

# Biometrical genetics

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## Outline

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1. Aim of this talk
2. Genetic concepts
3. Very basic statistical concepts
4. Biometrical model

## 1. Aim of this talk

- ▷ Revisit common genetic parameters - such as allele frequencies, genetic effects, dominance, variance components, etc
- ▷ Use these parameters to construct a biometrical genetic model



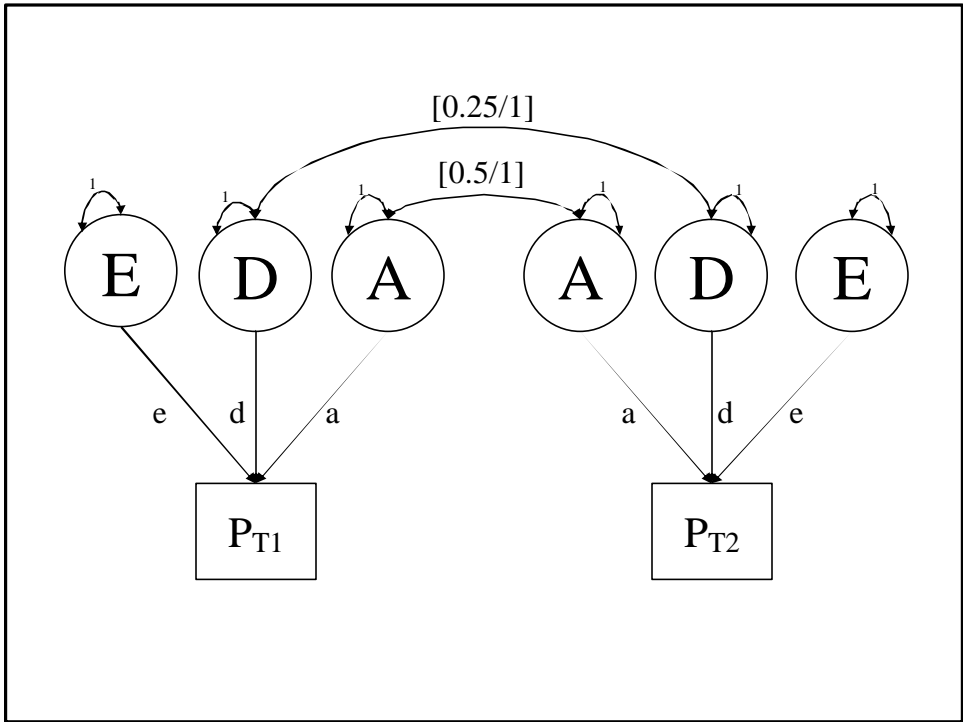
*Model that expresses the:*

(1) Mean

(2) Variance

(3) Covariance between individuals

*for a quantitative phenotype as a function of genetic parameters.*



## 2. Genetic concepts

▷ **Population level**

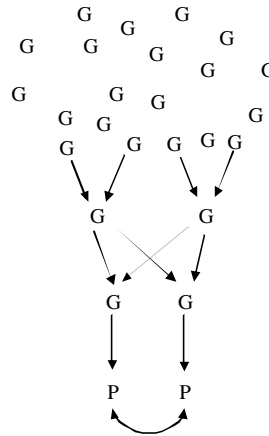
*Allele and genotype frequencies*

▷ **Transmission level**

*Mendelian segregation  
Genetic relatedness*

▷ **Phenotype level**

*Biometrical model  
Additive and dominance components*



## Population level

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### 1. Allele frequencies

▷ A single locus, with two alleles

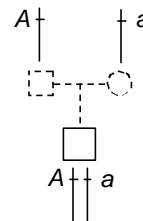
- Biallelic / diallelic
- Single nucleotide polymorphism, SNP

▷ Alleles **A** and **a**

- Frequency of **A** is  $p$
- Frequency of **a** is  $q = 1 - p$

▷ Every individual inherits two alleles

- A genotype is the combination of the two alleles
- e.g. **AA**, **aa** (the homozygotes) or **Aa** (the heterozygote)



## Population level

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### 2. Genotype frequencies (Random mating)

|          |              | Allele 1                  |                           |
|----------|--------------|---------------------------|---------------------------|
|          |              | <b>A (p)</b>              | <b>a (q)</b>              |
| Allele 2 | <b>A (p)</b> | <b>AA (p<sup>2</sup>)</b> | <b>Aa (pq)</b>            |
|          | <b>a (q)</b> | <b>aA (qp)</b>            | <b>aa (q<sup>2</sup>)</b> |

