

# BLUP Best Linear Unbiased Prediction-Estimation

## References

Searle, S.R. 1971 Linear Models, Wiley

Schaefer, L.R., Linear Models and Computer  
Strategies in Animal Breeding

Lynch and Walsh Chapter 26

# OLS Independently and Identically Distributed Errors with Mean 0 and variance $\sigma^2$

$$V(\varepsilon) = E[\varepsilon - E(\varepsilon)]^2$$

$$V(\varepsilon) = \begin{bmatrix} \sigma_e^2 & 0 & 0 & 0 \\ 0 & \sigma_e^2 & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \sigma_e^2 \end{bmatrix}$$

$$V(\varepsilon) = \mathbf{I}\sigma_e^2$$

The residual distribution from which each observation is sampled is the same

**Inbreeding Changes These**

**Related Individuals Cause these to be nonzero**

Residuals independent

# Solutions

- GLS
  - Fixes problem with changing variances and correlations in the data
- What about fixed effects?
  - How does one correct for
    - Environmental trend without a control
    - Herd effects
    - Year effects
    - Hatch effects

# Confounding of data

- Herd effects
  - Balanced design no problem
  - Require sample of every family in every herd
  - Old solution was within herd deviations
  - What if better herds have better genetics
- Fixed effects must be adjusted for genetic differences
- Random effects must be adjusted for fixed effects

# Simultaneous Adjustment of Fixed and Random effects

- Separate Independent variable into those that are

– Fixed  $\mathbf{Xb}$

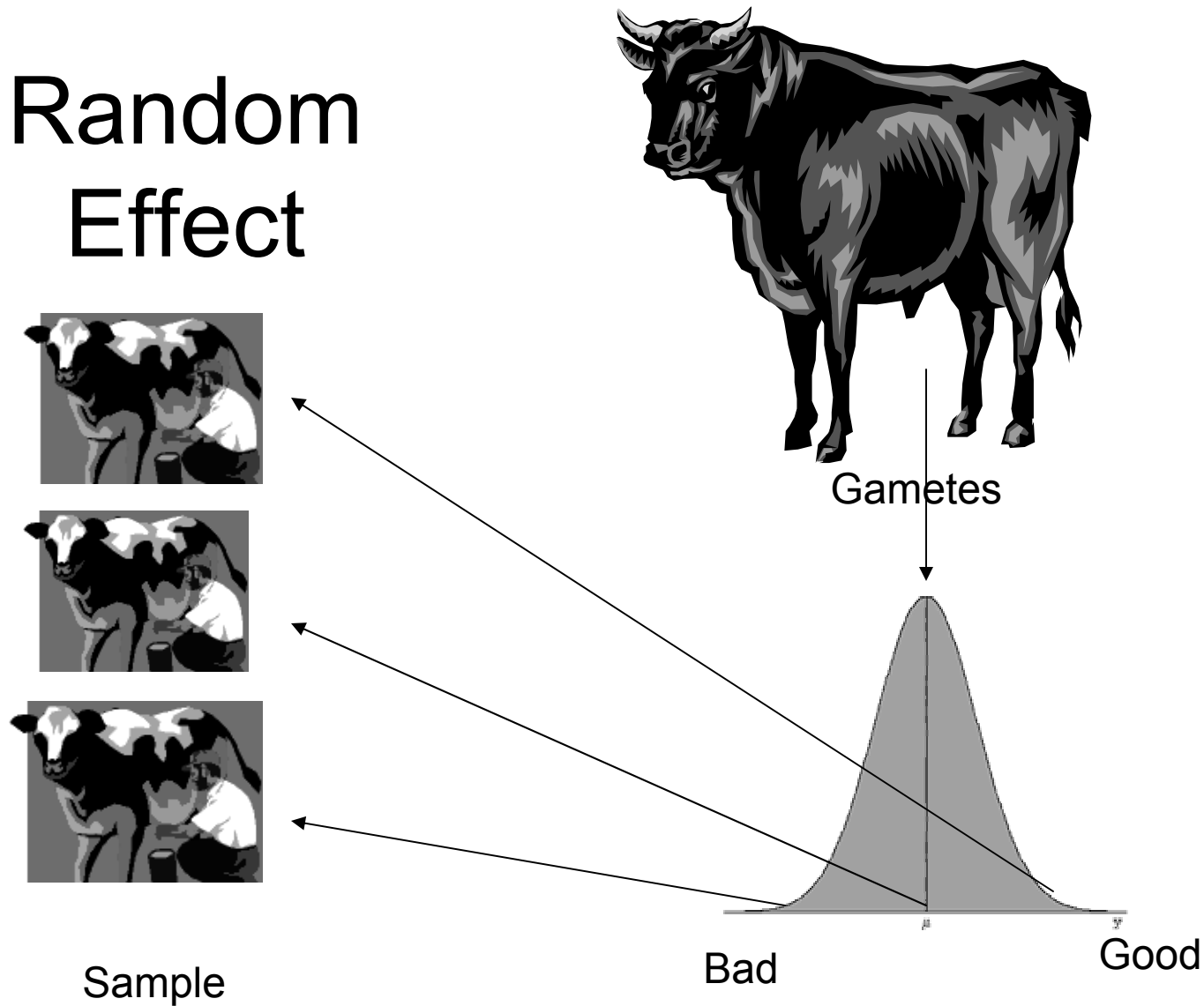
– Random  $\mathbf{Zu}$

$$\mathbf{Y} = \mathbf{Xb} + \mathbf{Zu} + \mathbf{e}$$

# Fixed and Random Effects

- Fixed Effect
  - Inference Space only to those levels
  - Herd, Year, Season, Parity, and Sex effect
- Random Effect
  - Effect Sampled From A Distribution Of Effects
  - Inference Space To The Population From Which The Random Effect Was Sampled

# Random Effect



Inference is to the genetic worth of the bull

# Variances In Mixed Models

$$\mathbf{Y} = \mathbf{X}\mathbf{b} + \mathbf{Z}\mathbf{u} + \mathbf{e} \quad V(\mathbf{b}) = \mathbf{0}$$

$$V(\mathbf{u}) = E(\mathbf{u}\mathbf{u}') = \mathbf{G}$$

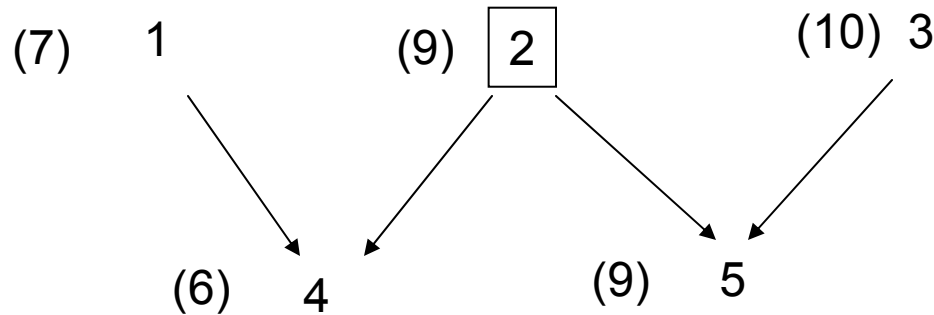
$$V(\mathbf{e}) = E(\mathbf{e}\mathbf{e}') = \mathbf{R}$$

$$V(\mathbf{Y}) = V(\mathbf{X}\mathbf{b} + \mathbf{Z}\mathbf{u} + \mathbf{e}) = \mathbf{Z}\mathbf{G}\mathbf{Z}' + \mathbf{R}$$



