# **Examples of Multiple Linear Regression Models**

Data: Stata tutorial data set in text file auto1.raw or auto1.txt.

Sample data: A cross-sectional sample of 74 cars sold in North America in 1978.

## Variable definitions:

```
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); foreign_{i} = 1 \ if \ the \ i-th \ car \ is \ manufactured \ outside \ North \ America, = 0 \ otherwise.
```

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## **Multiple Linear Regression Model (1)**

The PRE is:

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + u_i$$
. (1)  
 $k = 3$ ;  $k - 1 = 2$ 

• The regressor wgt<sub>i</sub><sup>2</sup> is called an *interaction* variable. It is the product of wgt<sub>i</sub> with itself; it is a *second-order* polynomial term in the variable wgt<sub>i</sub>.

## Marginal or partial effect of $wgt_i$

The marginal effect of  $wgt_i$  on price<sub>i</sub> is obtained by partially differentiating regression equation (2) with respect to  $wgt_i$ .

$$\frac{\partial \operatorname{price}_{i}}{\partial \operatorname{wgt}_{i}} = \frac{\partial \operatorname{E}(\operatorname{price}_{i} | \operatorname{wgt}_{i})}{\partial \operatorname{wgt}_{i}} = \frac{\partial \operatorname{E}(\operatorname{price}_{i} | \bullet)}{\partial \operatorname{wgt}_{i}} = \beta_{2} + 2\beta_{3} \operatorname{wgt}_{i}.$$

• Marginal effect of  $wgt_i$  on  $price_i$  is a linear function of  $wgt_i$ . It is not a constant.

# Hypotheses of interest

1. The *marginal* effect of  $wgt_i$  on  $price_i$  is zero: i.e.,  $wgt_i$  has no effect on  $price_i$ ; or car  $price_i$  is unrelated to car  $wgt_i$ .

• 
$$H_0$$
:  $\beta_2 = 0$  and  $\beta_3 = 0$   $\Rightarrow \frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i = 0$ .

**Restricted** model corresponding to  $H_0$ : set  $\beta_2 = 0$  and  $\beta_3 = 0$  in PRE (1).

price<sub>i</sub> = 
$$\beta_1 + u_i$$
.  
 $k_0 = 1$ ;  $k_0 - 1 = 0$ 

• 
$$H_1$$
:  $\beta_2 \neq 0$  and/or  $\beta_3 \neq 0$   $\Rightarrow$   $\frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i$ .

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (1).

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2. The *marginal* effect of  $wgt_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $wgt_i$ .

• 
$$H_0$$
:  $\beta_3 = 0$   $\Rightarrow$   $\frac{\partial \, price_i}{\partial \, wgt_i} = \beta_2 + 2\beta_3 wgt_i = \beta_2$ .

**Restricted** model corresponding to  $H_0$ : set  $\beta_3 = 0$  in PRE (1).

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + u_{i}.$$

$$k_0 = 2;$$
  $k_0 - 1 = 1$ 

• 
$$H_1$$
:  $\beta_3 \neq 0 \implies \frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i$ .

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (1).

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + u_i$$
. (1)  
 $k = 3$ ;  $k - 1 = 2$ 

## The OLS SRE for Model (1)

. regress price wgt wgtsq

Source	SS	df	MS		Number of obs	= 74
Model   Residual   	250285462 384779934 635065396	71 5419 	 142731 435.69  525.97		F( 2, 71) Prob > F R-squared Adj R-squared Root MSE	= 0.0000 $= 0.3941$
price	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
wgt   wgtsq   _cons	-7.273097 .0015142 13418.8	2.691747 .0004337 3997.822	-2.702 3.491 3.357	0.009 0.001 0.001	-12.64029 .0006494 5447.372	-1.905906 .002379 21390.23

. test wgt wgtsq

F-test of hypothesis 1

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3. The marginal effect of  $wgt_i$  on  $price_i$  is decreasing in  $wgt_i$ : i.e., the marginal effect of  $wgt_i$  on  $price_i$  exhibits decreasing marginal returns in  $wgt_i$ .

• 
$$H_0$$
:  $\beta_3 = 0$  or  $\beta_3 \ge 0$   $\Rightarrow$   $\frac{\partial^2 \operatorname{price}_i}{\partial \operatorname{wgt}_i^2} = 2\beta_3 \ge 0$ .