Hardy-Weinberg Equilibrium frequencies

$$P(AA) = p^2$$

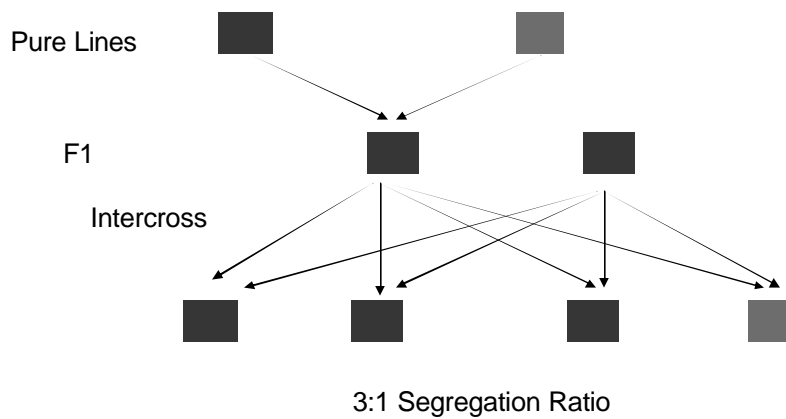
$$P(Aa) = 2pq \quad p^2 + 2pq + q^2 = 1$$

$$P(aa) = q^2$$

## Transmission level

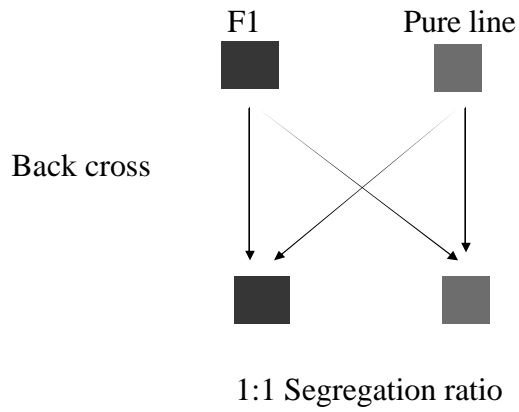
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### 1. Mendel's experiments



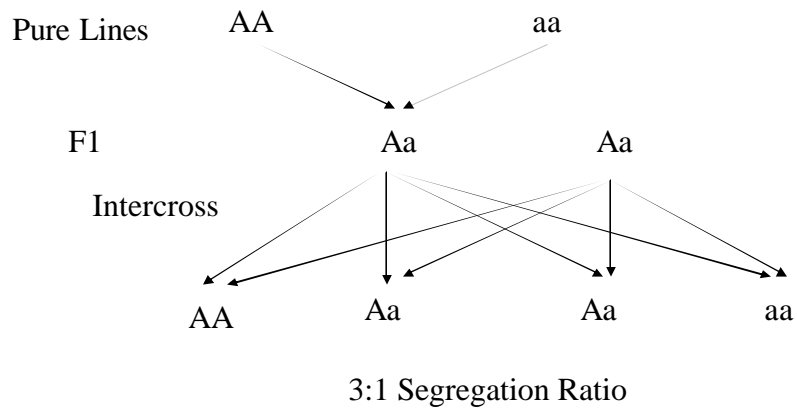
## Transmission level

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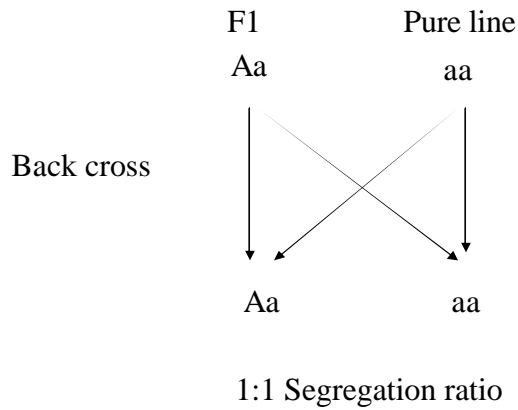
## Transmission level

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## Transmission level

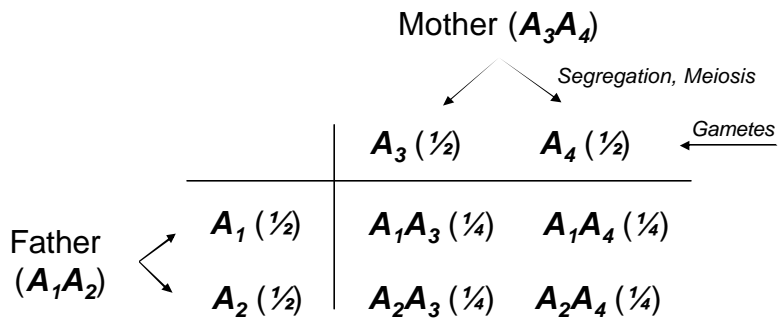
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## Transmission level

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### 1. Mendel's law of segregation



## Phenotype level

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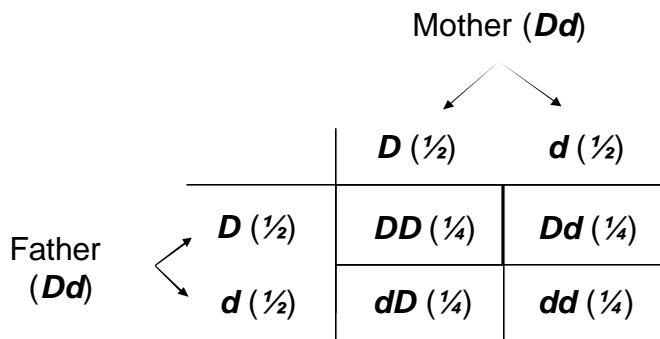
### 1. Classical Mendelian traits

