

ECON 351* -- NOTE 19

F-Tests of Exclusion Restrictions on Regression Coefficients:
Numerical Examples 2

Consider two alternative regression models of North American car prices.

- The **sample data** consist of **74 observations** on the following variables:

price_i = the price of the i-th car (in US dollars);
 wgt_i = the weight of the i-th car (in pounds);
 mpg_i = the fuel efficiency of the i-th car (in miles per gallon);
 N = 74 = the number of observations in the estimation sample.

Data Source: Stata-format data set **auto.dta** supplied with *Stata Release 8*.

- ◆ **Model 1:** is given by the PRE

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + u_i \tag{1}$$

This model contains two explanatory variables, wgt_i and mpg_i.

Model 1 assumes that the *marginal effect of each explanatory variable* is a *constant*; it is *linear* in the explanatory variables wgt_i and mpg_i.

marginal effect of wgt_i = $\partial \text{price}_i / \partial \text{wgt}_i = \beta_1 = \text{a constant}$;

marginal effect of mpg_i = $\partial \text{price}_i / \partial \text{mpg}_i = \beta_2 = \text{a constant}$.

. regress price wgt mpg

Source	SS	df	MS	Number of obs = 74		
Model	186321280	2	93160639.9	F(2, 71)	=	14.74
Residual	448744116	71	6320339.67	Prob > F	=	0.0000
				R-squared	=	0.2934
				Adj R-squared	=	0.2735
Total	635065396	73	8699525.97	Root MSE	=	2514.0

price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
wgt	1.746559	.6413538	2.723	0.008	.4677361	3.025382
mpg	-49.51222	86.15604	-0.575	0.567	-221.3025	122.278
_cons	1946.069	3597.05	0.541	0.590	-5226.244	9118.382

◆ **Model 2:** is given by the PRE

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

This model also contains the two explanatory variables wgt_i and mpg_i .

But Model 2 specifies that the *marginal effect of each explanatory variable is a linear function of both wgt_i and mpg_i* ; Model 2 is *nonlinear* in the explanatory variables wgt_i and mpg_i .

marginal effect of wgt_i =

$$\partial \text{price}_i / \partial \text{wgt}_i = \beta_1 + 2\beta_3 \text{wgt}_i + \beta_5 \text{mpg}_i = \text{a linear function of } \text{wgt}_i \text{ and } \text{mpg}_i;$$

marginal effect of mpg_i =

$$\partial \text{price}_i / \partial \text{mpg}_i = \beta_2 + 2\beta_4 \text{mpg}_i + \beta_5 \text{wgt}_i = \text{a linear function of } \text{wgt}_i \text{ and } \text{mpg}_i.$$

• The *unrestricted OLS SRE* is written in general as

$$\text{price}_i = \hat{\beta}_0 + \hat{\beta}_1 \text{wgt}_i + \hat{\beta}_2 \text{mpg}_i + \hat{\beta}_3 \text{wgt}_i^2 + \hat{\beta}_4 \text{mpg}_i^2 + \hat{\beta}_5 \text{wgt}_i \text{mpg}_i + \hat{u}_i \quad (2^*)$$

where the $\hat{\beta}_j$ ($j = 0, \dots, 5$) are the *unrestricted OLS coefficient estimates* and the \hat{u}_i ($i = 1, \dots, N$) are the *unrestricted OLS residuals*.

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. regress price wgt mpg wgtsq mpgsq wgtmpg
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Source	SS	df	MS	Number of obs =	74
Model	308384833	5	61676966.6	F(5, 68) =	12.84
Residual	326680563	68	4804125.93	Prob > F =	0.0000
				R-squared =	0.4856
				Adj R-squared =	0.4478
Total	635065396	73	8699525.97	Root MSE =	2191.8

price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
wgt	-31.88985	9.148215	-3.486	0.001	-50.14483 -13.63487
mpg	-3549.495	1126.464	-3.151	0.002	-5797.318 -1301.672
wgtsq	.0034574	.0008629	4.007	0.000	.0017355 .0051792
mpgsq	38.74472	12.62339	3.069	0.003	13.55514 63.9343
wgtmpg	.5421927	.1971854	2.750	0.008	.1487154 .9356701
_cons	92690.55	25520.53	3.632	0.001	41765.12 143616

- Recall from *Note 13* the **two alternative forms of the ANOVA F-statistic**:

$$\text{ANOVA} - F = \frac{\text{ESS}_1 / (K - 1)}{\text{RSS}_1 / (N - K)} = \frac{\text{ESS}_1 (N - K)}{\text{RSS}_1 (K - 1)} \quad (\text{A1})$$

$$\text{ANOVA} - F = \frac{R_U^2 / (K - 1)}{(1 - R_U^2) / (N - K)} = \frac{R_U^2 (N - K)}{(1 - R_U^2) (K - 1)} \quad (\text{A2})$$

- Recall from *Note 17* the **two alternative forms of the general F-statistic**:

$$F = \frac{(\text{RSS}_0 - \text{RSS}_1) / (\text{df}_0 - \text{df}_1)}{\text{RSS}_1 / \text{df}_1} = \frac{(\text{RSS}_0 - \text{RSS}_1)}{\text{RSS}_1} \frac{\text{df}_1}{(\text{df}_0 - \text{df}_1)}. \quad (\text{F1})$$

$$F = \frac{(R_U^2 - R_R^2) / (\text{df}_0 - \text{df}_1)}{(1 - R_U^2) / \text{df}_1} = \frac{(R_U^2 - R_R^2)}{(1 - R_U^2)} \frac{\text{df}_1}{(\text{df}_0 - \text{df}_1)}. \quad (\text{F2})$$

$\text{df}_0 = N - K_0 =$ degrees of freedom for the restricted RSS, RSS_0 ;

$\text{df}_1 = N - K_1 = N - K =$ degrees of freedom for the unrestricted RSS, RSS_1 ;

$\text{df}_0 - \text{df}_1 = N - K_0 - (N - K_1) = N - K_0 - N + K_1 = K_1 - K_0 = K - K_0$;

$K_0 =$ the number of free regression coefficients in the restricted model;

$K = K_1 =$ the number of free regression coefficients in the unrestricted model.

- Null distribution*** of F-statistic:

$$F \sim F[\text{df}_0 - \text{df}_1, \text{df}_1] = F[K - K_0, N - K] \text{ under } H_0.$$

1. The ANOVA F-Test of Zero Restrictions on All Slope Coefficients in Model 1

- ◆ **Model 1** is given by the population regression equation (PRE):

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + u_i \quad (1)$$

- ◆ The *null hypothesis* is:

$$H_0: \beta_j = 0 \quad \forall j = 1, 2; \quad \beta_1 = 0 \text{ and } \beta_2 = 0.$$

- ◆ The *alternative hypothesis* is:

$$H_1: \beta_j \neq 0 \quad j = 1, 2; \quad \beta_1 \neq 0 \text{ and/or } \beta_2 \neq 0.$$

Test statistic: For this test, we can use *either* the ANOVA F-statistic *or* the general F-statistic.

