

RESPONSES OF DIVERGENT SELECTION FOR LITTER SIZE RESIDUAL VARIANCE IN RABBITS

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Interests

Animal Production

Evolutionary Biology

- ❖ Uniformity
 - ❖ Heritability
 - ❖ Welfare
- ❖ Maintenance of residual variance along the time



Has the residual variance a genetic component?

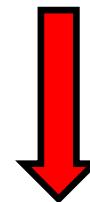
Indirect evidences

Data bases



Heteroscedastic models : $y_{ij} = \mu + a_i + (\sqrt{\exp(\mu^* + a_i^*)})\varepsilon_{ij}$

Estimates of $\sigma_{a^*}^2, h_{a^*}^2, \rho_{aa^*}$



PROBLEMS

Problems of heteroscedastics models

- ❖ Depend on a large number of parameters
- ❖ Are not robusts

Ye Yang *et al.*, 2011

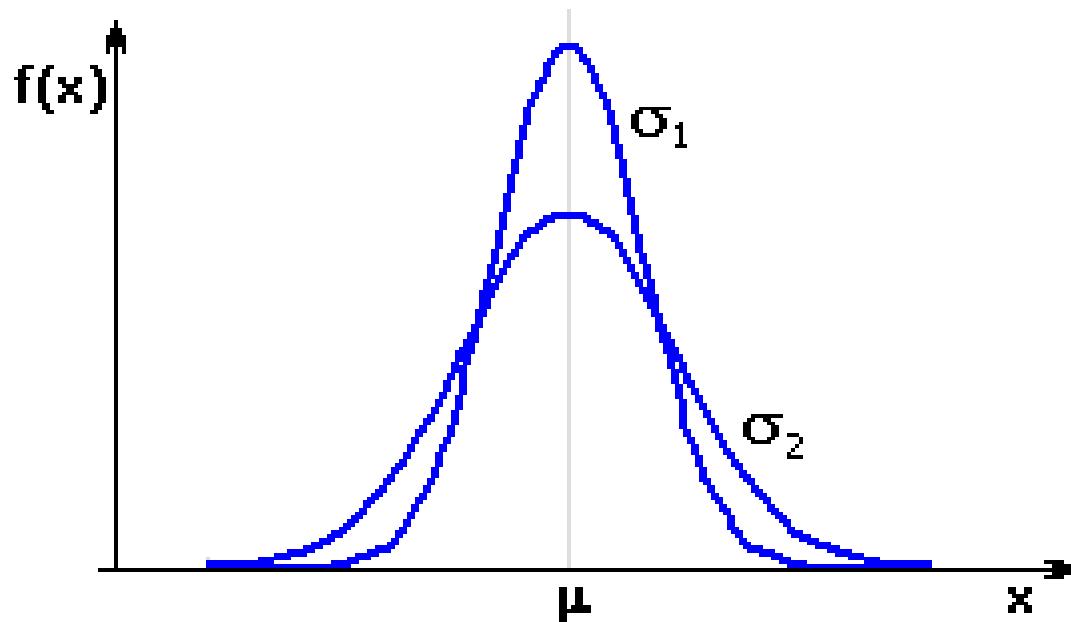
Data		$\hat{A} = \frac{\text{cov}(a, a^*)}{\sigma_a \sigma_{a^*}}$
Rabbits	untransformed	-0.73
	transformed	0.28
Pigs	untransformed	-0.64
	transformed	0.70

Residues Box-Cox transformation in litter size in rabbits



Objective

Estimate the response in a divergent selection experiment of residual variance of litter size (V_e) in rabbits.



Materials and Methods

Selection criterion



Residual variance of litter size (V_e)

