

Unbiased

$$E(\hat{\beta}) = \beta$$

→ Finite Sample Property

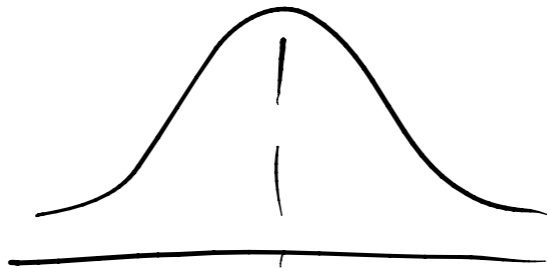
Consistent

$$\text{plim } \hat{\beta} = \beta$$

$$\lim_{n \rightarrow \infty} \text{Prob}(\hat{\beta} - \beta = \varepsilon) = 0$$

$\varepsilon \neq 0, \varepsilon \rightarrow 0$

→ Asymptotic Property  $n \rightarrow \infty$



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$$y_i = \beta_1 + \beta_2 x_i + u_i \quad u_i \sim N(0, \sigma^2)$$

★ Linear combination of normal distributed variable will also be normal distributed.

$$\hat{\beta}_2 = \frac{\sum x_i y_i}{\sum x_i^2} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}$$

$$\hat{\beta}_2 \sim N(\beta_2, \sigma_{\hat{\beta}_2}^2)$$

$$Z = \frac{\hat{\beta}_2 - \beta_2}{\sqrt{\sigma_{\hat{\beta}_2}^2}}$$

$$\sigma^2 \quad \sum (x_i - \mu)^2 \quad s^2 \quad \sum (x_i - \bar{x})^2$$

$$\sigma^2 = \frac{\sum (x_i - \mu)^2}{N}$$

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

$$Y_i = \beta_1 + \beta_2 X_{2i} + \beta_3 X_{3i} + u_i$$

### Individual Test

$$H_0: \beta_2 = 0$$

$$H_a: \beta_2 \neq 0$$

$$t = \frac{\hat{\beta}_2 - \beta_2^0}{\text{se}_{\hat{\beta}_2}}$$

$$H_0: \beta_3 = 0$$

$$H_a: \beta_3 \neq 0$$

$$t = \frac{\hat{\beta}_3 - \beta_3^0}{\text{se}_{\hat{\beta}_3}}$$

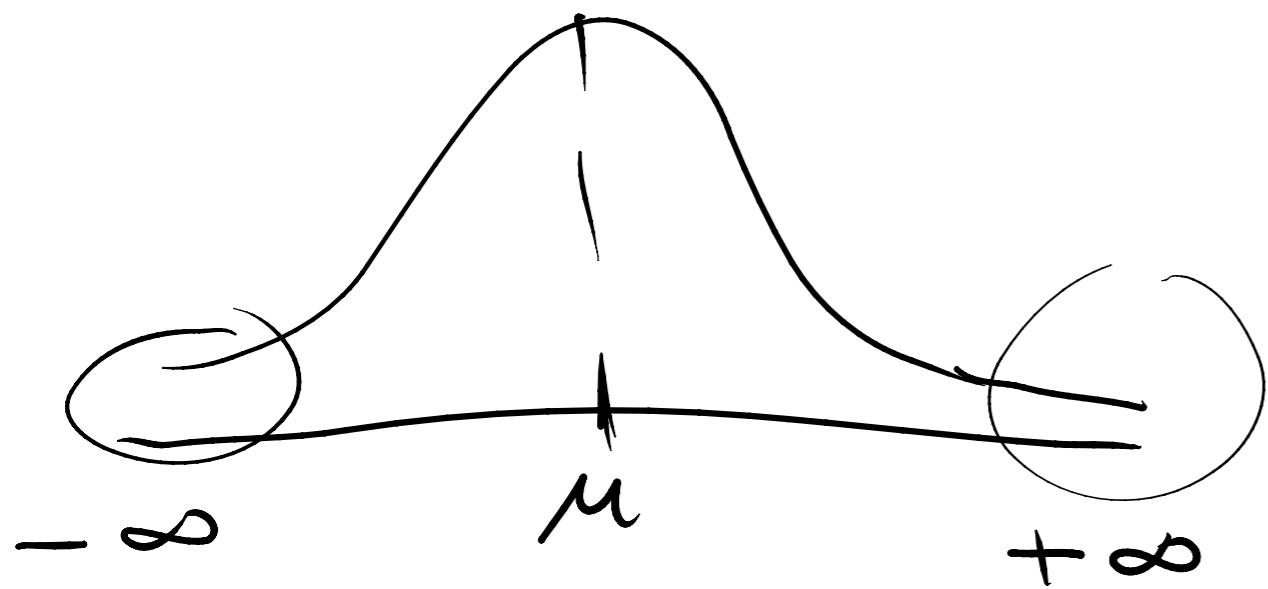
### Overall Test

$$H_0: \beta_2 = \beta_3 = 0$$

$$\beta_2^{\textcircled{1}} = 0 \text{ and } \beta_3^{\textcircled{2}} = 0$$

Chi-square.





