

for  $j$  in  $1 \dots J$

for  $i$  in  $1 \dots n_j$

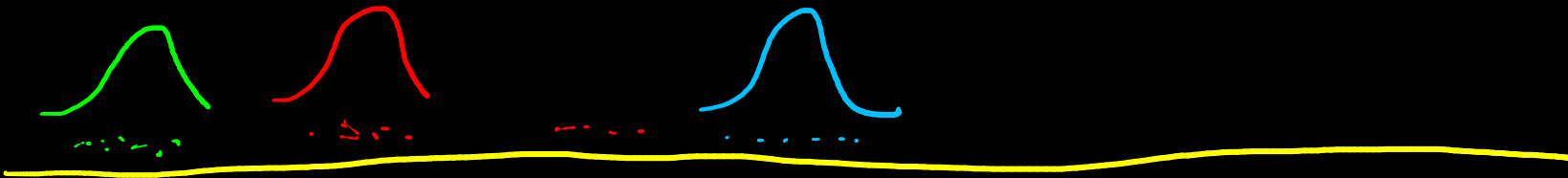
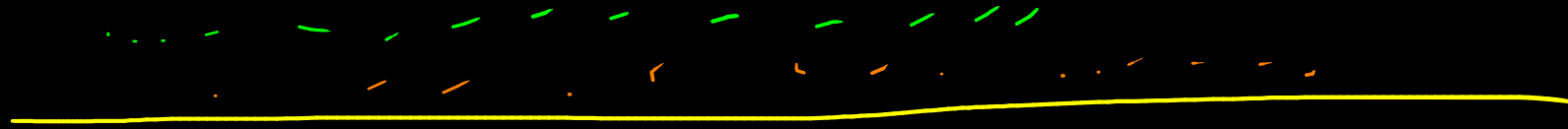
$$y_{ji} \sim N(\mu_j, \sigma^2)$$

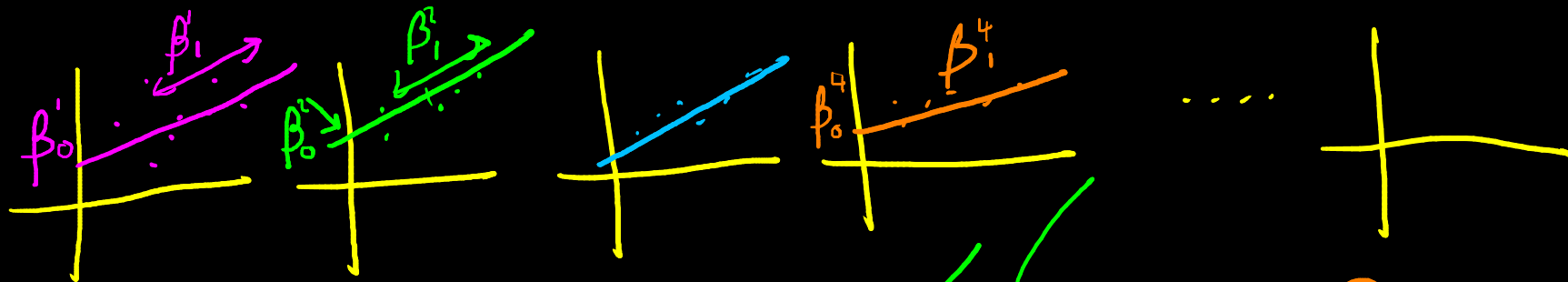
$$\mu_j \sim N(\theta, \tau^2)$$

$$n_j = 12$$

$$Z_j = \mu_j - \theta$$

$$\begin{cases} y_{ji} = \theta + Z_j + \varepsilon_{ji} \\ Z_j \sim N(0, \tau^2) ; \varepsilon_{ji} \sim N(0, \sigma^2) \end{cases}$$

