

# Quantitative genetics in structured populations

## Goals:

- 1) Estimate the amount of genetic and environmental variation in the population.
- 2) Estimate the genetic and environmental effects on an individual.
- 3) Predict the phenotype of individuals using genome, pedigree, and environment.

## Intermediate problems:

- 1) Genetic decompositions.
- 2) Estimating covariance matrices.

# Conventions

$g_n, g(n)$ : Scalars are italicized, non-bold, lower case letters with a subscript.

$\mathbf{g}$  vector are bold lower-case letters.

$\mathbf{V}_g$  matrices are bold upper-case letters, typically with a subscript.

$\langle \mathbf{g} \rangle$  denotes an expectation of the vector  $\mathbf{g}$ .

$\langle \beta, \delta \rangle$  denotes the inner product of two vectors.

$\sigma_\beta$  denotes the norm of a vector.

# Terms

$z$ : Phenotypic deviation.

$g$ : Genetic deviation.

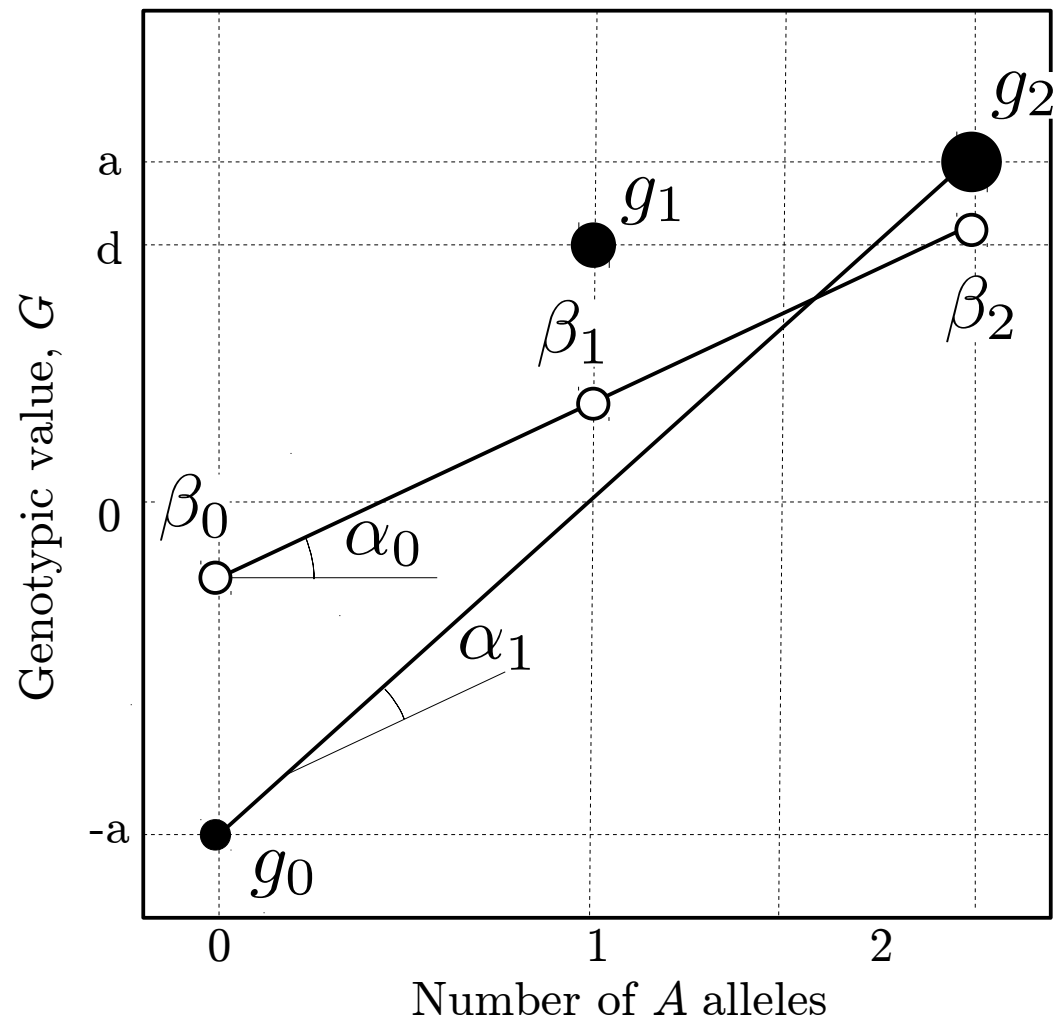
$\beta$ : Additive deviation.

$\delta$ : Random mating dominance deviation

$\iota$ : Inbred dominance deviation

$e$ : Environmental deviation.

$$\mathbf{z} = \mathbf{g} + \mathbf{e} = \boldsymbol{\beta} + \boldsymbol{\delta} + \boldsymbol{\iota} + \dots + \mathbf{e}$$