• 
$$H_1$$
:  $\beta_3 < 0 \implies \frac{\partial \, price_i}{\partial \, wgt_i} = \beta_2 + 2\beta_3 wgt_i \quad and \quad \frac{\partial^2 \, price_i}{\partial \, wgt_i^2} = 2\beta_3 < 0$ .  
 $\Rightarrow \quad a \, one\text{-sided} \, alternative \, hypothesis$   
 $\Rightarrow \quad a \, left\text{-tail} \, test$ 

Perform a *left-tail* **t-test** using the OLS coefficient estimate  $\hat{\beta}_3$  of  $\beta_3$  for the *unrestricted* **model** corresponding to H<sub>1</sub>, which is PRE (1):

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + u_i$$
. (1)  
 $k = 3; \quad k - 1 = 2; \quad N - k = N - 3.$ 

**4.** The *marginal* effect of  $wgt_i$  on *price*<sub>i</sub> is *increasing* in  $wgt_i$ : i.e., the marginal effect of  $wgt_i$  on  $price_i$  exhibits *increasing* marginal returns in  $wgt_i$ .

• 
$$H_0$$
:  $\beta_3 = 0$  or  $\beta_3 \le 0$   $\Rightarrow$   $\frac{\partial^2 \operatorname{price}_i}{\partial \operatorname{wgt}_i^2} = 2\beta_3 \le 0$ .

• 
$$H_1$$
:  $\beta_3 > 0 \implies \frac{\partial \, price_i}{\partial \, wgt_i} = \beta_2 + 2\beta_3 wgt_i \quad and \quad \frac{\partial^2 \, price_i}{\partial \, wgt_i^2} = 2\beta_3 > 0.$ 
 $\Rightarrow \quad a \, one\text{-sided} \, alternative \, hypothesis$ 
 $\Rightarrow \quad a \, right\text{-tail} \, test$ 

Perform a *right-tail* **t-test** using the OLS coefficient estimate  $\hat{\beta}_3$  of  $\beta_3$  for the *unrestricted* model corresponding to H<sub>1</sub>, which is PRE (1):

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + u_i$$
. (1)  
 $k = 3;$   $k - 1 = 2;$   $N - k = N - 3.$ 

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## **Multiple Linear Regression Model (2)**

### The PRE is:

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + \beta_4 mpg_i + u_i$$
. (2)  
 $k = 4$ ;  $k - 1 = 3$ .

# Marginal or partial effect of $wgt_i$

The marginal effect of  $wgt_i$  on price<sub>i</sub> is obtained by partially differentiating regression equation (2) with respect to  $wgt_i$ .

$$\frac{\partial \, price_{_{i}}}{\partial \, wgt_{_{i}}} = \frac{\partial \, E(price_{_{i}} \, \big| \, wgt_{_{i}}, \, mpg_{_{i}})}{\partial \, wgt_{_{i}}} = \frac{\partial \, E(price_{_{i}} \, \big| \, \bullet)}{\partial \, wgt_{_{i}}} = \beta_{_{2}} + 2\, \beta_{_{3}} wgt_{_{i}} \, .$$

• Marginal effect of  $wgt_i$  on  $price_i$  is a linear function of  $wgt_i$ ; it is not a constant.

## Marginal or partial effect of mpgi

The marginal or partial effect of mpg<sub>i</sub> mpg<sub>i</sub> on price<sub>i</sub> is obtained by partially differentiating regression equation (2) with respect to mpg<sub>i</sub>.

$$\frac{\partial \, price_{_{i}}}{\partial \, mpg_{_{i}}} = \frac{\partial \, E(price_{_{i}} \, \big| \, wgt_{_{i}}, \, mpg_{_{i}})}{\partial \, mpg_{_{i}}} = \frac{\partial \, E(price_{_{i}} \, \big| \, \bullet)}{\partial \, mpg_{_{i}}} = \beta_{_{4}} \, .$$

• Marginal effect of *mpg<sub>i</sub>* on *price<sub>i</sub>* is *constant*: it does not vary with any observable variable.

# Hypotheses of interest

- 1. The marginal effect of  $wgt_i$  on  $price_i$  is zero: i.e.,  $wgt_i$  has no effect on  $price_i$ ; or car  $price_i$  is unrelated to car  $wgt_i$ .
- 2. The marginal effect of  $wgt_i$  on  $price_i$  is constant: i.e., it does not depend on  $wgt_i$ .
- 3. The *marginal* effect of *mpg<sub>i</sub>* on *price<sub>i</sub>* is *zero*: i.e., *mpg<sub>i</sub>* has no effect on *price<sub>i</sub>*; or car *price<sub>i</sub>* is unrelated to fuel efficiency as measured by *mpg<sub>i</sub>*.

