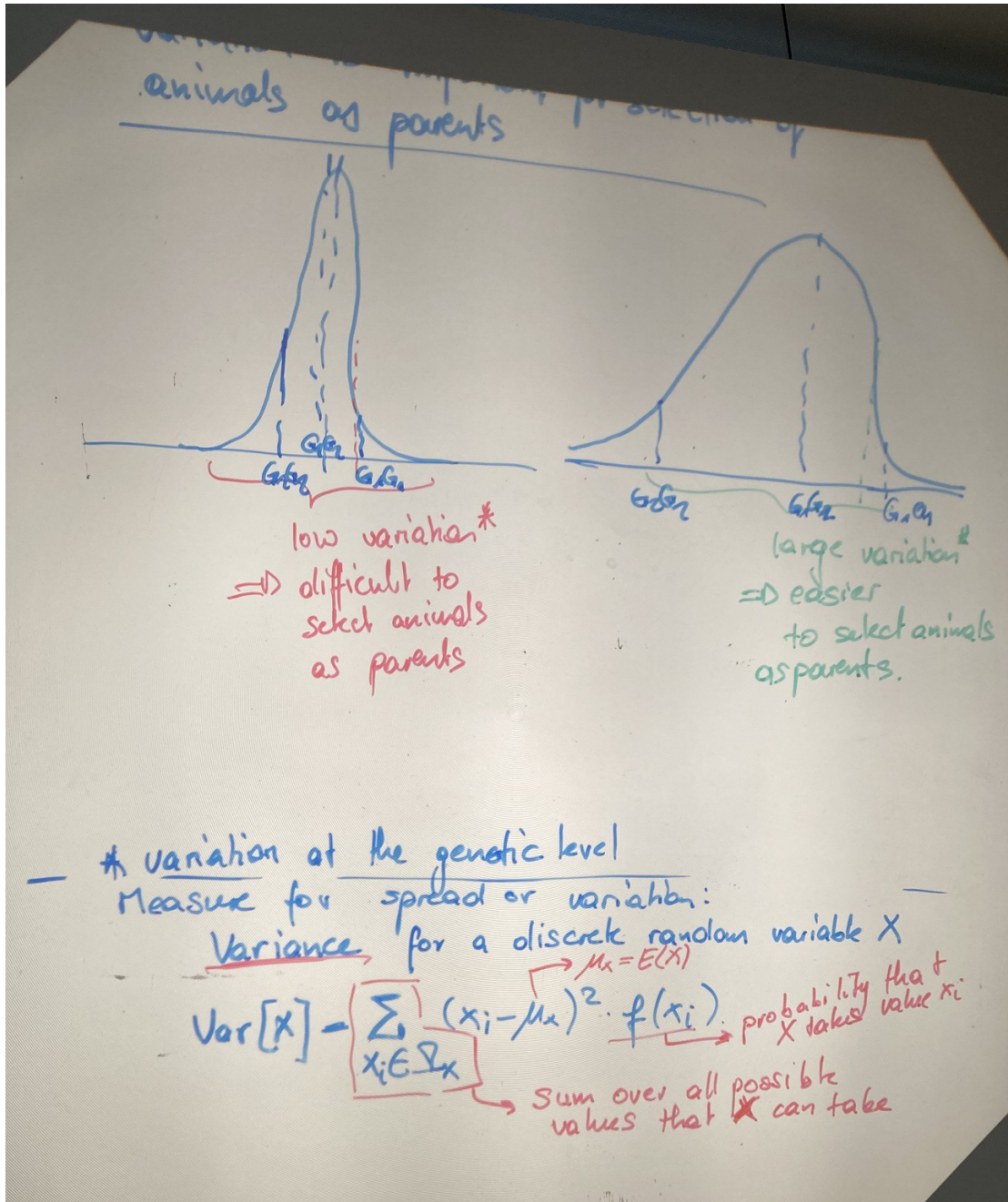
population mean

additive part, stands for the value due to the number of positive alleles (G_i) in a given genotype

! selection of parents

⊗ Deviation caused by potential dominance effects at a given locus.