$$\mu_1 = 3$$

$$n_2 = 100$$

$$y_i = x_i - \mu$$

$$z_i = \left( \frac{x_i - \mu}{\sigma} \right)$$

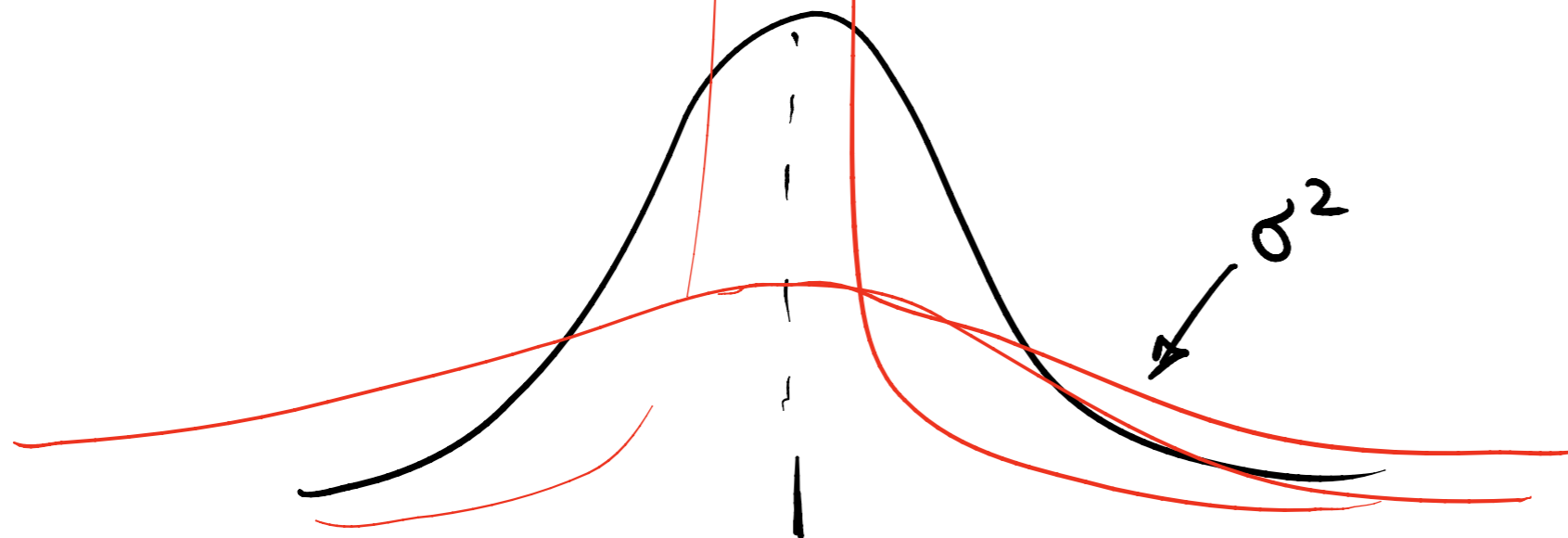
Chi-square.