ANOVA F-test: uses *only* the *unrestricted* OLS SRE.

- The **unrestricted model** corresponding to the alternative hypothesis H_1 is simply PRE (1):

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + u_i \quad (1)$$

Number of free (unrestricted) regression coefficients in (1) is $\mathbf{K} = \mathbf{K}_1 = 3$.

- OLS estimation of the unrestricted model (1) yields the ***unrestricted* OLS SRE:**

$$\text{price}_i = \hat{\beta}_0 + \hat{\beta}_1 \text{wgt}_i + \hat{\beta}_2 \text{mpg}_i + \hat{u}_i \quad (1^*)$$

(1) The **OLS decomposition equation for the *unrestricted* OLS SRE** is

$$\text{TSS} = \text{ESS}_1 + \text{RSS}_1 .$$

$$(N-1) \quad (K-1) \quad (N-K)$$

$$\begin{aligned} \text{TSS} &= 635065396.0 && \text{with } df = N - 1 = 74 - 1 = 73 \\ \text{ESS}_1 &= 186321280.0 && \text{with } df = K - 1 = 3 - 1 = 2 \\ \text{RSS}_1 &= \mathbf{448744116.0} && \text{with } df_1 = N - K = \mathbf{74 - 3 = 71} \end{aligned}$$

(2) The **R^2 for the *unrestricted* OLS SRE** is

$$R_U^2 = \frac{\text{ESS}_1}{\text{TSS}} = 1 - \frac{\text{RSS}_1}{\text{TSS}} = \mathbf{0.2934}.$$

- Compute the *sample value* of either of the ANOVA F-statistics.

$$F_0 = \frac{\text{ESS}_1/(K-1)}{\text{RSS}_1/(N-K)} = \frac{186321280/(3-1)}{448744116/(74-3)} = \frac{186321280/2}{448744116/71} = \mathbf{14.74}$$

$$F_0 = \frac{R_U^2/(K-1)}{(1-R_U^2)/(N-K)} = \frac{0.2934/(3-1)}{(1-0.2934)/(74-3)} = \frac{0.2934/2}{0.7066/71} = \mathbf{14.74}$$

- *Null distribution* of the F-statistic: $\mathbf{F} \sim \mathbf{F[2,71]}$ under H_0 .

The *critical values* of the **F[2,71]-distribution** are:

$$\text{for } \alpha = 0.05: \quad F_{\alpha}[2,71] = F_{0.05}[2,71] = 3.13.$$

$$\text{for } \alpha = 0.01: \quad F_{\alpha}[2,71] = F_{0.01}[2,71] = 4.92.$$

- **Inference at $\alpha = 0.05$ (5% significance level):**

Since $F_0 = 14.74 > 3.13 = F_{0.05}[2,71]$, **reject H_0** at the **5%** significance level.

- **Inference at $\alpha = 0.01$ (1% significance level):**

Since $F_0 = 14.74 > 4.92 = F_{0.01}[2,71]$, **reject H_0** at the **1%** significance level.

General F-test: uses *both the unrestricted and restricted OLS SREs*.

$$F = \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} = \frac{(RSS_0 - RSS_1)}{RSS_1} \frac{df_1}{(df_0 - df_1)} \quad (\text{F1})$$

$$F = \frac{(R_U^2 - R_R^2)/(df_0 - df_1)}{(1 - R_U^2)/df_1} = \frac{(R_U^2 - R_R^2)}{(1 - R_U^2)} \frac{df_1}{(df_0 - df_1)} \quad (\text{F2})$$

- The **unrestricted model** corresponding to the *alternative hypothesis* H_1 is simply PRE (1):

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + u_i \quad (1)$$

Number of free (unrestricted) regression coefficients in (1) is $K = K_1 = 3$.

- OLS estimation of the unrestricted model yields the **unrestricted OLS SRE**:

$$\text{price}_i = \hat{\beta}_0 + \hat{\beta}_1 \text{wgt}_i + \hat{\beta}_2 \text{mpg}_i + \hat{u}_i \quad (1^*)$$

(1) The **OLS decomposition equation for the unrestricted OLS SRE** is

$$\begin{array}{l} \text{TSS} = \text{ESS}_1 + \text{RSS}_1 \\ (N-1) \quad (K-1) \quad (N-K) \end{array}$$

$$\text{TSS} = 635065396.0 \quad \text{with} \quad df = N - 1 = 74 - 1 = 73$$

$$\text{ESS}_1 = 186321280.0 \quad \text{with} \quad df = K - 1 = 3 - 1 = 2$$

$$\text{RSS}_1 = 448744116.0 \quad \text{with} \quad df_1 = N - K = 74 - 3 = 71$$

(2) The R^2 for the **unrestricted OLS SRE** is

$$R_U^2 = \frac{\text{ESS}_1}{\text{TSS}} = 1 - \frac{\text{RSS}_1}{\text{TSS}} = \mathbf{0.2934}.$$

- The ***restricted model*** corresponding to the ***null hypothesis*** H_0 is obtained by setting $\beta_1 = 0$ and $\beta_2 = 0$ in the unrestricted model (1):

$$\text{price}_i = \beta_0 + u_i \quad (0)$$

Number of free (unrestricted) regression coefficients is $K_0 = 1$.

- OLS estimation of the restricted model yields the ***restricted OLS SRE***:

$$\text{price}_i = \tilde{\beta}_0 + \tilde{u}_i \quad (0^*)$$

(1) The **OLS decomposition equation for the *restricted OLS SRE*** is

$$\begin{array}{l} \text{TSS} = \text{ESS}_0 + \text{RSS}_0 \\ (N-1) \quad (K_0-1) \quad (N-K_0) \end{array}$$

$$\begin{array}{ll} \text{TSS} = 635065396.0 & \text{with } df = N - 1 = 74 - 1 = 73 \\ \text{ESS}_0 = 0.0 & \text{with } df = K_0 - 1 = 1 - 1 = 0 \\ \text{RSS}_0 = \mathbf{635065396.0} & \text{with } \mathbf{df_0 = N - K_0 = 74 - 1 = 73} \end{array}$$

(2) The R^2 for the ***restricted OLS SRE*** is

$$R_R^2 = \frac{\text{ESS}_0}{\text{TSS}} = 1 - \frac{\text{RSS}_0}{\text{TSS}} = \mathbf{0.0}.$$

- Compute the *sample value* of either of the general F-statistics.