# Example 1

$$\mathbf{Y} = \begin{bmatrix} 7 \\ 9 \\ 10 \\ 6 \\ 9 \end{bmatrix}$$



$$b = [\mu]$$

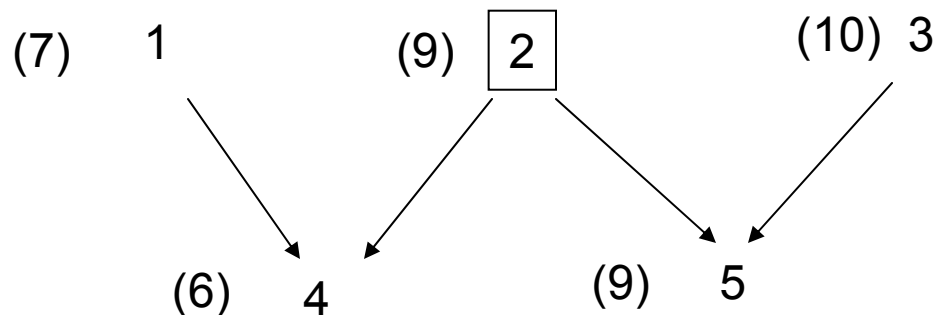
$$\mathbf{X} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\mathbf{Z} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{U} = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \end{bmatrix}$$

$$\mathbf{e} = \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \\ e_5 \end{bmatrix}$$

# Example 1



$$\begin{bmatrix} 7 \\ 9 \\ 10 \\ 6 \\ 9 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} [\mu] + \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \\ e_5 \end{bmatrix}$$

$$\mathbf{Y} = \mathbf{X}\mathbf{b} + \mathbf{Z}\mathbf{u} + \mathbf{e}$$

# ML Derivation of BLUP

Joint density of  $\mathbf{y}$  and  $\mathbf{u}$        $f(\mathbf{y}, \mathbf{u}) = g(\mathbf{y}/\mathbf{u})h(\mathbf{u})$

$$g(\mathbf{y}/\mathbf{u}) = g(\mathbf{e})$$

$$g(\mathbf{e}) = \frac{1}{(2\pi)^{\frac{1}{2}N} V(\mathbf{e})^{\frac{1}{2}}} e^{-\frac{1}{2}\mathbf{e}'V(\mathbf{e})^{-1}\mathbf{e}} \quad h(\mathbf{u}) = \frac{1}{(2\pi)^{\frac{1}{2}N} V(\mathbf{u})^{\frac{1}{2}}} e^{-\frac{1}{2}\mathbf{u}'V(\mathbf{u})^{-1}\mathbf{u}}$$

$$f(\mathbf{y}, \mathbf{u}) = \frac{1}{(2\pi)^{\frac{1}{2}N} \mathbf{R}^{\frac{1}{2}}} e^{-\frac{1}{2}\mathbf{e}'\mathbf{R}^{-1}\mathbf{e}} \frac{1}{(2\pi)^{\frac{1}{2}N} \mathbf{G}^{\frac{1}{2}}} e^{-\frac{1}{2}\mathbf{u}'\mathbf{G}^{-1}\mathbf{u}}$$

$$f(\mathbf{y}, \mathbf{u}) = c_1 e^{-\frac{1}{2}\mathbf{e}'\mathbf{R}^{-1}\mathbf{e}} c_2 e^{-\frac{1}{2}\mathbf{u}'\mathbf{G}^{-1}\mathbf{u}}$$

$$f(\mathbf{y}, \mathbf{u}) = c e^{-\frac{1}{2}\mathbf{e}'\mathbf{R}^{-1}\mathbf{e}} e^{-\frac{1}{2}\mathbf{u}'\mathbf{G}^{-1}\mathbf{u}}$$

$$f(\mathbf{y}, \mathbf{u}) = L = ce^{-\frac{1}{2}\mathbf{e}'\mathbf{R}^{-1}\mathbf{e}} e^{-\frac{1}{2}\mathbf{u}'\mathbf{G}^{-1}\mathbf{u}}$$

Maximize w.r.t **b** and **u**

$$\ln(L) = \ln(c) - \frac{1}{2}\mathbf{e}'\mathbf{R}^{-1}\mathbf{e} - \frac{1}{2}\mathbf{u}'\mathbf{G}^{-1}\mathbf{u}$$

$$\mathbf{e} = \mathbf{Y} - \mathbf{Xb} - \mathbf{Zu}$$

$$\begin{aligned} \ln(L) = \ln(c) - \frac{1}{2}(\mathbf{Y} - \mathbf{Xb} - \mathbf{Zu})'\mathbf{R}^{-1}(\mathbf{Y} - \mathbf{Xb} - \mathbf{Zu}) \\ - \frac{1}{2}\mathbf{u}'\mathbf{G}^{-1}\mathbf{u} \end{aligned}$$

Take Derivative w.r.t **b**

$$\begin{aligned} & (\mathbf{Y} - \mathbf{Xb} - \mathbf{Zu})' \mathbf{R}^{-1} (\mathbf{Y} - \mathbf{Xb} - \mathbf{Zu}) + \mathbf{u}' \mathbf{G}^{-1} \mathbf{u} \\ &= [\mathbf{Y}' - (\mathbf{Xb})' - (\mathbf{Zu})'] \mathbf{R}^{-1} (\mathbf{Y} - \mathbf{Xb} - \mathbf{Zu}) + \mathbf{u}' \mathbf{G}^{-1} \mathbf{u} \\ &= \mathbf{Y}' \mathbf{R}^{-1} \mathbf{Y} - \mathbf{Y}' \mathbf{R}^{-1} \mathbf{Xb} - \mathbf{Y}' \mathbf{R}^{-1} \mathbf{Zu} \\ &\quad - (\mathbf{Xb})' \mathbf{R}^{-1} \mathbf{Y} + (\mathbf{Xb})' \mathbf{R}^{-1} \mathbf{Xb} + (\mathbf{Xb})' \mathbf{R}^{-1} \mathbf{Zu} \\ &\quad - (\mathbf{Zu})' \mathbf{R}^{-1} \mathbf{Y} + (\mathbf{Zu})' \mathbf{R}^{-1} \mathbf{Xb} + (\mathbf{Zu})' \mathbf{R}^{-1} \mathbf{Zu} + \mathbf{u}' \mathbf{G}^{-1} \mathbf{u} \end{aligned}$$

