

How well the line fit the data?

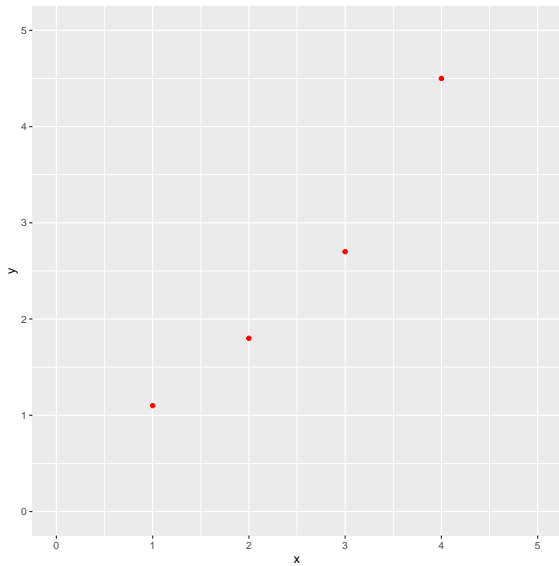
- ▶ Not every data can be fitted by a linear model
- ▶ We can measure a goodness of the fit in three ways
- ▶ Coefficient of Determination or R^2
- ▶ F-test
- ▶ t-test

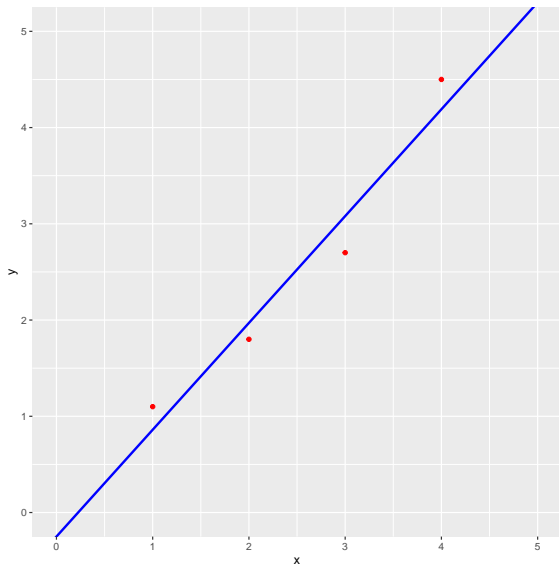
Coefficient of Determination

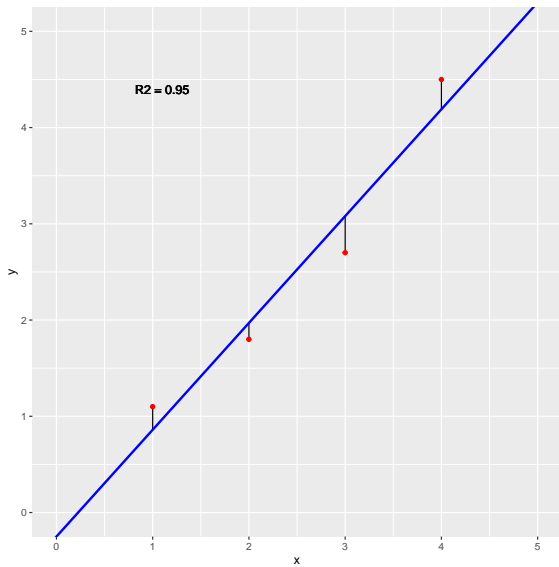
$$R^2 = \frac{RegSS}{TSS} = 1 - \frac{RSS}{TSS}$$

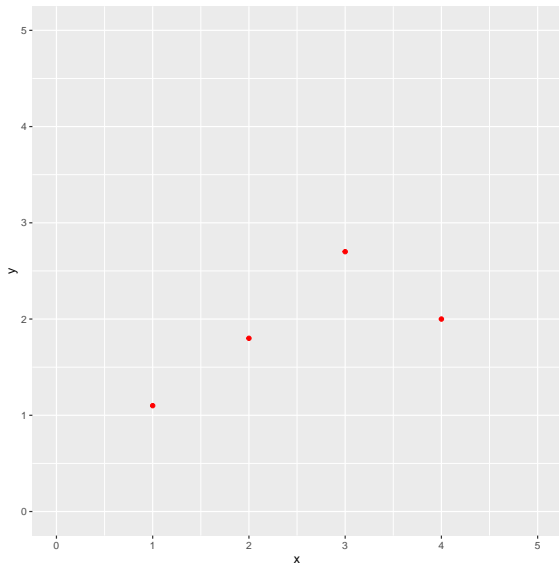
- ▶ RSS: Residual Sums Squares
- ▶ TSS: Total Sums Squares

