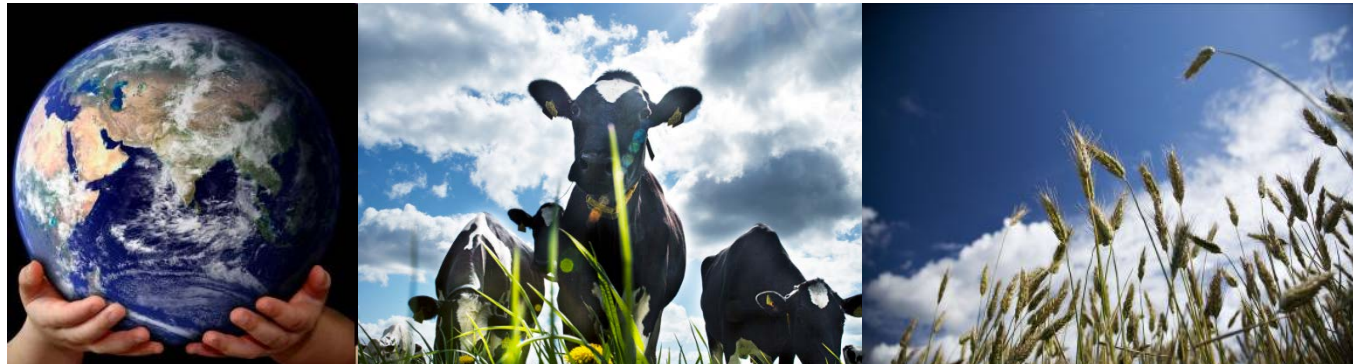
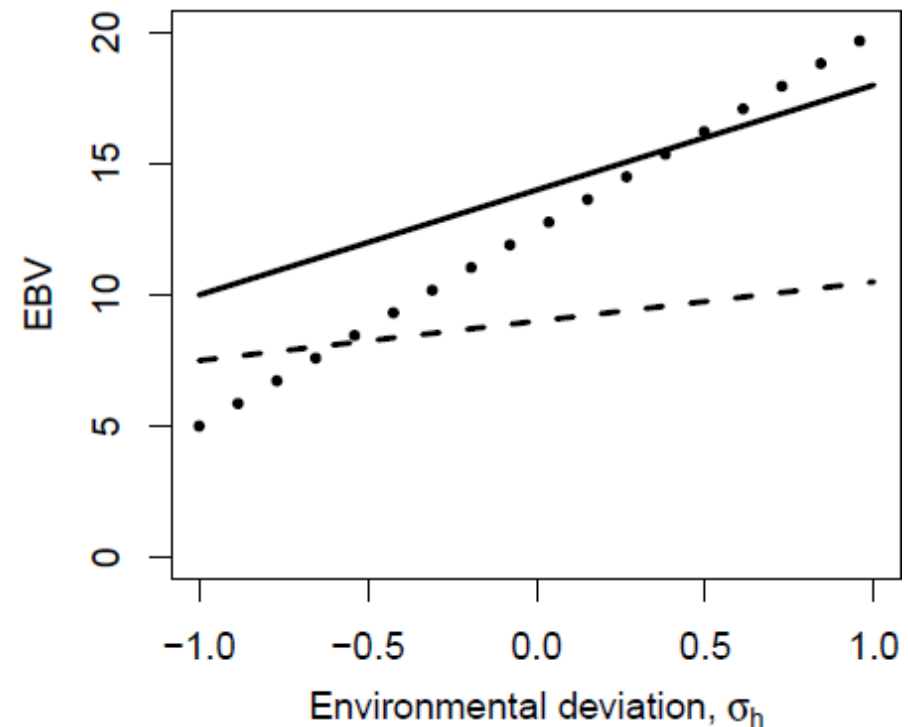
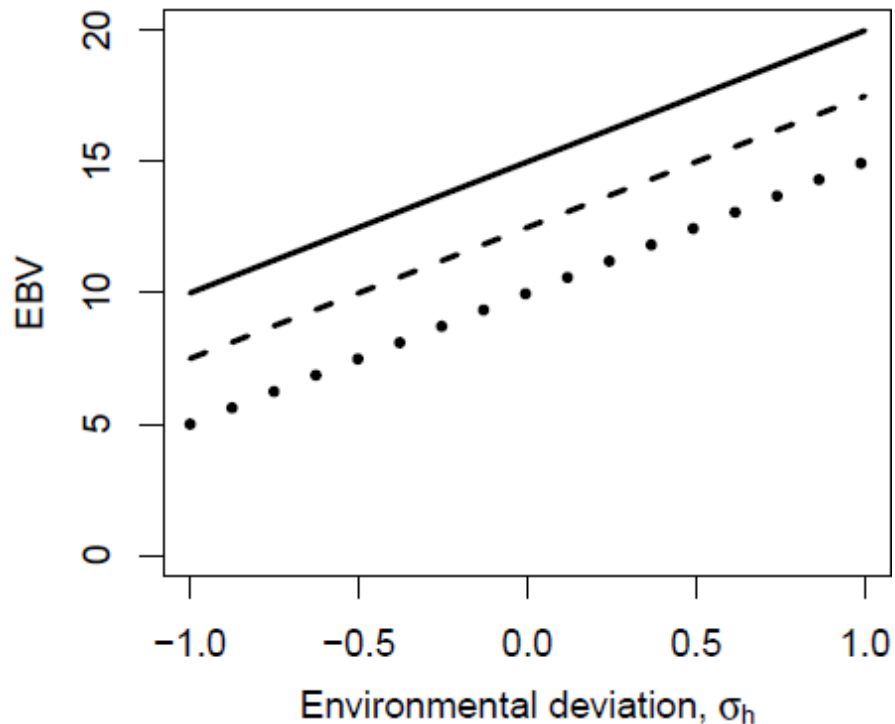


