

Quantitative characters - solutions to exercises

1. a) $Cov(HS) = \frac{1}{4}\sigma_{10}^2 + 0\sigma_{01}^2 + \frac{1}{16}\sigma_{20}^2 + 0\sigma_{11}^2 + 0\sigma_{02}^2$
- b) $Cov(FS) = \frac{1}{2}\sigma_{10}^2 + \frac{1}{4}\sigma_{01}^2 + \frac{1}{4}\sigma_{20}^2 + \frac{1}{8}\sigma_{11}^2 + \frac{1}{16}\sigma_{02}^2$
- c) $\sigma_D^2 = \sigma_{01}^2 \approx 4[Cov(FS) - 2Cov(HS)]$

Bias due to epistatic effects

2. a) $a_{XY} = \left(\frac{1}{2}\right)^{2+2}(1+F_A) + \left(\frac{1}{2}\right)^{2+2}(1+F_B) = \frac{1}{16} + \frac{1}{16} = \frac{1}{8}$
- $d_{XY} = \frac{1}{4}(a_{ED} \times a_{CF} + a_{EF} \times a_{CD}) = \frac{1}{4}\left(0 \times 0 + 0 \times \frac{1}{2}\right) = 0$

b)
$$Cov(X,Y) = \sum_{i=0}^n \sum_{j=0}^n \left(\frac{1}{8}\right)^i (0)^j \sigma_{ij}^2$$

$$1 \leq i+j \leq n$$

When maximum three loci are considered: $Cov(X,Y) =$

$$= \left(\frac{1}{8}\right)^1(0)^0 \sigma_{10}^2 + \left(\frac{1}{8}\right)^0(0)^1 \sigma_{01}^2 + \left(\frac{1}{8}\right)^2(0)^0 \sigma_{20}^2 + \left(\frac{1}{8}\right)^1(0)^1 \sigma_{11}^2 +$$

$$+ \left(\frac{1}{8}\right)^0(0)^2 \sigma_{02}^2 + \left(\frac{1}{8}\right)^3(0)^0 \sigma_{30}^2 + \left(\frac{1}{8}\right)^2(0)^1 \sigma_{21}^2 + \left(\frac{1}{8}\right)^1(0)^2 \sigma_{12}^2 +$$

$$+ \left(\frac{1}{8}\right)^0(0)^3 \sigma_{03}^2 + \dots$$

$$= \frac{1}{8}\sigma_{10}^2 + 0 + \frac{1}{64}\sigma_{20}^2 + 0 + 0 + \frac{1}{512}\sigma_{30}^2 + 0 + 0 + 0 \dots$$

The genetic covariance between cousins is composed only of additive variance and epistatic variance between additive gene effects

$$\begin{aligned}
3. \quad a_{XY} &= \left(\frac{1}{2}\right)^{2+2}(1+F_A) + \left(\frac{1}{2}\right)^{2+2}(1+F_B) + \left(\frac{1}{2}\right)^{2+2}(1+F_C) + \left(\frac{1}{2}\right)^{2+2}(1+F_D) \\
&= \frac{1}{16} + \frac{1}{16} + \frac{1}{16} + \frac{1}{16} = \frac{4}{16} = \frac{1}{4}
\end{aligned}$$

$$d_{XY} = \frac{1}{4}(a_{FE} \times a_{GH} + a_{FH} \times a_{GE}) = \frac{1}{4}\left(\frac{1}{2} \times \frac{1}{2} + 0 \times 0\right) = \frac{1}{4} \times \frac{1}{4} = \frac{1}{16}$$

$$\begin{aligned}
\text{Cov}(X,Y) &= \left(\frac{1}{4}\right)^1 \left(\frac{1}{16}\right)^0 \sigma_{10}^2 + \left(\frac{1}{4}\right)^0 \left(\frac{1}{16}\right)^1 \sigma_{01}^2 + \left(\frac{1}{4}\right)^2 \left(\frac{1}{16}\right)^0 \sigma_{20}^2 + \left(\frac{1}{4}\right)^1 \left(\frac{1}{16}\right)^1 \sigma_{11}^2 \\
&\quad + \left(\frac{1}{4}\right)^0 \left(\frac{1}{16}\right)^2 \sigma_{02}^2 \\
&= \frac{1}{4} \sigma_{10}^2 + \frac{1}{16} \sigma_{01}^2 + \frac{1}{16} \sigma_{20}^2 + \frac{1}{64} \sigma_{11}^2 + \frac{1}{256} \sigma_{02}^2
\end{aligned}$$

$$4. \quad \text{a) Parent-Offspring Cov(PO)} = \frac{1}{2} \sigma_{10}^2 + \frac{1}{4} \sigma_{20}^2$$

c)

