

# Basics of Quantitative Genetics (Part 2)

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# Variances

- ▶ selection can only be done with variation
- ▶ variation is assessed by variance

$$\text{Var} [X] = \sum_{x_i \in \mathcal{X}} (x_i - \mu_X)^2 * f(x_i)$$

where

$\mathcal{X}$ :	set of all possible $x$ -values
$f(x_i)$	probability that $x$ assumes the value of $x_i$
$\mu_X$	expected value $E[X]$ of $X$

## Variation of Genotypic Values

$$\begin{aligned}\sigma_G^2 &= \text{Var}[V] = (V_{11} - \mu)^2 * f(G_1 G_1) \\ &\quad + (V_{12} - \mu)^2 * f(G_1 G_2) \\ &\quad + (V_{22} - \mu)^2 * f(G_2 G_2) \\ &= (BV_{11} + D_{11})^2 * f(G_1 G_1) \\ &\quad + (BV_{12} + D_{12})^2 * f(G_1 G_2) \\ &\quad + (BV_{22} + D_{22})^2 * f(G_2 G_2) \\ &= 2pq\alpha^2 + (2pqd)^2 \\ &= \sigma_A^2 + \sigma_D^2\end{aligned}$$

where  $\mu = (p - q)a + 2pqd$  the population mean and  
 $V_{ij} - \mu = BV_{ij} + D_{ij}$

## Additive Genetic Variance

$$\begin{aligned}\sigma_A^2 &= \text{Var} [BV] = BV_{11}^2 * f(G_1 G_1) + BV_{12}^2 * f(G_1 G_2) + BV_{22}^2 * f(G_2 G_2) \\&= (2q\alpha)^2 * p^2 + ((q - p)\alpha)^2 * 2pq + (-2p\alpha)^2 * q^2 \\&= 4p^2 q^2 \alpha^2 + (q^2 \alpha^2 - 2pq\alpha^2 + p^2 \alpha^2) * 2pq + 4p^2 q^2 \alpha^2 \\&= 8p^2 q^2 \alpha^2 + 2pq^3 \alpha^2 - 4p^2 q^2 \alpha^2 + 2p^3 q \alpha^2 \\&= 4p^2 q^2 \alpha^2 + 2pq^3 \alpha^2 + 2p^3 q \alpha^2 \\&= 2pq\alpha^2 (2pq + q^2 + p^2) \\&= 2pq\alpha^2\end{aligned}$$

because  $E[BV] = 0$

## Dominance Variance

$$\begin{aligned}\sigma_D^2 &= D_{11}^2 * f(G_1 G_1) + D_{12}^2 * f(G_1 G_2) + D_{22}^2 * f(G_2 G_2) \\&= (-2q^2 d)^2 * p^2 + (2pqd)^2 * 2pq + (-2p^2 d)^2 * q^2 \\&= 4p^2 q^4 d^2 + 8p^3 q^3 d^2 + 4p^4 q^2 d^2 \\&= 4p^2 q^2 d^2 (q^2 + 2pq + p^2) \\&= 4p^2 q^2 d^2\end{aligned}$$

because  $E[D] = 0$

## Extension To More Loci

Population (N = 10)

$$\frac{A_1 \quad B_2}{A_1 \quad B_1}$$

$$\frac{A_1 \quad B_2}{A_1 \quad B_2}$$

$$\frac{A_1 \quad B_2}{A_1 \quad B_1}$$

$$\frac{A_1 \quad B_1}{A_2 \quad B_1}$$

$$\frac{A_2 \quad B_2}{A_2 \quad B_1}$$

$$\frac{A_1 \quad B_2}{A_2 \quad B_1}$$

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$$\frac{A_1 \quad B_2}{A_1 \quad B_1}$$

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$$\frac{A_2 \quad B_1}{A_2 \quad B_1}$$