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1. The marginal effect of  $wgt_i$  on  $price_i$  is zero: i.e.,  $wgt_i$  has no effect on  $price_i$ ; or car  $price_i$  is unrelated to car  $wgt_i$ .

• 
$$H_0$$
:  $\beta_2 = 0$  and  $\beta_3 = 0$   $\Rightarrow \frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i = 0$ .

**Restricted** model corresponding to  $H_0$ : set  $\beta_2 = 0$  and  $\beta_3 = 0$  in PRE (2).

$$price_{i} = \beta_{1} + \beta_{4}mpg_{i} + u_{i}.$$

$$k_0 = 2$$
;  $k_0 - 1 = 1$ .

• 
$$H_1$$
:  $\beta_2 \neq 0$  and/or  $\beta_3 \neq 0$   $\Rightarrow$   $\frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i$ .

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (2).

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + \beta_4 mpg_i + u_i$$
. (2)  
 $k = 4$ ;  $k - 1 = 3$ .

2. The marginal effect of  $wgt_i$  on  $price_i$  is constant: i.e., it does not depend on  $wgt_i$ .

• 
$$H_0$$
:  $\beta_3 = 0$   $\Rightarrow$   $\frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i = \beta_2$ .

**Restricted** model corresponding to  $H_0$ : set  $\beta_3 = 0$  in PRE (2).

$$price_i = \beta_1 + \beta_2 wgt_i + \beta_4 mpg_i + u_i$$
.

$$k_0 = 3;$$
  $k_0 - 1 = 2$ 

• 
$$H_1$$
:  $\beta_3 \neq 0 \implies \frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i$ .

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (2).

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + u_{i}$$
. (2)

$$k = 4;$$
  $k - 1 = 3.$ 

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3. The *marginal* effect of *mpg<sub>i</sub>* on *price<sub>i</sub>* is *zero*: i.e., *mpg<sub>i</sub>* has no effect on *price<sub>i</sub>*; or car *price<sub>i</sub>* is unrelated to fuel efficiency as measured by *mpg<sub>i</sub>*.

$$\bullet \quad H_0: \ \beta_4 = 0 \qquad \Rightarrow \qquad \frac{\partial \, price_i}{\partial \, mpg_i} = \beta_4 \, = \, 0 \, .$$

**Restricted** model corresponding to  $H_0$ : set  $\beta_4 = 0$  in PRE (2).

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + u_i$$
.  
 $k_0 = 3$ ;  $k_0 - 1 = 2$ 

• 
$$H_1$$
:  $\beta_4 \neq 0$   $\Rightarrow$   $\frac{\partial \, price_i}{\partial \, mpg_i} = \beta_4$ .

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (2).

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + \beta_4 mpg_i + u_i$$
. (2)  
 $k = 4$ ;  $k - 1 = 3$ .

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## The OLS SRE for Model (2)

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + u_{i}.$$
(2)

#### . regress price wgt wgtsq mpg

Source	SS S	df	MS		Number of obs F( 3, 70)	
Model Residual	262753599 372311797	70 531	84533.2 8739.95		Prob > F R-squared Adj R-squared	= 0.0000 = 0.4137 = 0.3886
Total	635065396	73 869	9525.97		Root MSE	= 2306.2
price	Coef.	Std. Err.	t 	P> t	[95% Conf.	Interval]
wgt wgtsq mpg _cons	-9.039563 .0016794 -124.7675 19804.79	2.905512 .000443 81.49014 5751.709	-3.111 3.791 -1.531 3.443	0.003 0.000 0.130 0.001	-14.83442 .0007958 -287.2945 8333.365	-3.244703 .002563 37.75945 31276.21

#### . test wgt wgtsq

### F-test of hypothesis 1

$$F(2, 70) = 11.59$$
  
Prob > F = 0.0000

#### . test wgtsq

## F-test of hypothesis 2

$$(1)$$
 wgtsq = 0.0

$$F( 1, 70) = 14.37$$
  
 $Prob > F = 0.0003$ 

#### . test mpg

## F-test of hypothesis 3

$$(1) mpg = 0.0$$

$$F(1, 70) = 2.34$$
  
 $Prob > F = 0.1303$ 

#### . lincom mpg

### t-test of hypothesis 3

(1) mpg = 0.0

- '	Coef.		 [95% Conf.	Interval]
	-124.7675		-287.2945	37.75945

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## **Multiple Linear Regression Model (3)**

The PRE is:

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + \beta_4 mpg_i + \beta_5 mpg_i^2 + u_i$$
. (3)  
 $k = 5$ ;  $k - 1 = 4$ .