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$$\frac{\partial(\ln L)}{\partial \mathbf{b}} = 0 \quad \begin{aligned} & -\mathbf{Y}' \mathbf{R}^{-1} \mathbf{X} - \mathbf{X}' \mathbf{R}^{-1} \mathbf{Y} + (\mathbf{Xb})' \mathbf{R}^{-1} \mathbf{X} + \\ & \mathbf{X}' \mathbf{R}^{-1} \mathbf{Xb} + \mathbf{X}' \mathbf{R}^{-1} \mathbf{Zu} + (\mathbf{Zu})' \mathbf{R}^{-1} \mathbf{X} = 0 \end{aligned}$$

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$$-2\mathbf{X}' \mathbf{R}^{-1} \mathbf{Y} + 2\mathbf{X}' \mathbf{R}^{-1} \mathbf{Xb} + 2\mathbf{X}' \mathbf{R}^{-1} \mathbf{Zu} = 0$$

$$\mathbf{X}' \mathbf{R}^{-1} \mathbf{Xb} + \mathbf{X}' \mathbf{R}^{-1} \mathbf{Zu} = \mathbf{X}' \mathbf{R}^{-1} \mathbf{Y}$$

Take Derivative w.r.t  $\mathbf{u}$

$$(\mathbf{Y} - \mathbf{Xb} - \mathbf{Zu})' \mathbf{R}^{-1} (\mathbf{Y} - \mathbf{Xb} - \mathbf{Zu}) + \mathbf{u}' \mathbf{G}^{-1} \mathbf{u}$$

$$= \mathbf{Y}' \mathbf{R}^{-1} \mathbf{Y} - \mathbf{Y}' \mathbf{R}^{-1} \mathbf{Xb} - \mathbf{Y}' \mathbf{R}^{-1} \mathbf{Zu}$$

$$- (\mathbf{Xb})' \mathbf{R}^{-1} \mathbf{Y} + (\mathbf{Xb})' \mathbf{R}^{-1} \mathbf{Xb} + (\mathbf{Xb})' \mathbf{R}^{-1} \mathbf{Zu}$$

$$- (\mathbf{Zu})' \mathbf{R}^{-1} \mathbf{Y} + (\mathbf{Zu})' \mathbf{R}^{-1} \mathbf{Xb} + (\mathbf{Zu})' \mathbf{R}^{-1} \mathbf{Zu} + \mathbf{u}' \mathbf{G}^{-1} \mathbf{u}$$

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$$\frac{\partial(\ln L)}{\partial \mathbf{u}} = 0 \quad \begin{aligned} & - \mathbf{Y}' \mathbf{R}^{-1} \mathbf{Z} + (\mathbf{Xb})' \mathbf{R}^{-1} \mathbf{Z} - (\mathbf{Z})' \mathbf{R}^{-1} \mathbf{Y} + (\mathbf{Z})' \mathbf{R}^{-1} \mathbf{Xb} \\ & + (\mathbf{Z})' \mathbf{R}^{-1} \mathbf{Zu} + (\mathbf{Zu})' \mathbf{R}^{-1} \mathbf{Z} + 2\mathbf{G}^{-1} \mathbf{u} = 0 \end{aligned}$$

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$$- 2\mathbf{Y}' \mathbf{R}^{-1} \mathbf{Z} + 2\mathbf{Z}' \mathbf{R}^{-1} \mathbf{Xb} + 2\mathbf{Z}' \mathbf{R}^{-1} \mathbf{Zu} + 2\mathbf{G}^{-1} \mathbf{u} = 0$$

$$\mathbf{Z}' \mathbf{R}^{-1} \mathbf{Xb} + \mathbf{Z}' \mathbf{R}^{-1} \mathbf{Zu} + \mathbf{G}^{-1} \mathbf{u} = \mathbf{Y}' \mathbf{R}^{-1} \mathbf{Z}$$

# Mixed Model Equations

$$\mathbf{X}'\mathbf{R}^{-1}\mathbf{X}\mathbf{b} + \mathbf{X}'\mathbf{R}^{-1}\mathbf{Z}\mathbf{u} = \mathbf{X}'\mathbf{R}^{-1}\mathbf{Y}$$

$$\mathbf{Z}'\mathbf{R}^{-1}\mathbf{X}\mathbf{b} + \mathbf{Z}'\mathbf{R}^{-1}\mathbf{Z}\mathbf{u} + \mathbf{G}^{-1}\mathbf{u} = \mathbf{Y}'\mathbf{R}^{-1}\mathbf{Z}$$

$$\begin{bmatrix} \mathbf{X}'\mathbf{R}^{-1}\mathbf{X} & \mathbf{X}'\mathbf{R}^{-1}\mathbf{Z} \\ \mathbf{Z}'\mathbf{R}^{-1}\mathbf{X} & \mathbf{Z}'\mathbf{R}^{-1}\mathbf{Z} + \mathbf{G}^{-1} \end{bmatrix} \begin{bmatrix} \mathbf{b} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{R}^{-1}\mathbf{Y} \\ \mathbf{Y}'\mathbf{R}^{-1}\mathbf{Z} \end{bmatrix}$$

Simplifications If  $\mathbf{R} = \mathbf{I}\sigma_e^2$

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z} \\ \mathbf{Z}'\mathbf{X} & \mathbf{Z}'\mathbf{Z} + \sigma_e^2\mathbf{G}^{-1} \end{bmatrix} \begin{bmatrix} \mathbf{b} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{Y} \\ \mathbf{Y}'\mathbf{Z} \end{bmatrix}$$

# Normal Distribution of Y Not Necessary

It is possible to show with alternative BLUE estimation techniques that the Same Results would be obtained without assuming normality  
See Schaffer Notes



# Simplifications

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z} \\ \mathbf{Z}'\mathbf{X} & \mathbf{Z}'\mathbf{Z} + \sigma_e^2 \mathbf{G}^{-1} \end{bmatrix} \begin{bmatrix} \mathbf{b} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{Y} \\ \mathbf{Y}'\mathbf{Z} \end{bmatrix}$$