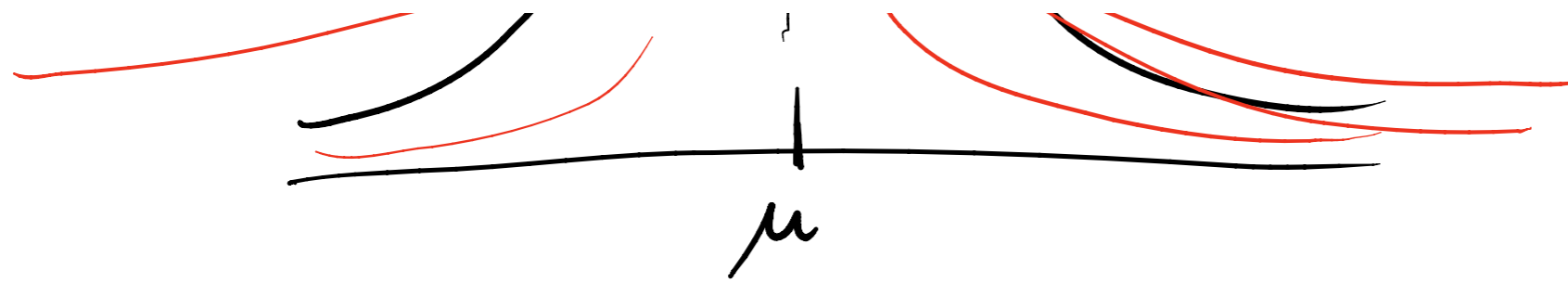


$$X \sim N(\mu, \sigma^2)$$

$$y \sim N(0, \sigma^2)$$

$$Z \sim N(0, 1)$$

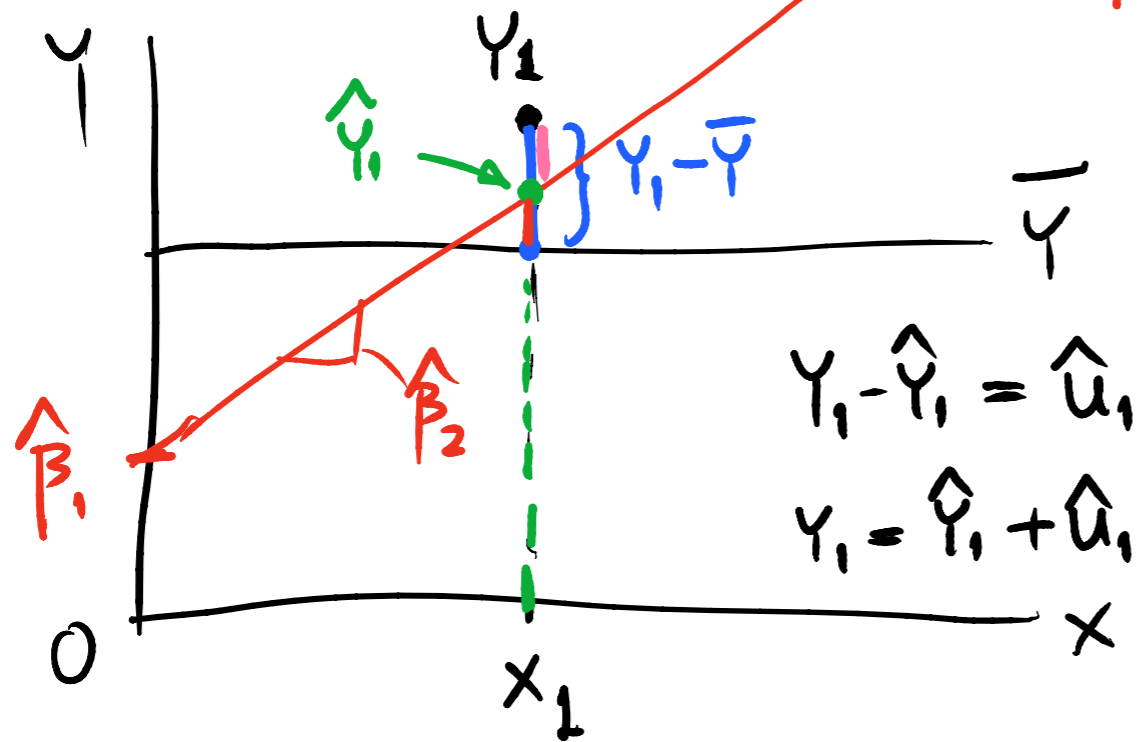




$$y_i = \beta_1 + \beta_2 x_{2i} + \beta_3 x_{3i} + u_i$$

$$\sum_{i=1}^n (y_i - \bar{y})^2 = \text{Total Variation in } Y \quad df = n - 1$$

$$\hat{y}_i = \hat{\beta}_1 + \hat{\beta}_2 x_i$$



$$y_i = \hat{y}_i + \hat{u}_i$$

$$\sum (y_i - \bar{y})^2 = \sum (\hat{y}_i - \bar{y})^2 + \sum (\hat{u}_i - \bar{u})^2$$

$$\sum_{i=1}^n (y_i - \bar{y})^2 = \sum_{i=1}^k (\hat{y}_i - \bar{y})^2 + \sum_{i=1}^{n-k} \hat{u}_i^2$$

Total Sum of Square = Explained SS + Residual SS  
 TSS = ESS + RSS

$$TSS = ESS + RSS$$

$\begin{matrix} n-1 & k-1 & n-k \end{matrix}$

$$F_{(k-1, n-k)} = \frac{(ESS / k-1)}{(RSS / n-k)}$$

13/9/2016

## Multiple Linear Regression

$$Y_i = \beta_0 + \beta_1 X_{2i} + \beta_2 X_{3i} + \dots$$

## Simple Linear Regression Model.

$$Y_i = \beta_1 + \beta_2 X_i + u_i$$

OLS  $\rightarrow \hat{\beta}_1, \hat{\beta}_2$

### Evaluate

1. Sign & Meaning of Coefficients
  2. Overall Test - F-test
  3.  $R^2$
  4. Individual Test - t-test
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## Multiple Linear Regression

