

Response to selection

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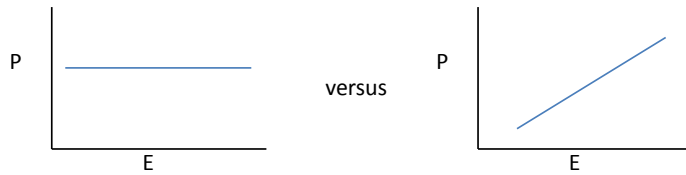
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Response to selection

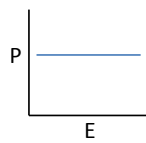
- Interest is in:
- Response in mean level of trait in each environment
- Response in environmental sensitivity

Environmental Sensitivity

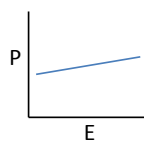


Environmental Sensitivity (“plasticity”)

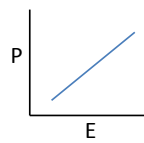
- Change of mean trait value of a genotype with a change in environment



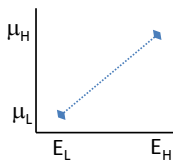
Not sensitive



A bit sensitive



Very sensitive



A measure of sensitivity: $s = \mu_H - \mu_L$

Scaled response in sensitivity: $s^* = \frac{\mu_H^* - \mu_L^*}{\mu_H - \mu_L}$ (WL38.5)

s^* = relative response in the difference

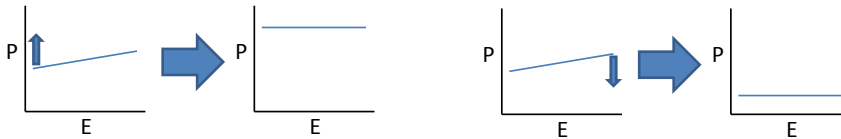
$s^* > 1$ → increase in sensitivity due to selection

$s^* < 1$ → decrease in sensitivity due to selection

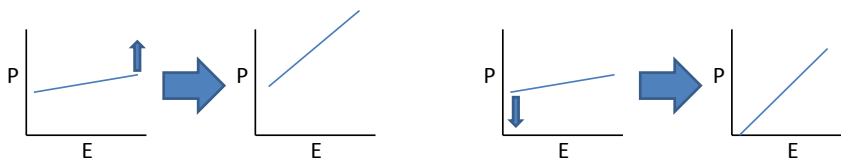
If E is quantitative (i.e. a value): $s = \frac{\mu_H - \mu_L}{E_H - E_L}$ = Slope of the reaction norm

Trends observed in response

- Jinks-Connolly rule
 - Antagonist selection reduces sensitivity



- Synergistic selection increases sensitivity



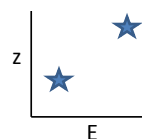
- A trend, not an absolute truth
- Falconer's rule:
 - Sensitivity is less after antagonistic than after synergistic selection: holds more general

Conclusion

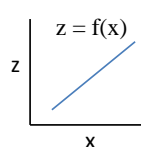
With GxE-interaction,
the environment and direction of selection matter
for the change in environmental sensitivity

Models for GxE-interaction:

Character state model
(multitrait model)

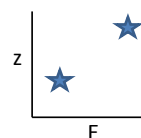


Reaction norm model
(random regression model)



Character State Model (Via and Lande)

- Multitrait approach (Falconer)
 - Population is split into discrete environments
 - Phenotype and genotype in each environment “is” a trait.
 - The different traits (environments) may have different genetic means, genetic (co)variances, and there is a **genetic correlation** between traits in different environments
 - Vector of breeding values in each environment: $\mathbf{g}'_i = [g_{1,i}, g_{2,i}, \dots, g_{m,i}]$
 - i = individual, m = no. of environments
- Applicable when you can split the environment into discrete classes
 - Captures *all* the GxE-interaction between these environments, in contrast to the reaction norm model.
 - Examples: purebreds vs crossbreds, temperate vs tropical climates, location, herd



Character state model: example

Yield of dairy cows in New Zealand vs. Australia



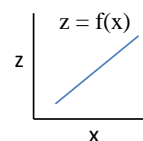
Two traits:

- Milk-yield in Australia
- Milk-yield in New Zealand

Key parameter:

Genetic correlation between both traits (r_g)
 $r_g < 1 \rightarrow$ GxE-interaction

Reaction norm model



- Trait expression is on a continuous environmental scale (x)
 - BV and genetic parameters change gradually along the environmental scale
 - At different locations on the environmental scale, traits may have different means and (co)variances, and there is a genetic correlation between trait expression at different points of the environmental scale
- Applicable when you have a continuous environmental variable that explains (part of) the GxE interaction
 - Temperature, average herd milk yield, HYS effect, Pasture contamination
- Note: Captures only the part of the GxE-interaction due to the environmental variable (the x-axis).
- Advantage RN-model:
 - Explicit model for environmental sensitivity (slope)
 - Prediction in other (non-recorded) environments.

Reaction norm model: basics

- Breeding value is a function of the environment $\rightarrow g(x)$
 - For example $g_i(x) = g_{0,i} + g_{1,i}x + g_{2,i}x^2 + g_{3,i}x^3 + \dots$
 - $g_i(x)$ = breeding values of individual i for environment x
 - x = value of the environment
 - $g_{0,i}, g_{1,i}, g_{2,i}$ are breeding values of individual i

– Vector notation: $g_i(x) = g_{0,i} + g_{1,i}x + g_{2,i}x^2 + \dots = \begin{bmatrix} g_{0,i} & g_{1,i} & g_{2,i} & \dots \end{bmatrix} \begin{bmatrix} 1 \\ x \\ x^2 \\ \vdots \end{bmatrix} = \mathbf{g}_i' \mathbf{x}$

- Phenotype
 - Analogy of $P = G + E \rightarrow P(x) = g(x) + e$
 - Expected phenotype given the genotype: $E[P(x) | g(x)] = g(x)$
 - Expected phenotype equals breeding value, $E[e] = 0$
 - Breeding value predicts the phenotype

Reaction norm model: example Growth rate (GR) and worm egg pasture contamination (PC)



- Idea:
- Sheep grow slower when pasture contamination is higher
 - But not all sheep are equally sensitive to PC (=GxE)

Linear reaction norm: $Y_{i,\text{flock}} = \text{HYS} + g_{0,i} + g_{1,i} \text{PC}_{\text{flock}} + e$

$g_{0,i}$ = breeding value for level

$g_{1,i}$ = breeding value for slope (with respect to PC)

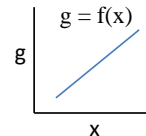
Breeding value: $g_i(PC) = \mathbf{g}_i' \mathbf{x} = \begin{bmatrix} g_{0,i} & g_{1,i} \end{bmatrix} \begin{bmatrix} 1 \\ \text{PC}_{\text{flock}} \end{bmatrix} = g_{0,i} + g_{1,i} \text{PC}_{\text{flock}}$

The linear reaction norm

$$\text{Breeding value: } g_i(x) = g_{0,i} + g_{1,i} x$$

Two BV → Three genetic parameters:

- Genetic variance in level: $\text{var}(g_0)$
- Genetic variance in slope: $\text{var}(g_1)$
- Genetic correlation between level and slope: $(r_{g_{01}})$



Q: What does the correlation between level and slope mean?

The correlation between level and slope

$$\text{Linear reaction norm: } g_i(x) = g_{0,i} + g_{1,i} x$$

$$\text{Genetic correlation level-slope: } r_{g_{01}} = \text{corr}(g_0, g_1)$$

$r_{g_{01}}$ indicates whether animals that have better breeding values for level are also more ($r_{g_{01}} > 0$) or less ($r_{g_{01}} < 0$) environmentally sensitive

Indicates whether selection for level
will increase or decrease environmental sensitivity

However:

RN: Meaning of the correlation between level and slope

- **Message:** Correlation depends on the location of the intercept on the environmental scale (*i.e.* where $x = 0$).
 - General RN: $g(x) = g_0 + g_1x$, where g_0 and g_1 are correlated with $r_{g,01}$
- Problem: Can we shift the intercept so that $r_{g,01}$ becomes zero?
 - Note: intercept is value of $g(x)$ at $x = 0$. Hence, find x where $g(x)$ is independent of g_1 , and put the intercept there.
 - $\text{Cov}[g(x), g_1] = \text{Cov}(g_0 + g_1x, g_1) = 0$

$$x = -r_{g,01} \frac{\sigma_{g_0}}{\sigma_{g_1}} \quad \begin{array}{l} r_{g,01} > 0 \rightarrow \text{shift intercept to the left (down)} \\ r_{g,01} < 0 \rightarrow \text{shift intercept to the right (up)} \end{array}$$

Conclusion: In itself, the correlation between level and slope is meaningless
Suggested convention: define the intercept in the mean environment

Modelling response to selection

Selection Index Theory

- How to predict response to selection?
 - Selection is usually on an index of EBV
 - But we don't have "genetic parameters" of EBV
 - However, EBV are predicted from phenotypes
- Hence, to predict response, we can mimic selection on EBV by selection on phenotypes
 - Villanueva, B., N. R. Wray, et al. (1993) Anim. Prod. 57: 1-13.
- Method: regress true breeding values on phenotypes

Selection index: response to selection (R)

- Selection is on an index: $I = b_1P_1 + b_2P_2 + \dots + b_nP_n = \mathbf{b}'\mathbf{p}$
 - \mathbf{p} are phenotypes, \mathbf{b} are weights.
 - We can use any \mathbf{b} or find the optimum \mathbf{b} .
- Interest is in response to selection in multiple traits
 - $\mathbf{g}' = [g_1, g_2, \dots, g_m]$, a vector of breeding values
- Regression of \mathbf{g} on I yields an expression for response:
 - Approximate normality is assumed

$$R = \begin{bmatrix} \Delta \bar{g}_1 \\ \Delta \bar{g}_2 \\ \vdots \\ \Delta \bar{g}_m \end{bmatrix} = \mathbf{G}'\mathbf{b} \frac{i}{\sigma_I}$$

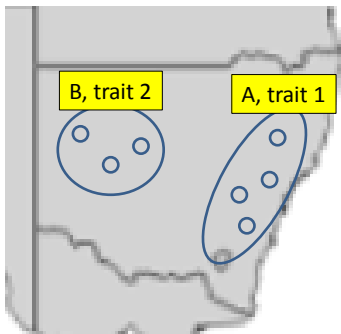
$\mathbf{G} = \text{cov}(\mathbf{p}, \mathbf{g})$ Phenotypes in rows, breeding values in columns

σ_I = standard deviation of the index

i = selection intensity

Response to Selection in the Character-state Model

Multi-environment selection with GxE



- Selection may occur in multiple flocks
- GxE between flocks
 - Mixing of flocks: exchange of rams

Full model is tedious → Illustration for 2-flock system

Character State Model: Response to Selection

- Analogy of the multitrait model
 - Two environments, A and B
 - Two traits, trait 1 expressed in A, trait 2 expressed in B
 - Each individual has a BV for **both** traits, **but** only one of those BV is expressed
 - Two groups of individuals: individuals in A, and individuals in B
 - Different (genetic) means between groups: $\bar{g}_{1,A}$, $\bar{g}_{2,A}$, $\bar{g}_{1,B}$, $\bar{g}_{2,B}$
- Response to selection: 2 steps
 - Step 1: Calculate genetic level of selected parents for both traits in each environment
 - Step 2: Follow the flow of individuals/genes between environments
 - Contribution of each environment to selected individuals
 - Contribution of selected individuals to each environment (mixing)

Step 1: Genetic selection differential per environment

- Selection of individuals that are in Environment A
 - Selection criterion for environment A $\rightarrow I_A = b_{i,A}P_{i,A} + b_{j,A}P_{j,A} + \dots$
 - Intensity of selection for environment A $\rightarrow i_A$
 - Covariances between P_A and true breeding values g_1 and $g_2 \rightarrow \mathbf{G}_A$
 - Response in both traits \rightarrow vector $\Delta\bar{\mathbf{g}}_A$