Assuming Additivity

$$\mathbf{G} = \mathbf{A} \sigma_a^2$$

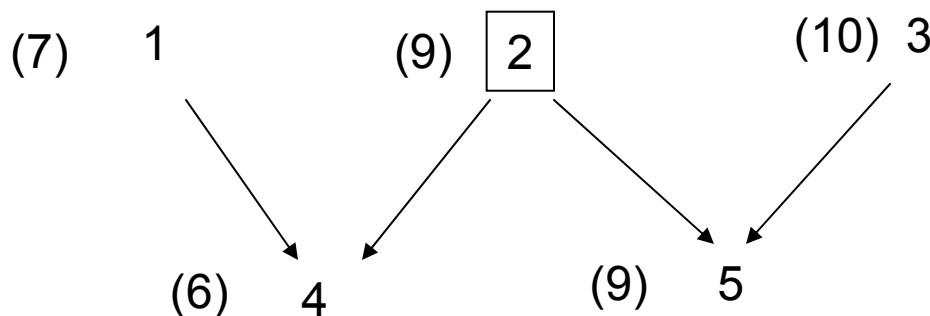
$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z} \\ \mathbf{Z}'\mathbf{X} & \mathbf{Z}'\mathbf{Z} + \frac{\sigma_e^2}{\sigma_a^2} \mathbf{A}^{-1} \end{bmatrix} \begin{bmatrix} \mathbf{b} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{Y} \\ \mathbf{Y}'\mathbf{Z} \end{bmatrix}$$

Only Estimate of Ratio is Needed

Only inverse is needed

# Example 1

$$\mathbf{Y} = \begin{bmatrix} 7 \\ 9 \\ 10 \\ 6 \\ 9 \end{bmatrix}$$



$$b = [\mu]$$

$$\mathbf{X} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\mathbf{Z} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{u} = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \end{bmatrix}$$

$$\mathbf{e} = \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \\ e_5 \end{bmatrix}$$

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z} \\ \mathbf{Z}'\mathbf{X} & \mathbf{Z}'\mathbf{Z} + \frac{\sigma_e^2}{\sigma_a^2} \mathbf{A}^{-1} \end{bmatrix} \begin{bmatrix} \mathbf{b} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{Y} \\ \mathbf{Y}'\mathbf{Z} \end{bmatrix}$$

$$\mathbf{X}'\mathbf{X} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \quad \mathbf{X}'\mathbf{Z} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

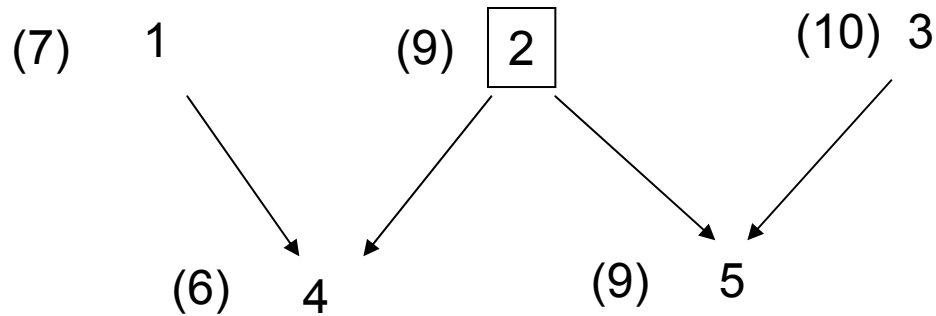
$$\mathbf{X}'\mathbf{X} = 5 \quad \mathbf{X}'\mathbf{Z} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z} \\ \mathbf{Z}'\mathbf{X} & \mathbf{Z}'\mathbf{Z} + \frac{\sigma_e^2}{\sigma_a^2} \mathbf{A}^{-1} \end{bmatrix} \begin{bmatrix} \mathbf{b} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{Y} \\ \mathbf{Y}'\mathbf{Z} \end{bmatrix}$$

$$\mathbf{Z}'\mathbf{X} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\mathbf{Z}'\mathbf{X} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\mathbf{Z}'\mathbf{Z} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$



$$\mathbf{A} = \begin{bmatrix} 1 & 0 & 0 & \frac{1}{2} & 0 \\ 0 & 1 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & 1 & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 & 1 & \frac{1}{4} \\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{4} & 1 \end{bmatrix}$$

Assume heritability=.5

$$\frac{\sigma_e^2}{\sigma_a^2} = 1 \quad \mathbf{Z}'\mathbf{Z} + \frac{\sigma_e^2}{\sigma_a^2} \mathbf{A}^{-1} = \begin{bmatrix} \frac{5}{2} & \frac{1}{2} & 0 & -1 & 0 \\ \frac{1}{2} & 3 & \frac{1}{2} & -1 & -1 \\ 0 & \frac{1}{2} & \frac{5}{2} & 0 & -1 \\ -1 & -1 & 0 & 3 & 0 \\ 0 & -1 & -1 & 0 & 3 \end{bmatrix}$$

$$\begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z} \\ \mathbf{Z}'\mathbf{X} & \mathbf{Z}'\mathbf{Z} + \frac{\sigma_e^2}{\sigma_a^2} \mathbf{A}^{-1} \end{bmatrix} \begin{bmatrix} \mathbf{b} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{Y} \\ \mathbf{Y}'\mathbf{Z} \end{bmatrix}$$

$$\mathbf{X}'\mathbf{Y} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 7 \\ 9 \\ 10 \\ 6 \\ 9 \end{bmatrix}$$

$$\mathbf{X}'\mathbf{Y} = [41]$$

$$\mathbf{Y}'\mathbf{Z} = \begin{bmatrix} 7 \\ 9 \\ 10 \\ 6 \\ 9 \end{bmatrix}$$

# MME

$$\begin{bmatrix} 5 & 1 & 1 & 1 & 1 & 1 \\ 1 & \frac{5}{2} & \frac{1}{2} & 0 & -1 & 0 \\ 1 & \frac{1}{2} & 3 & \frac{1}{2} & -1 & -1 \\ 1 & 0 & \frac{1}{2} & \frac{5}{2} & 0 & -1 \\ 1 & -1 & -1 & 0 & 3 & 0 \\ 1 & 0 & -1 & -1 & 0 & 3 \end{bmatrix} \begin{bmatrix} \mu \\ a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \end{bmatrix} = \begin{bmatrix} 41 \\ 7 \\ 9 \\ 10 \\ 6 \\ 9 \end{bmatrix}$$