OHP Picture 4



Variance with V_{ij} as random variable:

$$\begin{aligned} \sigma_G^2 = \text{Var}[V] &= (V_{11} - \mu)^2 \cdot f(G_1G_1) \\ &+ (V_{12} - \mu)^2 \cdot f(G_1G_2) \\ &+ (V_{22} - \mu)^2 \cdot f(G_2G_2) \end{aligned}$$

↓
population mean

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

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

$$\begin{aligned} \Rightarrow \text{Var}[V] &= (BV_{11} + D_{11})^2 \cdot p^2 + (BV_{12} + D_{12})^2 \cdot 2pq \\ &+ (BV_{22} + D_{22})^2 \cdot q^2 \end{aligned}$$

$$\dots \sigma_G^2 = \text{Var}[V] = 2pq\alpha^2 + (2pqd)^2$$

Variance:

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

□ BV_{ij} as random variables:

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

$$\begin{aligned} E[BV] &= 2p\alpha \cdot p^2 + (q-p)\alpha \cdot 2pq + \alpha \cdot q^2 \\ &= 2p^2q\alpha + q\alpha \cdot 2pq - p\alpha \cdot 2pq - q^2p\alpha \\ &= 2p^2q\alpha + 2pq^2\alpha - 2p^2q\alpha - 2pq^2\alpha \\ &= 0 \end{aligned}$$

OHP Picture 7

$$\begin{aligned}
 \text{Var}[BV] &= BV_{11}^2 \cdot p^2 + BV_{12}^2 \cdot 2pq + BV_{22}^2 \cdot q^2 \\
 &= (2q\alpha)^2 p^2 + ((q-p)\alpha)^2 2pq + (-2p\alpha)^2 q^2 \\
 &= 4p^2 q^2 \alpha^2 + (q^2 \alpha^2 - 2q\alpha p\alpha + p^2 \alpha^2) \cdot 2pq \\
 &\quad + 4p^2 q^2 \alpha^2 \\
 &= 4p^2 q^2 \alpha^2 + 2p^3 q \alpha^2 - 4p^2 q^2 \alpha^2 + 2p^3 q \alpha^2 \\
 &\quad + 4p^2 q^2 \alpha^2 \\
 &= 2p^3 q \alpha^2 + 2p^3 q \alpha^2 + 4p^2 q^2 \alpha^2 \\
 &= 2p^3 q \alpha^2 (q + p^2 + 2pq) \quad \left(\underbrace{q + p^2 + 2pq}_{(p+q)^2} \right) \\
 &= 2p^3 q \alpha^2 \cdot 1 \\
 &= 2p^3 q \alpha^2 = \sigma_A^2 \\
 \text{Var}[D] &= \dots = \sigma_D^2
 \end{aligned}$$

OHP Picture 8

Two and more loci

Interaction between Loci G and H

Population

□ Two loci G and H with two alleles each

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

I_{GH} interaction

V_{ijkl} , corresponding to genotypic value of genotype $G_iG_j H_kH_l$, 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:

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

$$= \mu + U + D + I_{GH}$$

□ More than 2 loci : A, B, C, ...

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

$$= \mu_A + BV_A + D_A + \mu_B + BV_B + D_B$$

$$+ \mu_C + BV_C + D_C + \dots + I_{ABC\dots}$$

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

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

Genetic Model:

- $P = G + E$ for phenotypic observation P
↓
genotypic value V
- With observation y (instead of P)

$$y = V + E$$
- Insert decomposition of V :

$$y = \underbrace{\mu + U + D + \sum_{ABC} \dots}_{=V} + E$$
- Perspective of Livestock Breeding, breeding values (U) are of primary interest, hence

$$y = \mu + U + \underbrace{D + \sum_{ABC} \dots + E}_{E^*}$$

OHP Picture 11

Definition of Breeding Value

$BV_{sire1} = 2(\mu_{sire1} - \mu)$

Influence on quality of BV

□ New Model: $y = \mu + u + e$

□ Properties of components

u : - Sum of a large number of single-locus breeding values. Each single-locus breeding value is expected to have a small effect on u . \rightarrow Central Limit Theorem

- From one-locus model: $E[BV] = 0$
 $\Rightarrow E(u) = E[BV_A] + E[BV_B] + \dots = 0$

Central Limit Theorem (CLT)

- Take 10'000 random samples
- 10 times :

		1	2	3	...	10'000
		0.01	0.25	...		
left	} middle right	1				SUM
		2				
⋮						
10						
11						
⋮						
100						
⋮						
1000						

↓ histogram

- CLT is used for μ and σ
- From distribution theory, any sum of normal distributions, is normal $\Rightarrow y \sim \text{Normal}$

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})

$$\hat{u} = b * (y - \mu)$$

Step 2 → (y) ← Observation

→ appropriate mean performance ← (μ)

Step 1