Genetic selection differential:
$$\Delta\bar{\mathbf{g}}_A = \begin{bmatrix} \Delta\bar{g}_{A,1} \\ \Delta\bar{g}_{A,2} \end{bmatrix} = \mathbf{G}'_A \mathbf{b}_A \frac{i_A}{\sigma_{I_A}}$$

For environment B, the equation is analogous

Step 1: Genetic mean of selected in each environment

Genetic mean of selected indiv. in environment A:

$$\bar{\mathbf{g}}_{A,selected} = \begin{bmatrix} \bar{g}_{1,A,selected} \\ \bar{g}_{2,A,selected} \end{bmatrix} = \begin{bmatrix} \bar{g}_{1,A} \\ \bar{g}_{2,A} \end{bmatrix} + \begin{bmatrix} \Delta\bar{g}_{1,A} \\ \Delta\bar{g}_{2,A} \end{bmatrix}$$

Genetic mean of selected indiv. in environment B:

$$\bar{\mathbf{g}}_{B,selected} = \begin{bmatrix} \bar{g}_{1,B,selected} \\ \bar{g}_{2,B,selected} \end{bmatrix} = \begin{bmatrix} \bar{g}_{1,B} \\ \bar{g}_{2,B} \end{bmatrix} + \begin{bmatrix} \Delta\bar{g}_{1,B} \\ \Delta\bar{g}_{2,B} \end{bmatrix}$$

Step 2: Flow of animals → genetic mean next generation

Breeding value offspring is mean of its parents:

$$\bar{\mathbf{g}}_{A,t+1} = f_{A \rightarrow A} \bar{\mathbf{g}}_{A,selected} + f_{B \rightarrow A} \bar{\mathbf{g}}_{B,selected}$$

$$\begin{bmatrix} \bar{g}_{A,1,t+1} \\ \bar{g}_{A,2,t+1} \end{bmatrix} = f_{A \rightarrow A} \begin{bmatrix} \bar{g}_{A,1,selected} \\ \bar{g}_{A,2,selected} \end{bmatrix} + f_{B \rightarrow A} \begin{bmatrix} \bar{g}_{B,1,selected} \\ \bar{g}_{B,2,selected} \end{bmatrix}$$

$$f_{A \rightarrow A} + f_{B \rightarrow A} = 1$$

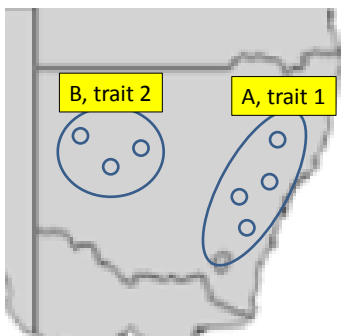
$f_{A \rightarrow A}$ = contribution of selected from A to offspring in A

$f_{B \rightarrow A}$ = contribution of selected from B to offspring in A

Simplest example: correlated response

- Mass selection in environment A, no selection in B
 - E.g. breeding nucleus and production environment
 - Flow of animals: $f_{B \rightarrow A} = 0, f_{B \rightarrow B} = 0, f_{A \rightarrow A} = 1, f_{A \rightarrow B} = 1$

Multi-environment selection with GxE



Selection in environment A

Producers in B get their rams from A

→ Response in both A and B

Simplest example: correlated response

- Mass selection in environment A, no selection in B
 - E.g. breeding nucleus and production environment
 - Flow of animals: $f_{B \rightarrow A} = 0, f_{B \rightarrow B} = 0, f_{A \rightarrow A} = 1, f_{A \rightarrow B} = 1$
 - Interest in response in both environments: two traits
 - Suppose starting level is equal (say zero)

$$\begin{bmatrix} \bar{g}_{A,1,t+1} \\ \bar{g}_{A,2,t+1} \end{bmatrix} = f_{A \rightarrow A} \begin{bmatrix} \bar{g}_{A,1,selected} \\ \bar{g}_{A,2,selected} \end{bmatrix} + f_{B \rightarrow A} \begin{bmatrix} \bar{g}_{B,1,selected} \\ \bar{g}_{B,2,selected} \end{bmatrix} = 1 \times \begin{bmatrix} \bar{g}_{A,1,selected} \\ \bar{g}_{A,2,selected} \end{bmatrix} + 0$$