- ▷ Dominant trait (**D** - presence, **R** - absence)
  - **AA, Aa** **D**
  - **aa** **R**
  
- ▷ Recessive trait (**D** - absence, **R** - presence)
  - **AA, Aa** **D**
  - **aa** **R**
  
- ▷ Codominant trait (**X, Y, Z**)
  - **AA** **X**
  - **Aa** **Y**
  - **aa** **Z**

## Phenotype level

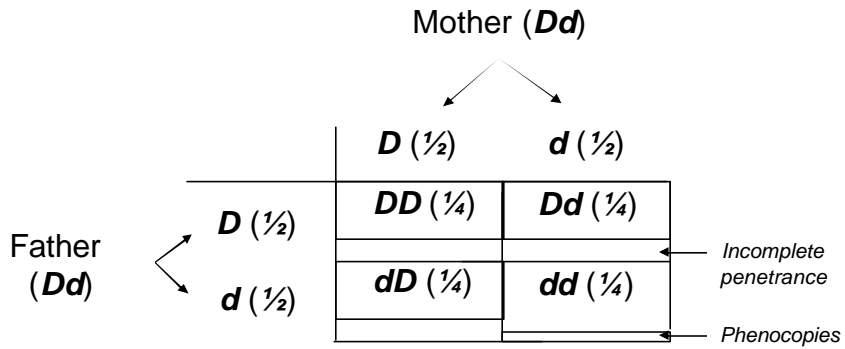
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### 2. Dominant Mendelian inheritance



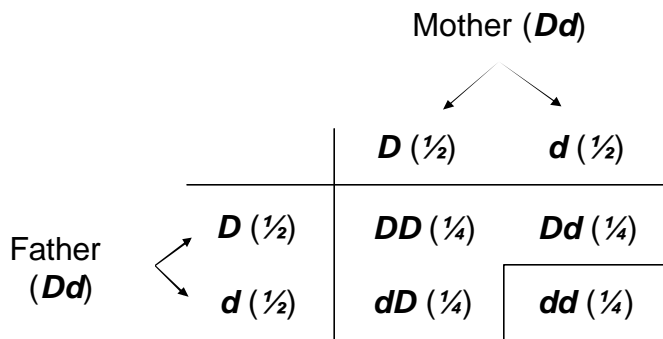
## Phenotype level

### 3. Dominant Mendelian inheritance with incomplete penetrance and phenocopies



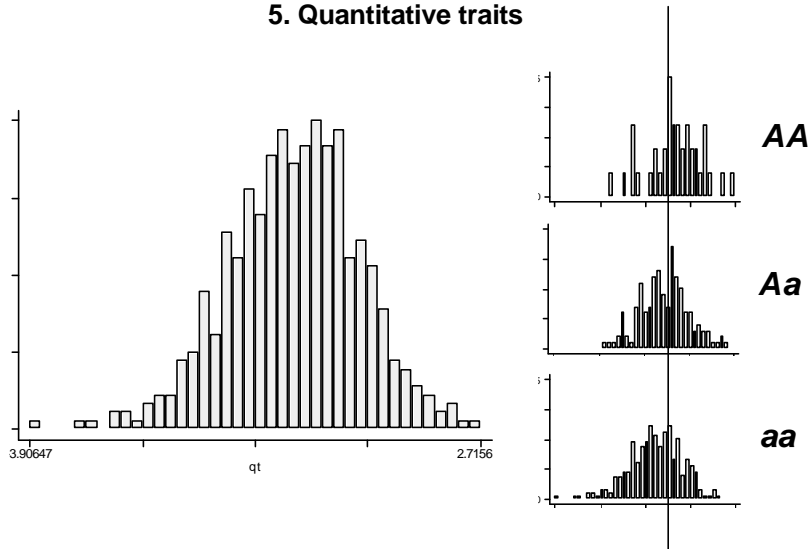
## Phenotype level

### 4. Recessive Mendelian inheritance

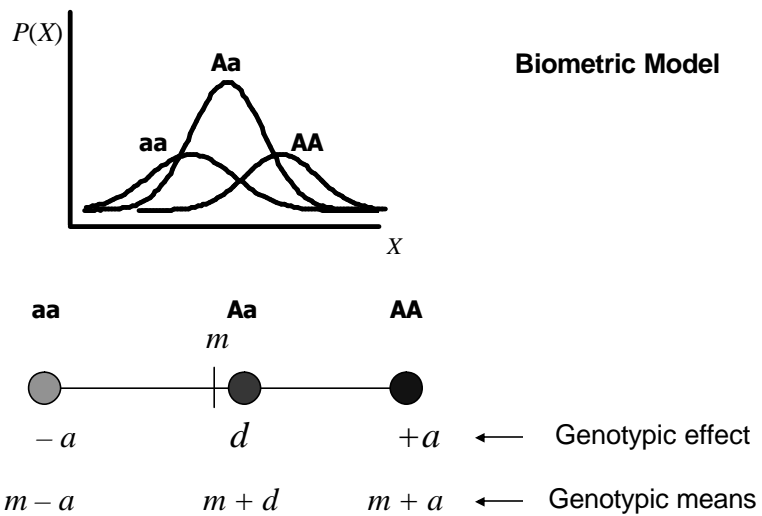


## Phenotype level

### 5. Quantitative traits



## Phenotype level



### **3. Very basic statistical concepts**

#### **Mean, variance, covariance**

##### **1. Mean ( $X$ )**

$$\mathbf{m} = E(X) = \frac{\sum_i x_i}{n} = \sum_i x_i f(x_i)$$

## Mean, variance, covariance

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### 2. Variance (X)

$$\text{Var}(X) = E(X - \mathbf{m})^2 = \frac{\sum_i (x_i - \mathbf{m})^2}{n-1} = \sum_i (x_i - \mathbf{m})^2 f(x_i)$$