# The genetic deviations of an individual's phenotype at a locus.

Gene Cont. $n$	Gene dev. $g(n)$	$\bar{f}$	$\text{Pr}(n)$	Additive dev., $\beta(n)$	Dominance Dev., $\delta(n)$	Dom. Dev. Inbred $\iota(n)$
0	$-a - \mu$	0	$q^2$	$-2\alpha p$	$-2dp^2$	0
1	$d - \mu$	0	$2pq$	$\alpha(q - p)$	$2dpq$	0
2	$a - \mu$	0	$p^2$	$2\alpha q$	$-2dq^2$	0
0	$-a - \mu$	1	$q$	$-2\alpha p$	0	$-2dp(q - p)$
1	$d - \mu$	1	0	$\alpha(q - p)$	$d$	$d(q - p)^2$
2	$a - \mu$	1	$p$	$2\alpha q$	0	$2dq(q - p)$
0	$-a - \mu$	$-q/p$	0	$-2\alpha p$	$-2d$	$-2dp(q - p)$
1	$d - \mu$	$-q/p$	$2(pq + q^2)$	$\alpha(q - p)$	0	$d(q - p)^2$
2	$a - \mu$	$-q/p$	$p^2 - q^2$	$2\alpha q$	0	$2dq(q - p)$

$$\mu = a(p - q) + 2pqd(1 - \bar{f})$$

# Constructing the genetic deviations

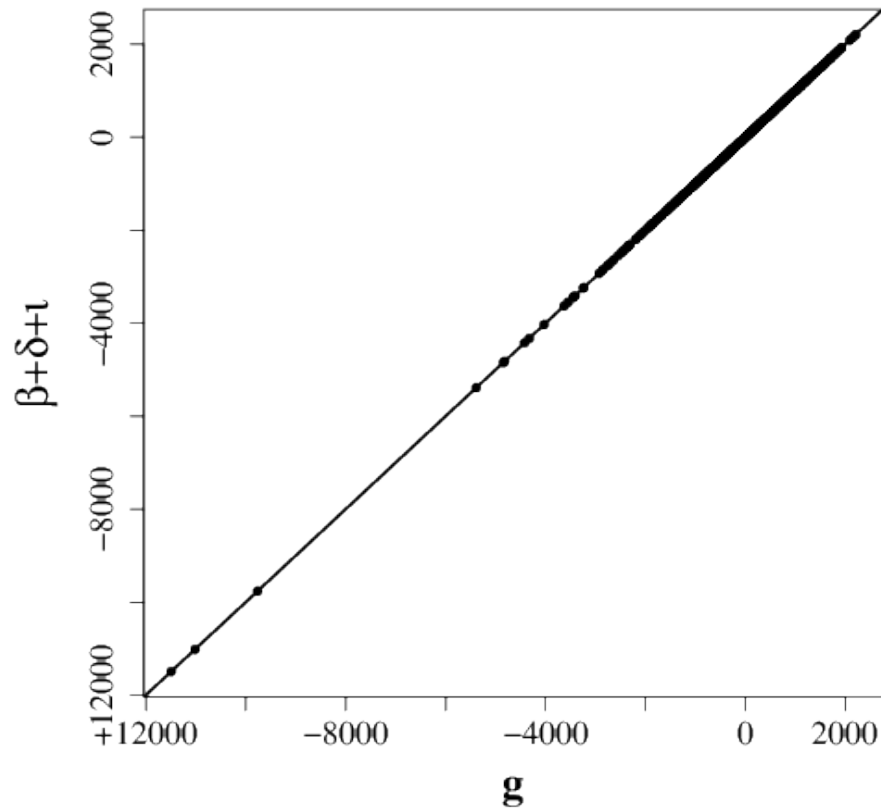
Gene Content $n$	Freq. (sample)	Dominance Deviation, $\delta(n)$	Inbred Dom. dev, $\iota(n)$
0	$q^2 + pq\bar{f}$	$\frac{-2dp^2}{c} \frac{(q-f_i q)}{(q+f_i p)}$	$\frac{-4p\alpha_1 f_i}{1+f_i} + \Delta\mu$
1	$2pq(1 - \bar{f})$	$\frac{2dpq}{c}$	$\frac{2(q-p)\alpha_1 f_i}{1+f_i} + \Delta\mu$
2	$p^2 + pq\bar{f}$	$\frac{-2dq^2}{c} \frac{(p-f_i p)}{(p+f_i q)}$	$\frac{4q\alpha_1 f_i}{1+f_i} + \Delta\mu$

$$\mathbf{n}, \mathbf{f} \rightarrow \boldsymbol{\beta}, \boldsymbol{\delta}, \boldsymbol{\iota}$$

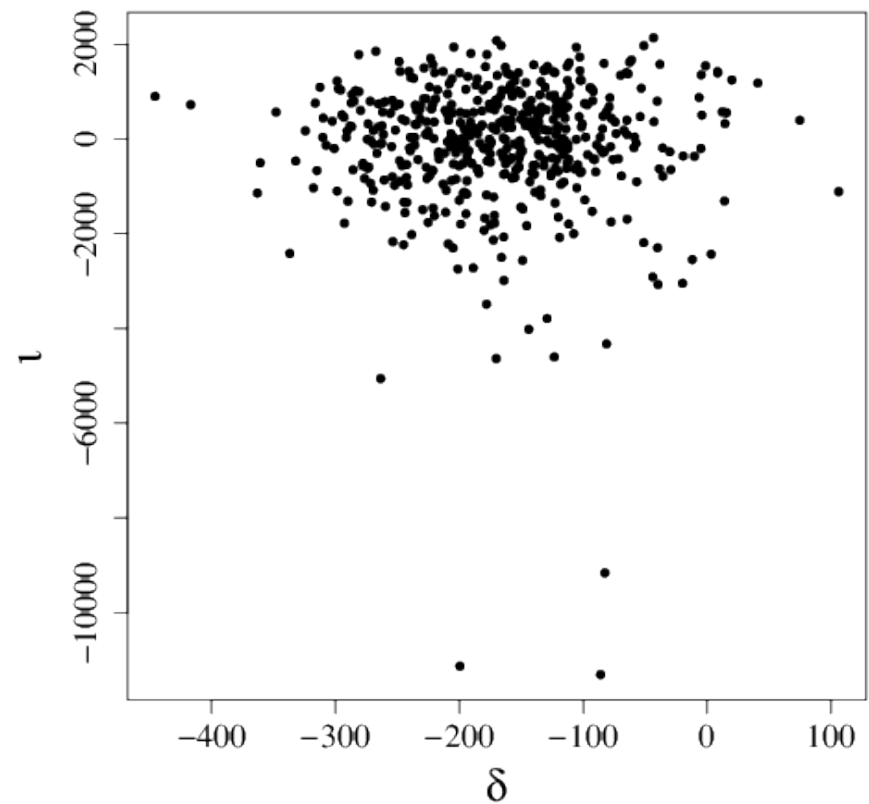
$$c = \frac{(1 + f_i)pq}{(f_i p + q)(f_i q + p)}$$