$$\blacksquare F = \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} \sim F[df_0 - df_1, df_1] = F[K - K_0, N - K]$$

$$\begin{aligned} RSS_0 &= 635065396.0 & \text{and} & \quad df_0 = N - K_0 = 74 - 1 = 73 \\ RSS_1 &= 448744116.0 & \text{and} & \quad df_1 = N - K = 74 - 3 = 71 \\ RSS_0 - RSS_1 &= 635065396.0 - 448744116.0 = 186321280 \\ df_0 - df_1 &= 73 - 71 = 2 & \text{or} & \quad K - K_0 = 3 - 1 = 2. \end{aligned}$$

$$F_0 = \frac{(635065396 - 448744116)/(73 - 71)}{448744116/(74 - 3)} = \frac{186321280/2}{448744116/71} = \mathbf{14.74}$$

$$\blacksquare F = \frac{(R_U^2 - R_R^2)/(df_0 - df_1)}{(1 - R_U^2)/df_1} \sim F[df_0 - df_1, df_1] = F[K - K_0, N - K]$$

$$R_U^2 = \frac{ESS_1}{TSS} = 1 - \frac{RSS_1}{TSS} = \mathbf{0.2934}; \quad R_R^2 = \frac{ESS_0}{TSS} = 1 - \frac{RSS_0}{TSS} = \mathbf{0.0}.$$

$$\begin{aligned} df_0 &= N - K_0 = 74 - 1 = 73 \\ df_1 &= N - K = 74 - 3 = 71 \\ df_0 - df_1 &= K - K_0 = 3 - 1 = 2 \text{ or } K - K_0 = 3 - 1 = 2. \end{aligned}$$

$$F_0 = \frac{(0.2934 - 0.0)/(3 - 1)}{(1 - 0.2934)/(74 - 3)} = \frac{0.2934/2}{0.7066/71} = \mathbf{14.74}$$

- *Null distribution* of the F-statistic: $F \sim F[2,71]$ under H_0 .

The *critical values* of the $F[2,71]$ -distribution are:

$$\begin{aligned} \text{for } \alpha = 0.05: & \quad F_{\alpha}[2,71] = F_{0.05}[2,71] = 3.13. \\ \text{for } \alpha = 0.01: & \quad F_{\alpha}[2,71] = F_{0.01}[2,71] = 4.92. \end{aligned}$$

- **Inference at $\alpha = 0.05$ (5% significance level):**

Since $F_0 = 14.74 > 3.13 = F_{0.05}[2,71]$, **reject H_0** at the **5%** significance level.

- **Inference at $\alpha = 0.01$ (1% significance level):**

Since $F_0 = 14.74 > 4.92 = F_{0.01}[2,71]$, **reject H_0** at the **1%** significance level.

□ **Conclusion -- ANOVA F-test**

Reject the *restricted* model corresponding to H_0 (Model 0)

$$\text{price}_i = \beta_0 + u_i \quad (0)$$

in favour of the ***unrestricted* model corresponding to H_1 (Model 1)**

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + u_i \quad (1)$$

2. Tests of Exclusion Restrictions on *Subsets* of Slope Coefficients

TEST 1: Test Model 1 against Model 2

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + u_i \quad (1)$$

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

- What coefficient restrictions does PRE (1) impose on PRE (2)?

Answer: $\beta_3 = \beta_4 = \beta_5 = 0$

- ◆ The *null hypothesis* is:

$$H_0: \beta_j = 0 \quad \forall j = 3, 4, 5; \quad \beta_3 = 0 \text{ and } \beta_4 = 0 \text{ and } \beta_5 = 0.$$

The *restricted model implied by H₀* is obtained by setting $\beta_3 = 0$ and $\beta_4 = 0$ and $\beta_5 = 0$ in the unrestricted model (2):

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + u_i \quad (1)$$

- ◆ The *alternative hypothesis* is:

$$H_1: \beta_j \neq 0 \quad j = 3, 4, 5; \quad \beta_3 \neq 0 \text{ and/or } \beta_4 \neq 0 \text{ and/or } \beta_5 \neq 0.$$

The *unrestricted model implied by H₁* is:

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

Test statistic: For this test, we can use *only* the general **F-test**.

- The ***unrestricted model*** corresponding to the ***alternative hypothesis*** H_1 is simply PRE (2):

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

Number of free (unrestricted) regression coefficients in (2) is $K = K_1 = 6$.

- OLS estimation of the unrestricted model yields the ***unrestricted OLS SRE***:

$$\text{price}_i = \hat{\beta}_0 + \hat{\beta}_1 \text{wgt}_i + \hat{\beta}_2 \text{mpg}_i + \hat{\beta}_3 \text{wgt}_i^2 + \hat{\beta}_4 \text{mpg}_i^2 + \hat{\beta}_5 \text{wgt}_i \text{mpg}_i + \hat{u}_i \quad (2^*)$$

(1) The **OLS decomposition equation** for the ***unrestricted OLS SRE*** is

$$\begin{array}{l} \text{TSS} = \text{ESS}_1 + \text{RSS}_1 \\ (N-1) \quad (K-1) \quad (N-K) \end{array}$$

$$\begin{array}{ll} \text{TSS} = 635065396.0 & \text{with } df = N - 1 = 74 - 1 = 73 \\ \text{ESS}_1 = 308384833.0 & \text{with } df = K - 1 = 6 - 1 = 5 \\ \text{RSS}_1 = 326680563.0 & \text{with } df_1 = N - K = 74 - 6 = 68 \end{array}$$

(2) The R^2 for the ***unrestricted OLS SRE*** is

$$R_U^2 = \frac{\text{ESS}_1}{\text{TSS}} = 1 - \frac{\text{RSS}_1}{\text{TSS}} = 0.4856.$$

- The **restricted model** corresponding to the **null hypothesis H_0** is obtained by setting $\beta_3 = 0$ and $\beta_4 = 0$ and $\beta_5 = 0$ in the unrestricted model (2):

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + u_i \quad (1)$$

Number of free (unrestricted) regression coefficients is $\mathbf{K_0 = 3}$.