# Estimation of Error Variance if the Ratio is Known

$$\lambda = \frac{\sigma_e^2}{\sigma_a^2}$$

$$\hat{\sigma}_e^2 = MSE = \left[ \frac{\mathbf{Y}'\mathbf{Y} - \hat{\mathbf{b}}\mathbf{X} - \hat{\mathbf{u}}\mathbf{Z}}{N - R(X)} \right]$$

$$\hat{\sigma}_a^2 = \frac{\hat{\sigma}_e^2}{\lambda}$$



```
proc iml;  
start main;
```

```
y={ 7,  
    9,  
    10,  
    6,  
    9};
```

```
X={1,  
    1,  
    1,  
    1,  
    1};
```

```
A={1 0 0 .5 0,  
    0 1 0 .5 .5,  
    0 0 1 0 .5,  
    .5 .5 0 1 .25,  
    0 .5 .5 .25 1};
```

```
lam=1;
```

```
Z={1 0 0 0 0,  
    0 1 0 0 0,  
    0 0 1 0 0,  
    0 0 0 1 0,  
    0 0 0 0 1};
```

```
LHS=((X`*X)||((X`*Z))//((Z`*X)||((Z`*  
    Z+INV(A)#LAM)));
```

```
RHS=(X`*Y)//(Z`*Y);  
C=INV(LHS);
```

```
BU=C*RHS;  
print C BU;
```

```
finish main;  
run;  
quit;
```

# Estimates

BU

$$\begin{array}{r} 8.3018868 \\ -0.960813 \\ 0.0754717 \\ 0.8853411 \\ -1.062409 \\ 0.5529753 \end{array} \quad \left\{ \begin{array}{l} \mathbf{U} = \begin{bmatrix} \hat{a}_1 \\ \hat{a}_2 \\ \hat{a}_3 \\ \hat{a}_4 \\ \hat{a}_5 \end{bmatrix} \end{array} \right.$$

$b = [\hat{\mu}]$

# Variance of the Estimates

$$\begin{bmatrix} \mathbf{C}_{11} & \mathbf{C}_{12} \\ \mathbf{C}_{21} & \mathbf{C}_{22} \end{bmatrix} = \begin{bmatrix} \mathbf{X}'\mathbf{X} & \mathbf{X}'\mathbf{Z} \\ \mathbf{Z}'\mathbf{X} & \mathbf{Z}'\mathbf{Z} + \frac{\sigma_e^2}{\sigma_a^2} \mathbf{A}^{-1} \end{bmatrix}^{-1}$$

$$V(\hat{\mathbf{b}}) = \mathbf{C}_{11} \sigma_e^2$$

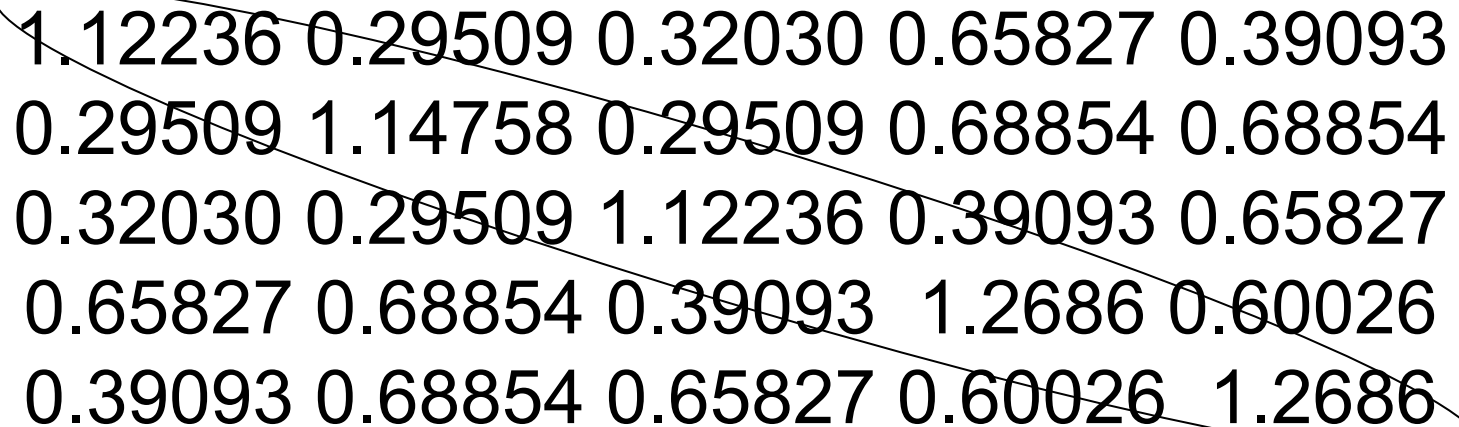
$$V(\hat{\mathbf{u}} - \mathbf{u}) = \mathbf{C}_{22} \sigma_e^2$$

Prediction Error Variance

$$V(\hat{\mathbf{u}}) = \mathbf{A} \sigma_a^2 + \mathbf{C}_{22} \sigma_e^2$$

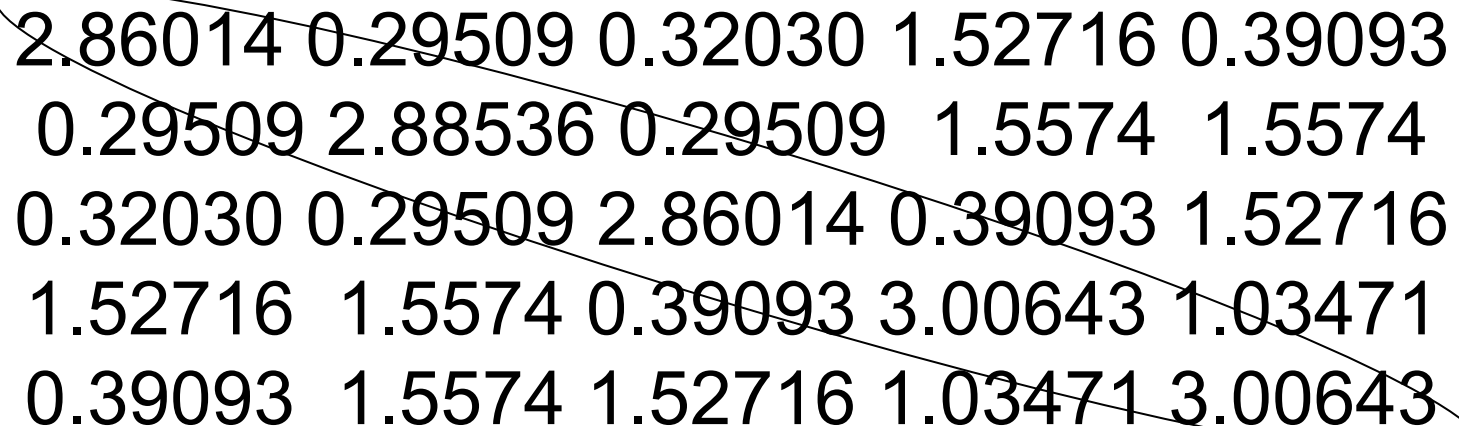
Prediction Error Variance  
Including Drift Variance

## PEV



1.12236	0.29509	0.32030	0.65827	0.39093
0.29509	1.14758	0.29509	0.68854	0.68854
0.32030	0.29509	1.12236	0.39093	0.65827
0.65827	0.68854	0.39093	1.2686	0.60026
0.39093	0.68854	0.65827	0.60026	1.2686