## Mean, variance, covariance

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### 3. Covariance (X, Y)

$$\begin{aligned} \text{Cov}(X, Y) &= E(X - \mathbf{m}_X)(Y - \mathbf{m}_Y) = \frac{\sum_i (x_i - \mathbf{m}_X)(y_i - \mathbf{m}_Y)}{n-1} \\ &= \sum_i (x_i - \mathbf{m}_X)(y_i - \mathbf{m}_Y) f(x_i, y_i) \end{aligned}$$

## 4. Biometrical model

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### Biometrical model for single biallelic QTL

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- ▷ Biallelic locus
  - Genotypes: **AA, Aa, aa**
  - Genotype frequencies:  **$p^2, 2pq, q^2$**
- ▷ Alleles at this locus are transmitted from P-O according to Mendel's law of segregation
- ▷ Genotypes for this locus influence the expression of a quantitative trait  $X$  (i.e. locus is a QTL)



**Biometrical genetic model** that estimates the contribution of this QTL towards the **(1) Mean**, **(2) Variance** and **(3) Covariance between individuals** for this quantitative trait  $X$

## Biometrical model for single biallelic QTL

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### 1. Contribution of the QTL to the Mean ( $X$ )

$$m = \sum_i x_i f(x_i)$$

|                     |                         |                         |                         |
|---------------------|-------------------------|-------------------------|-------------------------|
| Genotypes           | <b>AA</b>               | <b>Aa</b>               | <b>aa</b>               |
| Effect, $x$         | <b><math>a</math></b>   | <b><math>d</math></b>   | <b><math>-a</math></b>  |
| Frequencies, $f(x)$ | <b><math>p^2</math></b> | <b><math>2pq</math></b> | <b><math>q^2</math></b> |

$$\text{Mean}(X) = a(p^2) + d(2pq) - a(q^2) = a(p-q) + 2pqd$$

## Biometrical model for single biallelic QTL

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### 2. Contribution of the QTL to the Variance ( $X$ )

$$\text{Var} = \sum_i (x_i - m)^2 f(x_i)$$

|                     |                         |                         |                         |
|---------------------|-------------------------|-------------------------|-------------------------|
| Genotypes           | <b>AA</b>               | <b>Aa</b>               | <b>aa</b>               |
| Effect, $x$         | <b><math>a</math></b>   | <b><math>d</math></b>   | <b><math>-a</math></b>  |
| Frequencies, $f(x)$ | <b><math>p^2</math></b> | <b><math>2pq</math></b> | <b><math>q^2</math></b> |

$$\begin{aligned} \text{Var}(X) &= (a-m)^2 p^2 + (d-m)^2 2pq + (-a-m)^2 q^2 \\ &= V_{QTL} \end{aligned}$$

$$\text{Broad-sense heritability of } X \text{ at this locus} = V_{QTL} / V_{\text{Total}}$$

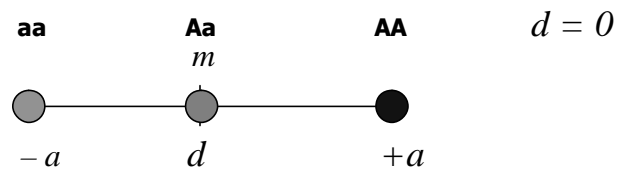
$$\text{Broad-sense total heritability of } X = SV_{QTL} / V_{\text{Total}}$$

## Biometrical model for single biallelic QTL

$$\begin{aligned} \text{Var}(X) &= (a-m)^2p^2 + (d-m)^22pq + (-a-m)^2q^2 \\ &= \underline{2pq[a+(q-p)d]^2} + \underline{(2pqd)^2} \\ &= V_{A_{QTL}} + V_{D_{QTL}} \end{aligned}$$

Additive effects: the main effects of individual alleles

Dominance effects: represent the interaction between alleles

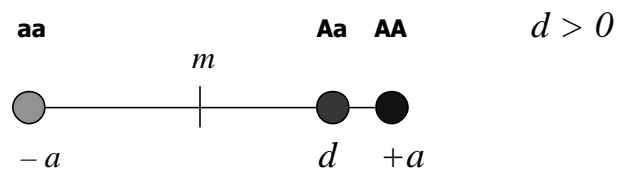


## Biometrical model for single biallelic QTL

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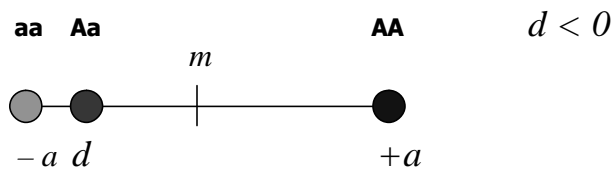


