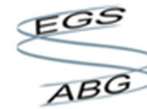
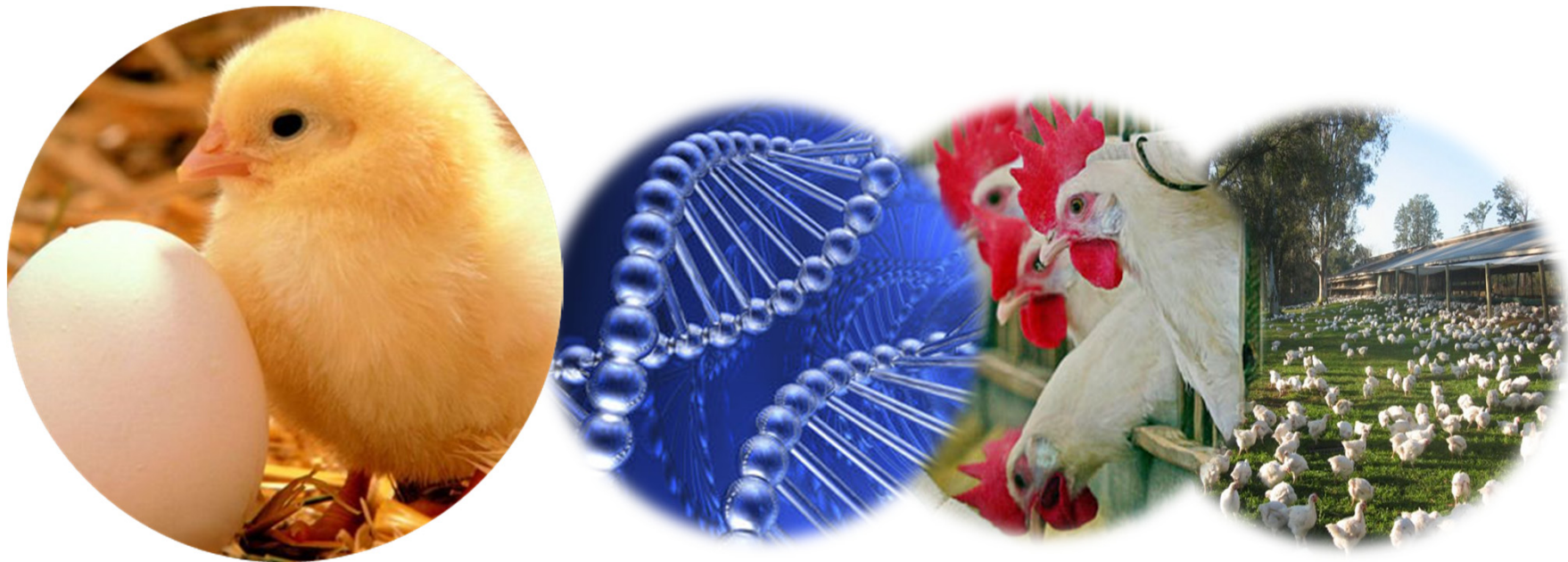


# Sire-based genomic prediction of heterosis in White Leghorn crossbreeds

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- Heterosis is an essential element of crossbreeding schemes
- Which parental lines should be mated to maximize benefits from heterosis?
- Can we predict heterosis?
  - long history with inconsistent results
  - based on limited number of markers/small data sets



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# Prediction of heterosis with 60K SNPs:

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- $\frac{\sum_{n=1}^N (p_i - p_j)^2}{N}$  predicts heterosis with an accuracy of 0.5  
(Amuzu-Aweh et al., 2013)

- We are able to identify suitable pure-line combinations

Exploit line-level heterosis

Reduce dependency on field trials of crossbreds

$p_i$  = allele frequency in sire line,  $p_j$  = allele frequency in dam line



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# Research question

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Can we exploit heterosis at the sire level?

- Utilise the variation between sires from the same pure-line
- Benefit more from heterosis



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# Derive equation to predict heterosis

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- Assumptions:

Heterosis is due to dominance deviations ( $d$ )

$d$  at each locus is independent of the allele frequency

$$\text{Heterosis} = \left[ (p_i - p_j)^2 + (p_{s_i} - p_i)(1 - 2p_j) \right] \cdot d$$

- Heterosis is a function of heterozygosity excess due to between- and within-line differences in allele frequencies

$p_i$  = allele frequency in sire line ,  $p_{s_i}$  = allele frequency in the sire

$p_j$  = allele frequency in dam line



## Data:

- ~3 450 sires from 4 lines genotyped at 60K SNPs
- 6 dam lines with known allele frequencies
- Phenotypic records from 16 crosses
  - 35 000 egg number records (EN)
  - 26 000 egg weight records (EW)

## Model:

$$y_{s_{ij}} = \text{sire\_line}_i + \text{dam\_line}_j + \beta \bar{x}_{s_{ij}} + \dots + e_{s_{ij}}$$

$$\bar{x}_{s_{ij}} = \frac{\sum_{n=1}^N [(p_i - p_j)^2 + (p_{s_i} - p_i)(1 - 2p_j)]}{N}$$



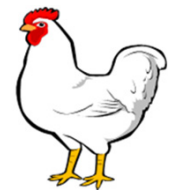
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# Results