**Estimated as
the variance
within doe**



Minimum quadratic risk estimator

Total number of parities of the doe

$$LS_{ij} = YS_i + L_j + e_{ij}$$

$$\frac{1}{n+1} \sum^n (x - \bar{x})^2$$

LS in V_r
 e in V_e

Materials and Methods

ANALYZED
TRAITS

Bayesian statistical
analysis

V_e

V_r

Sd_e

Sd_r

LS

V_e

V_r

Genetic effects
 $\boldsymbol{a} \sim N(\mathbf{0}, \mathbf{A}\sigma_a^2)$

Diagonal weight matrix
Element in the diagonal

$$\frac{2(n-1)}{(n+1)^2} \quad n: \text{num. of parities of each}$$

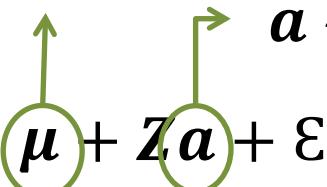
Materials and Methods

ANALYZED
TRAITS

Bayesian statistical
analysis

Sd_e
 Sd_r

Mean Genetic effects

$$Sd_e = \mu + Z\alpha + \varepsilon$$

$$\alpha \sim N(\mathbf{0}, A\sigma_a^2)$$

$$(Sd_e | \mu, \alpha, \sigma_\varepsilon^2) \sim N(\mu + Z\alpha, I\sigma_\varepsilon^2)$$