Marginal or partial effect of wgti

$$\frac{\partial \operatorname{price}_{i}}{\partial \operatorname{wgt}_{i}} = \frac{\partial \operatorname{E}(\operatorname{price}_{i} \big| \operatorname{wgt}_{i}, \operatorname{mpg}_{i})}{\partial \operatorname{wgt}_{i}} = \frac{\partial \operatorname{E}(\operatorname{price}_{i} \big| \bullet)}{\partial \operatorname{wgt}_{i}} = \beta_{2} + 2\beta_{3} \operatorname{wgt}_{i}.$$

• Marginal effect of  $wgt_i$  on  $price_i$  is a linear function of  $wgt_i$ ; it is not a constant.

Marginal or partial effect of mpgi

$$\frac{\partial \operatorname{price}_{i}}{\partial \operatorname{mpg}_{i}} = \frac{\partial \operatorname{E}(\operatorname{price}_{i} \big| \operatorname{wgt}_{i}, \operatorname{mpg}_{i})}{\partial \operatorname{mpg}_{i}} = \frac{\partial \operatorname{E}(\operatorname{price}_{i} \big| \bullet)}{\partial \operatorname{mpg}_{i}} = \beta_{4} + 2\beta_{5} \operatorname{mpg}_{i}.$$

• Marginal effect of  $mpg_i$  on  $price_i$  is a linear function of  $mpg_i$ ; it is not a constant.

# Hypotheses of interest

- 1. The marginal effect of  $wgt_i$  on  $price_i$  is zero: i.e.,  $wgt_i$  has no effect on  $price_i$ ; or car  $price_i$  is unrelated to car  $wgt_i$ .
- 2. The *marginal* effect of  $wgt_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $wgt_i$  or  $mpg_i$ .
- 3. The *marginal* effect of *mpg<sub>i</sub>* on *price<sub>i</sub>* is *zero*: i.e., *mpg<sub>i</sub>* has no effect on *price<sub>i</sub>*; or car *price<sub>i</sub>* is unrelated to fuel efficiency as measured by *mpg<sub>i</sub>*.
- **4.** The *marginal* effect of  $mpg_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $mpg_i$  or  $wgt_i$ .

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1. The *marginal* effect of  $wgt_i$  on  $price_i$  is zero: i.e.,  $wgt_i$  has no effect on  $price_i$ ; or  $car\ price_i$  is unrelated to  $car\ wgt_i$ .

• 
$$H_0$$
:  $\beta_2 = 0$  and  $\beta_3 = 0$   $\Rightarrow \frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i = 0$ .

**Restricted** model corresponding to  $H_0$ : set  $\beta_2 = 0$  and  $\beta_3 = 0$  in PRE (3).

price<sub>i</sub> = 
$$\beta_1 + \beta_4 mpg_i + \beta_5 mpg_i^2 + u_i$$
.  
 $k_0 = 3$ ;  $k_0 - 1 = 2$ .

• 
$$H_1$$
:  $\beta_2 \neq 0$  and/or  $\beta_3 \neq 0$   $\Rightarrow$   $\frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i$ 

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (3).

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + u_{i}.$$

$$k = 5; \quad k - 1 = 4.$$
(3)

2. The *marginal* effect of  $wgt_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $wgt_i$  or  $mpg_i$ .

• 
$$H_0$$
:  $\beta_3 = 0$   $\Rightarrow$   $\frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2$ 

**Restricted** model corresponding to  $H_0$ : set  $\beta_3 = 0$  in PRE (3).

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_4 mpg_i + \beta_5 mpg_i^2 + u_i$$
.  
 $k_0 = 4$ ;  $k_0 - 1 = 3$ .

• 
$$H_1$$
:  $\beta_3 \neq 0 \implies \frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i$ 

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (3).

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + u_{i}.$$

$$k = 5; \quad k - 1 = 4.$$
(3)

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3. The marginal effect of  $mpg_i$  on  $price_i$  is zero: i.e.,  $mpg_i$  has no effect on  $price_i$ ; or car  $price_i$  is unrelated to fuel efficiency as measured by  $mpg_i$ .