## Epistatic Interaction

- ▶ Decomposition of overall genotypic value  $V$

$$V = V_A + V_B + I_{AB}$$

where  $I_{AB}$ : deviation from additively combining  $A$  and  $B$

- ▶ Decomposition of  $V_A$  and  $V_B$  as before and re-grouping additive and dominance terms

$$V = \mu + A + D + I_{AB}$$

where  $A$  : breeding values and  $D$ : Dominance deviations

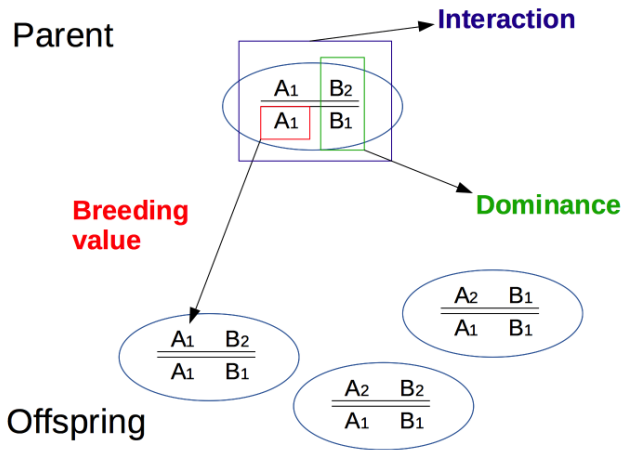
# Interaction Variance

- Interaction deviation lead to new variance component  $\sigma_I^2$

$$\begin{aligned}\sigma_G^2 &= \text{var}(V) = \text{var}(\mu + A + D + I_{AB}) \\ &= \text{var}(A) + \text{var}(D) + \text{var}(I_{AB}) \\ &= \sigma_A^2 + \sigma_D^2 + \sigma_I^2\end{aligned}$$



# Summary of Effects



# Importance for Livestock Breeding

- ▶ Breeding values measure effects of alleles
- ▶ Dominance deviations measure effects of genotypes
- ▶ Interaction deviations measure effects of different loci

→ for livestock breeding

- ▶ parents pass random sample of alleles to offspring

→ **breeding value** and associated **genetic additive variance** are important

→ dominance and interaction deviations and associated variance components rarely considered

# Genetic Models

- ▶ **polygenic model**

- ▶ large but finite number of loci affect phenotype
- ▶ used in genomic selection

- ▶ **infinitesimal model**

- ▶ infinite number of loci affect phenotype
- ▶ breeding value is the sum of infinitely small single locus breeding values
- ▶ breeding value converges to normal distribution (Central Limit Theorem)
- ▶ used in traditional breeding value estimation

# Central Limit Theorem

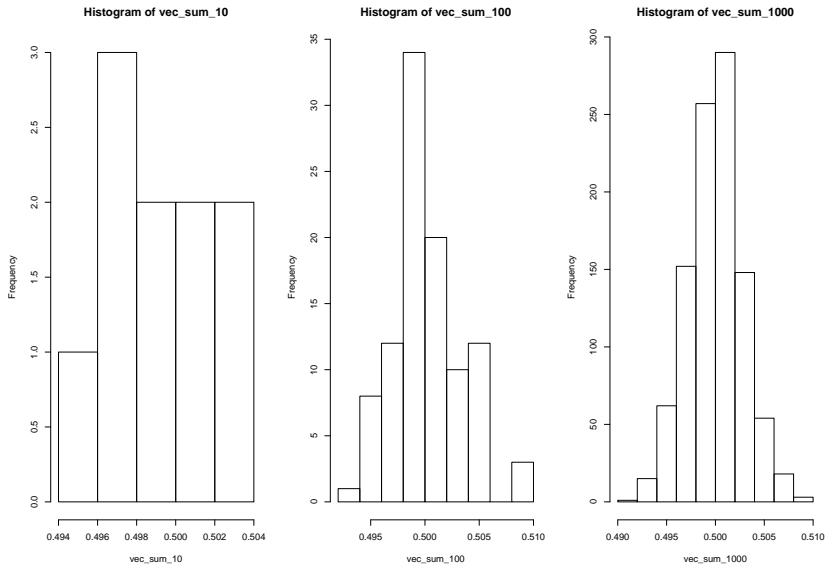


Figure 1: Distribution of Sums of Different Numbers of Components