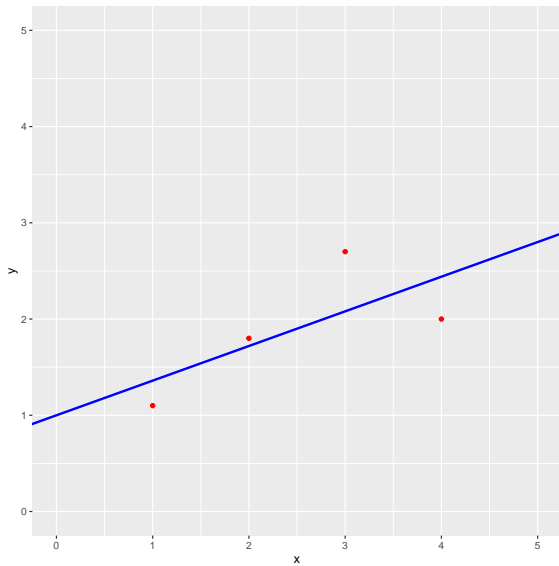
- ▶ If $RSS = 0$, $R^2 = 1$. This is a perfect fit.
- ▶ If $RSS = TSS$, $R^2 = 0$ means. This is the lowest R^2 could be.
- ▶ The closer R^2 to 1, the better the fit.

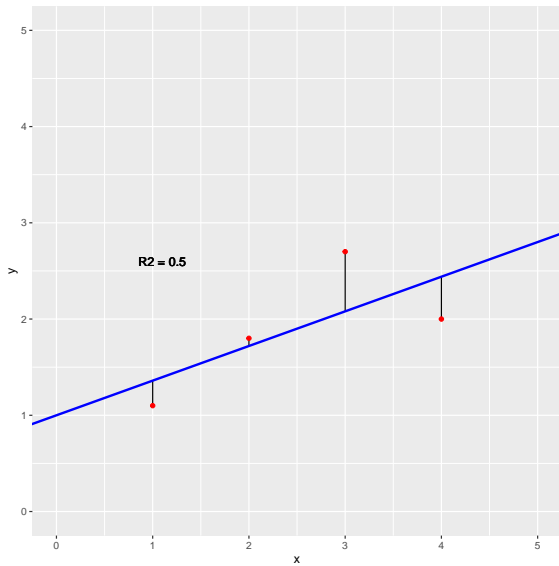












► Some examples of R^2

F-test

$$H_0 : \beta_1 = 0$$

$$H_0 : y = \beta_0 + \epsilon$$

H_0 : The linear model is not a good fit

$$H_\alpha : \beta_1 \neq 0$$

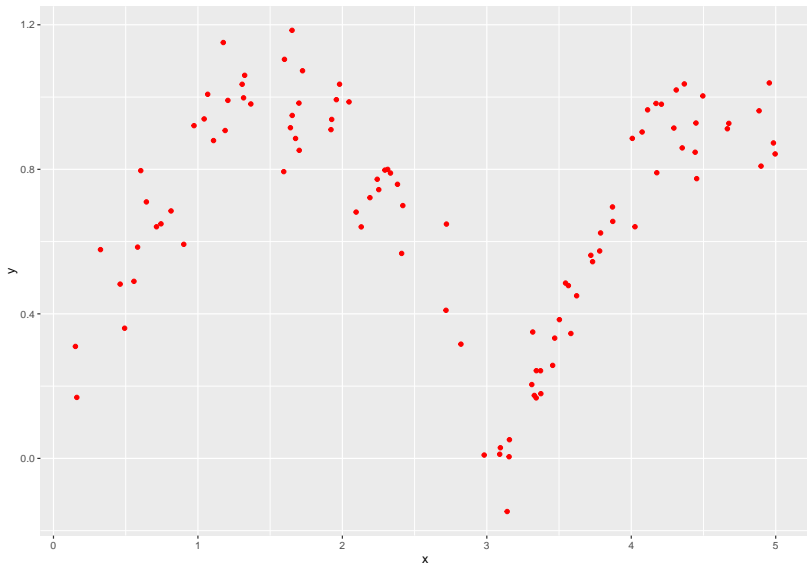
$$H_\alpha : y = \beta_0 + \beta_1 x + \epsilon$$