$$\Delta\mu = 2pqd(f - f_i)$$

Components sum to one; inbreeding and dom. are almost orthogonal.

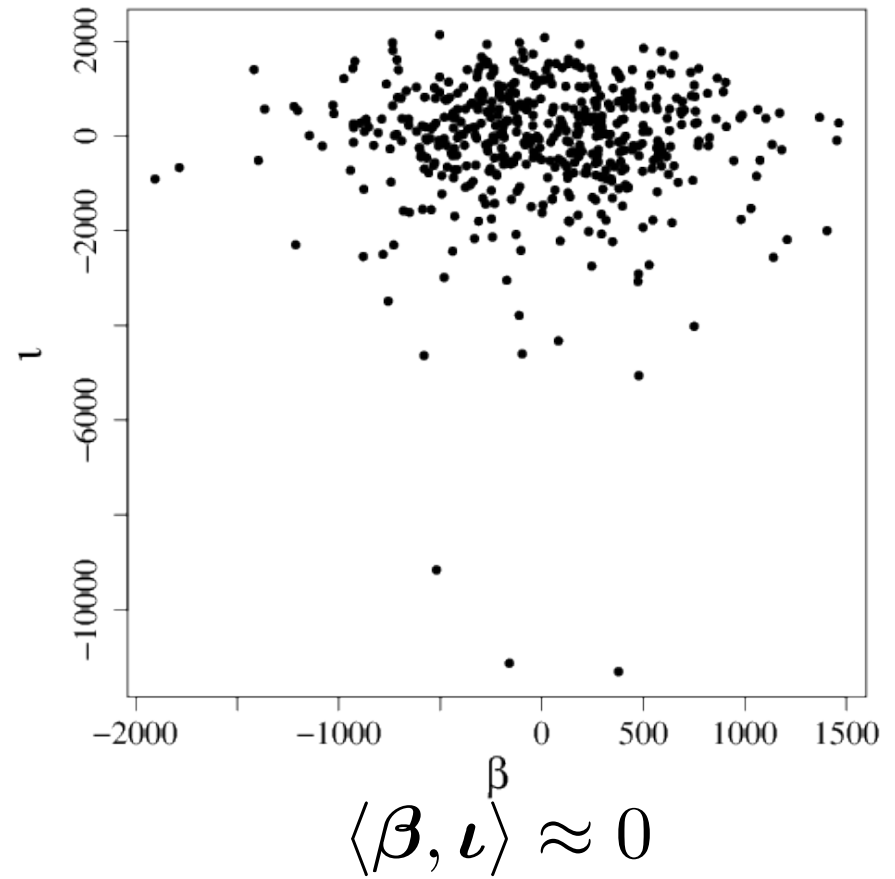
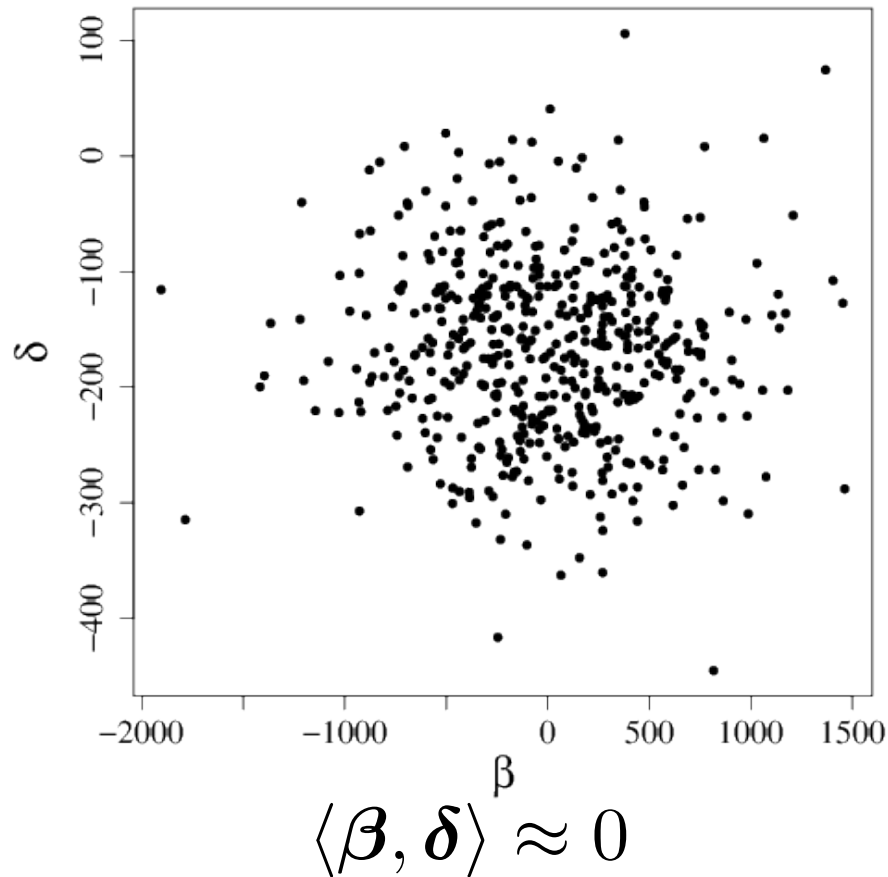