- OLS estimation of the restricted model yields the **restricted OLS SRE**:

$$\text{price}_i = \tilde{\beta}_0 + \tilde{\beta}_1 \text{wgt}_i + \tilde{\beta}_2 \text{mpg}_i + \tilde{u}_i \quad (1^*)$$

(1) The **OLS decomposition equation for the restricted OLS SRE** is

$$\begin{array}{l} \text{TSS} = \text{ESS}_0 + \text{RSS}_0 \\ (N-1) \quad (K_0-1) \quad (N-K_0) \end{array}$$

$$\begin{array}{ll} \text{TSS} = 635065396.0 & \text{with } df = N - 1 = 74 - 1 = 73 \\ \text{ESS}_0 = 186321280.0 & \text{with } df = K_0 - 1 = 3 - 1 = 2 \\ \text{RSS}_0 = \mathbf{448744116.0} & \text{with } \mathbf{df_0 = N - K_0 = 74 - 3 = 71} \end{array}$$

(2) The **R^2 for the restricted OLS SRE** is

$$R_R^2 = \frac{\text{ESS}_0}{\text{TSS}} = 1 - \frac{\text{RSS}_0}{\text{TSS}} = \mathbf{0.2934}.$$

- Compute the *sample value* of either of the general F-statistics.

$$\blacksquare F = \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} \sim F[df_0 - df_1, df_1] = F[K - K_0, N - K]$$

$$RSS_0 = 448744116.0 \quad \text{with} \quad df_0 = N - K_0 = 74 - 3 = 71$$

$$RSS_1 = 326680563.0 \quad \text{with} \quad df_1 = N - K = 74 - 6 = 68$$

$$RSS_0 - RSS_1 = 448744116.0 - 326680563.0 = 186321280$$

$$df_0 - df_1 = 71 - 68 = 3 \quad \text{or} \quad K - K_0 = 6 - 3 = 3.$$

$$F_0 = \frac{(448744116 - 326680563)/(71 - 68)}{326680563/68} = \frac{122063553/3}{326680563/68} = \mathbf{8.469}$$

$$\blacksquare F = \frac{(R_U^2 - R_R^2)/(df_0 - df_1)}{(1 - R_U^2)/df_1} \sim F[df_0 - df_1, df_1] = F[K - K_0, N - K]$$

$$R_U^2 = \frac{ESS_1}{TSS} = 1 - \frac{RSS_1}{TSS} = \mathbf{0.4856}; \quad R_R^2 = \frac{ESS_0}{TSS} = 1 - \frac{RSS_0}{TSS} = \mathbf{0.2934}.$$

$$df_0 = N - K_0 = 74 - 3 = 71$$

$$df_1 = N - K = 74 - 6 = 68$$

$$df_0 - df_1 = 71 - 68 = 3 \quad \text{or} \quad K - K_0 = 6 - 3 = 3.$$

$$F_0 = \frac{(0.4856 - 0.2934)/(71 - 68)}{(1 - 0.4856)/68} = \frac{0.1922/3}{0.5144/68} = \mathbf{8.469}$$

- *Null distribution* of the F-statistic: $F \sim F[3,68]$ under H_0 .

The *critical values* of the $F[3,68]$ -distribution are:

$$\text{for } \alpha = 0.05: \quad F_{\alpha}[3,68] = F_{0.05}[3,68] = 2.74.$$

$$\text{for } \alpha = 0.01: \quad F_{\alpha}[3,68] = F_{0.01}[3,68] = 4.08.$$

- **Inference at $\alpha = 0.05$ (5% significance level):**

Since $F_0 = 8.469 > 2.74 = F_{0.05}[3,68]$, **reject H_0** at the **5%** significance level.

- **Inference at $\alpha = 0.01$ (1% significance level):**

Since $F_0 = 8.469 > 4.08 = F_{0.01}[3,68]$, **reject H_0** at the **1%** significance level.

□ **Conclusion -- TEST 1**

Reject the *restricted* model (Model 1)

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + u_i \quad (1)$$

in favour of the ***unrestricted* model (Model 2)**

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

TEST 2: Test that *marginal effect of wgt_i* equals zero in Model 2

Proposition: The variable wgt_i has no effect on car price; car price is unrelated to car weight when the effect of fuel efficiency (mpg_i) on car price is controlled for.

$$price_i = \beta_0 + \beta_1 wgt_i + \beta_2 mpg_i + \beta_3 wgt_i^2 + \beta_4 mpg_i^2 + \beta_5 wgt_i mpg_i + u_i \quad (2)$$

□ **Marginal effect of wgt_i in Model 2 =**

$$\partial price_i / \partial wgt_i = \beta_1 + 2\beta_3 wgt_i + \beta_5 mpg_i$$

A sufficient condition for $\partial price_i / \partial wgt_i = 0$ is: $\beta_1 = \beta_3 = \beta_5 = 0$.

◆ The **null hypothesis** is:

$$H_0: \beta_j = 0 \quad \forall j = 1, 3, 5; \quad \beta_1 = 0 \text{ and } \beta_3 = 0 \text{ and } \beta_5 = 0.$$

The **restricted model** implied by H_0 is:

$$price_i = \beta_0 + \beta_2 mpg_i + \beta_4 mpg_i^2 + u_i \quad (3)$$

◆ The **alternative hypothesis** is:

$$H_1: \beta_j \neq 0 \quad j = 1, 3, 5; \quad \beta_1 \neq 0 \text{ and/or } \beta_3 \neq 0 \text{ and/or } \beta_5 \neq 0.$$

The **unrestricted model** implied by H_1 is:

$$price_i = \beta_0 + \beta_1 wgt_i + \beta_2 mpg_i + \beta_3 wgt_i^2 + \beta_4 mpg_i^2 + \beta_5 wgt_i mpg_i + u_i \quad (2)$$

Test statistic: For this test, we can use *only* the general F-test.

- The ***unrestricted model*** corresponding to the ***alternative hypothesis H₁*** is simply PRE (2):

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

Number of free (unrestricted) regression coefficients in (2) is **$K = K_1 = 6$** .

- OLS estimation of the unrestricted model yields the ***unrestricted OLS SRE***:

$$\text{price}_i = \hat{\beta}_0 + \hat{\beta}_1 \text{wgt}_i + \hat{\beta}_2 \text{mpg}_i + \hat{\beta}_3 \text{wgt}_i^2 + \hat{\beta}_4 \text{mpg}_i^2 + \hat{\beta}_5 \text{wgt}_i \text{mpg}_i + \hat{u}_i \quad (2^*)$$

(1) The **OLS decomposition equation for the *unrestricted OLS SRE*** is

$$\begin{array}{ccc} \text{TSS} & = & \text{ESS}_1 + \text{RSS}_1 \\ (N-1) & & (K-1) \quad (N-K) \end{array}$$

$$\begin{array}{lll} \text{TSS} = 635065396.0 & \text{with} & \text{df} = N - 1 = 74 - 1 = 73 \\ \text{ESS}_1 = 308384833.0 & \text{with} & \text{df} = K - 1 = 6 - 1 = 5 \\ \text{RSS}_1 = \mathbf{326680563.0} & \text{with} & \mathbf{df_1 = N - K = 74 - 6 = 68} \end{array}$$

(2) The **R^2** for the ***unrestricted OLS SRE*** is

$$R_U^2 = \frac{\text{ESS}_1}{\text{TSS}} = 1 - \frac{\text{RSS}_1}{\text{TSS}} = \mathbf{0.4856}.$$

- The ***restricted model*** corresponding to the ***null hypothesis*** H_0 is:

$$\text{price}_i = \beta_0 + \beta_2 \text{mpg}_i + \beta_4 \text{mpg}_i^2 + u_i \quad (3)$$

Number of free (unrestricted) regression coefficients in (3) is $K_0 = 3$.