• 
$$H_0$$
:  $\beta_4 = 0$  and  $\beta_5 = 0$   $\Rightarrow$   $\frac{\partial \, price_i}{\partial \, mpg_i} = \beta_4 + 2 \beta_5 mpg_i = 0$ .

**Restricted** model corresponding to  $H_0$ : set  $\beta_4 = 0$  and  $\beta_5 = 0$  in PRE (3).

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + u_i$$
.  
 $k_0 = 3$ ;  $k_0 - 1 = 2$ .

• 
$$H_1$$
:  $\beta_4 \neq 0$  and/or  $\beta_5 \neq 0$   $\Rightarrow$   $\frac{\partial \operatorname{price}_i}{\partial \operatorname{mpg}_i} = \beta_4 + 2\beta_5 \operatorname{mpg}_i$ .

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (3).

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + u_{i}.$$

$$k = 5; \quad k - 1 = 4.$$
(3)

**4.** The *marginal* effect of  $mpg_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $mpg_i$  or  $wgt_i$ .

• 
$$H_0$$
:  $\beta_5 = 0$   $\Rightarrow$   $\frac{\partial \operatorname{price}_i}{\partial \operatorname{mpg}_i} = \beta_4 + 2\beta_5 \operatorname{mpg}_i = \beta_4$ .

**Restricted** model corresponding to  $H_0$ : set  $\beta_5 = 0$  in PRE (3).

price<sub>i</sub> = 
$$\beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + \beta_4 mpg_i + u_i$$
.  
 $k_0 = 4$ :  $k_0 - 1 = 3$ .

• 
$$H_1$$
:  $\beta_5 \neq 0 \implies \frac{\partial \operatorname{price}_i}{\partial \operatorname{mpg}_i} = \beta_4 + 2\beta_5 \operatorname{mpg}_i$ 

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (3).

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + u_{i}.$$

$$k = 5; \quad k - 1 = 4.$$
(3)

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## Model (3)

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + u_{i}.$$
 (3)

## The OLS SRE for Model (3)

#### . regress price wgt wgtsq mpg mpgsq

Source	SS S	df	MS		Number of obs $F(4, 69)$	
Model Residual	272062621 363002775		5655.2 909.79		Prob > F R-squared Adj R-squared	= 0.0000 = 0.4284
Total	635065396	73 8699!	525.97		Root MSE	= 2293.7
price	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
wgt wgtsq mpg mpgsq cons	-7.99723   .0014407   -615.2419   9.323582	2.994029 .0004757 377.5204 7.009083 6878.057	-2.671 3.028 -1.630 1.330 3.618	0.009 0.003 0.108 0.188 0.001	-13.97015 .0004916 -1368.375 -4.659156 11163.61	-2.024306 .0023898 137.8907 23.30632 38606.3

#### . test wgt wgtsq

### F-test of hypothesis 1

$$(1)$$
 wgt = 0.0

(2) wgtsq = 0.0

$$F(2, 69) = 5.34$$
  
Prob > F = 0.0070

### . test wgtsq

## F-test of hypothesis 2

$$(1)$$
 wgtsq = 0.0

$$F(1, 69) = 9.17$$
  
 $Prob > F = 0.0035$ 

### . test mpg mpgsq

### F-test of hypothesis 3

$$(1)$$
 mpg =  $0.0$ 

(2) mpgsq = 0.0

$$F(2, 69) = 2.07$$
  
Prob > F = 0.1340

. test mpgsq

## F-test of hypothesis 4

(1) mpgsq = 0.0

$$F(1, 69) = 1.77$$
  
Prob > F = 0.1878

## **Multiple Linear Regression Model (4)**

### The PRE is:

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + \beta_{6}wgt_{i}mpg_{i} + u_{i}.$$
 (4)  

$$k = 6; \quad k - 1 = 5.$$

# Marginal or partial effect of $wgt_i$

$$\frac{\partial \, price_{_{i}}}{\partial \, wgt_{_{i}}} = \frac{\partial \, E(price_{_{i}} \big| \, wgt_{_{i}}, \, mpg_{_{i}})}{\partial \, wgt_{_{i}}} = \frac{\partial \, E(price_{_{i}} \big| \, \bullet)}{\partial \, wgt_{_{i}}} = \beta_{_{2}} + 2\, \beta_{_{3}} wgt_{_{i}} + \beta_{_{6}} mpg_{_{i}}.$$

• Marginal effect of  $wgt_i$  on  $price_i$  is a linear function of  $wgt_i$  and  $mpg_i$ ; it is not a constant.