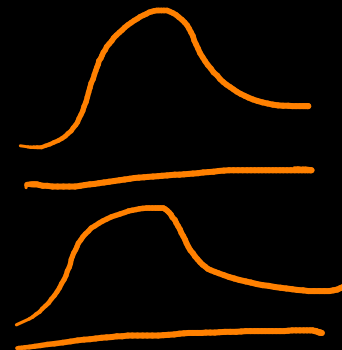




for  $j$  in  $1 \dots J$

$$\beta_0^j \sim N(b_0, \tau_0^2)$$

$$\beta_1^j \sim N(b_1, \tau_1^2)$$



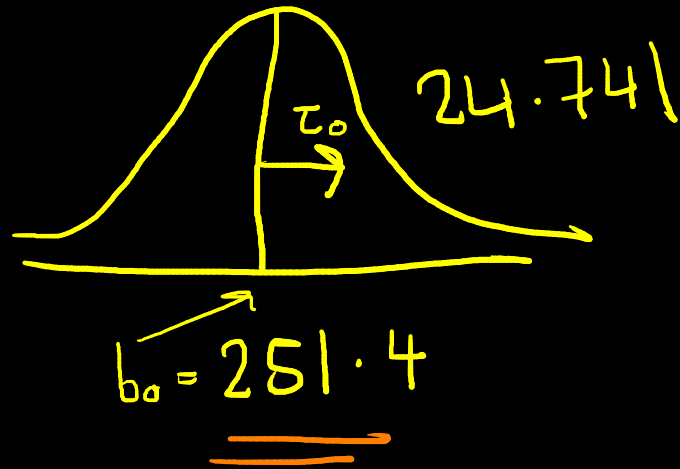
for  $i$  in  $1 \dots n$

$$y_i \sim N(\mu_i, \sigma^2)$$

$$\mu_i = \beta_0 + \beta_1 x_i$$

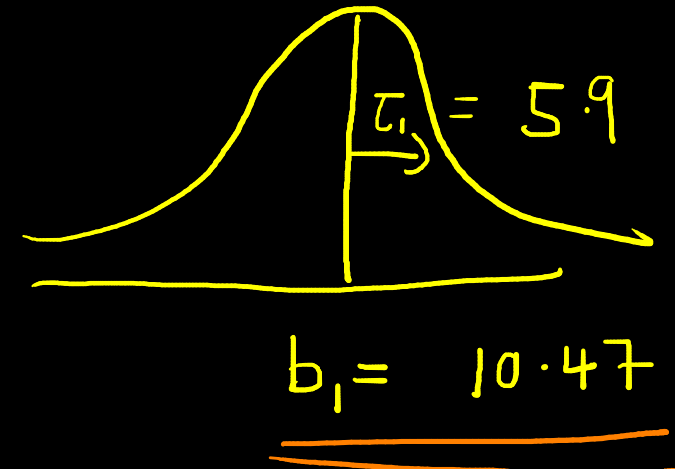
$$\begin{aligned}\mu &= \beta_0 + \beta_1 x \\ &= \beta_0 \times 1 + \beta_1 \times X\end{aligned}$$

intercepts



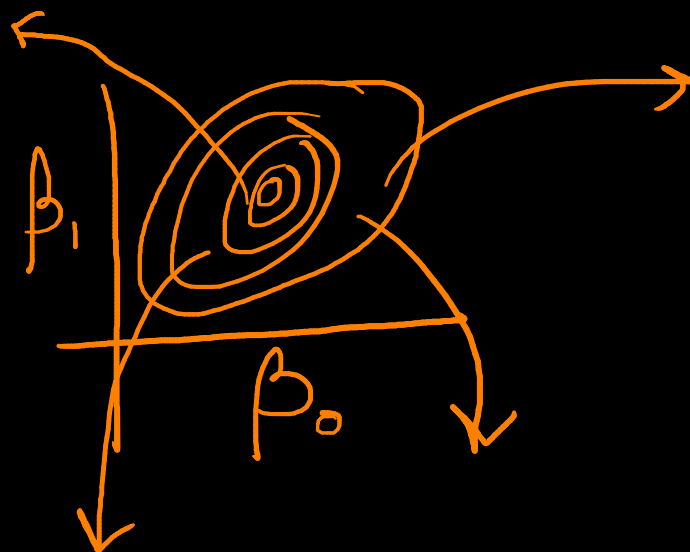
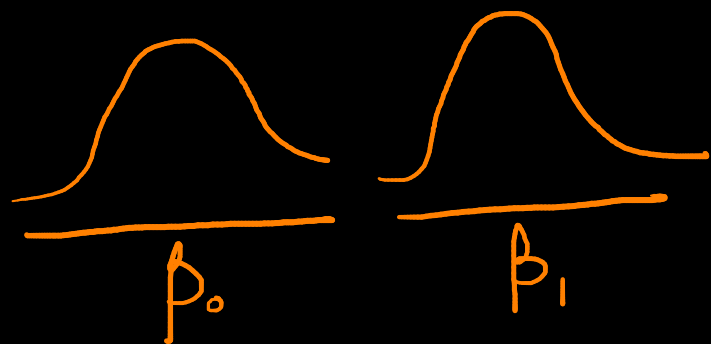
95%  $\approx$  (200, 300)  
interval  
over subject  
specific  
intercepts