- OLS estimation of the restricted model (3) yields the ***restricted OLS SRE***:

$$\text{price}_i = \tilde{\beta}_0 + \tilde{\beta}_2 \text{mpg}_i + \tilde{\beta}_4 \text{mpg}_i^2 + \tilde{u}_i \quad (3^*)$$

(1) The **OLS decomposition equation for the *restricted OLS SRE*** is

$$\begin{array}{ccc} \text{TSS} & = & \text{ESS}_0 + \text{RSS}_0 \\ (N-1) & & (K_0-1) \quad (N-K_0) \end{array}$$

$$\begin{array}{lll} \text{TSS} = 635065396.0 & \text{with} & \text{df} = N - 1 = 74 - 1 = 73 \\ \text{ESS}_0 = 215835615.0 & \text{with} & \text{df} = K_0 - 1 = 3 - 1 = 2 \\ \text{RSS}_0 = \mathbf{419229781.0} & \text{with} & \text{df}_0 = N - K_0 = \mathbf{74 - 3 = 71} \end{array}$$

(2) The R^2 for the ***restricted OLS SRE*** is

$$R_R^2 = \frac{\text{ESS}_0}{\text{TSS}} = 1 - \frac{\text{RSS}_0}{\text{TSS}} = \mathbf{0.3399}.$$

- Compute the *sample value* of either of the general F-statistics.

$$\blacksquare F = \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} \sim F[df_0 - df_1, df_1] = F[K - K_0, N - K]$$

$$RSS_0 = 419229781.0 \quad \text{with} \quad df_0 = N - K_0 = 74 - 3 = 71$$

$$RSS_1 = 326680563.0 \quad \text{with} \quad df_1 = N - K = 74 - 6 = 68$$

$$RSS_0 - RSS_1 = 419229781.0 - 326680563.0 = 92549218$$

$$df_0 - df_1 = 71 - 68 = 3 \quad \text{or} \quad K - K_0 = 6 - 3 = 3.$$

$$F_0 = \frac{(419229781 - 326680563)/(71 - 68)}{326680563/68} = \frac{92549218/3}{326680563/68} = \mathbf{6.422}$$

$$\blacksquare F = \frac{(R_U^2 - R_R^2)/(df_0 - df_1)}{(1 - R_U^2)/df_1} \sim F[df_0 - df_1, df_1] = F[K - K_0, N - K]$$

$$R_U^2 = \frac{ESS_1}{TSS} = 1 - \frac{RSS_1}{TSS} = \mathbf{0.4856}; \quad R_R^2 = \frac{ESS_0}{TSS} = 1 - \frac{RSS_0}{TSS} = \mathbf{0.3399}.$$

$$df_0 = N - K_0 = 74 - 3 = 71$$

$$df_1 = N - K = 74 - 6 = 68$$

$$df_0 - df_1 = 71 - 68 = 3 \quad \text{or} \quad K - K_0 = 6 - 3 = 3.$$

$$F_0 = \frac{(0.4856 - 0.3399)/(71 - 68)}{(1 - 0.4856)/68} = \frac{0.1457/3}{0.5144/68} = \mathbf{6.422}$$

- *Null distribution* of the F-statistic: $F \sim F[3,68]$ under H_0 .

The *critical values* of the $F[3,68]$ -distribution are:

$$\text{for } \alpha = 0.05: \quad F_{\alpha}[3,68] = F_{0.05}[3,68] = 2.74.$$

$$\text{for } \alpha = 0.01: \quad F_{\alpha}[3,68] = F_{0.01}[3,68] = 4.08.$$

- **Inference at $\alpha = 0.05$ (5% significance level):**

Since $F_0 = 6.422 > 2.74 = F_{0.05}[3,68]$, **reject H_0** at the **5%** significance level.

- **Inference at $\alpha = 0.01$ (1% significance level):**

Since $F_0 = 6.422 > 4.08 = F_{0.01}[3,68]$, **reject H_0** at the **1%** significance level.

□ **Conclusion -- TEST 2**

Reject the restricted model (Model 3)

$$\text{price}_i = \beta_0 + \beta_2 \text{mpg}_i + \beta_4 \text{mpg}_i^2 + u_i \quad (3)$$

in favour of the **unrestricted model (Model 2)**

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

TEST 3: Test that *marginal effect of wgt_i* is constant in Model 2

Proposition: The variable wgt_i has a constant marginal effect on car price; the marginal effect wgt_i on car price does not vary with wgt_i and mpg_i .

$$price_i = \beta_0 + \beta_1 wgt_i + \beta_2 mpg_i + \beta_3 wgt_i^2 + \beta_4 mpg_i^2 + \beta_5 wgt_i mpg_i + u_i \quad (2)$$

□ **Marginal effect of wgt_i in Model 2 =**

$$\partial price_i / \partial wgt_i = \beta_1 + 2\beta_3 wgt_i + \beta_5 mpg_i.$$

A sufficient condition for $\partial price_i / \partial wgt_i = \beta_1 = \text{a constant}$ is: $\beta_3 = \beta_5 = 0$.

◆ The ***null hypothesis*** is:

$$H_0: \beta_j = 0 \quad \forall j = 3, 5; \quad \beta_3 = 0 \text{ and } \beta_5 = 0.$$

The ***restricted model implied by H₀*** is:

$$price_i = \beta_0 + \beta_1 wgt_i + \beta_2 mpg_i + \beta_4 mpg_i^2 + u_i \quad (4)$$

◆ The ***alternative hypothesis*** is:

$$H_1: \beta_j \neq 0 \quad j = 3, 5; \quad \beta_3 \neq 0 \text{ and/or } \beta_5 \neq 0.$$

The ***unrestricted model implied by H₁*** is Model 2:

$$price_i = \beta_0 + \beta_1 wgt_i + \beta_2 mpg_i + \beta_3 wgt_i^2 + \beta_4 mpg_i^2 + \beta_5 wgt_i mpg_i + u_i \quad (2)$$

Test statistic: For this test, we can use *only* the general F-statistic.

- The ***unrestricted model*** corresponding to the ***alternative hypothesis H₁*** is simply PRE (2):

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

Number of free (unrestricted) regression coefficients in (2) is **$K = K_1 = 6$** .