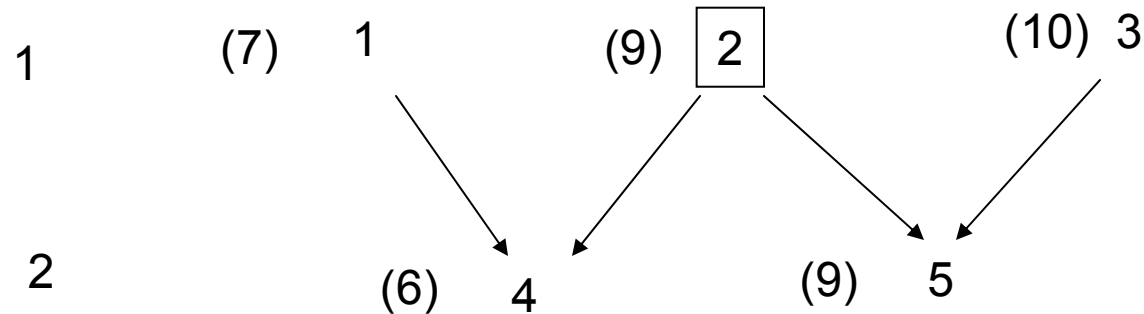
## EV



2.86014	0.29509	0.32030	1.52716	0.39093
0.29509	2.88536	0.29509	1.5574	1.5574
0.32030	0.29509	2.86014	0.39093	1.52716
1.52716	1.5574	0.39093	3.00643	1.03471
0.39093	1.5574	1.52716	1.03471	3.00643

# Example 1

Generation



Generation	Mean	Variance
1	8.66	2.860
2	7.5	3.006

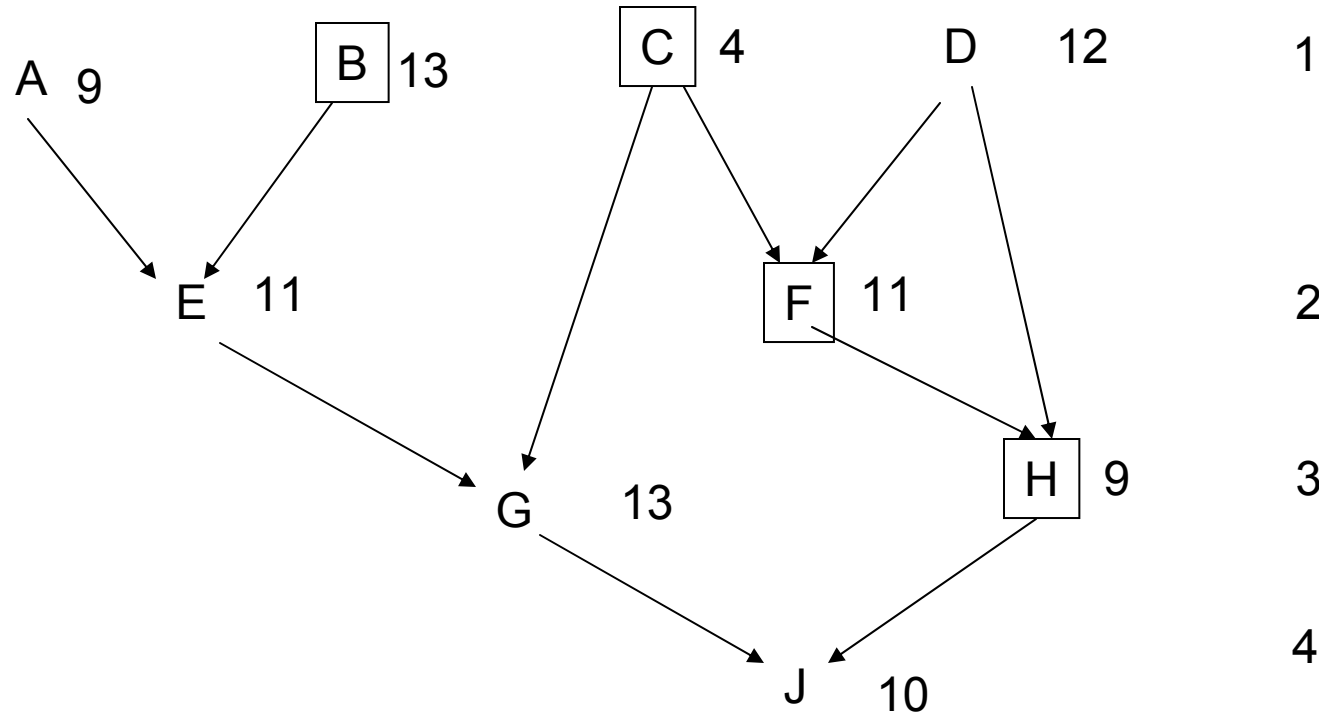
# Lab Problem 6.1

- How does changing the heritability affect the estimates and PEV and PE variance? Set to each of the following and compare results

$$\frac{\sigma_e^2}{\sigma_a^2} = 100 \qquad \frac{\sigma_e^2}{\sigma_a^2} = .1$$