slope



10-12, 10+12  
(-2, 22)





$$\begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix} \sim N\left(\begin{bmatrix} b_0 \\ b_1 \end{bmatrix}, \Sigma\right)$$

$$\Sigma = \begin{bmatrix} \sigma_0^2 & \sigma_0 \sigma_1 \rho \\ \sigma_0 \sigma_1 \rho & \sigma_1^2 \end{bmatrix}$$

Annotations for the covariance matrix  $\Sigma$ :

- A green arrow points to the top-left element  $\sigma_0^2$ .
- A green arrow points to the bottom-right element  $\sigma_1^2$ .
- A green arrow points to the top-right element  $\sigma_0 \sigma_1 \rho$ .
- A green arrow points to the bottom-left element  $\sigma_0 \sigma_1 \rho$ .



$x_1 \quad x_2 \quad \dots \quad x_n$

MLE  $\mu$ :  $\bar{x}$

$$s = \sqrt{\frac{(x_i - \bar{x})^2}{n-1}}$$

←

MLE  $\sigma$ :  $s$

REML  $\sigma$ :  $n-1$

$M_0, LL_0, D_0$

$M_1, LL_1, D_1$

$$\boxed{D_0 - D_1} \quad \underline{LRT}$$

$$= -2(LL_0 - LL_1)$$
$$= -2 \log \left( \frac{\text{Likelihood } M_0}{\text{Likelihood } M_1} \right)$$

$$D_0 = -2 LL_0$$

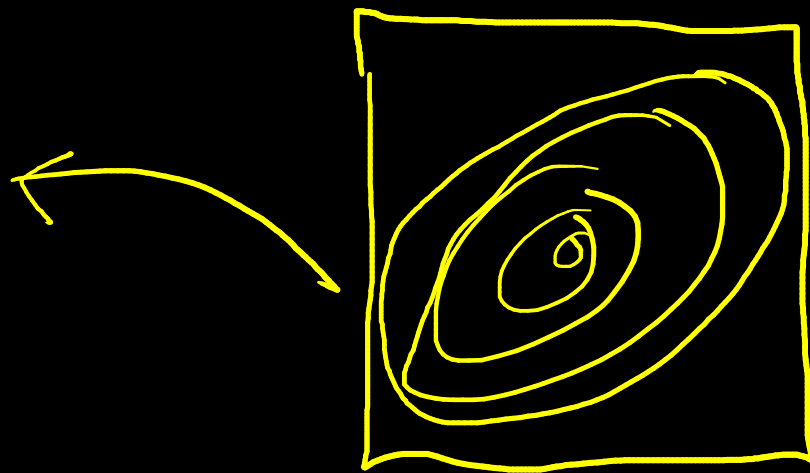
$$LL_0 = \log \text{Likelihood } M_0$$

$$LL_1 = \log \text{Likelihood } M_1$$

$$\log \left( \frac{\text{likelihood } M_0}{\text{likelihood } M_1} \right)$$

$$= \log \text{like } M_0 - \log \text{likelihood } M_1$$

$$\begin{bmatrix} \beta_{0i} \\ \beta_{1i} \end{bmatrix} = \vec{\beta}_i$$



for  $i$  in  $1 \dots n$

$$y_i \sim N(\overset{[s_0]}{\beta_0} + \overset{[s_1]}{\beta_1} x_i, \sigma^2)$$

$$\vec{\beta}_i \sim N(\vec{b}, \Sigma)$$

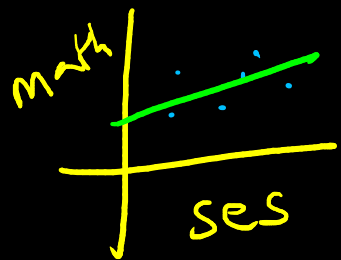
$$\underline{\underline{\vec{\beta}_i = \vec{b} + \vec{Z}_i, \quad \vec{Z}_i \sim N(0, \Sigma)}}$$

$$\rightarrow X \sim N(\mu, \sigma^2)$$

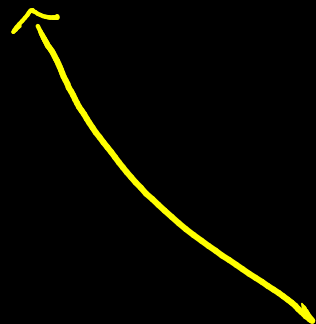
$$\rightarrow X = \mu + \epsilon, \epsilon \sim N(0, \sigma^2)$$



