

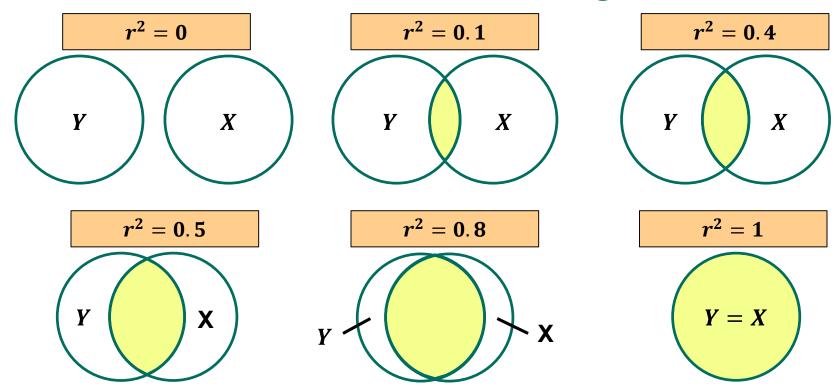
## Lesson Goal

• Assess the goodness of fit of regression models.

## Coefficient of Determination $r^2$

- We look at the goodness of fit of the fitted regression line to a set of data.
- The coefficient of determination  $r^2$  (two-variable case) or  $R^2$  (multiple regression) is a summary measure of how well the sample regression line fits the data.

# Coefficient of Determination $r^2$ - Venn Diagram/Ballentine



Source: Peter Kennedy, "Ballentine: A Graphical Aid for Econometrics," Australian Economics Papers, vol. 20, 1981, pp. 414–416.

## Coefficient of Determination $r^2$ - Proof

Recall, 
$$Y_i = \hat{Y}_i + \hat{U}_i$$
  
beviation form,  $y_i = \hat{y}_i + \hat{u}_i$   
where  $y_i = Y_i - Y$ ,  $\hat{y}_i = \hat{Y}_i - Y$ ,  $\hat{u}_i = \hat{U}_i - 0$   
 $= Y_i - Y_i + \hat{u}_i Y_i^2$ 

$$\Sigma y_{i}^{2} = \Sigma (\hat{y}_{i} + \hat{u}_{i})^{2}$$

$$\Sigma y_{i}^{2} = \Sigma (\hat{y}_{i}^{2} + 2\hat{y}_{i}\hat{u}_{i} + \hat{u}_{i}^{2})$$

$$\Sigma y_{i}^{2} = \Sigma (\hat{y}_{i}^{2} + 2\Sigma \hat{y}_{i}\hat{u}_{i} + \Sigma \hat{u}_{i}^{2})$$

$$\Sigma y_{i}^{2} = \Sigma \hat{y}_{i}^{2} + 2\Sigma \hat{y}_{i}\hat{u}_{i} + \Sigma \hat{u}_{i}^{2}$$

$$\Omega t, \Sigma \hat{y}_{i}\hat{u}_{i} = 0$$

$$\sum y_i^2 = \sum \hat{y}_i^2 + \sum \hat{u}_i^2$$

TSS 
$$\Leftarrow$$
  $\Sigma y_i^2 = \Sigma (Y_i - Y)^2$  — the total variation of the actual Y values about their sample means squares

Ess = 
$$\Sigma \hat{y}_{i}^{2} = \Sigma (\hat{y}_{i} - \hat{y}_{i})^{2}$$
 regression)

Ess =  $\Sigma \hat{y}_{i}^{2} = \Sigma (\hat{y}_{i} - \hat{y}_{i})^{2}$  regression very due to regression/

sum of squares explained by regression)

#### **ECONOMETRICS**

## Coefficient of Determination $r^2$ - Proof

$$\Sigma y_i^2 = \Sigma y_i^2 + \Sigma y_i^2$$

$$TSS = ESS + RSS$$

$$TSS = ESS + RSS$$

$$TSS = TSS + RSS$$

$$I = \frac{ESS}{TSS} + \frac{RSS}{TSS}$$

$$I = \frac{ESS}{TSS} + \frac{RSS}{TSS}$$

$$I = \frac{ESS}{TSS} = \frac{\Sigma (\hat{Y}_i - \hat{Y}_i)^2}{\Sigma (\hat{Y}_i - \hat{Y}_i)^2}$$

$$I = r^2 + \frac{RSS}{TSS}$$

$$I = r^2 + \frac{RSS}{TSS}$$

$$r^{2} = 1 - \frac{RSS}{TSS}$$

$$= 1 - \frac{\Sigma \hat{V}_{i}^{2}}{\Sigma (Y_{i} - \bar{Y}_{i})^{2}}$$

$$= 1 - \frac{\Sigma (Y_{i} - \hat{Y}_{i})^{2}}{\Sigma (Y_{i} - \bar{Y}_{i})^{2}}$$

## Coefficient of Determination $r^2$ - Definition

•  $r^2$  measures the proportion or percentage of the total variation in Y explained by the variation in X in the regression model.

$$r^2 = \frac{ESS}{TSS} = \frac{\sum (\widehat{Y}_i - \overline{Y})^2}{\sum (Y_i - \overline{Y})^2}$$

$$r^2 = 1 - \frac{RSS}{TSS} = 1 - \frac{\sum (Y_i - \widehat{Y}_i)^2}{\sum (Y_i - \overline{Y})^2}$$

- Simply put,  $r^2$  is the ratio of explained variation to the total variation.
  - $\circ$   $r^2$  lies between 0 and 1.

## Correlation Coefficient (r)

• Another formula for  $r^2$  is:

$$\mathbf{r^2} = \frac{(\sum x_i y_i)^2}{\sum x_i^2 \cdot \sum y_i^2}$$

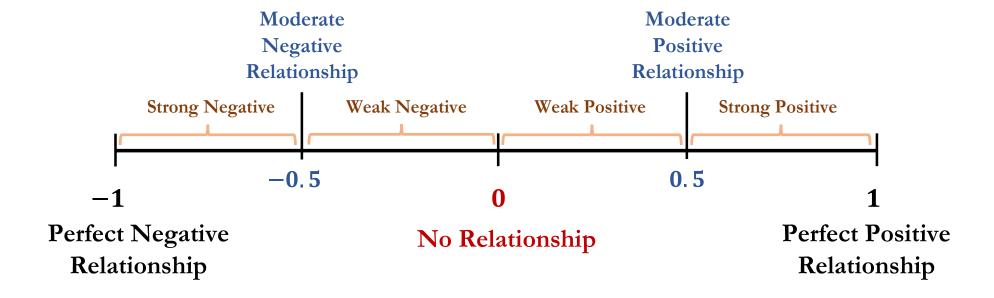
• The square root of  $r^2$  is the correlation coefficient r.

$$r = \pm \sqrt{r^2}$$

$$r = \frac{\sum x_i y_i}{\sqrt{\sum x_i^2 \cdot \sum y_i^2}} = \frac{\sum (X_i - \overline{X})(Y_i - \overline{Y})}{\sqrt{\sum (X_i - \overline{X})^2 \cdot \sum (Y_i - \overline{Y})^2}}$$

## Correlation Coefficient (r)

• Correlation coefficient r measures the strength and direction of the relationship between two variables. It ranges from -1 to 1.



# THANK YOU!

Next Lesson: Numerical Example of Regression
Analysis

**ECONOMETRICS**