Adjusting for macro-environmental sensitivity in growth rate



Macro-ES \neq $G \times E$

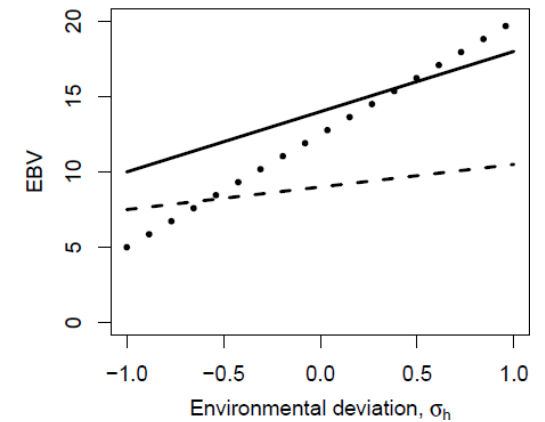
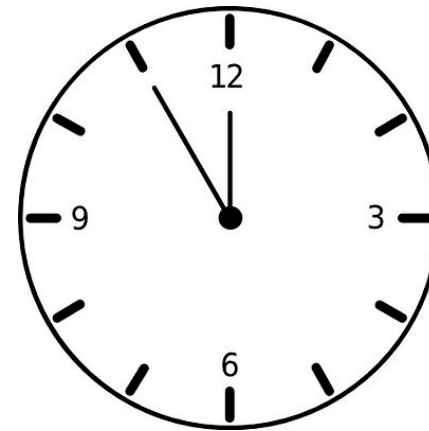
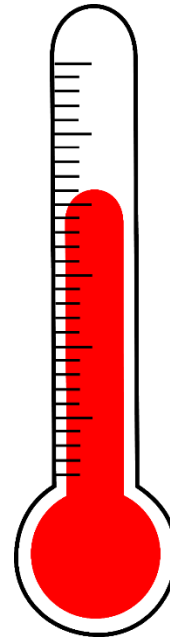
- Macro-ES \rightarrow change in EBV across environments
- $G \times E$ \rightarrow differences in change between genotypes



Statistical approaches

- Character state models
 - Discrete environments

- Reaction norm models
 - Continuous environments



The problem of knowing the environment

- Phenotypic means as covariate
- Pre-estimated covariates
- Iteratively updated covariate