## Biometrical model for single biallelic QTL

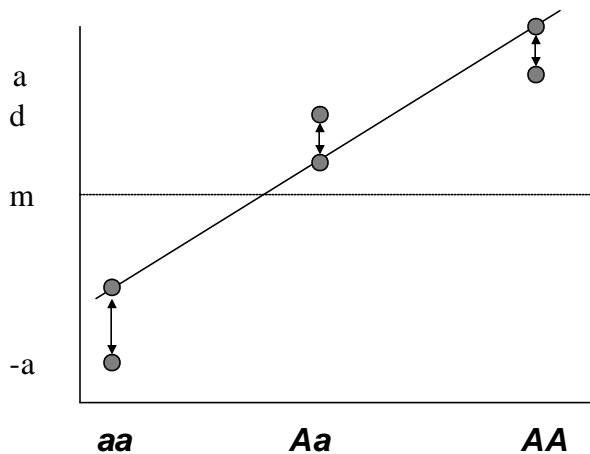
$$\begin{aligned} \text{Var}(X) &= (a-m)^2p^2 + (d-m)^22pq + (-a-m)^2q^2 \\ &= \underline{2pq[a+(q-p)d]^2} + \underline{(2pqd)^2} \\ &= V_{A_{QTL}} + V_{D_{QTL}} \end{aligned}$$

Additive effects: the main effects of individual alleles

Dominance effects: represent the interaction between alleles



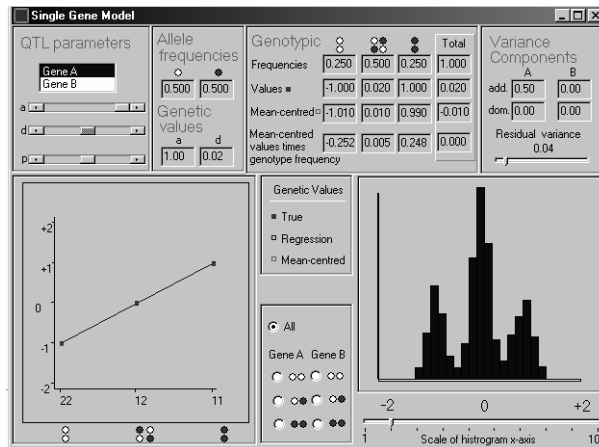
## Biometrical model for single biallelic QTL



$$\begin{aligned} \text{Var}(X) &= \text{Regression Variance} + \text{Residual Variance} \\ &= \text{Additive Variance} + \text{Dominance Variance} \end{aligned}$$

# Practical

H:\manuel\Biometric\sgene.exe



# Practical

- ▷ **Aim** Visualize graphically how allele frequencies, genetic effects, dominance, etc, influence trait mean and variance

## Ex1

$a=0$ ,  $d=0$ ,  $p=0.4$ , Residual Variance = 0.04, Scale = 2.  
Vary  $\underline{a}$  from 0 to 1.

## Ex2

$a=1$ ,  $d=0$ ,  $p=0.4$ , Residual Variance = 0.04, Scale = 2.  
Vary  $\underline{d}$  from -1 to 1.

## Ex3

$a=1$ ,  $d=0$ ,  $p=0.4$ , Residual Variance = 0.04, Scale = 2.  
Vary  $\underline{p}$  from 0 to 1.

**Look at scatter-plot, histogram and variance components.**

## Some conclusions

1. Additive genetic variance depends on

*allele frequency*  $p$

& *additive genetic value*  $a$

as well as

*dominance deviation*  $d$

2. Additive genetic variance typically greater than dominance variance

## Biometrical model for single biallelic QTL

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$$\text{Var}(X) = \frac{2pq[a+(q-p)d]^2}{\downarrow \text{Demonstrate}} + \frac{(2pqd)^2}{V_{A_{QTL}} + V_{D_{QTL}}}$$

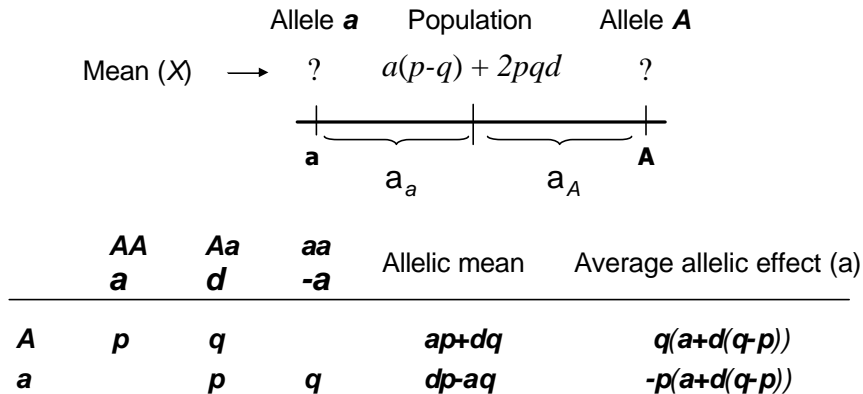
2A. Average allelic effect

2B. Additive genetic variance

## Biometrical model for single biallelic QTL

### 2A. Average allelic effect ( $a$ )

The deviation of the allelic mean from the population mean



## Biometrical model for single biallelic QTL

- ▷ Denote the average allelic effects
  - $a_A = q(a+d(q-p))$
  - $a_a = -p(a+d(q-p))$
- ▷ If only two alleles exist, we can define the *average effect of allele substitution*
  - $a = a_A - a_a$
  - $a = (q(-p))(a+d(q-p)) = (a+d(q-p))$
- ▷ Therefore:
  - $a_A = qa$
  - $a_a = -pa$