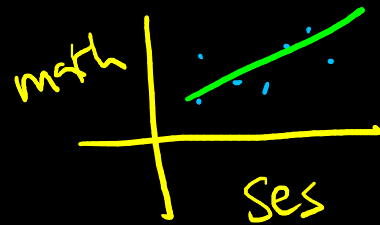
classid = 160



$\vec{\beta}_{160}$



classid = 161



$\vec{\beta}_{161}$

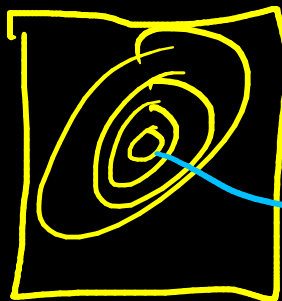
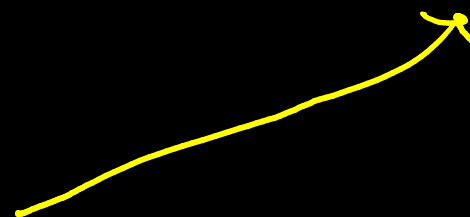


...

classid = J

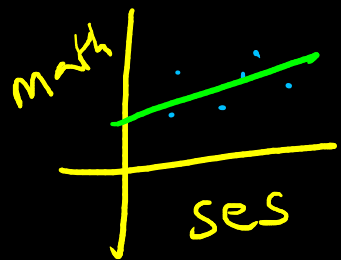


$\vec{\beta}_J$



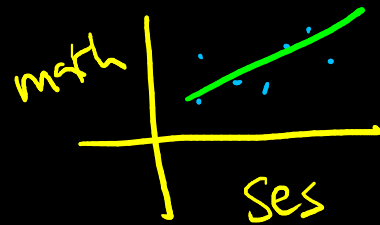
$\vec{\sigma}, \Sigma$

classid = 160



$\vec{\beta}_{160}$

classid = 161



$\vec{\beta}_{161}$

...

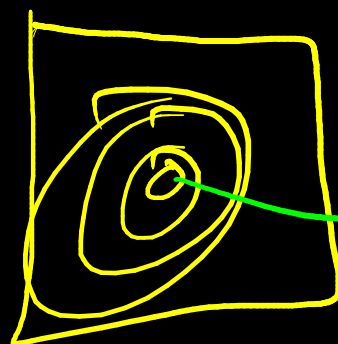
classid = J



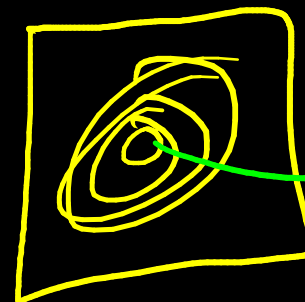
$\vec{\beta}_J$

School 1

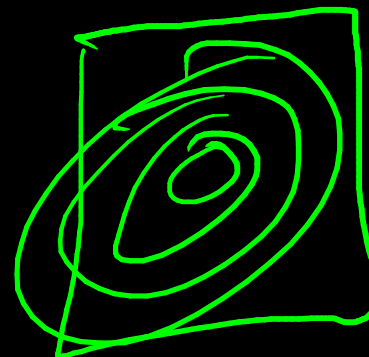
School K



$\vec{\beta}_1$

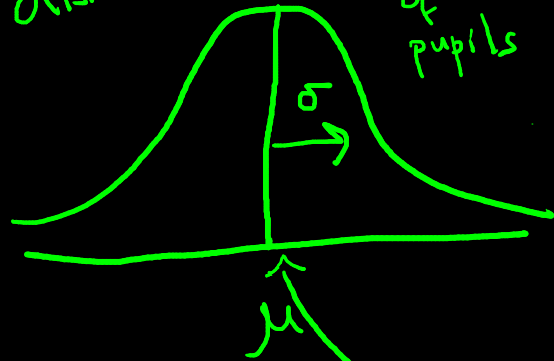


$\vec{\beta}_K$





dist over math scores  
of pupils



class  $i$   
school A

dist over classroom average  
in school A