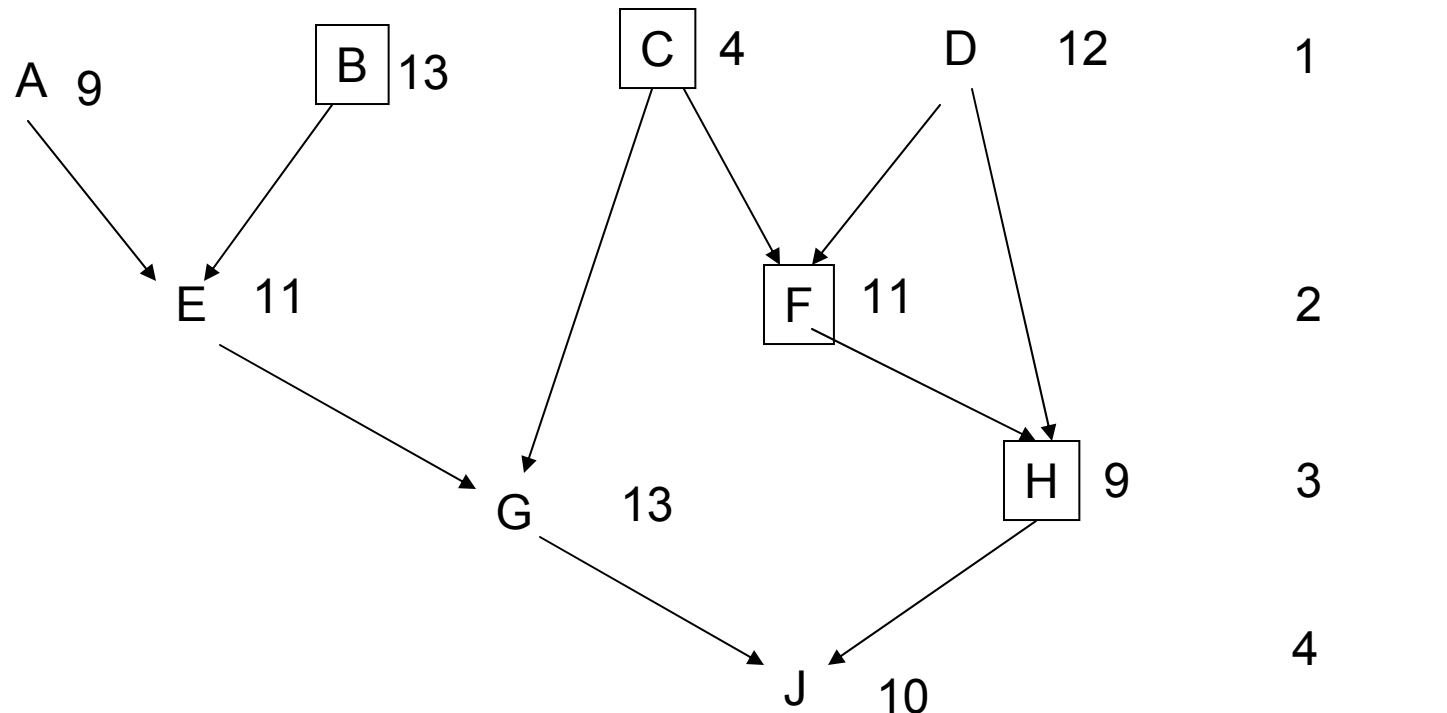
Interpret the results

# Lab Problem 6.2a



Find the best estimate of the genetic worth of each animal, additive and error variance, PEV, and PV. Assume a heritability of .5

# Lab Problem 6.2b

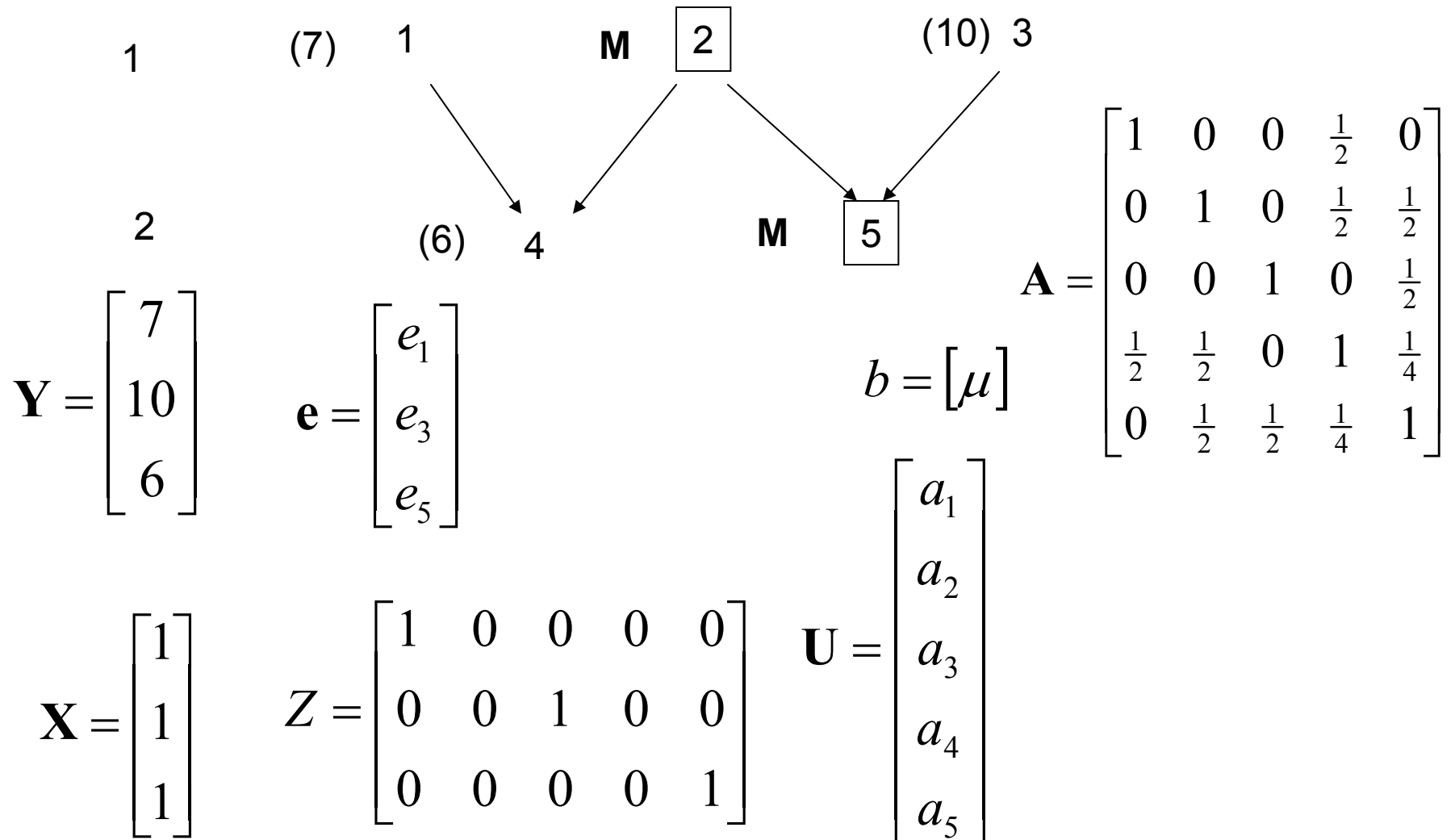


Environmental trend can be found by fitting generation number as a covariate. Genetic trend is found by taking the average of all EBV's in that generation and fitting the means to a linear regression. What are the genetic and environmental trend for this data



# Missing Values (Sex Limited Traits)

Generation



```
proc iml;  
start main;
```

```
y={ 7,  
    10,  
    6};
```

```
X={1,  
    1,  
    1};
```

```
A={1 0 0 .5 0,  
    0 1 0 .5 .5,  
    0 0 1 0 .5,  
    .5 .5 0 1 .25,  
    0 .5 .5 .25 1};
```

```
lam=1;
```

```
Z={1 0 0 0 0,  
    0 0 1 0 0,  
    0 0 0 0 1};
```