Materials and Methods

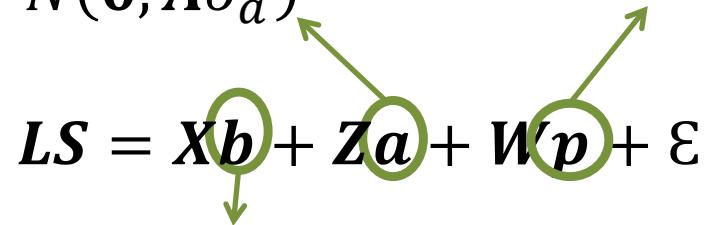
ANALYZED
TRAITS

Bayesian statistical
analysis

LS

Genetic effects
 $a \sim N(\mathbf{0}, A\sigma_a^2)$

Permanent effect
 $p \sim N(\mathbf{0}, I\sigma_p^2)$

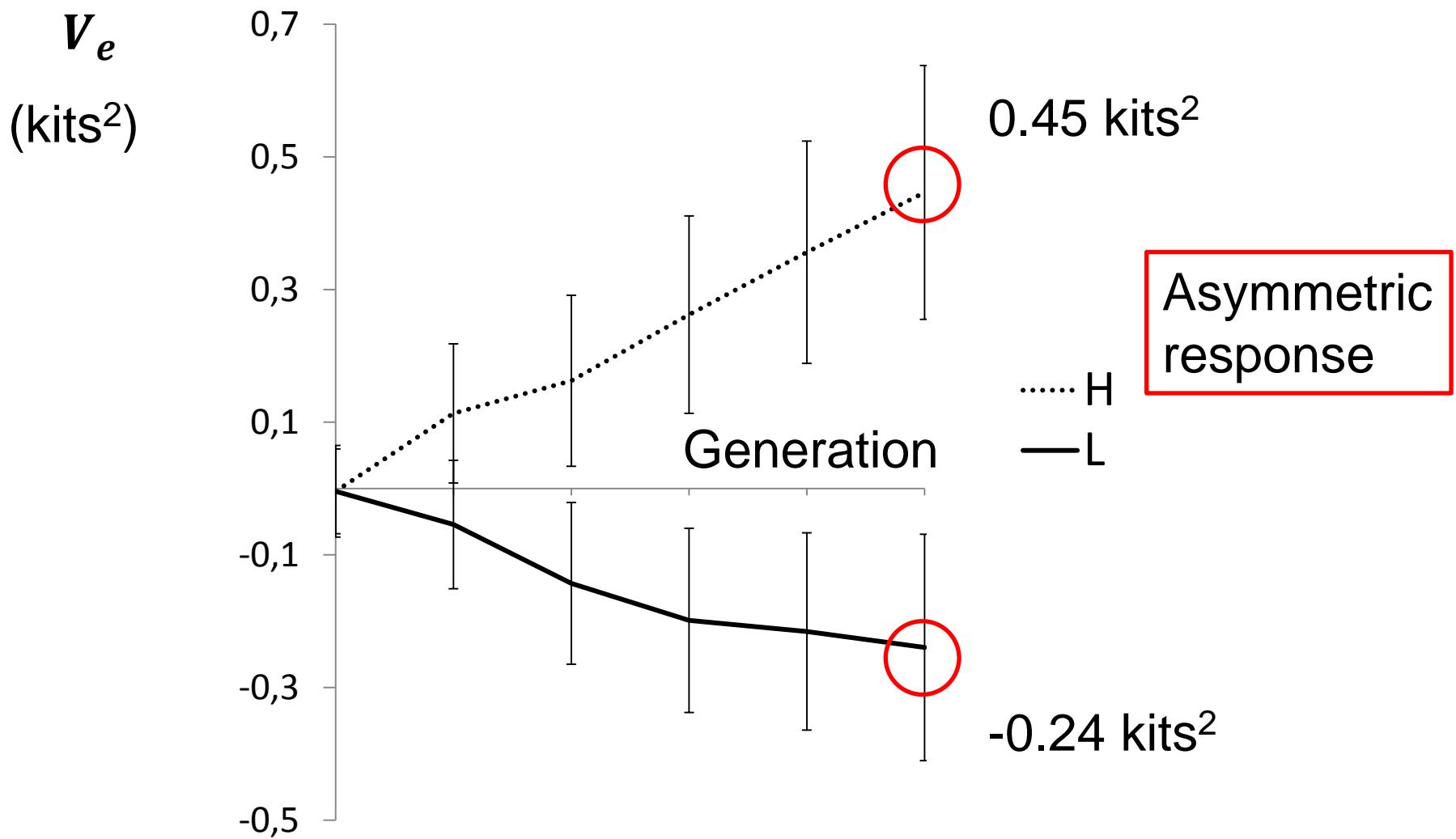
$$LS = Xb + Za + Wp + \varepsilon$$


Year-season and lactation status effects

$$(LS | b, a, p, \sigma_\varepsilon^2) \sim N(Xb + Za + Wp, I\sigma_\varepsilon^2)$$