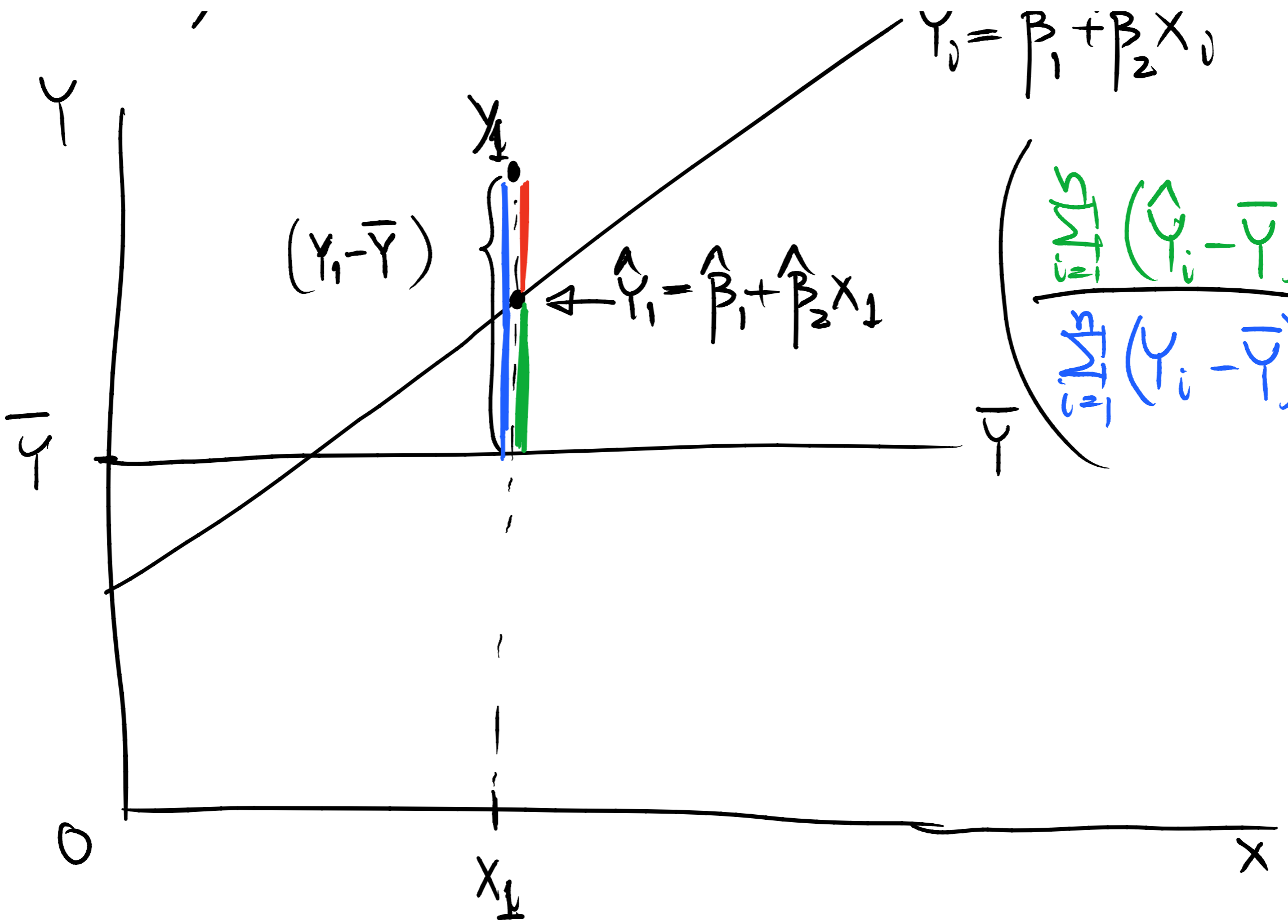
$$Y_i = \beta_1 + \beta_2 X_{2i} + \beta_3 X_{3i} + \dots + \beta_k X_{ki} + u_i$$

### Covariance

$$\sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x})$$

$$\Delta x \rightarrow \Delta y$$

$$\hat{Y}_i = \hat{\beta}_1 + \hat{\beta}_2 X_i$$



$$\left( \frac{\sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2}{\sum_{i=1}^n (Y_i - \bar{Y})^2} \right) = R^2$$