Bayesian analysis of the linear reaction norm model with unknown covariates

G. Su, P. Madsen, M. S. Lund, D. Sorensen, I. R. Korsgaard and J. Jensen

2006

Macro-environmental sensitivity for growth rate in Danish Duroc pigs is under genetic control¹

Mette D. Madsen,^{*,2} Per Madsen,^{*} Bjarne Nielsen,[†] Torsten N. Kristensen,^{‡,§} Just Jensen,^{*} and Mahmoud Shirali^{*,⊕}

2018



Data



- Phenotype: ADG (g/day)
 - Sex specific trait, $r_g=0.88$ (Nielsen et al., 2018)

	Boars	Gilts
#Records	32,297	42,724
Mean (g/day)	1,184	1,117
SD (g/day)	123	108
#Herds	16	19

		Boars			Gilts		
		HYM	Group	Litter	HYM	Group	Litter
#Levels		1212	3398	13802	1280	4200	14973
#Animals per level	Mean (SD)	27 (16)	10 (2)	2 (1)	33 (21)	10 (2)	3 (2)
	Min	2	1	1	3	1	1
	Max	103	15	10	139	15	14

Model

- Univariate reaction norm model

$$y = \mathbf{Xb} + \mathbf{Za}_0 + \mathbf{Ha}^* + \mathbf{Wh} + \mathbf{Vp} + \mathbf{Ll} + \mathbf{e}$$

y: Response variable - ADG

b: Fixed parameters

a₀: Intercept of additive genetic variance

a^{*}: Slope of additive genetic variance

h: Herd-Year-Month on test (HYM)

p: Group

l: Litter

e: Residual variance

X, Z, W, V and **L**: Design matrices

H: Design matrix with environmental covariate

$$\begin{bmatrix} \mathbf{a}_0 \\ \mathbf{a}^* \end{bmatrix} \sim N \left(\mathbf{0}, \mathbf{A} \otimes \begin{bmatrix} \sigma_{a_0}^2 & \sigma_{a_0, a^*} \\ \sigma_{a_0, a^*} & \sigma_{a^*}^2 \end{bmatrix} \right)$$

$$\mathbf{h} \sim N(\mathbf{0}, \mathbf{I}\sigma_h^2)$$

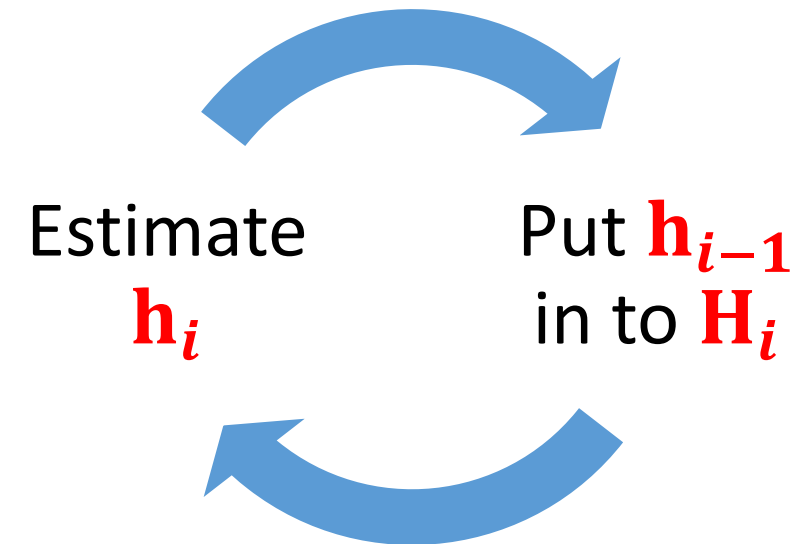
$$\mathbf{p} \sim N(\mathbf{0}, \mathbf{I}\sigma_p^2)$$

$$\mathbf{l} \sim N(\mathbf{0}, \mathbf{I}\sigma_l^2)$$

$$\begin{bmatrix} \mathbf{e}_1 \\ \vdots \\ \mathbf{e}_m \end{bmatrix} \sim N \left(\mathbf{0}, \begin{bmatrix} \mathbf{I}\sigma_{e_1}^2 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & \mathbf{I}\sigma_{e_m}^2 \end{bmatrix} \right)$$