- OLS estimation of the unrestricted model yields the ***unrestricted OLS SRE***:

$$\text{price}_i = \hat{\beta}_0 + \hat{\beta}_1 \text{wgt}_i + \hat{\beta}_2 \text{mpg}_i + \hat{\beta}_3 \text{wgt}_i^2 + \hat{\beta}_4 \text{mpg}_i^2 + \hat{\beta}_5 \text{wgt}_i \text{mpg}_i + \hat{u}_i \quad (2^*)$$

(1) The **OLS decomposition equation for the *unrestricted OLS SRE*** is

$$\begin{array}{l} \text{TSS} = \text{ESS}_1 + \text{RSS}_1 \\ (N-1) \quad (K-1) \quad (N-K) \end{array}$$

$$\begin{array}{ll} \text{TSS} = 635065396.0 & \text{with } df = N - 1 = 74 - 1 = 73 \\ \text{ESS}_1 = 308384833.0 & \text{with } df = K - 1 = 6 - 1 = 5 \\ \text{RSS}_1 = \mathbf{326680563.0} & \text{with } \mathbf{df_1 = N - K = 74 - 6 = 68} \end{array}$$

(2) The **R^2** for the ***unrestricted OLS SRE*** is

$$R_U^2 = \frac{\text{ESS}_1}{\text{TSS}} = 1 - \frac{\text{RSS}_1}{\text{TSS}} = \mathbf{0.4856}.$$

- The ***restricted model*** corresponding to the ***null hypothesis*** H_0 is:

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_4 \text{mpg}_i^2 + u_i \quad (4)$$

Number of free (unrestricted) regression coefficients in (4) is $K_0 = 4$.

- OLS estimation of the restricted model (4) yields the ***restricted OLS SRE***:

$$\text{price}_i = \tilde{\beta}_0 + \tilde{\beta}_1 \text{wgt}_i + \tilde{\beta}_2 \text{mpg}_i + \tilde{\beta}_4 \text{mpg}_i^2 + \tilde{u}_i \quad (4^*)$$

(1) The **OLS decomposition equation for the *restricted OLS SRE*** is

$$\begin{array}{l} \text{TSS} = \text{ESS}_0 + \text{RSS}_0 \\ (N-1) \quad (K_0-1) \quad (N-K_0) \end{array}$$

$$\begin{array}{ll} \text{TSS} = 635065396.0 & \text{with } df = N - 1 = 74 - 1 = 73 \\ \text{ESS}_0 = 223815416.0 & \text{with } df = K_0 - 1 = 4 - 1 = 3 \\ \text{RSS}_0 = \mathbf{411249980.0} & \text{with } df_0 = N - K_0 = \mathbf{74 - 4 = 70} \end{array}$$

(2) The R^2 for the ***restricted OLS SRE*** is

$$R_R^2 = \frac{\text{ESS}_0}{\text{TSS}} = 1 - \frac{\text{RSS}_0}{\text{TSS}} = \mathbf{0.3524}.$$

- Compute the *sample value* of either of the general F-statistics.

$$\blacksquare F = \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} \sim F[df_0 - df_1, df_1] = F[K - K_0, N - K]$$

$$RSS_0 = 411249980.0 \quad \text{with} \quad df_0 = N - K_0 = 74 - 4 = 70$$

$$RSS_1 = 326680563.0 \quad \text{with} \quad df_1 = N - K = 74 - 6 = 68$$

$$RSS_0 - RSS_1 = 411249980.0 - 326680563.0 = 84569417$$

$$df_0 - df_1 = 70 - 68 = 2 \quad \text{or} \quad K - K_0 = 6 - 4 = 2.$$

$$F_0 = \frac{(411249980 - 326680563)/(70 - 68)}{326680563/68} = \frac{84569417/2}{326680563/68} = \mathbf{8.80}$$

$$\blacksquare F = \frac{(R_U^2 - R_R^2)/(df_0 - df_1)}{(1 - R_U^2)/df_1} \sim F[df_0 - df_1, df_1] = F[K - K_0, N - K]$$

$$R_U^2 = \frac{ESS_1}{TSS} = 1 - \frac{RSS_1}{TSS} = \mathbf{0.4856}; \quad R_R^2 = \frac{ESS_0}{TSS} = 1 - \frac{RSS_0}{TSS} = \mathbf{0.3524}.$$

$$df_0 = N - K_0 = 74 - 4 = 70$$

$$df_1 = N - K = 74 - 6 = 68$$

$$df_0 - df_1 = 70 - 68 = 2 \quad \text{or} \quad K - K_0 = 6 - 4 = 2.$$

$$F_0 = \frac{(0.4856 - 0.3524)/(70 - 68)}{(1 - 0.4856)/68} = \frac{0.1332/2}{0.5144/68} = \mathbf{8.80}$$

- *Null distribution* of the F-statistic: $F \sim F[2,68]$ under H_0 .

The *critical values* of the $F[2,68]$ -distribution are:

$$\text{for } \alpha = 0.05: \quad F_{\alpha}[2,68] = F_{0.05}[2,68] = 3.13$$

$$\text{for } \alpha = 0.01: \quad F_{\alpha}[2,68] = F_{0.01}[2,68] = 4.93$$

- **Inference at $\alpha = 0.05$ (5% significance level):**

Since $F_0 = 8.80 > 3.13 = F_{0.05}[2,68]$, **reject H_0** at the **5%** significance level.

- **Inference at $\alpha = 0.01$ (1% significance level):**

Since $F_0 = 8.80 > 4.93 = F_{0.01}[2,68]$, **reject H_0** at the **1%** significance level.

□ **Conclusion -- TEST 3**

Reject the restricted model (Model 4)

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_4 \text{mpg}_i^2 + u_i \quad (4)$$

in favour of the **unrestricted model (Model 2)**

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

TEST 4: Test that the *marginal effect of wgt_i* in Model 2 does not depend on *mpg_i*, and that the *marginal effect of mpg_i* in Model 2 does not depend on *wgt_i*.

Model 2:

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

□ **Marginal effect of wgt_i in Model 2 =**

$$\partial \text{price}_i / \partial \text{wgt}_i = \beta_1 + 2\beta_3 \text{wgt}_i + \beta_5 \text{mpg}_i$$

A sufficient condition for $\partial \text{price}_i / \partial \text{wgt}_i$ to be independent of mpg_i is: $\beta_5 = 0$.

□ **Marginal effect of mpg_i in Model 2 =**

$$\partial \text{price}_i / \partial \text{mpg}_i = \beta_2 + 2\beta_4 \text{mpg}_i + \beta_5 \text{wgt}_i$$

A sufficient condition for $\partial \text{price}_i / \partial \text{mpg}_i$ to be independent of wgt_i is: $\beta_5 = 0$.

◆ The *null hypothesis* is:

$$H_0: \beta_5 = 0$$

◆ The *alternative hypothesis* is:

$$H_1: \beta_5 \neq 0$$

Test statistics: For this test, we can use the general F-statistic, the F-statistic for $\hat{\beta}_5$, or the t-statistic for $\hat{\beta}_5$.