	σ_{10}^2	σ_{01}^2	σ_{20}^2	σ_{11}^2	σ_{02}^2
2 x Cov(OP)	1		$\frac{1}{2}$		
2 x Cov(FS)	1	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{4}$	$\frac{1}{8}$
4 x Cov(HS)	1		$\frac{1}{4}$		

* Cov(OP) and Cov(HS) only give bias due to epistatic (add. x add.) gene effects. This bias is smaller for Cov(HS) compared to Cov(OP).

* Cov(FS) gives biased estimates of σ_A^2 due to influence of dominance effects and also all types of epistatic effects.

The estimates of σ_{10}^2 calculated from the genetic covariance between half sibs contains the least bias as regards non-additive components.

5. a) $Y_{ij} = \mu + s_i + e_{ij} \quad E(S_i) = 0 \quad E(S_i^2) = \sigma_s^2$

$E(e_{ij}) = 0 \quad E(e_{ij}^2) = \sigma_e^2$

Source of var.	d.f.	SS ^I	MS	E(MS)
Between sires	4	C-B	4299	$\sigma_e^2 + 8 \sigma_s^2$
Within sires	35	A-C	2334	σ_e^2
Total	39	A-B		

b)

1) $A = \sum_i \sum_j Y_{ij}^2 = 18773473$

$B = (\sum_{ij} Y_{ij})^2 / N = (27331)^2 / 40 = 18674589$

$C = \sum_i (\sum_j Y_{ij})^2 / n_i = \frac{(5720)^2}{8} = + \dots + \frac{(5466)^2}{8} = 18691786$

c) $\sigma_e^2 = 2334 \quad \sigma_s^2 = \frac{4299 - 2334}{8} = 246$

$\sigma_A^2 = 4 \sigma_s^2 = 4 \times 246 = 984$

$\sigma_p^2 = \sigma_s^2 + \sigma_e^2 = 2334 + 246 = 2580$

d) The genetic covariance (HS) has the following composition:

$$\frac{1}{4} \sigma_{10}^2 + \frac{1}{16} \sigma_{20}^2 + \frac{1}{64} \sigma_{30}^2 + \dots$$

- Interaction between non-allelic genes, i.e. σ_{20}^2 , σ_{30}^2 give some bias as an

overstimation of σ_A^2 .

- Interaction between non-allelic genes (dominance), i.e. σ_{01}^2 , is not included in Cov(HS)

and thus do not affect the estimate of σ_A^2 .

e) $h^2 = \frac{\sigma_A^2}{\sigma_p^2} = \frac{984}{2580} = 0.38$

6.

Source of var.	d.f.	SS ¹	MS	E(MS)
Between sires	39	5000	128.20	$\sigma_e^2 + 88\sigma_s^2$
Within sires	3480	46800	13.45	σ_e^2
Total	3519	51800		

$$a) \quad n = \frac{3520}{40} = 88$$

$$\sigma_e^2 = 13.45; \quad \sigma_s^2 = \frac{128.20 - 13.45}{88} = 1.30$$

$$\sigma_A^2 = \frac{1}{a_{HS}} \times \sigma_s^2 = \frac{1}{0.28} \times 1.30 = 4.64$$

$$\sigma_p^2 = \sigma_s^2 + \sigma_e^2 = 1.30 + 13.34 = 14.75$$

- b) Yes. The trait is sex limited and the covariance sire-offspring can not be used. The covariance dam-offspring is not so good since the environmental covariance dam-daughter is usually $\neq 0$.