$$\begin{bmatrix} \bar{g}_{1,A,selected} \\ \bar{g}_{2,A,selected} \end{bmatrix} = \begin{bmatrix} \bar{g}_{1,A} \\ \bar{g}_{2,A} \end{bmatrix} + \begin{bmatrix} \Delta \bar{g}_{1,A} \\ \Delta \bar{g}_{2,A} \end{bmatrix}$$

Hence: We need selection differentials
for both traits
due to selection in environment A

Simplest application: correlated response

$$\text{Selection differentials: } \begin{bmatrix} \Delta \bar{g}_{A,1} \\ \Delta \bar{g}_{A,2} \end{bmatrix} = \mathbf{G}'_A \mathbf{b}_A \frac{i_A}{\sigma_{I_A}}$$

Mass selection in environment A

Selection criterion: $l = P_1 (=1 \times P_1)$

$\mathbf{b} = \mathbf{1}$; $\sigma_l = \sigma_{P,1}$

$$\begin{aligned} \begin{bmatrix} \Delta \bar{g}_{A,1} \\ \Delta \bar{g}_{A,2} \end{bmatrix} &= \begin{bmatrix} \text{Cov}(P_1, g_1) \\ \text{Cov}(P_1, g_2) \end{bmatrix} \frac{i}{\sigma_{P,1}} \\ &= \begin{bmatrix} \sigma_{A,1}^2 \\ r_g \sigma_{A,1} \sigma_{A,2} \end{bmatrix} \frac{i}{\sigma_{P,1}} \\ &= \begin{bmatrix} ih_1 \sigma_{A,1} \\ ih_1 r_g \sigma_{A,2} \end{bmatrix} \quad (\text{see e.g. Falconer \& Mackaybook}) \end{aligned}$$

Response in BV in environment B
is the same ($f_{A \rightarrow B} = 1, f_{B \rightarrow B} = 0$)

Simplest example: correlated response

- $h^2_1 = 0.30$, $h^2_2 = 0.25$, $\sigma^2_{p,1} = 1$, $\sigma^2_{p,2} = 1.33$, $r_g = 0.84$,
selected proportion in env. A = 20% $\rightarrow i_1 = 1.4$

$$\begin{bmatrix} \Delta \bar{g}_{A,1} \\ \Delta \bar{g}_{A,2} \end{bmatrix} = \begin{bmatrix} i_1 h_1 \sigma_{A,1} \\ i_1 h_1 r_g \sigma_{A,2} \end{bmatrix} = \begin{bmatrix} 1.4 \times \sqrt{0.3} \times \sqrt{0.3 \times 1} \\ 1.4 \times \sqrt{0.3} \times 0.84 \times \sqrt{0.25 \times 1.33} \end{bmatrix} = \begin{bmatrix} 0.42 \\ 0.37 \end{bmatrix} = \begin{bmatrix} \Delta \bar{g}_{B,1} \\ \Delta \bar{g}_{B,2} \end{bmatrix}$$

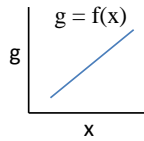
Expressed

Question:

What does this mean for environmental sensitivity / plasticity?

Assignment 1: GxE in sheep breeding

- Breeding program for meat sheep (body weight)
 - Two flocks, one in each environment (High and Low)
 - Both environments have a breeding program, there is exchange of animals
 - Phenotypic means: $\mu_H = 60\text{kg}$, $\mu_L = 50\text{kg}$
 - In generation 0, both breeds are a random sample from a single breed imported from elsewhere
 - In each environment:
 - 50% of the selected parents come from the environment itself (e.g. the dams)
 - 50% come from the other environment (e.g. the sires)
 - Mass selection in both environments
 - Selected proportions in both sexes: $p_H = 20\%$, $p_L = 50\%$
 - $h^2_H = 0.4$; $h^2_L = 0.2$; $r_g = 0.7$; $\sigma_{p,H} = 6\text{kg}$, $\sigma_{p,L} = 5\text{kg}$,
 - Calculate the change of phenotypic means for 10 generations (e.g use Excel)
 - Consequences for environmental sensitivity?
 - Can you achieve zero change in environmental sensitivity? (using mass selection)
 - See Excelsheet “sheepbreeding GxE.xlsx”



