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## ECON 351\* -- Addendum to NOTE 13

### Goodness-of-Fit in the Multiple Linear Regression Model

#### 3.4 Limitations of $R^2$

The  $R^2$  can be used to compare the goodness-of-fit of alternative sample regression equations only if the regression models satisfy **two conditions**.

(1) The models must have the **same regressand**, or **same dependent variable**.

**Reason:** TSS, ESS, and RSS depend on the units in which the regressand  $Y_i$  is measured.

(2) The models must have the **same number of regressors and regression coefficients** -- i.e., the **same value of  $k$** .

**Reason:** Adding additional regressors to a regression equation -- i.e., increasing the value of  $k$  -- always increases the value of  $R^2$ .

- ESS is an **increasing function** of the number of regressors  $k$ .
- RSS is a **decreasing function** of the number of regressors  $k$ .
- Therefore,  $R^2$  is an **increasing function of the number of regressors  $k$** .

- Consider three alternative regression models of North American car prices

price<sub>i</sub> = price of car i, in US dollars

wgt<sub>i</sub> = weight of car i, in pounds

mpg<sub>i</sub> = miles per gallon of car i

foreign<sub>i</sub> = 1 if car i is foreign, = 0 if car i is domestic

length<sub>i</sub> = length of car i, in inches

turn<sub>i</sub> = turning radius of car i, in inches

**Model 1:**

PRE: price<sub>i</sub> = β<sub>1</sub> + β<sub>2</sub>wgt<sub>i</sub> + β<sub>3</sub>mpg<sub>i</sub> + β<sub>4</sub>foreign<sub>i</sub> + β<sub>5</sub>length<sub>i</sub> + β<sub>6</sub>turn<sub>i</sub> + u<sub>i</sub>

**Model 2:** Drops regressor mpg<sub>i</sub> from Model 1;  
sets β<sub>3</sub> = 0 in Model 1.

PRE: price<sub>i</sub> = β<sub>1</sub> + β<sub>2</sub>wgt<sub>i</sub> + β<sub>4</sub>foreign<sub>i</sub> + β<sub>5</sub>length<sub>i</sub> + β<sub>6</sub>turn<sub>i</sub> + u<sub>i</sub>

**Model 3:** Drops regressors mpg<sub>i</sub> and turn<sub>i</sub> from Model 1;  
sets β<sub>3</sub> = 0 and β<sub>6</sub> = 0 in Model 1.

PRE: price<sub>i</sub> = β<sub>1</sub> + β<sub>2</sub>wgt<sub>i</sub> + β<sub>4</sub>foreign<sub>i</sub> + β<sub>5</sub>length<sub>i</sub> + u<sub>i</sub>

• **Values of R<sup>2</sup> for Models 1, 2 and 3**

For the OLS sample regression equations of Models 1, 2 and 3:

$$R^2 \text{ for Model 1} > R^2 \text{ for Model 2} > R^2 \text{ for Model 3}$$

This would be true even if β<sub>3</sub> = 0 and β<sub>6</sub> = 0; i.e., even if the true population values of the slope coefficients β<sub>3</sub> and β<sub>6</sub> were zero.

**OLS SRE for Model 1 (k = 6)**

```
. regress price wgt mpg foreign length turn
```

Source	SS	df	MS			
Model	355491949	5	71098389.9	Number of obs =	74	
Residual	279573447	68	4111374.22	F( 5, 68) =	17.29	
Total	635065396	73	8699525.97	Prob > F =	0.0000	
				R-squared =	0.5598	
				Adj R-squared =	0.5274	
				Root MSE =	2027.7	

  

price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
wgt	5.918179	1.023498	5.78	0.000	3.87582	7.960538
mpg	-22.7625	72.13908	-0.32	0.753	-166.7138	121.1888
foreign	3273.408	687.0594	4.76	0.000	1902.402	4644.413
length	-78.78337	35.09348	-2.24	0.028	-148.8113	-8.755416
turn	-149.7014	116.5489	-1.28	0.203	-382.2711	82.86829
_cons	8548.568	5726.899	1.49	0.140	-2879.28	19976.42

**OLS SRE for Model 2 (k = 5): omits *mpg<sub>i</sub>* from Model 1**

```
. regress price wgt foreign length turn
```

Source	SS	df	MS			
Model	355082608	4	88770652.0	Number of obs =	74	
Residual	279982788	69	4057721.57	F( 4, 69) =	21.88	
Total	635065396	73	8699525.97	Prob > F =	0.0000	
				R-squared =	0.5591	
				Adj R-squared =	0.5336	
				Root MSE =	2014.4	