$$g = \beta + \delta + \iota$$



$$\langle \iota, \delta \rangle \approx 0$$

Additive dev. is nearly orthogonal to  
dominance dev. and inbred dev.

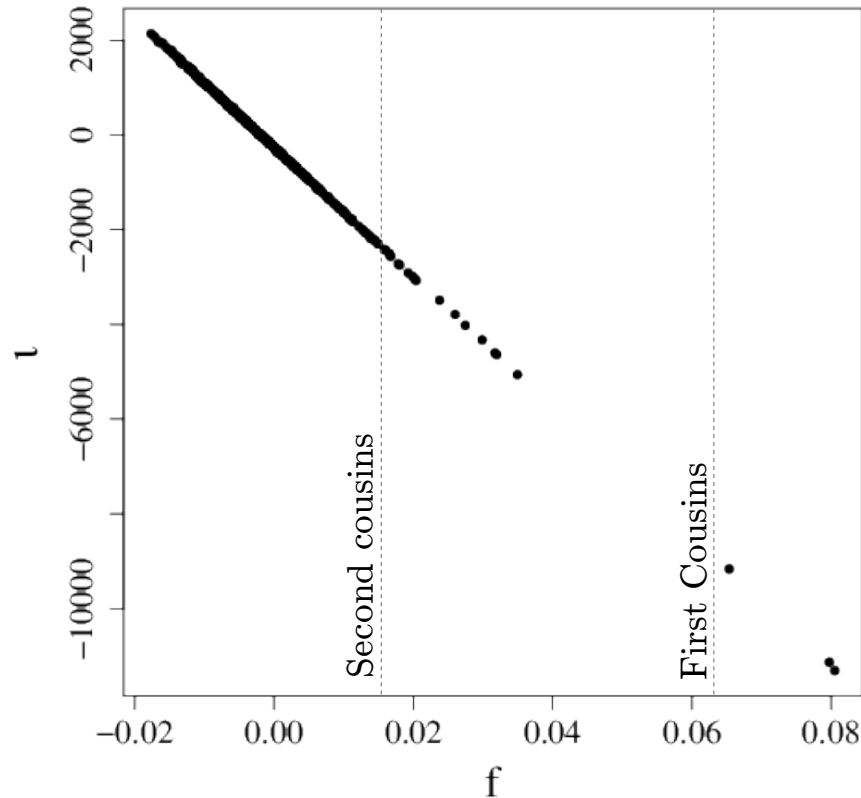




Inbred dom. dev. are a projection of dominance dev. onto inbreeding.