$$g_i(x) = g_{0,i} + g_{1,i} x$$

Response to Selection in the Reaction Norm Model

$$R(x) = \Delta \bar{g}_i(x) = \Delta \bar{g}_{0,i} + \Delta \bar{g}_{1,i} x$$

Reaction norm model: response to sel.

- Response is in the average parameters of the RN (mean values of the g 's)
- $\Delta \bar{g}_0, \Delta \bar{g}_1, \Delta \bar{g}_2, \Delta \bar{g}_3, \dots$
- Response in the g 's depends on the environment of selection
- Observed phenotypic response depends on the environment in which the offspring perform

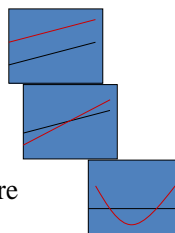
$$\Delta \bar{g}(x_{off}) = \Delta \bar{g}_0 + \Delta \bar{g}_1 x_{off} + \Delta \bar{g}_2 x_{off}^2 + \Delta \bar{g}_3 x_{off}^3 + \dots$$

$$E[\Delta \bar{P}(x_{off})] = \Delta \bar{g}(x_{off})$$

$\Delta \bar{g}_0 \rightarrow$ change in level

$\Delta \bar{g}_1 \rightarrow$ change in slope

$\Delta \bar{g}_{\geq 2} \rightarrow$ change in curvature



$\Delta g(x)$ depends on the changes in level, slope, quadratic, etc., and on the environment in which performance is expressed

Reaction norm model: response to sel.

- Mass selection in a single environment x_s
 - Selection is for phenotype $P(x_s)$
 - Hence, the “index” is: $I = P(x_s) = g_0 + g_1 x_s + g_2 x_s^2 + \dots + e$
 - Hence, the index weights are: $\mathbf{b}' = (1 \quad x_s \quad x_s^2 \quad \dots)$
 - I ignore e here, as it does not correlate with \mathbf{g}
 - Response is in the average genetic coefficients of the reaction norm: $\Delta \bar{\mathbf{g}}' = (\Delta \bar{g}_0 \quad \Delta \bar{g}_1 \quad \Delta \bar{g}_2 \quad \dots)$

$$\Delta \bar{\mathbf{g}} = \begin{bmatrix} \Delta \bar{g}_0 \\ \Delta \bar{g}_1 \\ \vdots \\ \Delta \bar{g}_n \end{bmatrix} = \mathbf{G}' \mathbf{b} \frac{i}{\sigma_I} \quad \text{where } \sigma_I = \sigma_{P(x_s)} \text{ the phenotypic standard deviation in the selection environment}$$

Hence, selection on a reaction norm trait can be interpreted as index selection, the weights (\mathbf{b}) being the powers of the environmental variable (x)

Reaction norm model: response to sel.

- Response to selection

$$\Delta \bar{\mathbf{g}} = \begin{bmatrix} \Delta \bar{g}_0 \\ \Delta \bar{g}_1 \\ \vdots \\ \Delta \bar{g}_n \end{bmatrix} = \mathbf{G}' \mathbf{b} \frac{i}{\sigma_I} \quad \text{with} \quad \mathbf{G} = \text{cov}(\mathbf{g}, \mathbf{g}) = \text{var}(\mathbf{g}) = \begin{bmatrix} \sigma_{g_0}^2 & \sigma_{g_0, g_1} & \dots & \sigma_{g_0, g_n} \\ \sigma_{g_0, g_1} & \sigma_{g_1}^2 & & \vdots \\ \vdots & & \ddots & \vdots \\ \sigma_{g_0, g_n} & \dots & \dots & \sigma_{g_n}^2 \end{bmatrix}$$

$$\text{and } \mathbf{b} = \begin{bmatrix} 1 \\ x_s \\ x_s^2 \\ \vdots \\ x_s^n \end{bmatrix} \quad \text{where } x_s \text{ is the selection environment.}$$