  

price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
wgt	6.011528	.9733945	6.18	0.000	4.069659	7.953397
foreign	3318.753	667.4639	4.97	0.000	1987.198	4650.307
length	-77.25884	34.53175	-2.24	0.028	-146.1478	-8.369885
turn	-145.9884	115.1943	-1.27	0.209	-375.7946	83.81788
_cons	7334.719	4214.683	1.74	0.086	-1073.344	15742.78

**OLS SRE for Model 3 (k = 4): omits  $mpg_i$  and  $turn_i$  from Model 1**

```
. regress price wgt foreign length
```

Source	SS	df	MS			
Model	348565467	3	116188489	Number of obs =	74	
Residual	286499930	70	4092856.14	F( 3, 70) =	28.39	
Total	635065396	73	8699525.97	Prob > F =	0.0000	
				R-squared =	0.5489	
				Adj R-squared =	0.5295	
				Root MSE =	2023.1	

  

	price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
	wgt	5.774712	.9594168	6.02	0.000	3.861215	7.688208
	foreign	3573.092	639.328	5.59	0.000	2297.992	4848.191
	length	-91.37083	32.82833	-2.78	0.007	-156.8449	-25.8968
	_cons	4838.021	3742.01	1.29	0.200	-2625.183	12301.22

## 4. The Adjusted $R^2$

### 4.1 Definition of Adjusted $R^2$

$$\bar{R}^2 \equiv 1 - \frac{\text{RSS}/(N-k)}{\text{TSS}/(N-1)} = 1 - \frac{\hat{\sigma}^2}{s_Y^2}$$

where

$$\hat{\sigma}^2 = \frac{\text{RSS}}{N-k} = \text{the unbiased estimator of the error variance } \sigma^2;$$

$$s_Y^2 = \frac{\text{TSS}}{N-1} = \frac{\sum_{i=1}^N (Y_i - \bar{Y})^2}{N-1} = \text{the sample variance of the } Y_i \text{ values.}$$

### 4.2 Relationship Between $R^2$ and Adjusted $R^2$

$$\bar{R}^2 = 1 - (1 - R^2) \frac{N-1}{N-k}.$$

(1) For values of  $k > 1$ ,  $\bar{R}^2 < R^2$ .

(2)  $\bar{R}^2$  can be negative, even though  $R^2$  is non-negative.

### 4.3 Guidelines for Using Adjusted $R^2$

1.  $\bar{R}^2$  can be used to compare the goodness-of-fit of two regression models **only if the models have the same regressand**.
2.  $\bar{R}^2$  **should never be the sole criterion** for choosing between two or more sample regression equations.

## Example of Using Adjusted $R^2$

### Model 1:

$$\text{PRE: } \text{price}_i = \beta_1 + \beta_2 \text{wgt}_i + \beta_3 \text{mpg}_i + \beta_4 \text{foreign}_i + \beta_5 \text{length}_i + \beta_6 \text{turn}_i + u_i$$

### Model 2: Sets $\beta_3 = 0$ in Model 1.

$$\text{PRE: } \text{price}_i = \beta_1 + \beta_2 \text{wgt}_i + \beta_4 \text{foreign}_i + \beta_5 \text{length}_i + \beta_6 \text{turn}_i + u_i$$

### Model 3: Sets $\beta_3 = 0$ and $\beta_6 = 0$ in Model 1.

$$\text{PRE: } \text{price}_i = \beta_1 + \beta_2 \text{wgt}_i + \beta_4 \text{foreign}_i + \beta_5 \text{length}_i + u_i$$

- Compare *unadjusted*  $R^2$  values for Models 1, 2 and 3

$$\mathbf{R^2 \text{ for Model 1 (k = 6) = 0.5598}}$$

$$\mathbf{R^2 \text{ for Model 2 (k = 5) = 0.5591}}$$

$$\mathbf{R^2 \text{ for Model 3 (k = 4) = 0.5489}}$$

Note:  $R^2$  for Model 1 >  $R^2$  for Model 2 >  $R^2$  for Model 3

- Compare *adjusted*  $R^2$  values for Models 1, 2 and 3

$$\mathbf{Adjusted R^2 \text{ for Model 1 (k = 6) = 0.5274}} \quad \text{worst fit to sample data}$$

$$\mathbf{Adjusted R^2 \text{ for Model 2 (k = 5) = 0.5336}} \quad \text{best fit to sample data}$$

$$\mathbf{Adjusted R^2 \text{ for Model 3 (k = 4) = 0.5295}}$$