- The ***unrestricted model*** corresponding to the alternative hypothesis H_1 is simply PRE (2):

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

Number of free (unrestricted) regression coefficients is $\mathbf{K} = \mathbf{K}_1 = 6$.

- OLS estimation of the unrestricted model yields the ***unrestricted OLS SRE***:

$$\text{price}_i = \hat{\beta}_0 + \hat{\beta}_1 \text{wgt}_i + \hat{\beta}_2 \text{mpg}_i + \hat{\beta}_3 \text{wgt}_i^2 + \hat{\beta}_4 \text{mpg}_i^2 + \hat{\beta}_5 \text{wgt}_i \text{mpg}_i + \hat{u}_i \quad (2^*)$$

(1) The **OLS decomposition equation for the *unrestricted OLS-SRE*** is

$$\text{TSS} = \text{ESS}_1 + \text{RSS}_1 .$$

(N-1) (K -1) (N-K)

$$\begin{aligned} \text{TSS} &= 635065396.0 && \text{with } df = N - 1 = 74 - 1 = 73 \\ \text{ESS}_1 &= 308384833.0 && \text{with } df = K - 1 = 6 - 1 = 5 \\ \text{RSS}_1 &= \mathbf{326680563.0} && \text{with } df_1 = N - K = 74 - 6 = \mathbf{68} \end{aligned}$$

(2) The $\mathbf{R^2}$ for the *unrestricted OLS SRE* is

$$R_U^2 = \frac{\text{ESS}_1}{\text{TSS}} = 1 - \frac{\text{RSS}_1}{\text{TSS}} = \mathbf{0.4856}.$$

. regress price wgt mpg wgtsq mpgsq wgtmpg

Source	SS	df	MS	Number of obs =	74
Model	308384833	5	61676966.6	F(5, 68) =	12.84
Residual	326680563	68	4804125.93	Prob > F =	0.0000
Total	635065396	73	8699525.97	R-squared =	0.4856
				Adj R-squared =	0.4478
				Root MSE =	2191.8

price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
wgt	-31.88985	9.148215	-3.486	0.001	-50.14483 -13.63487
mpg	-3549.495	1126.464	-3.151	0.002	-5797.318 -1301.672
wgtsq	.0034574	.0008629	4.007	0.000	.0017355 .0051792
mpgsq	38.74472	12.62339	3.069	0.003	13.55514 63.9343
wgtmpg	.5421927	.1971854	2.750	0.008	.1487154 .9356701
_cons	92690.55	25520.53	3.632	0.001	41765.12 143616

- The **restricted model** corresponding to the null hypothesis H_0 is obtained by setting $\beta_5 = 0$ in the unrestricted model (2):

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + u_i \tag{5}$$

Number of free (unrestricted) regression coefficients is $K_0 = 5$.

- OLS estimation of the restricted model (5) yields the **restricted OLS SRE**:

$$\text{price}_i = \tilde{\beta}_0 + \tilde{\beta}_1 \text{wgt}_i + \tilde{\beta}_2 \text{mpg}_i + \tilde{\beta}_3 \text{wgt}_i^2 + \tilde{\beta}_4 \text{mpg}_i^2 + \tilde{u}_i \tag{5*}$$

(1) The **OLS decomposition equation for the restricted OLS-SRE** is

$$\begin{matrix} \text{TSS} & = & \text{ESS}_0 & + & \text{RSS}_0 \\ (N-1) & & (K_0-1) & & (N-K_0) \end{matrix}$$

$$\begin{aligned} \text{TSS} &= 635065396.0 & \text{with } df &= N - 1 = 74 - 1 = 73 \\ \text{ESS}_0 &= 272062621.0 & \text{with } df &= K_0 - 1 = 5 - 1 = 4 \\ \text{RSS}_0 &= 363002775.0 & \text{with } df_0 &= N - K_0 = 74 - 5 = 69 \end{aligned}$$

(2) The R^2 for the **restricted OLS-SRE** is

$$R_R^2 = \frac{\text{ESS}_0}{\text{TSS}} = 1 - \frac{\text{RSS}_0}{\text{TSS}} = 0.4284.$$

```
. regress price wgt mpg wgtsq mpgsq
```

Source	SS	df	MS	Number of obs =	74
Model	272062621	4	68015655.2	F(4, 69) =	12.93
Residual	363002775	69	5260909.79	Prob > F =	0.0000
				R-squared =	0.4284
				Adj R-squared =	0.3953
Total	635065396	73	8699525.97	Root MSE =	2293.7

price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
wgt	-7.99723	2.994029	-2.671	0.009	-13.97015 -2.024306
mpg	-615.2419	377.5204	-1.630	0.108	-1368.375 137.8907
wgtsq	.0014407	.0004757	3.028	0.003	.0004916 .0023898
mpgsq	9.323582	7.009083	1.330	0.188	-4.659156 23.30632
_cons	24884.95	6878.057	3.618	0.001	11163.61 38606.3

- Compute the *sample value* of either of the general F-statistics.

$$\blacksquare F = \frac{(RSS_0 - RSS_1)/(df_0 - df_1)}{RSS_1/df_1} \sim F[df_0 - df_1, df_1] = F[K - K_0, N - K]$$

$$RSS_0 = 363002775.0 \quad \text{with} \quad df_0 = N - K_0 = 74 - 5 = 69$$

$$RSS_1 = 326680563.0 \quad \text{with} \quad df_1 = N - K = 74 - 6 = 68$$

$$RSS_0 - RSS_1 = 363002775.0 - 326680563.0 = 36322212.0$$

$$df_0 - df_1 = 69 - 68 = 1 \quad \text{or} \quad K - K_0 = 6 - 5 = 1.$$

$$F_0 = \frac{(363002775 - 326680563)/(69 - 68)}{326680563/68} = \frac{36322212/1}{326680563/68} = \mathbf{7.561}$$

$$\blacksquare F = \frac{(R_U^2 - R_R^2)/(df_0 - df_1)}{(1 - R_U^2)/df_1} \sim F[df_0 - df_1, df_1] = F[K - K_0, N - K]$$

$$R_U^2 = \frac{ESS_1}{TSS} = 1 - \frac{RSS_1}{TSS} = \mathbf{0.4856}; \quad R_R^2 = \frac{ESS_0}{TSS} = 1 - \frac{RSS_0}{TSS} = \mathbf{0.4284}.$$

$$df_0 = N - K_0 = 74 - 5 = 69$$

$$df_1 = N - K = 74 - 6 = 68$$

$$df_0 - df_1 = 69 - 68 = 1 \quad \text{or} \quad K - K_0 = 6 - 5 = 1.$$

$$F_0 = \frac{(0.4856 - 0.4284)/(69 - 68)}{(1 - 0.4856)/68} = \frac{0.0572/1}{0.5144/68} = \mathbf{7.561}$$

- *Null distribution* of the F-statistic: $F \sim F[1,68]$ under H_0 .