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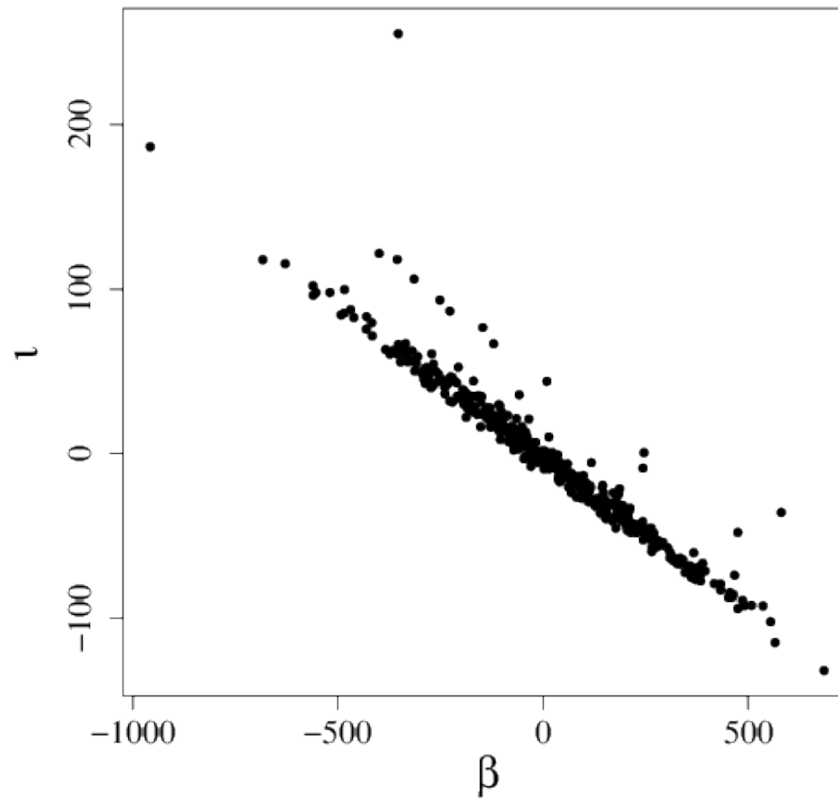
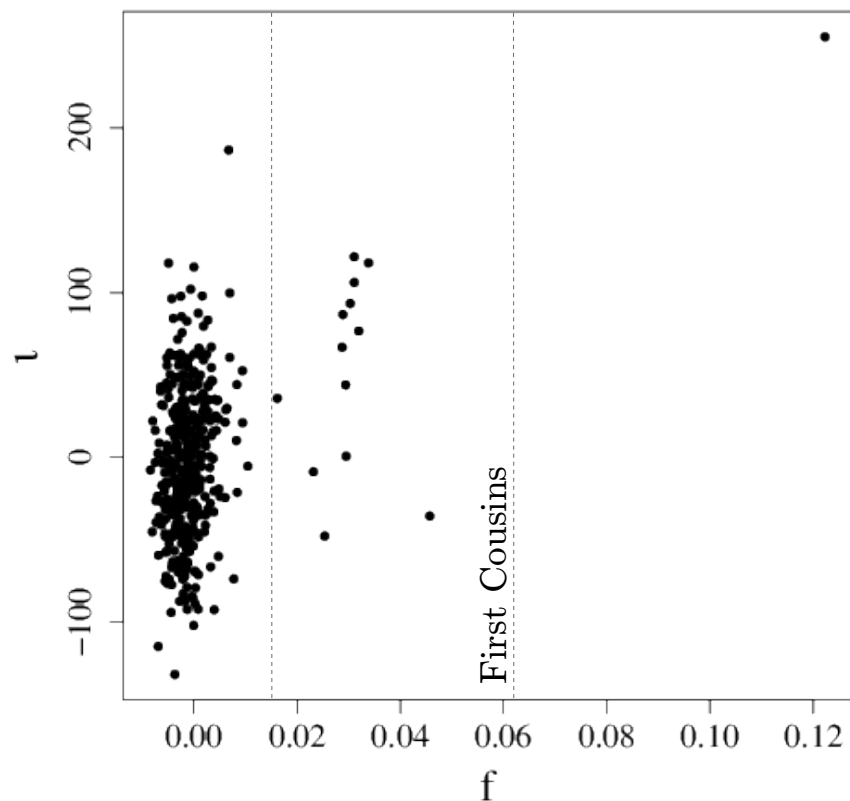
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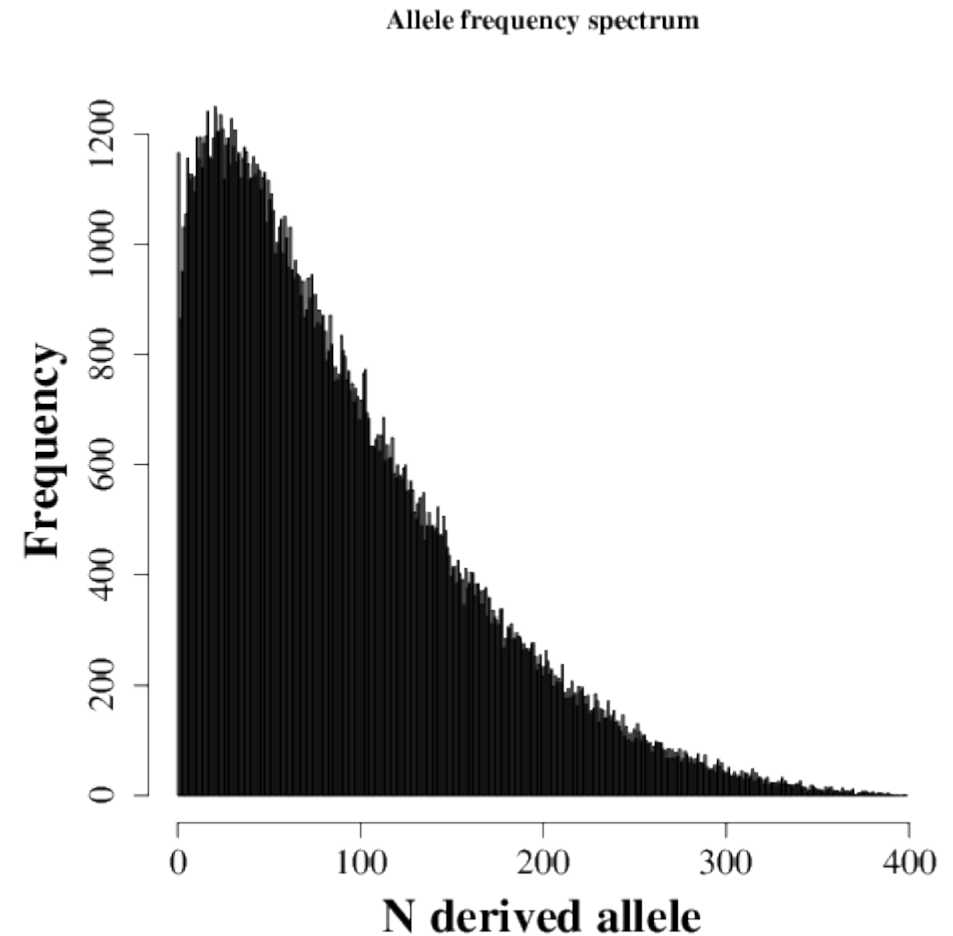
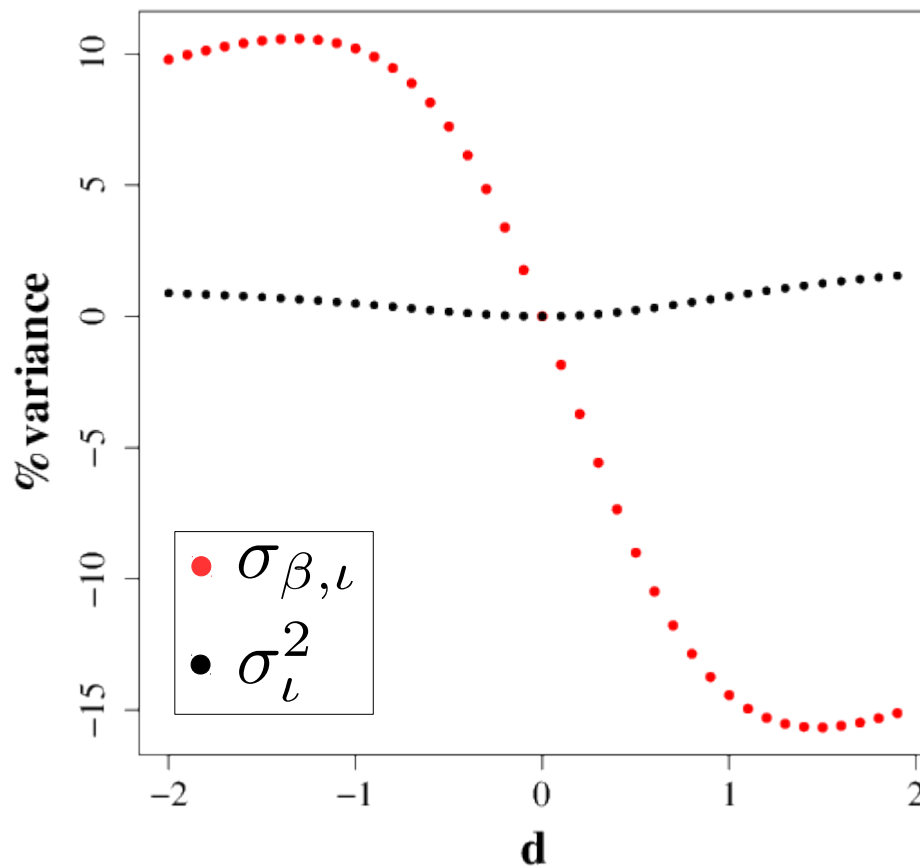
$$\frac{\langle \delta', \mathbf{f} \rangle}{\sigma_f^2} \mathbf{f} = \iota$$