$$c) \quad \hat{h}^2 = \frac{\sigma_A^2}{\sigma_p^2} = \frac{4.64}{14.75} = 0.31$$

$$7. \quad a) \quad Y_{ijk} = \mu + s_i + d_{ij} + v_K + e_{ijk}$$

$$E(s_i) = 0 \quad E(s_i^2) = \sigma_s^2$$

$$E(d_{ij}) = 0 \quad E(d_{ij}^2) = \sigma_d^2$$

$$E(e_{ijk}) = 0 \quad E(e_{ijk}^2) = \sigma_e^2$$

b)

Source of var.	d.f.	SS	MS	EMS
Between sires	137	184.9	1.35	$\sigma_e^2 + 2\sigma_d^2 + 6\sigma_s^2$
Between sows	276	223.5	0.81	$\sigma_e^2 + 2\sigma_d^2$
Within weights	1	9.4	9.40	$\sigma_e^2 + K \kappa^2$
Residual	413	251.9	0.61	σ_e^2
Total	827	669.7		

$$c) \quad \sigma_e^2 = 0.61 \qquad \sigma_s^2 = \frac{1.35 - 0.81}{6} = 0.09$$

$$\sigma_d^2 = \frac{0.81 - 0.61}{2} = 0.10$$

$$d) \quad \sigma_A^2 = 4\sigma_s^2 = 4 \times 0.09 = 0.36$$

σ_s^2 does not give any bias due to dominance or maternal effects and smaller bias due to epistatic effects compared to σ_d^2 .

$$e) \quad \sigma_P^2 = \sigma_s^2 + \sigma_d^2 + \sigma_e^2 = 0.09 + 0.10 + 0.61 = 0.80$$

$$f) \quad h^2 = \frac{\sigma_A^2}{\sigma_P^2} = \frac{0.36}{0.80} = 0.45$$

$$g) \quad \hat{r}_g = \frac{\sigma_{\tilde{s}\tilde{s}}}{\sqrt{\sigma_s^2 \times \sigma_{\tilde{s}}^2}} = \frac{0.06}{\sqrt{0.09 \times 0.40}} = 0.32$$

$$\hat{r}_p = \frac{\sigma_{s\tilde{s}} + \sigma_{d\tilde{d}} + \sigma_{e\tilde{e}}}{\sqrt{(\sigma_s^2 + \sigma_d^2 + \sigma_e^2)(\sigma_{\tilde{s}}^2 + \sigma_{\tilde{d}}^2 + \sigma_{\tilde{e}}^2)}} = \frac{0.65}{\sqrt{0.8 \times 2.1}} = 0.50$$

8. a) Weight gain: $\sigma_e^2 = 0.0110$; $\sigma_s^2 = \frac{0.0187 - 0.0110}{17} = 0.00045$

Height at withers: $\sigma_e^2 = 1258.5$; $\sigma_s^2 = \frac{3097.2 - 1258.5}{17} = 108.159$

Covariance comp.: $\sigma_{e\tilde{e}} = 1.1746$; $\sigma_{s\tilde{s}} = \frac{3.1978 - 1.1746}{17} = 0.119$

b) $\sigma_A^2 = 4\sigma_s^2 = 4 \times 0.00045 = 0.0018$

$\sigma_{\tilde{A}}^2 = 4\sigma_{\tilde{s}}^2 = 4 \times 108.159 = 432.636$

$\sigma_{A\tilde{A}} = 4\sigma_{s\tilde{s}} = 4 \times 0.119 = 0.476$

c) $\sigma_{E\tilde{E}} = \sigma_{e\tilde{e}} - 3\sigma_{s\tilde{s}} = 1.1746 - 3 \times 0.119 = 0.8176$

$\sigma_{P\tilde{P}} = \sigma_{s\tilde{s}} + \sigma_{e\tilde{e}} = 0.119 + 1.1746 = 1.2936$

d) $0.476 + 0.8176 = 1.2936 \quad \therefore \sigma_{P\tilde{P}} = \sigma_{A\tilde{A}} + \sigma_{E\tilde{E}}$

e) $\hat{r}_g = \frac{\sigma_{A\tilde{A}}^2}{\sqrt{\sigma_A^2 \times \sigma_{\tilde{A}}^2}} = \frac{0.476}{\sqrt{0.0018 \times 432.636}} = 0.54$