The *critical values* of the $F[1,68]$ -distribution are:

$$\text{for } \alpha = 0.05: \quad F_{\alpha}[1,68] = F_{0.05}[1,68] = 3.98$$

$$\text{for } \alpha = 0.01: \quad F_{\alpha}[1,68] = F_{0.01}[1,68] = 7.02$$

- **Inference at $\alpha = 0.05$ (5% significance level):**

Since $F_0 = 7.561 > 3.98 = F_{0.05}[1,68]$, **reject H_0** at the **5%** significance level.

- **Inference at $\alpha = 0.01$ (1% significance level):**

Since $F_0 = 7.561 > 7.02 = F_{0.01}[1,68]$, **reject H_0** at the **1%** significance level.

□ **Conclusion -- TEST 4**

Reject the restricted model (Model 5)

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + u_i \quad (5)$$

in favour of the **unrestricted model (Model 2)**

$$\text{price}_i = \beta_0 + \beta_1 \text{wgt}_i + \beta_2 \text{mpg}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i^2 + \beta_5 \text{wgt}_i \text{mpg}_i + u_i \quad (2)$$

- **Alternative formula for the F-statistic for testing $H_0: \beta_5 = 0$ vs. $H_1: \beta_5 \neq 0$:**

$$F(\hat{\beta}_5) = \frac{(\hat{\beta}_5 - \beta_5)^2}{\text{Var}(\hat{\beta}_5)} \sim F[1, N - K] = F[1, 74 - 6] = F[1, 68].$$

- The *sample value* of $F(\hat{\beta}_5)$ under $H_0: \beta_5 = 0$ is calculated by substituting into this formula the following values:

$$\hat{\beta}_5 = 0.54219;$$

$$\beta_5 = 0;$$

$$\text{Var}(\hat{\beta}_5) = [\text{s}\hat{\text{e}}(\hat{\beta}_5)]^2 = 0.19718539^2 = 0.0388821.$$

Thus, the *sample value* of $F(\hat{\beta}_5)$ under $H_0: \beta_5 = 0$ is:

$$F_0(\hat{\beta}_5) = \frac{(\hat{\beta}_5 - \beta_5)^2}{\text{Var}(\hat{\beta}_5)} = \frac{(0.54219 - 0)^2}{0.0388821} = \frac{0.29397}{0.0388821} = \mathbf{7.561} = \mathbf{F_0}.$$

- **Result: $F_0(\hat{\beta}_5) = F_0$ using formula (F1) or (F2) = 7.561**

- **A two-tailed t-test of $H_0: \beta_5 = 0$ vs. $H_1: \beta_5 \neq 0$:**

- The **t-statistic** is:

$$t(\hat{\beta}_5) = \frac{\hat{\beta}_5 - \beta_5}{\hat{s}e(\hat{\beta}_5)} \sim t[N - K] = t[74 - 6] = t[68].$$

- The **sample value** of this t-statistic under $H_0: \beta_5 = 0$ is calculated by substituting into this formula the following values:

$$\begin{aligned}\hat{\beta}_5 &= 0.54219; \\ \beta_5 &= 0; \\ \hat{s}e(\hat{\beta}_5) &= 0.197185.\end{aligned}$$

Thus, the **sample value of $t(\hat{\beta}_5)$** under $H_0: \beta_5 = 0$ is:

$$t_0(\hat{\beta}_5) = \frac{\hat{\beta}_5 - \beta_5}{\hat{s}e(\hat{\beta}_5)} = \frac{0.54219 - 0}{0.197185} = \frac{0.54219}{0.197185} = \mathbf{2.750}.$$

- **Null distribution** of t-statistic: $t(\hat{\beta}_5) \sim t[N - K] = t[70]$

The **two-tailed critical values of the $t[N - K] = t[70]$ -distribution** are:

$$\begin{aligned}\text{for } \alpha = 0.10, \alpha/2 = 0.05: & \quad t_{\alpha/2}[68] = t_{0.05}[68] = 1.667. \\ \text{for } \alpha = 0.05, \alpha/2 = 0.025: & \quad t_{\alpha/2}[68] = t_{0.025}[68] = 1.995. \\ \text{for } \alpha = 0.01, \alpha/2 = 0.005: & \quad t_{\alpha/2}[68] = t_{0.005}[68] = 2.650.\end{aligned}$$

- **Inference at $\alpha = 0.01$ (1% significance level):**

Since $|t_0(\hat{\beta}_5)| = 2.750 > 2.650 = t_{\alpha/2}[68] = t_{0.005}[68]$, **reject H_0** at the **1%** significance level.

(1) **Relationship between values of the two test statistics $t_0(\hat{\beta}_5)$ and $F_0(\hat{\beta}_5)$:**

$$F_0(\hat{\beta}_5) = \frac{(\hat{\beta}_5 - \beta_5)^2}{\text{V}\hat{\text{ar}}(\hat{\beta}_5)} = \left[\frac{\hat{\beta}_5 - \beta_5}{\sqrt{\text{V}\hat{\text{ar}}(\hat{\beta}_5)}} \right]^2 = \left[\frac{\hat{\beta}_5 - \beta_5}{\text{s}\hat{\text{e}}(\hat{\beta}_5)} \right]^2 = (t_0(\hat{\beta}_5))^2.$$

Example -- Test 4: For this particular test,

$$(t_0(\hat{\beta}_5))^2 = (2.750)^2 = 7.56 = F_0(\hat{\beta}_5).$$

(2) The **null distributions of the two test statistics** are related according to a similar equality.

$$(t[N - K])^2 \sim F[1, N - K].$$

The square of a $t[N - K]$ distribution has the $F[1, N - K]$ distribution.

Implication 1: The square of the $\alpha/2$ critical value of the $t[N - K]$ distribution equals the α -level critical value of the $F[1, N - K]$ distribution:

$$(t_{\alpha/2}[N - K])^2 = F_{\alpha}[1, N - K].$$

Example -- Test 4: For $\alpha = 0.01$ (the 1% significance level)

$$F_{\alpha}[1, 68] = F_{0.01}[1, 68] = 7.023; \quad t_{\alpha/2}[68] = t_{0.01}[68] = 2.65$$

$$\therefore (t_{0.025}[68])^2 = 2.65^2 = 7.023 = F_{0.01}[1, 68].$$

Implication 2:

the two-tailed p-value for $t_0(\hat{\beta}_5)$ = the p-value for $F_0(\hat{\beta}_5)$.

Example -- Test 4:

the two-tailed p-value for $t_0(\hat{\beta}_5)$ = 0.0076

the p-value for $F_0(\hat{\beta}_5)$ = 0.0076