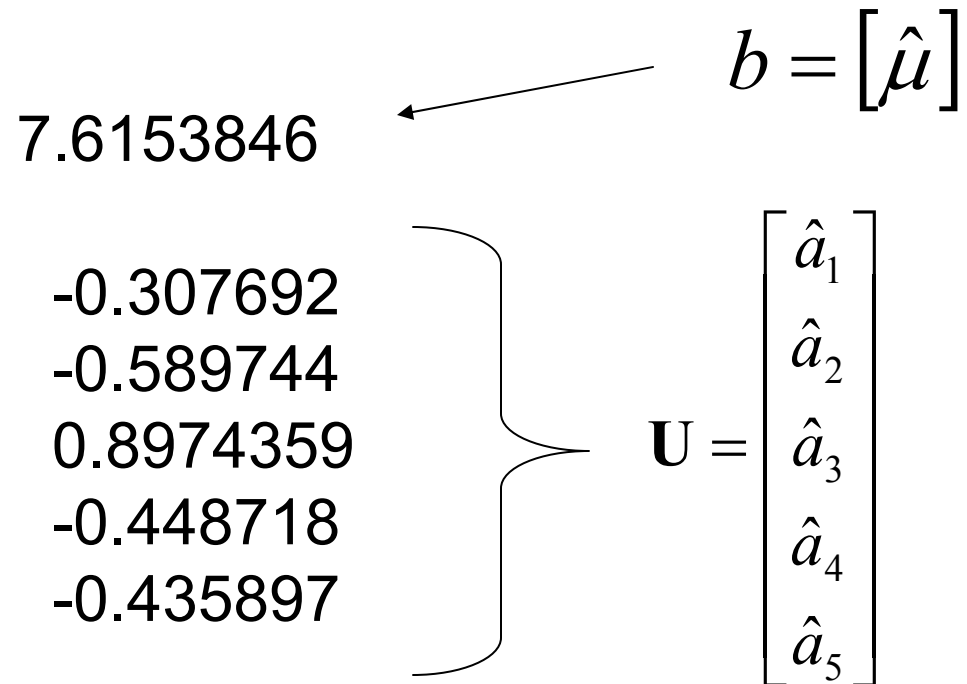
```
LHS=((X`*X)|| (X`*Z))//((Z`*X)|| (Z`*Z+INV(A)#LAM));
```

```
RHS=(X`*Y)//(Z`*Y);  
C=INV(HS);
```

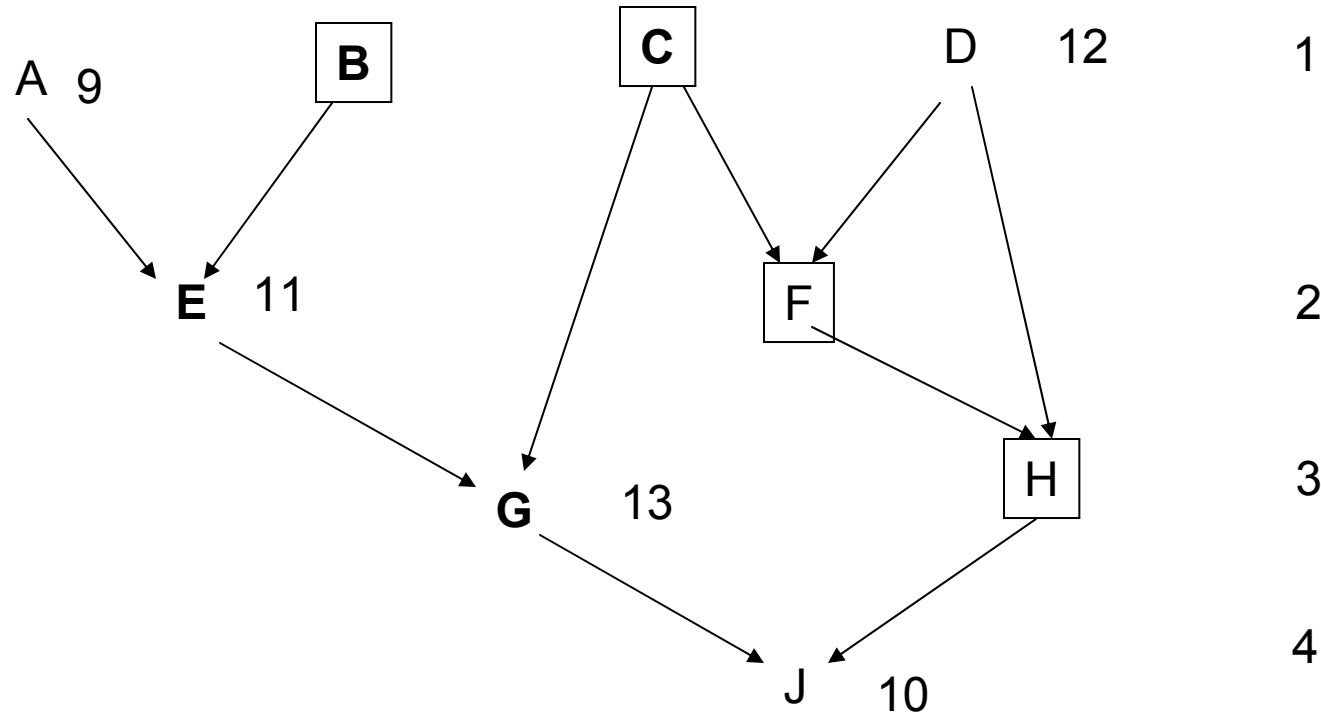
```
BU=C*RHS;  
print C BU;
```

```
finish main;  
run;  
quit;
```

# Estimates

$$\begin{array}{c} 7.6153846 \\ -0.307692 \\ -0.589744 \\ 0.8974359 \\ -0.448718 \\ -0.435897 \end{array} \quad \left\{ \begin{array}{c} \mathbf{U} = \begin{bmatrix} \hat{a}_1 \\ \hat{a}_2 \\ \hat{a}_3 \\ \hat{a}_4 \\ \hat{a}_5 \end{bmatrix} \end{array} \right. \quad b = [\hat{\mu}]$$


## Lab Problem 6.3: Sex Limited Trait



## Estimate breeding values for the males

$$\frac{\sigma_e^2}{\sigma_a^2} = 1$$

# Extensions of Model

- Inclusion of Dominance and Epistasis
  - Dominance relationship needed
    - Reflect the probability that animals have the same pair of alleles in common
    - Epistatic genetic effects are the result of interactions of among additive and dominance genetic effects
  - Useful to determine in crossbreeding programs but generally not useful in pure breeding programs
    - An individual does not pass on a dominance or epistatic effect, it is the results of both parents

# Limitations

- Based on infinitesimal model
  - Does not work for traits determined by small number of loci
  - Genetic variance is assumed constant except Bulmer effect (reduction in variance due to disequilibrium)
- Model needs to be correct
  - Garbage In Garbage Out (GIGO)
  - Typical Animal Model Assumes Additivity and Independence of residuals