## Biometrical model for single biallelic QTL

2A. Average allelic effect ( $a$ )

2B. Additive genetic variance

The variance of the average allelic effects

$$\begin{aligned} a_A &= qa \\ a_a &= -pa \end{aligned}$$

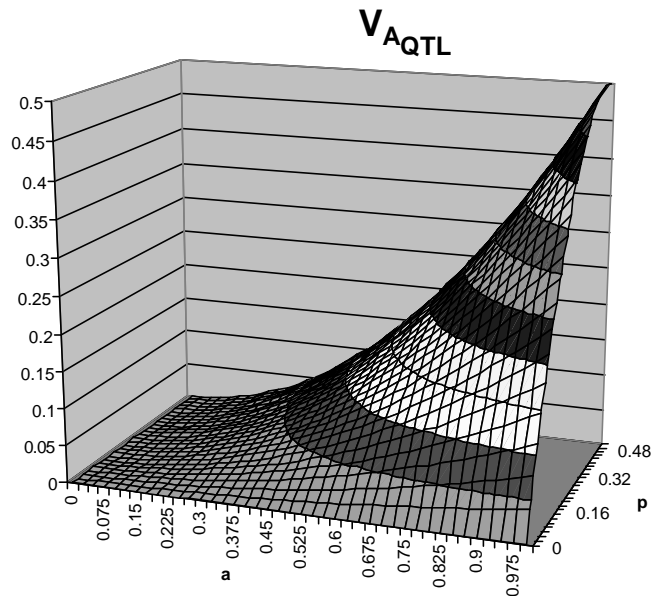
|           | Freq. | Additive effect |            |
|-----------|-------|-----------------|------------|
| <b>AA</b> | $p^2$ | $2a_A$          | $= 2qa$    |
| <b>Aa</b> | $2pq$ | $a_A + a_a$     | $= (q-p)a$ |
| <b>aa</b> | $q^2$ | $2a_a$          | $= -2pa$   |

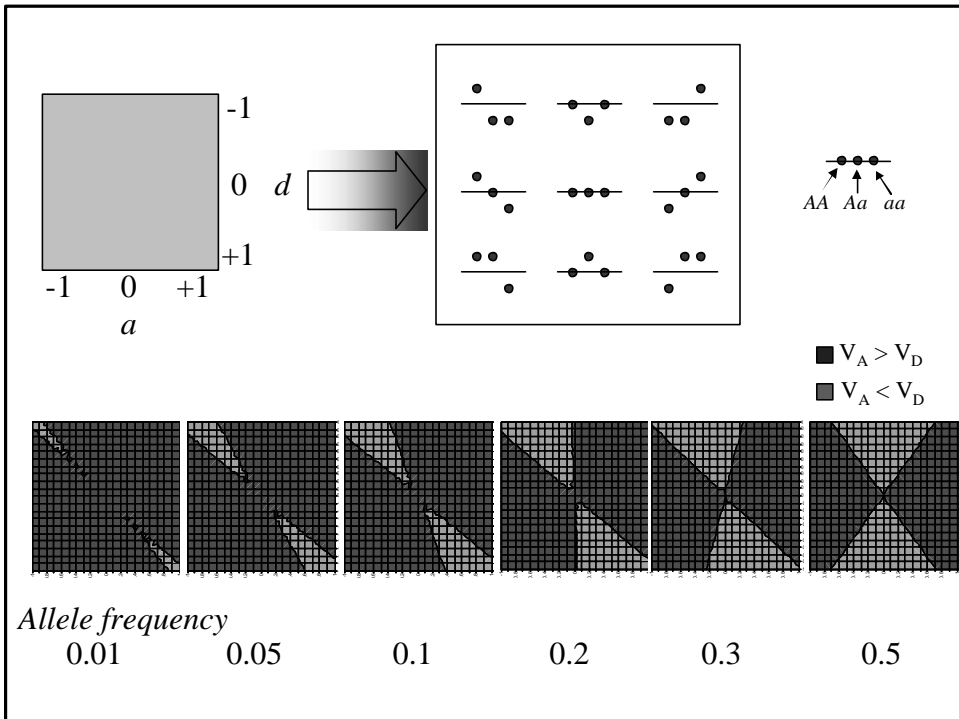
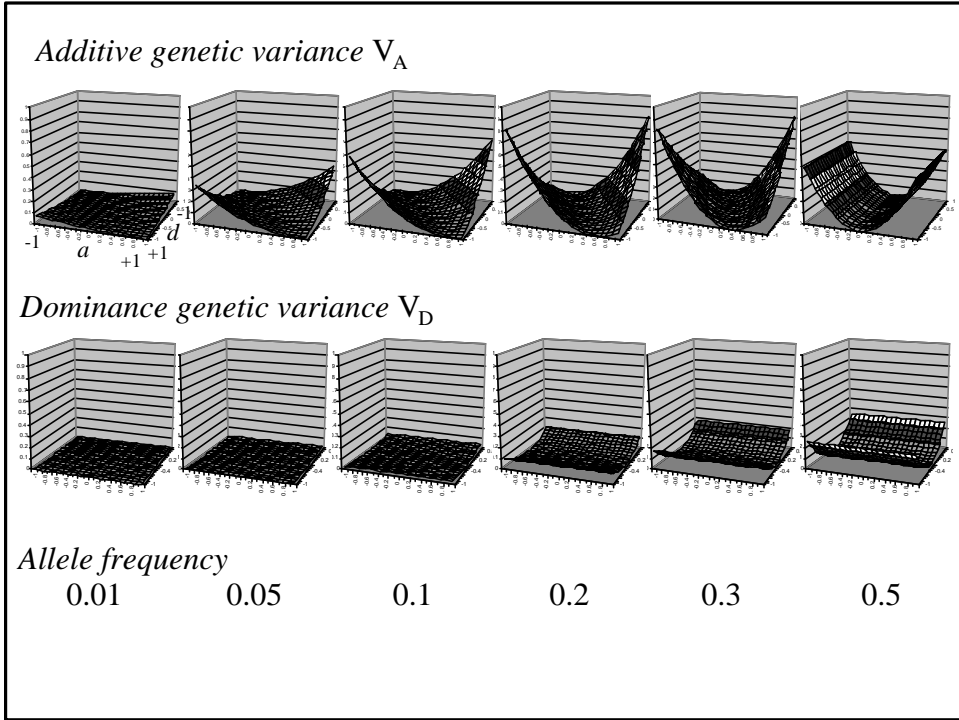
$$\begin{aligned} V_{A_{QTL}} &= (2qa)^2 p^2 + ((q-p)a)^2 2pq + (-2pa)^2 q^2 \\ &= 2pqa^2 \\ &= 2pq[a + d(q-p)]^2 \end{aligned}$$