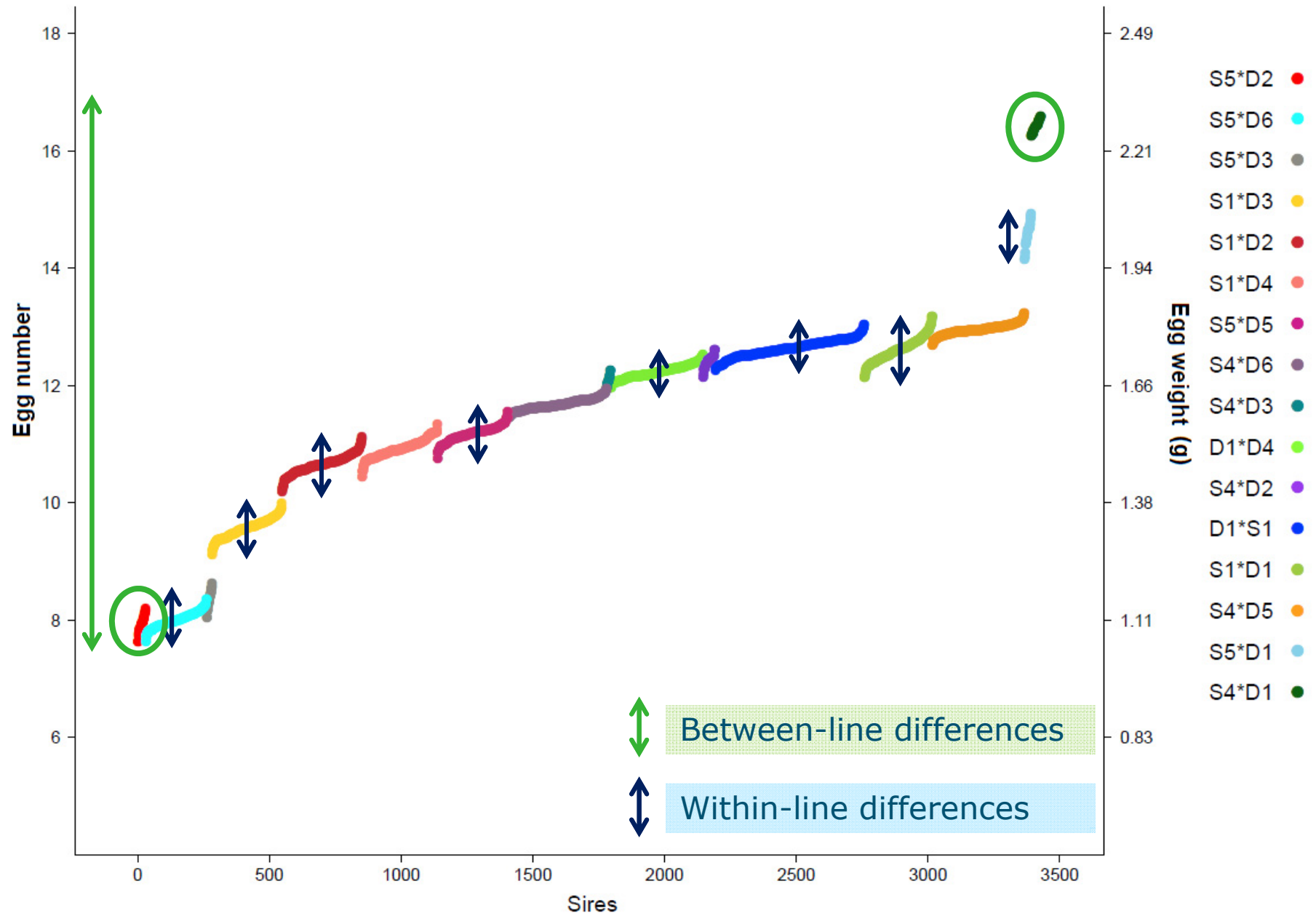
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- $\bar{x}_{s_{ij}}$  ranged from 0.08 to 0.18
- $\hat{\beta}_{EN} = 93.5$  (s.e = 18.3)
- $\hat{\beta}_{EW} = 12.9$  (s.e = 2.7)

$$\text{Predicted heterosis}_{\text{trait}, s_{ij}} = \hat{\beta}_{\text{trait}} \cdot \bar{x}_{s_{ij}}$$



Predicted heterosis for Egg number and Egg weight





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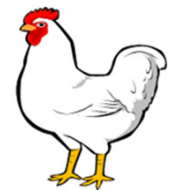
## Variance explained by components of $x$

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$$x_{s_{ij}} = \left[ (p_i - p_j)^2 + (p_{s_{ij}} - p_i)(1 - 2p_j) \right]$$

- Between-line accounted for 99.00%
- Within-line accounted for 0.72%

Individual sire information did not explain much



# Variance explained by components of $x$

- Regression on the **between-** and **within-**line components

	Egg number				Egg weight		
$s_{ij}$	Estimate	Se	p-value	$s_{ij}$	Estimate	se	p-value

	Full x			Components		
$\beta$	93.5	18.3	3.4 E-7	12.9	2.7	1.1 E-6
$\beta_1$	92.5	19.3	2.2 E-6	12.9	2.8	4.7 E-7



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# Ongoing...

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Single SNP regression on  $x_{s_i j}$

- Identified ~12K SNPs significant for predicting heterosis
- Larger variance between sires
- Bigger values of predicted heterosis



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# Conclusions

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- Heterosis can be predicted at the sire level
- Within-line sire variance contributes little to heterosis
- However, if sire genotypes are available, variance can be exploited
- Method may work better if genomic regions affecting heterosis are identified



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# Thanks for your attention

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## Acknowledgements:



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# Conclusions

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