Reaction norm: Example



- Linear reaction norm for milk yield, mass selection
 - Environmental variable (x) = mean milk yield in an environment
 - E.g. HYS = Finlay Wilkinson-regression
 - $g(x) = g_0 + g_1 x$
 - $\text{Var}(g_0) = 550^2$; $\text{Var}(g_1) = 0.05$
 - Meta-population mean milk yield = 7000kg
 - Genetic correlation **mean** milk yield and slope = -0.2
- Q: How does the mean RN before selection look like?
- where to put the intercept?
 - what is the mean slope?

Reaction norm: Example



- Linear reaction norm for milk yield, mass selection
 - Environmental variable = mean milk yield in an environment
 - E.g. HYS = Finlay Wilkinson-regression
 - $g(x) = g_0 + g_1 x$
 - $\text{Var}(g_0) = 550^2$; $\text{Var}(g_1) = 0.05$, $\text{Corr}(g_0, g_1) = -0.2$
 - Meta-population mean milk yield = 7000kg
 - Two environments
 - Selection environment 9000kg,

$$I = P(x_s) = g_0 + g_1(9000 - 7000) + e = (1 \quad 2000) \begin{pmatrix} g_0 \\ g_1 \end{pmatrix} + e$$
 - $\mathbf{b}' = (1, 2000)$
 - $\text{Var}(P_s) = 1308^2 \text{ kg}^2$ (assumed given here)
 - Selected proportion = 20% $\rightarrow i = 1.4$
 - Production environment 8000kg
 - $x_{\text{off}}' = (1, 1000)$

Reaction norm: Example

$$\Delta \bar{\mathbf{g}} = \begin{bmatrix} \Delta \bar{g}_0 \\ \Delta \bar{g}_1 \\ \vdots \\ \Delta \bar{g}_n \end{bmatrix} = \mathbf{G}' \mathbf{b} \frac{i}{\sigma_I}$$

$$\Delta \bar{\mathbf{g}} = \begin{bmatrix} \Delta \bar{g}_0 \\ \Delta \bar{g}_1 \end{bmatrix} = \begin{bmatrix} \sigma_{g_0}^2 & \sigma_{g_0 g_1} \\ \sigma_{g_0 g_1} & \sigma_{g_1}^2 \end{bmatrix} \begin{bmatrix} 1 \\ x_s \end{bmatrix} \frac{i}{\sigma_{P, x_s}}$$

$$\begin{bmatrix} \Delta \bar{g}_0 \\ \Delta \bar{g}_1 \end{bmatrix} = \begin{bmatrix} 550^2 & -24.5 \\ -24.5 & 0.05 \end{bmatrix} \begin{bmatrix} 1 \\ 2000 \end{bmatrix} \frac{1.4}{1308} = \begin{bmatrix} +271 \\ +0.081 \end{bmatrix}$$

$$\Delta g(x_{off}) = \Delta \bar{\mathbf{g}}' \mathbf{x}_{off} = [271 \quad 0.081] \begin{bmatrix} 1 \\ 1000 \end{bmatrix} = 352 \text{kg}$$

$$\Delta g(x_s) = \Delta \bar{\mathbf{g}}' \mathbf{x}_s = [271 \quad 0.081] \begin{bmatrix} 1 \\ 2000 \end{bmatrix} = 433 \text{kg}$$

- An increase in level (+271)
- An increase in slope (+0.081)

More response in better environments
 Increased sensitivity
 Despite $r_{g_{01}} = -0.2$



Assignment 2: Selection for milk yield when GxE is modelled as a linear RN

- See word document: "Single trait selection on a linear reaction norm trait.docx"
- See R-code: "univariate reaction norm milk yield.R"
- Investigate:
 - Consequences of selection for environmental sensitivity
 - Effect of the selection environment