

Problem: convert the breeder's equation into the form used in the QG of correlated traits.

$R = h^2 S$                       The breeder's equation

$\Delta \bar{z}_1 = h_1^2 S_1$                       substitution:  $R = \Delta \bar{z}_1$ , where the 1 stands for trait 1.

$\Delta \bar{z}_1 = \frac{V_{A(1)}}{V_{P(1)}} S_1$                       substitution:  $h_1 = \frac{V_{A(1)}}{V_{P(1)}}$

$\Delta \bar{z}_1 = \frac{V_{A(1)}}{V_{P(1)}} \text{COV}(w, z_1)$                       substitution:  $S_1 = \text{COV}(w, z_1)$

$\Delta \bar{z}_1 = V_{A(1)} \frac{\text{COV}(w, z_1)}{V_{P(1)}}$                       re-arrangement.

$\Delta \bar{z}_1 = V_{A(1)} B(w, z_1)$                       substitution:  $\frac{\text{COV}(w, z_1)}{V_{P(1)}} = B(w, z_1)$  the selection gradient.

$\Delta \bar{z}_1 = G_{11} B_1$                        $V_{A(1)} = G_{11}$  and  $B(w, z_1) = B_1$

$\Delta \bar{z} = \mathbf{GB}$                       the standard form for correlated traits, where **G** is the additive genetic variance/covariance matrix, and **B** is the vector of selection gradients

$$\begin{bmatrix} \Delta \bar{z}_1 \\ \Delta \bar{z}_2 \\ \cdot \\ \cdot \\ \cdot \\ \Delta \bar{z}_p \end{bmatrix} = \begin{bmatrix} G_{11} & G_{12} & \cdot & \cdot & \cdot \\ G_{21} & G_{22} & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ G_{p1} & G_{p2} & \cdot & \cdot & \cdot \end{bmatrix} \begin{bmatrix} B_1 \\ B_2 \\ \cdot \\ \cdot \\ \cdot \\ B_p \end{bmatrix} \quad \text{the matrix/vector expansion}$$

Thus, for two traits, we get

$\Delta \bar{z}_1 = G_{11} B_1 + G_{12} B_2$

and

$\Delta \bar{z}_2 = G_{21} B_1 + G_{22} B_2$

where  $G_{11}$  = the additive genetic variance for trait 1, and  $G_{22}$  = the additive genetic variance for trait 2. And:  $G_{12} = G_{21}$  = Additive genetic covariance between traits 1 and 2.

Now, the selection gradients here are not simple regression coefficients as in the single-trait case. The selection gradients are coefficients in a multiple regression.

$$B_1 = \frac{S_1 V_{P(1)} - S_2 \text{COV}(z_1, z_2)}{V_{P(1)} V_{P(2)} - \text{COV}(z_1, z_2)^2} \quad \text{expansion of a partial regression coefficient}$$

Note that if the  $\text{COV}=0$ , then

$$B_1 = \frac{S_1 V_{P(1)} - 0}{V_{P(1)} V_{P(2)} - 0} = \frac{S_1}{V_{P(1)}}$$

$$\Delta \bar{z}_1 = G_{11} B_1 = G_{11} \frac{S_1}{V_{P(1)}} = \frac{V_A}{V_{P(1)}} S = h^2 S$$

But if the covariance is not equal to zero, selection on trait 2 affects the selection gradient for trait 1.

Here is the main point. Selection on a trait can affect other traits. This could affect the magnitude and even the direction of change.

See article on web by Steve Arnold.