$$d = 0, V_{A_{QTL}} = 2pqa^2$$

$$p = q, V_{A_{QTL}} = \frac{1}{2}a^2$$

$$d = 0, V_{A_{QTL}} = 2pqa^2$$





## Biometrical model for single biallelic QTL

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### 1. Contribution of the QTL to the Mean ( $X$ )

### 2. Contribution of the QTL to the Variance ( $X$ )

2A. Average allelic effect ( $a$ )

2B. Additive genetic variance

### 3. Contribution of the QTL to the Covariance ( $X, Y$ )

## Biometrical model for single biallelic QTL

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### 3. Contribution of the QTL to the Cov ( $X, Y$ )

$$\text{Cov}(X, Y) = \sum_i (x_i - m_x)(y_i - m_y) f(x_i, y_i)$$

|               | AA ( $a-m$ )  | Aa ( $d-m$ )  | aa ( $-a-m$ ) |
|---------------|---------------|---------------|---------------|
| AA ( $a-m$ )  | $(a-m)^2$     |               |               |
| Aa ( $d-m$ )  | $(a-m)(d-m)$  | $(d-m)^2$     |               |
| aa ( $-a-m$ ) | $(a-m)(-a-m)$ | $(d-m)(-a-m)$ | $(-a-m)^2$    |

## Biometrical model for single biallelic QTL

### 3A. Contribution of the QTL to the Cov (X, Y) – MZ twins

$$\text{Cov}(X, Y) = \sum_i (x_i - m_x)(y_i - m_y) f(x_i, y_i)$$

|           | AA (a-m)       | Aa (d-m)       | aa (-a-m)     |
|-----------|----------------|----------------|---------------|
| AA (a-m)  | $p^2(a-m)^2$   |                |               |
| Aa (d-m)  | $0(a-m)(d-m)$  | $2pq(d-m)^2$   |               |
| aa (-a-m) | $0(a-m)(-a-m)$ | $0(d-m)(-a-m)$ | $q^2(-a-m)^2$ |

$$\begin{aligned} \text{Covar}(X_i, X_j) &= (a-m)^2 p^2 + (d-m)^2 2pq + (-a-m)^2 q^2 \\ &= 2pq[a+(q-p)d]^2 + (2pqd)^2 = V_{A_{QTL}} + V_{D_{QTL}} \end{aligned}$$

## Biometrical model for single biallelic QTL

### 3B. Contribution of the QTL to the Cov (X, Y) – Parent-Offspring

|           | AA (a-m)         | Aa (d-m)          | aa (-a-m)     |
|-----------|------------------|-------------------|---------------|
| AA (a-m)  | $p^3(a-m)^2$     |                   |               |
| Aa (d-m)  | $p^2q(a-m)(d-m)$ | $pq(d-m)^2$       |               |
| aa (-a-m) | $0(a-m)(-a-m)$   | $pq^2(d-m)(-a-m)$ | $q^3(-a-m)^2$ |