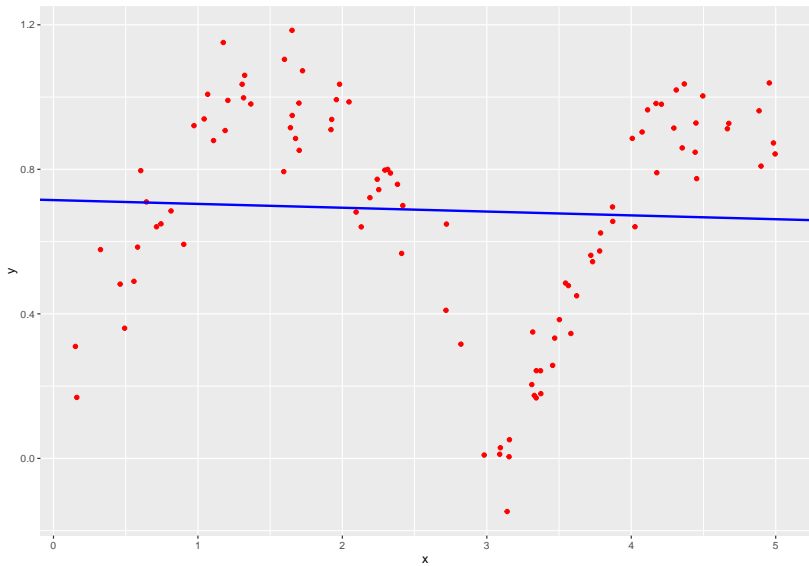
H_α : The linear model is a good fit

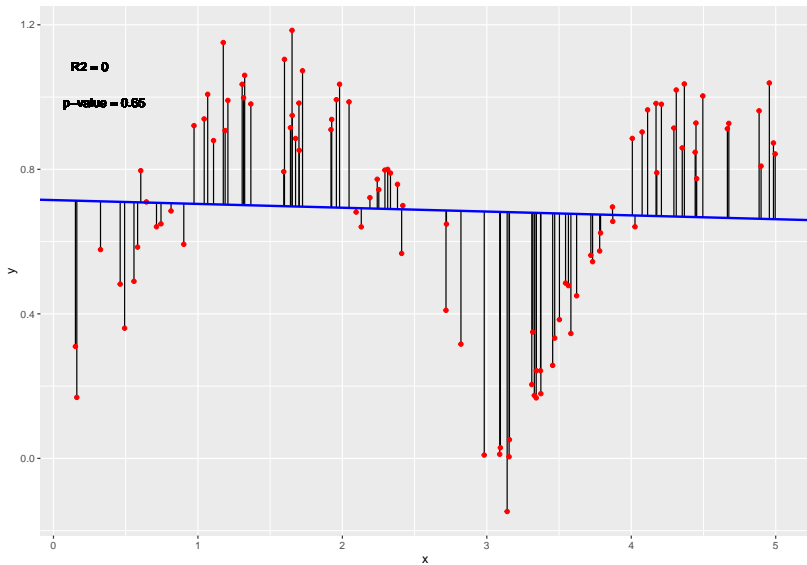
$$F = \frac{RegSS}{RSS/(n-2)}$$

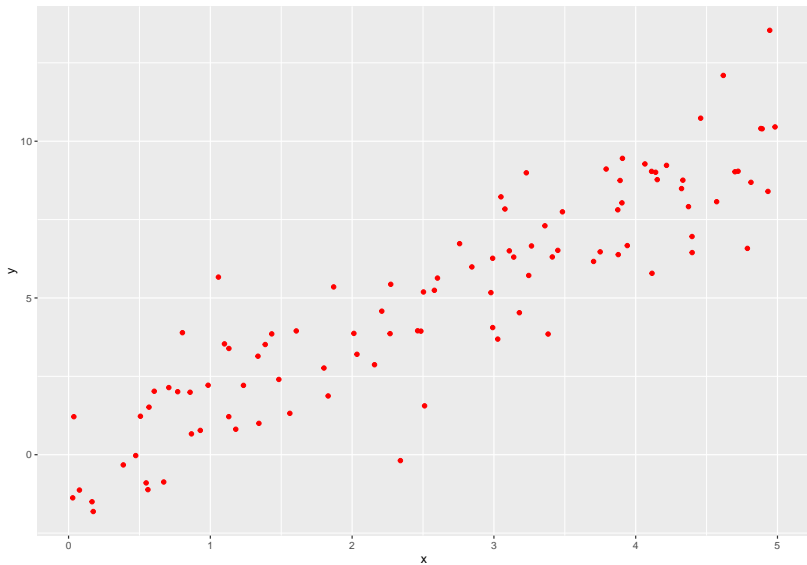
These are equivalent

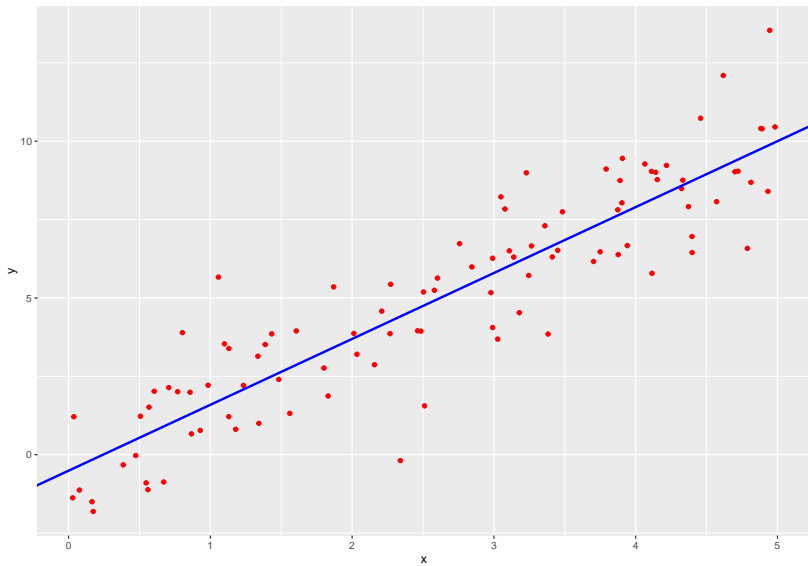
- ▶ Larger F
- ▶ Larger Reg SS
- ▶ Smaller Residual SS (RSS)
- ▶ Smaller p-value
- ▶ More support for H_α or More useful the model

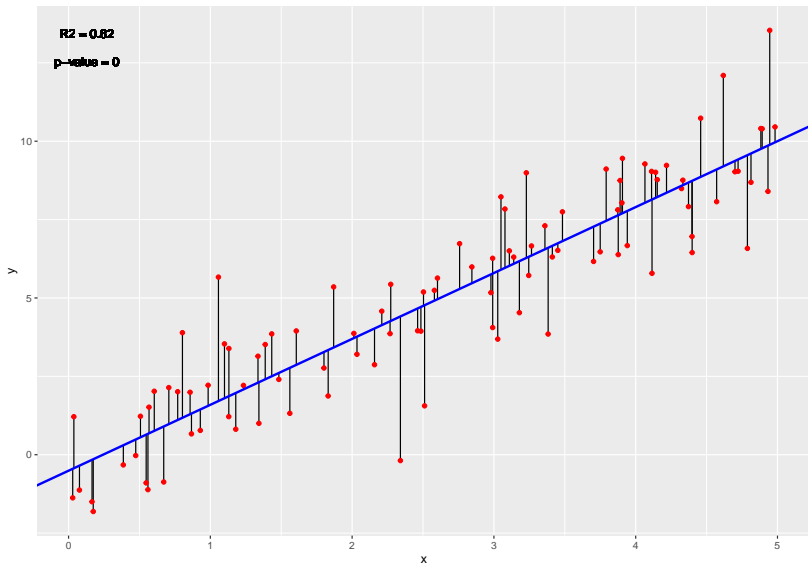












t-test

$$H_0 : \beta_1 = 0$$

$$H_0 : y = \beta_0 + \epsilon$$

H_0 : The linear model is not a good fit

$$H_\alpha : \beta_1 \neq 0$$

$$H_\alpha : y = \beta_0 + \beta_1 x + \epsilon$$

H_α : The linear model is a good fit

$$t = \frac{\hat{\beta}_1}{sd(\hat{\beta}_1)}$$

Relation between F-test, t-test and and R2

► We have

$$F = t^2 = \frac{(n-2)R^2}{(1-R^2)}$$

$$\Rightarrow \frac{1}{F} = \left(\frac{1}{R^2} - 1 \right) \cdot \frac{1}{n-2}$$

- This means when $R^2 \nearrow$, $\frac{1}{R^2} \searrow$, $\frac{1}{F} \searrow$ and $F \nearrow$.
- The p-values of F-test and t-test are the same.