Approach

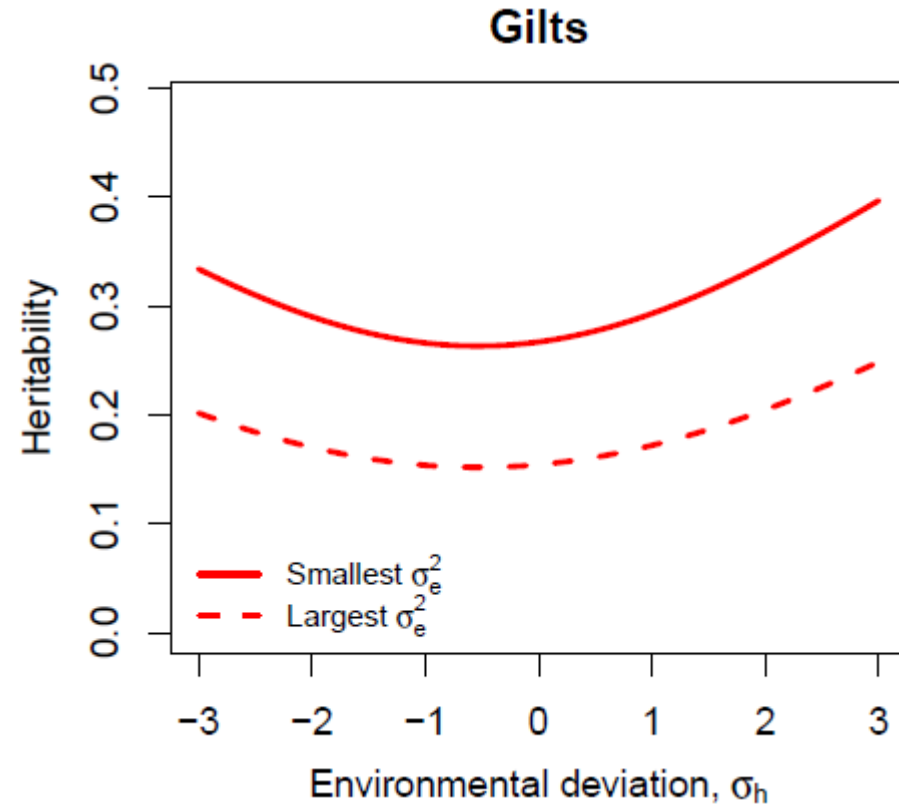
- $\mathbf{y} = \mathbf{Xb} + \mathbf{Za}_0 + \mathbf{Ha}^* + \mathbf{Wh} + \mathbf{Vp} + \mathbf{Ll} + \mathbf{e}$
- Covariate (\mathbf{H}) updated in each iteration based on the HYM effect (\mathbf{h})
- RJMC module in DMU (Madsen and Jensen, 2013)
- Bayesian setting (Gibbs sampling)
 - 2.5 M rounds
 - 500k burn-in
 - 200 interleave