$$\delta' - \iota = \delta$$

# Not quite right....

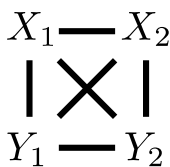


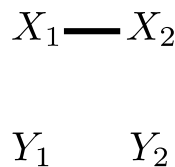
# Inbred dominance deviation in a randomly mating population

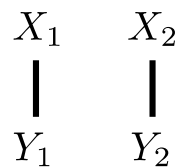


# Covariance Matrices

$$V_z(x, y) = 2\sigma_\beta^2 V_\beta(x, y) + \sigma_\delta^2 V_\delta(x, y) + \sigma_{\beta\iota} (V_{\beta\iota}(x, y) + V_{\iota\beta}(x, y)) + \sigma_\iota^2 V_\iota(x, y)$$

$$V_\beta(x, y) = \Delta_1 + \frac{\Delta_3 + \Delta_5 + \Delta_7}{2} + \frac{\Delta_8}{4} \Delta_1$$


$$\Delta_4$$


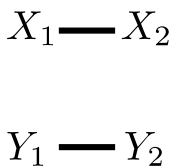
$$\Delta_7$$


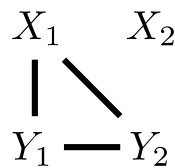
$$V_\delta(x, y) = \Delta_7 + \Delta_2$$

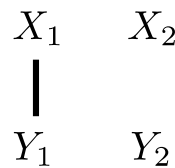
$$V_{\beta\iota}(x, y) = \Delta_1 + \Delta_3$$

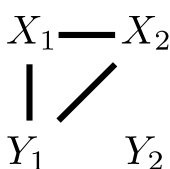
$$V_{\iota\beta}(x, y) = \Delta_1 + \Delta_5$$

