

15.5: Testing for significance

→ Model : $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_p X_p + \epsilon$

$H_0 : \beta_1 = \beta_2 = \dots = \beta_p = 0$

$H_1 : \text{Not all } \beta_j \text{ are zero.}$

F-test

→ ANOVA table :

Source of Variation	df	SS	MS	F	F _α	p-value
Regression	p	SSR	MSR	F	F _α	p-value
Error	n-p-1	SSE	MSE			
Total	n-1	SST				

• $MSR = \frac{SSR}{p}$

• $MSE = \frac{SSE}{n-p-1}$

• $F = \frac{MSR}{MSE}$ with $df_1 = p$ and $df_2 = n-p-1$

• Reject H_0 if $F \geq F_{\alpha}$ or p-value $\leq \alpha$.

F-test only 1-times

H_0 reject
او

do significance of the Model

By F-test

$n \neq p+1$
 sample size
 $n=6$
 $p=5$
 $\rightarrow n-p-1 = 0$
 وهو خطأ
 يجب ان يكون أكبر من
 نسبة د م

$$\rightarrow H_0: \beta_j = 0$$

$$H_1: \beta_j \neq 0$$

t-test

5% significance level variable α

By t-test also valid

→ Test statistic:

$$t = \frac{b_j}{s_{b_j}}$$

→ Rejection Rule:

Reject H_0 if $|t| \geq t_{\alpha/2}$ or p-value $\leq \alpha$

with $df = n - p - 1$

→ p-times

p-times, $\alpha/2$

→ Remark:

$$y = \beta_0 + \sum_{j=1}^p \beta_j X_j + \epsilon$$

• SSR with $p \rightarrow MSR = \frac{SSR}{p}$

• SSE with $n-p-1 \rightarrow s^2 = MSE = \frac{SSE}{n-p-1}$

• $F = \frac{MSR}{MSE}$, F_{α} with $df_1 = p$ and $df_2 = n-p-1$

• $t^{(j)} = \frac{b_j}{s_{b_j}}$, b_j : estimator for β_j

• $S = \sqrt{S^2} = \sqrt{MSE}$: standard error of the estimate.

• $\sigma_{b_j} = \sqrt{\text{Var}(b_j)}$: standard deviation of b_j .

• $S_{b_j} =$ estimated standard deviation of b_j

Remark :

Excel :
- multiple $R = \sqrt{R^2}$
- standard error = $S = \sqrt{MSE}$.
- significance $F =$ p-value of F-test.
} on Regression statistic.

on Exp : p-value = 0.0003 < $\alpha = 0.01$

\Rightarrow so Reject $H_0 (\beta_1 = \dots = \beta_p = 0) (\alpha = 0.01)$

\Rightarrow The Model is significance.

on Table of coefficients.

- standard error : estimated standard deviation of $b_j = S_{b_j}$.

- $t^j = \frac{b_j}{S_{b_j}}$ with $df = n - p - 1$.

- $(1 - \alpha) CI \beta_j = b_j \pm t_{\frac{\alpha}{2}} S_{b_j}$ with $df = n - p - 1$.

exp on Excel : on table of coeff.

Sheet 3

$$H_0^1 : \beta_1 = 0 \rightarrow p\text{-value} = 0.0005 < \alpha = 0.01$$

\rightarrow Reject $H_0^1 (\alpha = 0.01)$

$$\rightarrow \beta_1 \neq 0 (\alpha = 0.01)$$

$\rightarrow X_1$ significance variable. ($\alpha = 0.01$)

$$H_0^2 : \beta_2 = 0 \rightarrow p\text{-value} = 0.0004 < \alpha = 0.01$$

\rightarrow Reject $H_0^2 (\alpha = 0.01)$

$$\rightarrow \beta_2 \neq 0 (\alpha = 0.01)$$

$\rightarrow X_2$ significance variable ($\alpha = 0.01$)

\rightarrow Multicollinearity .

input variable X_1, X_2, \dots, X_p

Some times some X_i is dependent on the other X_j s, this case is

Known as multicollinearity.

بعض المتغيرات input تعتمد على الأخرى

\rightarrow Variance inflation factor (VIF) .

$$VIF = \frac{1}{1 - R_i^2}$$

R_i^2 : Multiple coefficient of determination for X_i as a function of the other X_j s .

• function means multiple regression .

وقتها يتم إبعاد X_i عن النموذج $VIF(X_i) \geq 10 \Rightarrow X_i$ should be eliminated.

Significance level $\alpha = 0.01$

Model: $y = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p + \varepsilon$.

* Goodness of fit of Model.

* Significance of Model Variable.

* Validity of assumption.

طرق التقييم في الانحدار

Model

Multicollinearity $VIF(x_i) = \frac{1}{1-R_i^2}$

R_i^2 : Goodness of fit ($x_i = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p + \varepsilon$).

on ex9 excel (sheet 6).

$$VIF(x_2) = \frac{1}{1-R_2^2} = \frac{1}{1-0.03} = 1.03$$

$$\hat{x}_2 = 2.32 + 0.007 x_1$$

$$VIF(x_1) = \frac{1}{1-R_1^2} = \frac{1}{1-0.03} = 1.03$$

$$\hat{x}_1 = 69.49 + 3.62(x_2)$$

$VIF(x_2) < 10$: there is No collinearity between x_2 and the other variables (x_1).

$VIF(x_1) < 10$: there is No collinearity between x_1 and the other variable (x_2).