\hat{r} can also be calculated directly from the statistical components according to:

$\hat{r}_g = \frac{\sigma_{s\tilde{s}}}{\sqrt{\sigma_s^2 \times \sigma_{\tilde{s}}^2}} = \frac{0.119}{\sqrt{0.00045 \times 108.159}} = 0.54$

Conclusions: When we select for weight gain we will also get a correlated increase in height at withers or vice versa. The correlation is high enough for an indirect selection.

f)

$$\hat{r}_P = \frac{\sigma_{P\tilde{P}}}{\sqrt{\sigma_P^2 \times \sigma_{\tilde{P}}^2}} = \frac{\sigma_{s\tilde{s}} + \sigma_{e\tilde{e}}}{\sqrt{(\sigma_s^2 + \sigma_e^2)(\sigma_{\tilde{s}}^2 + \sigma_{\tilde{e}}^2)}} = \frac{0.119 + 1.1746}{\sqrt{(0.00045 + 0.0110)(108.159 + 1258.5)}} = 0.33$$

g)
$$\hat{r}_e = \frac{\sigma_{e\tilde{e}}}{\sqrt{\sigma_e^2 \times \sigma_{\tilde{e}}^2}}$$

$$\sigma_{e\tilde{e}} = 0.8176$$

$$\sigma_E^2 = \sigma_e^2 - 3\sigma_s^2 = 0.0110 - 3 \times 0.00045 = 0.00965$$

$$\sigma_{\tilde{E}}^2 = \sigma_{\tilde{e}}^2 - 3\sigma_{\tilde{s}}^2 = 1258.5 - 3 \times 108.159 = 934.023$$

$$\hat{r}_e = \frac{0.8176}{\sqrt{0.00965 \times 934.023}} = 0.27$$

9. a)
$$\sigma_{P_1'P_2'} = \underbrace{\sigma_{G_1'G_2'}}_{\text{Genetic covariance between halfsibs}} + \underbrace{\sigma_{G_1'E_2} + \sigma_{G_2'E_1}}_{\text{Covariance between genotype and environment of halfsibs}} + \underbrace{\sigma_{E_1E_2}}_{\text{Covariance between environments of the halfsibs}}$$

$$\sigma_{G_1'G_2'} = Cov(XY) = \underbrace{\frac{1}{4}\sigma_{10}^2}_{\text{Add. genetic variance}} + \underbrace{\frac{1}{16}\sigma_{20}^2}_{\text{Add. x Add. genetic variance}} + \underbrace{\frac{1}{64}\sigma_{30}^2}_{\text{Add. x Add. x Add. genetic variance}} +$$

Variance due to epistatic effects, σ_I^2

b) $\sigma_{G_1'E_2}$ $\sigma_{G_2'E_1}$ $\sigma_{E_1E_2}$ and σ_I^2

10. Quantitative traits are influenced both by a number of genes and by environment. They can be described by μ and σ^2 but gene frequencies and genotypes can not be estimated. The estimated biological variances covariances are used for:

- * Calculation of genetic parameters \Rightarrow description of the traits.
- * Breeding values: Calculations of selection index and BLUP assumes known values of

σ_A^2 , σ_P^2 (or quotas), $\sigma_{A\tilde{A}}$ and $\sigma_{P\tilde{P}}$.

- * Prediction of genetic gain and changes in correlated traits assumes known values of

σ_A^2 and $\sigma_{A\tilde{A}}$.

- 11.** * σ_A^2 constitutes the major part of the genetic variance for most of the traits.
- * σ_A^2 is that part of the genetic variance which is utilized in selection. Individuals submit genes and not genotypes to their offspring.
- 12.** * The traits are measured at the same time and on the same type of animals.
- * Also sex limited and carcass traits can be measured
- * Effects of $\sigma_M^2 + \sigma_C^2$ are omitted for paternal halfsibs, but often occur for example between mothers and daughters.
- * Large materials of sibs are available in field data
- * Half sibs give less epistatic bias than parent-offspring