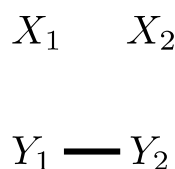
$$V_\iota(x, y) = \Delta_1$$

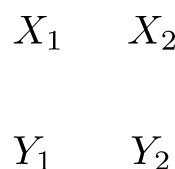
$$\Delta_2$$


$$\Delta_5$$


$$\Delta_8$$


$$\Delta_3$$


$$\Delta_6$$


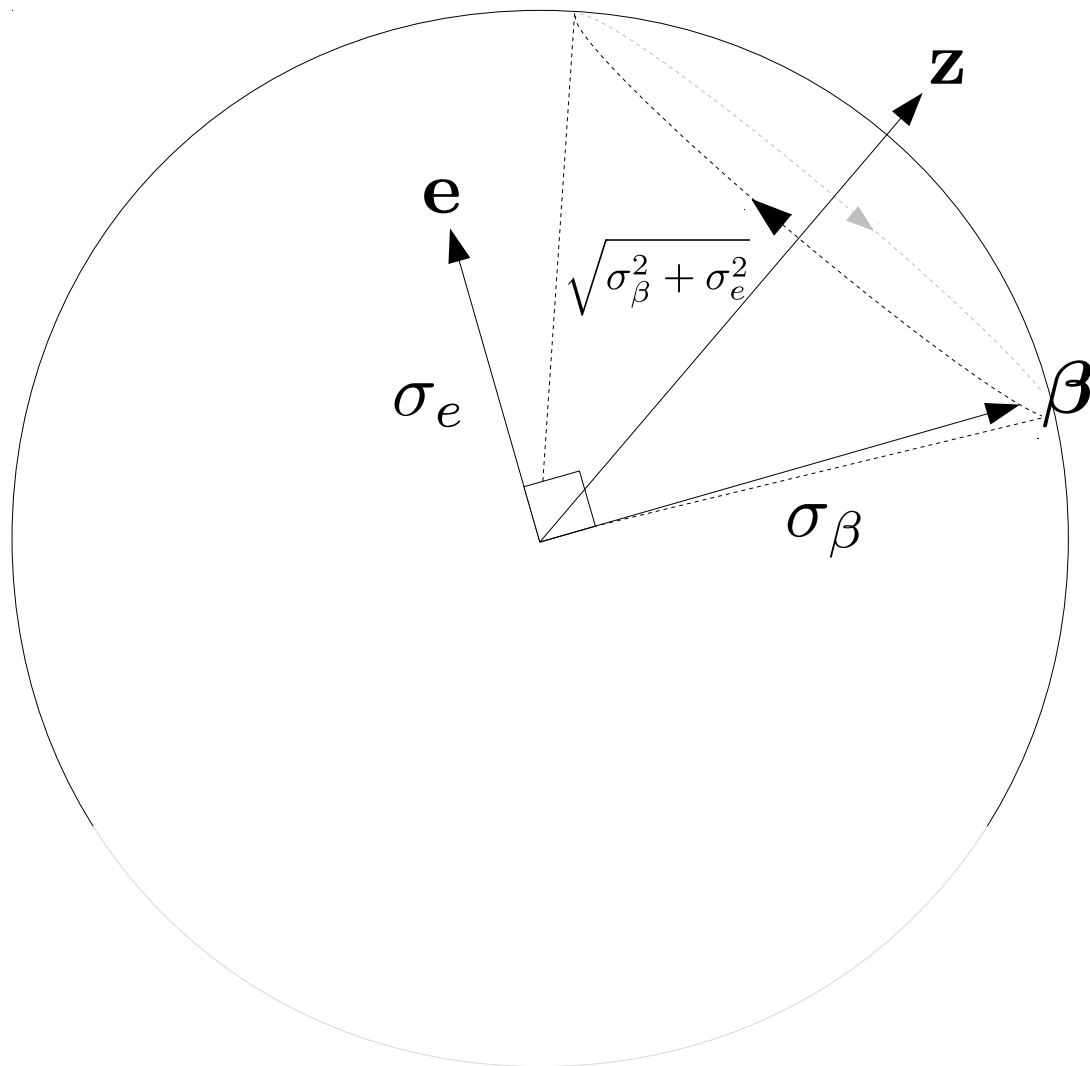
$$\Delta_9$$


Diagonal isn't variance???