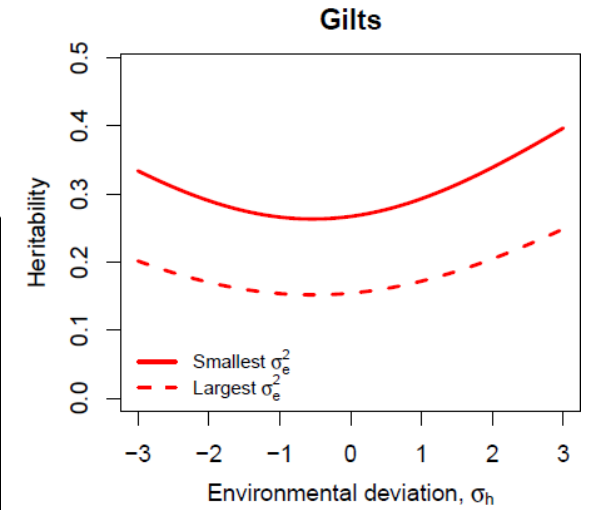
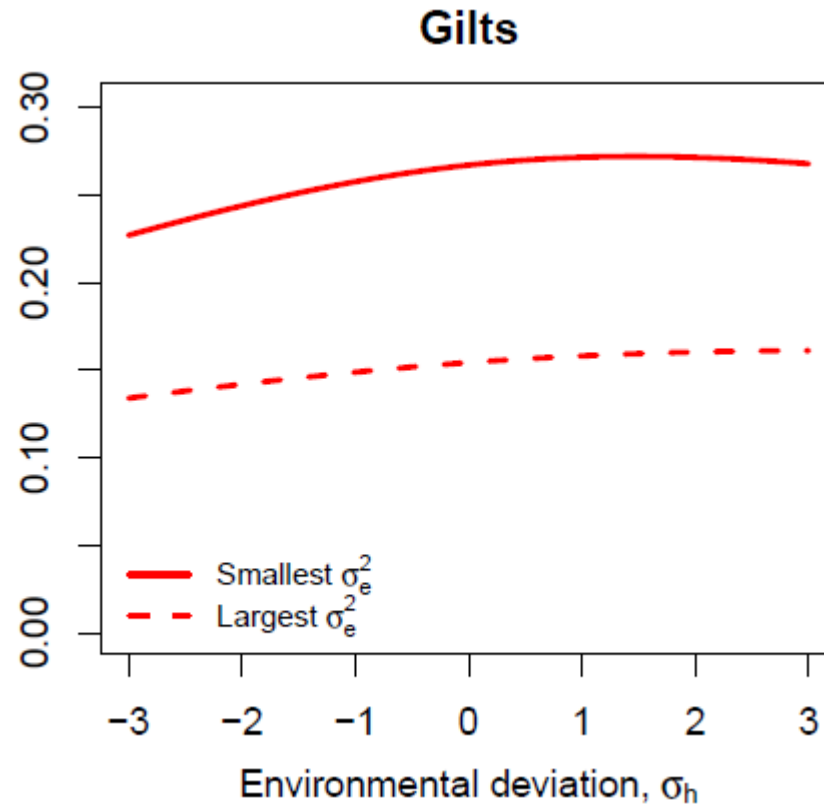
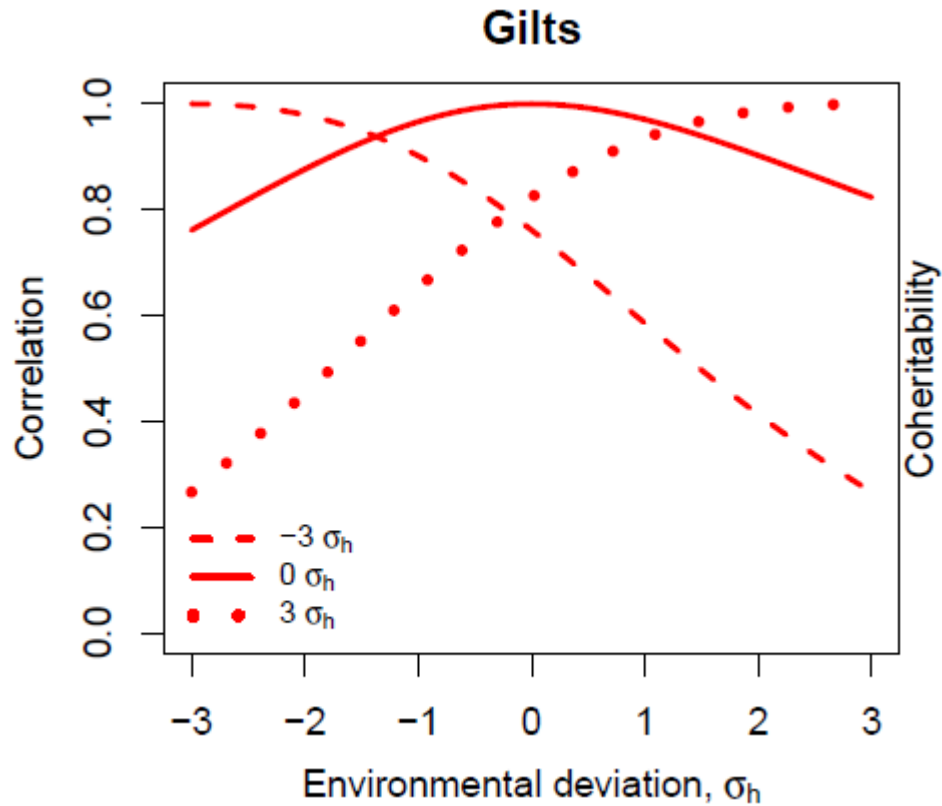
σ_a^2 and h^2

