



Correlated traits: Index selection

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The correlation between two traits

$$r_P = r_A h_X h_Y + r_E e_X e_Y$$

- r_P : the phenotypic correlation between two traits X and Y
- r_A : the genetic correlation due to breeding values between X and Y
- r_E : the environmental correlation between X and Y, including non-additive genetic effects
- h^2 : heritability
- e^2 : $1 - h^2$

Genetic and **environmental correlation** come together to create the phenotypic correlation.

Correlated response to selection

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In the formula:

- $h_X h_Y r_A$ is referred to as the **coheritability**, as it takes the place of the heritability in the direct response equation.

Indirect selection

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$$\begin{aligned}\frac{CR_Y}{R_Y} &= \frac{i_X h_X r_A \sigma_{A_Y}}{i_Y h_Y \sigma_{A_Y}} \\ &= \frac{i_X h_X r_A}{i_Y h_Y}\end{aligned}$$

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 - Note that this exact same property applies to usefulness of molecular markers for **marker-assisted selection**

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Practical considerations

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- If trait Y is very **expensive** and **difficult** to measure, but trait X is very cheap and easy to measure.
 - e.g. high-throughput phenotyping technologies
- Or the desired traits is measurable **in one sex** only, but the secondary traits is measurable in both.
 - e.g. milk yield and body weight in dairy cow

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Genotype-by-environment interaction

- Performance in different environments can be regarded as **two separate, but correlated traits**.
 - Improvement in one environment by selection in another environment can be predicted by knowing the heritability in each environment and the genetic correlation between them.

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- **Independent culling levels:**
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- **Index selection:**
 - Select for multiple traits simultaneously by constructing an index value.
 - Index value is then treated as a single economic trait.

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The **true value of merit** is represented as:

$$\begin{aligned} T &= a_1G_1 + a_2G_2 + \dots + a_mG_m \\ &= \sum_{i=1}^m a_iG_i \end{aligned}$$

- G_i is the genetic value for trait i
- a_i is the economic weight placed on trait i
 - The economic weights are set by the breeder according to production needs and value.

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Correlation between T and I

The goal is to **find the values of the b_i s** that could **maximize the correlations** between T and I , or r_{TI} .

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Finally, we get:

$$\begin{bmatrix} Var(G_1) & Cov(G_1, G_2) & \cdots & Cov(G_m, G_1) \\ Cov(G_1, G_2) & Var(G_2) & \cdots & Cov(G_m, G_2) \\ \vdots & \vdots & \ddots & \vdots \\ Cov(G_1, G_m) & Cov(G_2, G_m) & \cdots & Var(G_m) \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_m \end{bmatrix} = \begin{bmatrix} Var(P_1) & Cov(P_1, P_2) & \cdots & Cov(P_m, P_1) \\ Cov(P_1, P_2) & Var(P_2) & \cdots & Cov(P_m, P_2) \\ \vdots & \vdots & \ddots & \vdots \\ Cov(P_1, P_m) & Cov(P_2, P_m) & \cdots & Var(P_m) \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

Matrix Notation

$$\mathbf{Ga} = \mathbf{Pb}$$

- \mathbf{P} is the phenotypic variance-covariance matrix
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In matrix algebra, taking the inverse of a matrix and multiplying is equivalent to division in scalar algebra.

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- Then the resulting vector of weights will give a selection index that maximized genetic gain in T , the true genetic merit.
- The drawback of this index is that the genetic variance and covariances are often estimated with large amounts error, which may reduce the correlation between I and T .

A problem

This litter size of mice could be increased:

- 1) by selection of females for their litter size
- 2) by selection of both parents for body weight

Which would be the better of these two simple procedures, given the following parameters?

- h^2 of litter size: 0.22
- h^2 of body weight: 0.35
- Genetic correlation r_A : 0.43
- Proportion of selection:
 - Females: 25% ($i_f = 1.271$)
 - Males: 10% ($i_m = 1.755$)