- e.g. given an AA father, an AA offspring can come from either AA x AA or AA x Aa parental mating types

AA x AA will occur  $p^2 \times p^2 = p^4$   
and have AA offspring Prob() $=1$

AA x Aa will occur  $p^2 \times 2pq = 2p^3q$   
and have AA offspring Prob() $=0.5$   
and have Aa offspring Prob() $=0.5$

$$\begin{aligned} \text{Therefore, P(AA father \& AA offspring)} &= p^4 + p^3q \\ &= p^3(p+q) \\ &= p^3 \end{aligned}$$

## Biometrical model for single biallelic QTL

### 3B. Contribution of the QTL to the Cov (X, Y) – Parent-Offspring

|           | AA (a-m)         | Aa (d-m)          | aa (-a-m)     |
|-----------|------------------|-------------------|---------------|
| AA (a-m)  | $p^3(a-m)^2$     |                   |               |
| Aa (d-m)  | $p^2q(a-m)(d-m)$ | $pq(d-m)^2$       |               |
| aa (-a-m) | $0(a-m)(-a-m)$   | $pq^2(d-m)(-a-m)$ | $q^3(-a-m)^2$ |

$$\begin{aligned} \text{Cov}(X_i, X_j) &= (a-m)^2p^3 + \dots + (-a-m)^2q^3 \\ &= pq[a+(q-p)d]^2 = \frac{1}{2}V_{A_{QTL}} \end{aligned}$$

## Biometrical model for single biallelic QTL

### 3C. Contribution of the QTL to the Cov (X, Y) – Unrelated individuals

|           | AA (a-m)              | Aa (d-m)             | aa (-a-m)      |
|-----------|-----------------------|----------------------|----------------|
| AA (a-m)  | $p^4 (a-m)^2$         |                      |                |
| Aa (d-m)  | $2p^3q (a-m) (d-m)$   | $4p^2q^2 (d-m)^2$    |                |
| aa (-a-m) | $p^2q^2 (a-m) (-a-m)$ | $2pq^3 (d-m) (-a-m)$ | $q^4 (-a-m)^2$ |

$$\begin{aligned} \text{Cov}(X_i, X_j) &= (a-m)^2 p^4 + \dots + (-a-m)^2 q^4 \\ &= 0 \end{aligned}$$

## Biometrical model for single biallelic QTL

### 3D. Contribution of the QTL to the Cov (X, Y) – DZ twins and full sibs

|  | ¼ genome                          | ¼ genome                         | ¼ genome                         | ¼ genome                          |
|--|-----------------------------------|----------------------------------|----------------------------------|-----------------------------------|
| # identical alleles inherited from parents | <b>2</b>                          | <b>1</b><br>(father)             | <b>1</b><br>(mother)             | <b>0</b>                          |
|  | $\frac{1}{4} (2 \text{ alleles})$ | $\frac{1}{2} (1 \text{ allele})$ | $\frac{1}{2} (1 \text{ allele})$ | $\frac{1}{4} (0 \text{ alleles})$ |
|  | MZ twins                          | P-O                              |                                  | Unrelateds                        |

$$\begin{aligned} \text{Cov}(X_i, X_j) &= \frac{1}{4} \text{Cov}(MZ) + \frac{1}{2} \text{Cov}(P-O) + \frac{1}{4} \text{Cov}(Unrel) \\ &= \frac{1}{4} (V_{A_{QTL}} + V_{D_{QTL}}) + \frac{1}{2} (\frac{1}{2} V_{A_{QTL}}) + \frac{1}{4} (0) \\ &= \frac{1}{2} V_{A_{QTL}} + \frac{1}{4} V_{D_{QTL}} \end{aligned}$$

## Summary

- ▷ Biometrical model predicts contribution of a QTL to the mean, variance and covariances of a trait

### **1 QTL**

$$\text{Var}(X) = V_{A_{QTL}} + V_{D_{QTL}}$$

$$\text{Cov}(MZ) = V_{A_{QTL}} + V_{D_{QTL}}$$

$$\text{Cov}(DZ) = \frac{1}{2}V_{A_{QTL}} + \frac{1}{4}V_{D_{QTL}}$$

### **Multiple QTL**

$$\text{Var}(X) = S(V_{A_{QTL}}) + S(V_{D_{QTL}}) = V_A + V_D$$

$$\text{Cov}(MZ) = S(V_{A_{QTL}}) + S(V_{D_{QTL}}) = V_A + V_D$$

$$\text{Cov}(DZ) = S(\frac{1}{2}V_{A_{QTL}}) + S(\frac{1}{4}V_{D_{QTL}}) = \frac{1}{2}V_A + \frac{1}{4}V_D$$

- ▷ Biometrical model underlies the variance components estimation performed in Mx

$$\text{Var}(X) = V_A + V_D + V_E$$

$$\text{Cov}(MZ) = V_A + V_D$$

$$\text{Cov}(DZ) = \frac{1}{2}V_A + \frac{1}{4}V_D$$