	Boars	Gilts
$\sigma_{a_0}^2$	1385 ^a	1333 ^a
$\sigma_{a^*}^2$	0.014 ^a	0.024 ^a
r_{a_0,a^*}	-0.227	0.144
σ_h^2	5076 ^a	3755 ^a

^a significantly different from zero



r_g and ch^2



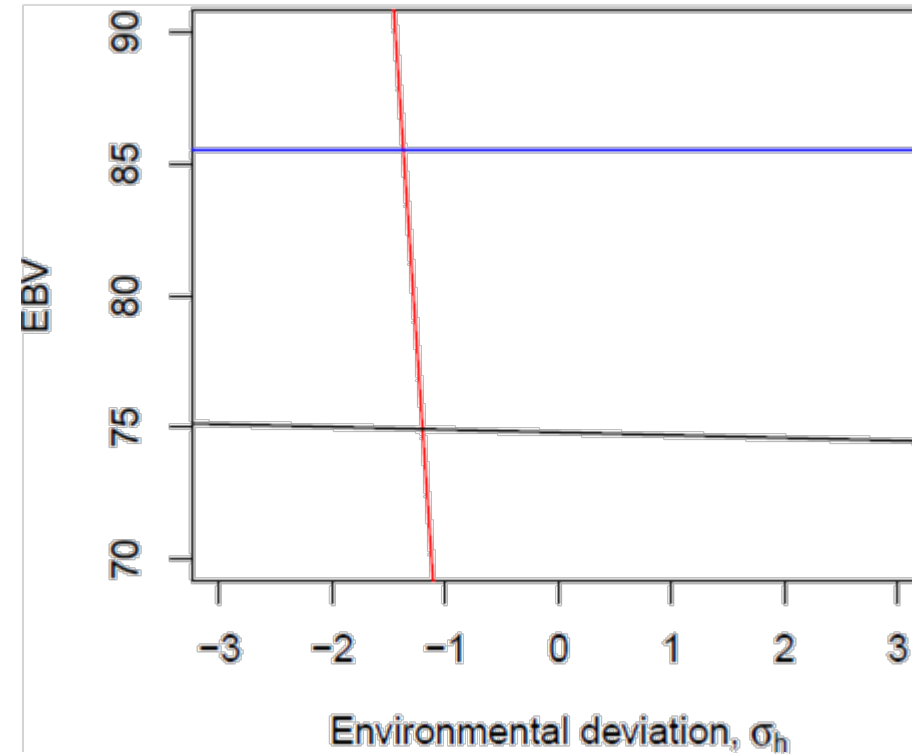
Benefits of the iterative update

- Comparison between
 - Iteratively updated covariate model (I)
 - Phenotypic means as covariate model (PM)

Prediction ability	Boars		Gilts		Bias	Boars		Gilts	
	I	PM	I	PM		I	PM	I	PM
Direct EBV	0.69	0.59	0.74	0.72	Direct EBV	0.98 (0.03)	0.73 (0.03)	1.00 (0.03)	0.73 (0.02)
Macro-ES EBV	0.77	0.58	0.90	0.73	Macro-ES EBV	0.69 (0.02)	0.70 (0.03)	0.86 (0.01)	0.74 (0.02)

How to use it

- Decrease macro-ES
 - Less response to change
- Decrease G×E
 - Similar response to change
- Environment specific selection
 - InterBull



Take home message

- G×E allows for adjusting macro-ES
- Growth rate in Danish Duroc exhibits G×E
- Reducing macro-ES of ADG in Danish Duroc will not reduce the level
- Coheritability can increase even if genetic correlations decrease
- The iteratively updated model performed better than a model with phenotypic means.

Thank you