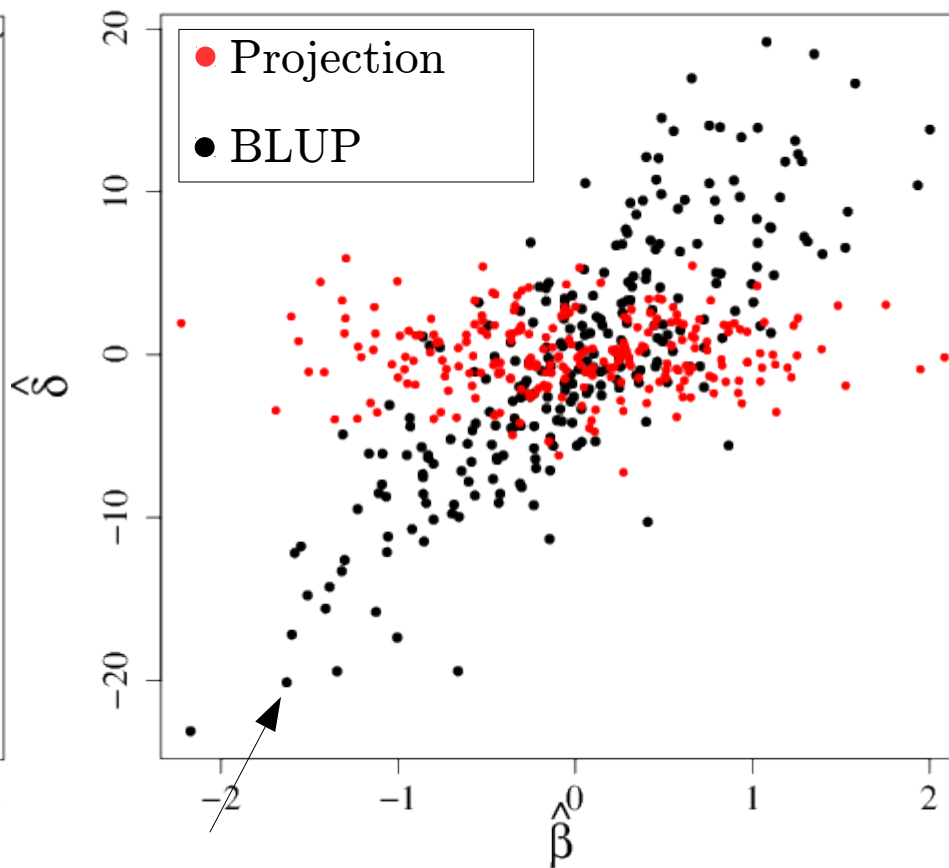
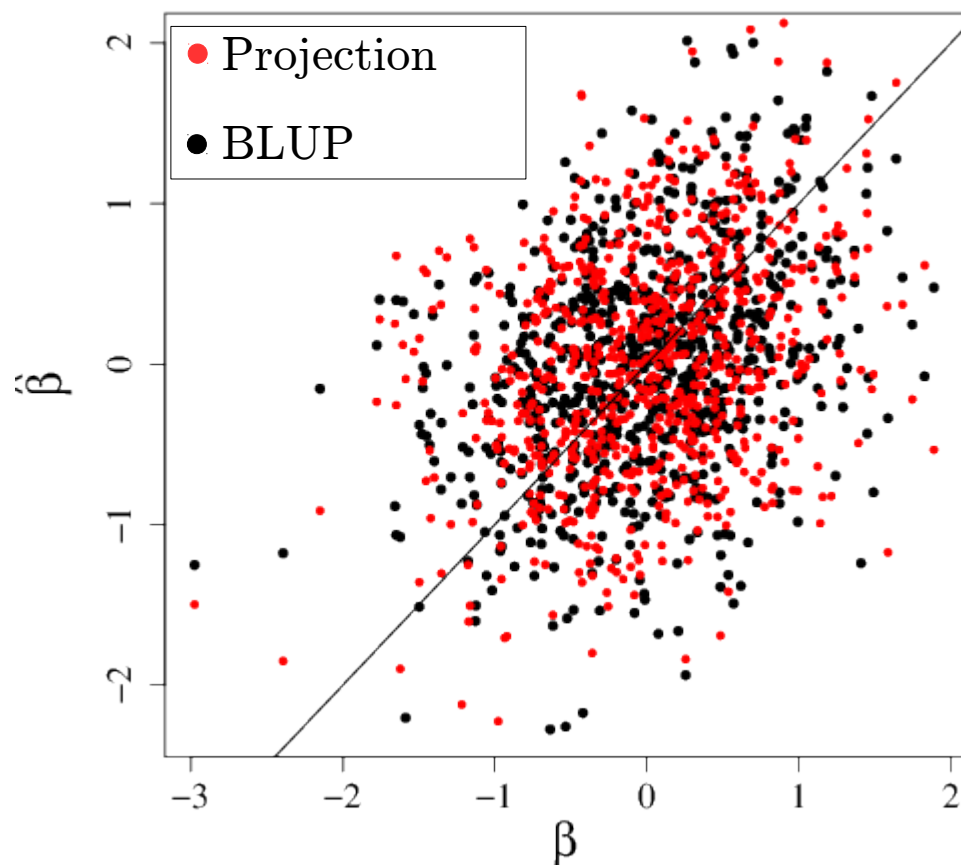
# How are deviations shared between individuals?

$\Pr(n_x, n_y)$	$\Sigma$	$p_x$	$p_y$	$f_x$	$f_y$	$V_\beta(x, y)$
$\Pr(0, 0)$	1	0	0	$\frac{1}{2(1-p_x)}$	$\frac{1}{2(1-p_y)}$	
$\Pr(1, 0)$	1	0.5	0	$-\frac{1}{2p_x(1-p_x)}$	$\frac{1}{2(1-p_y)}$	
$\Pr(2, 0)$	1	1	0	$\frac{1}{2p_x}$	$\frac{1}{2(1-p_y)}$	
$\Pr(0, 1)$	1	0	0.5	$\frac{1}{2(1-p_x)}$	$-\frac{1}{2p_y(1-p_y)}$	
$\Pr(1, 1)$	1	0.5	0.5	$-\frac{1}{2p_x(1-p_x)}$	$-\frac{1}{2p_y(1-p_y)}$	
$\Pr(2, 1)$	1	1	0.5	$\frac{1}{2p_x}$	$-\frac{1}{2p_y(1-p_y)}$	
$\Pr(0, 2)$	1	0	1	$\frac{1}{2(1-p_x)}$	$\frac{1}{2p_y}$	
$\Pr(1, 2)$	1	0.5	1	$-\frac{1}{2p_x(1-p_x)}$	$\frac{1}{2p_y}$	
$\Pr(2, 2)$	1	1	1	$\frac{1}{2p_x}$	$\frac{1}{2p_y}$	

# Slide outlining problem



# Similar performance.



BLUP variance way too high!?!?

# Predicting phenotypes from genotypic deviations.

$$\begin{pmatrix} \mathbf{v}_\beta \\ \mathbf{v}_\delta \\ \mathbf{v}_\iota \end{pmatrix}^T \begin{pmatrix} \lambda_\beta \mathbf{V}_\beta^{-1} & \mathbf{0} & \lambda_{\beta,\iota} \mathbf{V}_{\beta,\iota}^{-1} \\ \mathbf{0} & \lambda_\delta \mathbf{V}_\delta^{-1} & \mathbf{0} \\ \lambda_{\beta,\iota} \mathbf{V}_{\beta,\iota}^{-1} & \mathbf{0} & \lambda_\iota \mathbf{V}_\iota^{-1} \end{pmatrix} \begin{pmatrix} \hat{\beta} \\ \hat{\delta} \\ \hat{\iota} \end{pmatrix} = z_i$$



# TO DO LIST

Formally derive covariance matrices.

Estimate variances/covariances.

Get the BLUP right.

Implement model of DFE.

Implement population structure in simulations.

Check against GCTA in simulations.

Check against GCTA on human data.