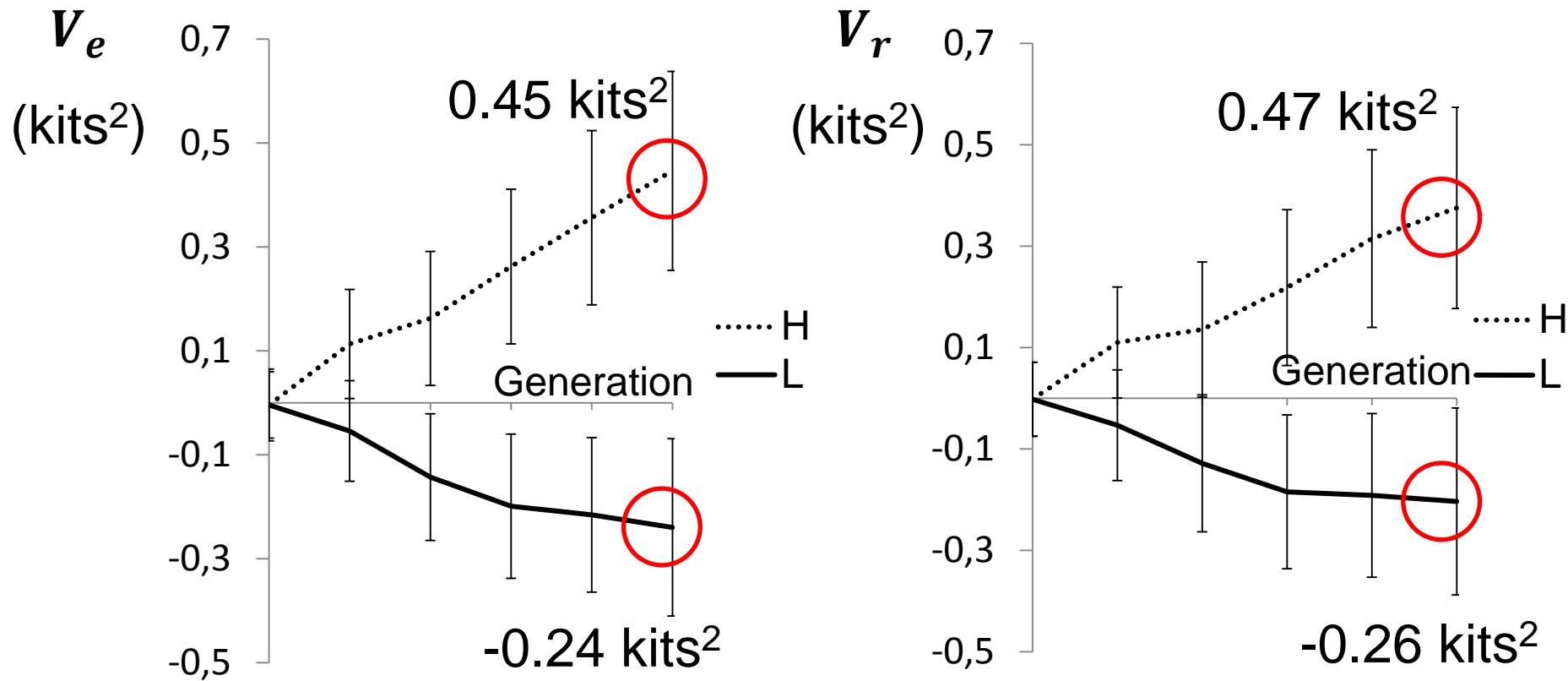
Results and Discussion

Response to selection



Results and Discussion

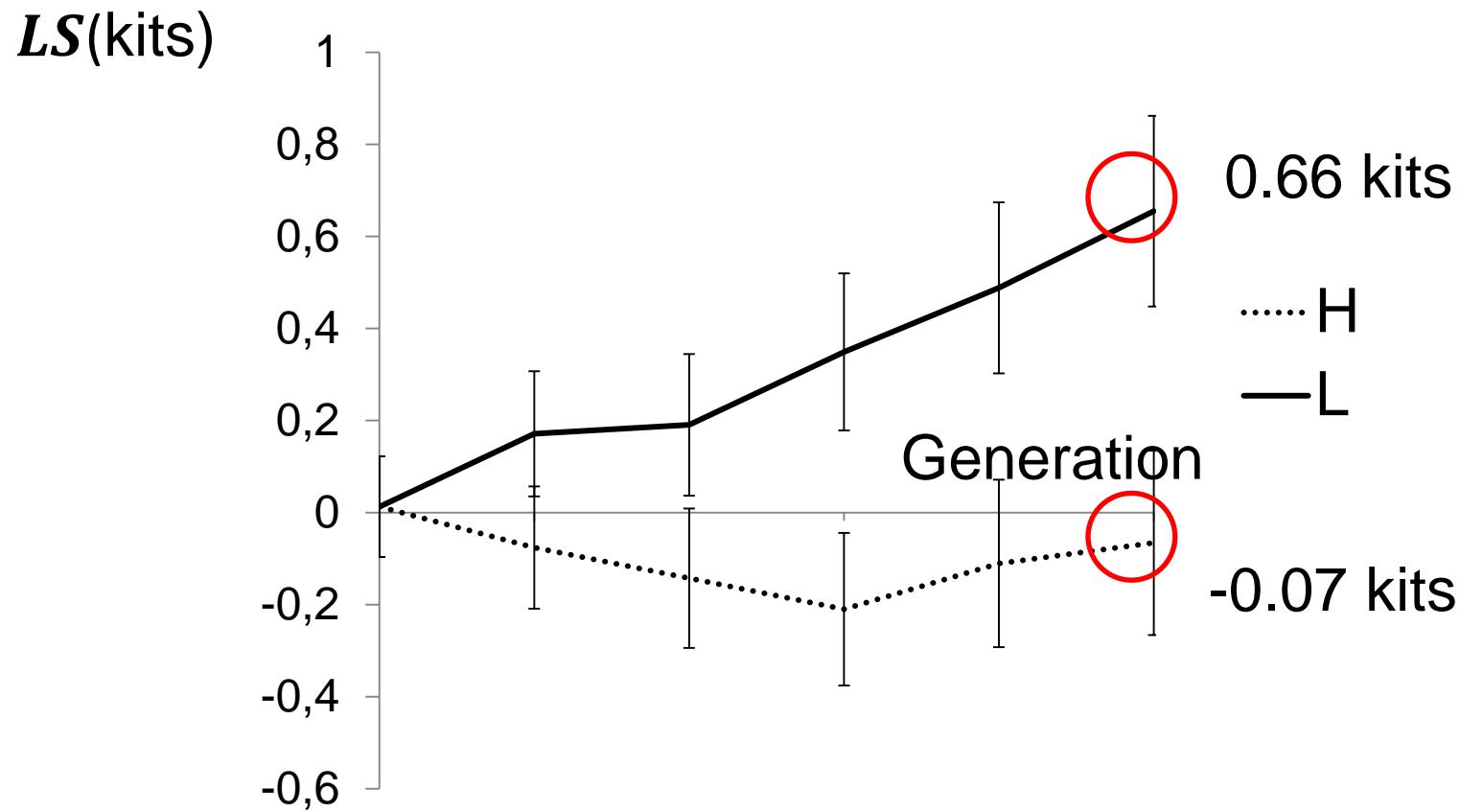
The effect of precorrecting LS data



Response is almost not affected by the data precorrection

Results and Discussion

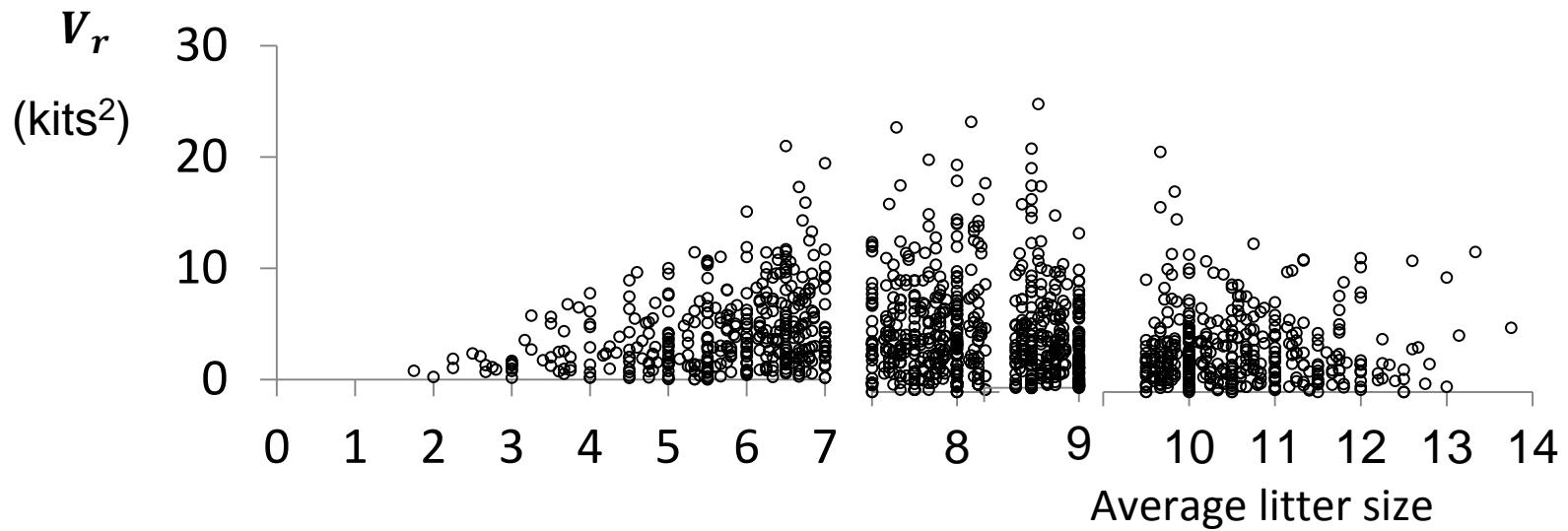
Correlated response



Negative correlated response when selecting by homogeneity of litter size

Results and Discussion

Data distribution by quartiles



Results and Discussion

Heritabilities of the traits

Trait	Median	CV
V_e	0.06	0.90
V_r	0.05	0.90
LS	0.12	0.36

Responses in V_e and V_r should be due to the high variability of the traits

Gutiérrez *et al.*, 2006, Damgaard *et al.* 2003

Blasco *et al.*, 1993

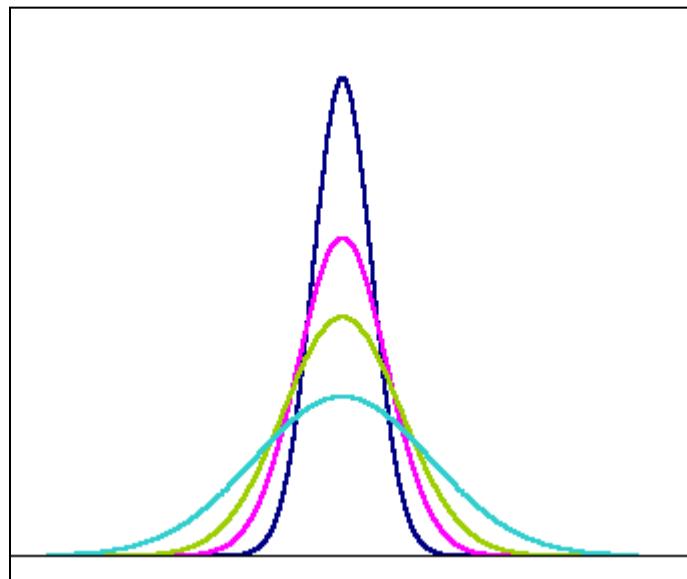
Conclusions

Asymmetric response for V_e .

Correlated negative response for litter size in homogeneous does.



Thank you for your attention



Materiales y métodos

- ❖ 5 generaciones de selección
- ❖ Cada línea : 125 @y 25

SELECCIÓN DIVERGENTE POR V_e

Materiales y métodos

Análisis estadístico
bayesiano

La varianza de un estimador de mínimo riesgo (Searle, 1982)

$$\begin{aligned} \text{var}(s_{n+1}^2) &= \text{var}\left(\frac{1}{n+1} \mathbf{X}' \mathbf{A} \mathbf{X}\right) \\ &= \frac{1}{(n+1)^2} (2\text{tr}(\mathbf{A} \mathbf{V} \mathbf{A} \mathbf{V}) + 4 * 1' * \mathbf{A} \mathbf{V} \mathbf{A} 1 \mu^2) \\ &= \frac{1}{(n+1)^2} 2\text{tr}(\mathbf{A} \mathbf{A}) \sigma^4 = \boxed{\frac{2(n-1)}{(n+1)^2} \sigma^4} \\ &\quad \downarrow \\ \text{tr}(\mathbf{A}) &= n \frac{n-1}{n} = n - 1 \end{aligned}$$

SELECCIÓN DIVERGENTE POR V_e

Asunción de normalidad del carácter V_e

Cualquier función, cuando n es suficientemente grande tiende a la normalidad

$$f(\theta|y) \quad f(y|\theta)f(\theta) \rightarrow \prod_1^n f(y|\theta) = f(y|\theta)^n$$

$$\ln f(y|\theta)^n = n \ln f(y|\theta)$$

Desarrollando por las series de Taylor:

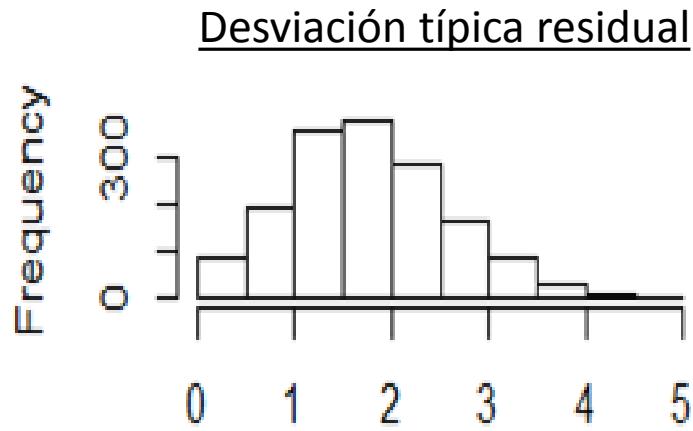
$$n \ln f(y|\theta) = k_0 + \left(\frac{d(n \ln f(y|\theta))}{d\theta} \right)_{\theta_0} (\theta - \theta_0) + \left(\frac{d^2 (n \ln f(y|\theta))}{d\theta^2} \right)_{\theta_0} (\theta - \theta_0)^2$$

$$f(y|\theta) = e^{\ln f(y|\theta)^n} = e^{n \ln f(y|\theta)} \propto e^{k(\theta - \theta_0)^2} \propto ke^{k^2}$$

Función normal

$$f(\theta|y) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(y-\mu)^2}$$

Distribución de los datos



Respuestas fenotípicas