## Marginal or partial effect of mpg<sub>i</sub>

$$\frac{\partial \, price_{_{i}}}{\partial \, mpg_{_{i}}} = \frac{\partial \, E(price_{_{i}} \, \big| \, wgt_{_{i}}, \, mpg_{_{i}})}{\partial \, mpg_{_{i}}} = \frac{\partial \, E(price_{_{i}} \, \big| \, \bullet)}{\partial \, mpg_{_{i}}} = \beta_{_{4}} + 2\beta_{_{5}} mpg_{_{i}} + \, \beta_{_{6}} wgt_{_{i}}.$$

• Marginal effect of  $mpg_i$  on  $price_i$  is a linear function of  $mpg_i$  and  $wgt_i$ ; it is not a constant.

# Hypotheses of interest

- 1. The *marginal* effect of  $wgt_i$  on  $price_i$  is zero: i.e.,  $wgt_i$  has no effect on  $price_i$ ; or car  $price_i$  is unrelated to car  $wgt_i$ .
- **2.** The *marginal* effect of  $wgt_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $wgt_i$  and/or  $mpg_i$ .
- 3. The *marginal* effect of  $mpg_i$  on  $price_i$  is zero: i.e.,  $mpg_i$  has no effect on  $price_i$ ; or car  $price_i$  is unrelated to fuel efficiency as measured by  $mpg_i$ .
- **4.** The *marginal* effect of  $mpg_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $mpg_i$  and/or  $wgt_i$ .

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1. The marginal effect of  $wgt_i$  on  $price_i$  is zero: i.e.,  $wgt_i$  has no effect on  $price_i$ ; or  $car\ price_i$  is unrelated to  $car\ wgt_i$ .

• 
$$H_0$$
:  $\beta_2 = 0$  and  $\beta_3 = 0$  and  $\beta_6 = 0$   $\Rightarrow$   $\frac{\partial \, price_i}{\partial \, wgt_i} = \beta_2 + 2\beta_3 wgt_i + \beta_6 mpg_i = 0$ 

**Restricted** model corresponding to  $H_0$ : set  $\beta_2 = 0$  and  $\beta_3 = 0$  and  $\beta_6 = 0$  in PRE (4).

$$price_i = \beta_1 + \beta_4 mpg_i + \beta_5 mpg_i^2 + u_i$$
.

$$k_0 = 3$$
;  $k_0 - 1 = 2$ .

• 
$$H_1$$
:  $\beta_2 \neq 0$  and/or  $\beta_3 \neq 0$  and/or  $\beta_6 \neq 0 \Rightarrow \frac{\partial \, price_i}{\partial \, wgt_i} = \beta_2 + 2 \, \beta_3 wgt_i + \beta_6 mpg_i$ 

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (4).

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + \beta_{6}wgt_{i}mpg_{i} + u_{i}.$$
 (4)  
  $k = 6$ ;  $k - 1 = 5$ .

**2.** The *marginal* effect of  $wgt_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $wgt_i$  and/or  $mpg_i$ .

• 
$$H_0$$
:  $\beta_3 = 0$  and  $\beta_6 = 0$   $\Rightarrow \frac{\partial \, price_i}{\partial \, wgt_i} = \beta_2 + 2 \beta_3 wgt_i + \beta_6 mpg_i = \beta_2$ 

**Restricted** model corresponding to  $H_0$ : set  $\beta_3 = 0$  and  $\beta_6 = 0$  in PRE (4).

$$price_i = \beta_1 + \beta_2 wgt_i + \beta_4 mpg_i + \beta_5 mpg_i^2 + u_i.$$

$$k_0 = 4;$$
  $k_0 - 1 = 3.$ 

• 
$$H_1$$
:  $\beta_3 \neq 0$  and/or  $\beta_6 \neq 0$   $\Rightarrow \frac{\partial \operatorname{price}_i}{\partial \operatorname{wgt}_i} = \beta_2 + 2\beta_3 \operatorname{wgt}_i + \beta_6 \operatorname{mpg}_i$ 

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (4).

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + \beta_{6}wgt_{i}mpg_{i} + u_{i}$$
. (4)

$$k = 6;$$
  $k - 1 = 5.$ 

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3. The *marginal* effect of  $mpg_i$  on  $price_i$  is zero: i.e.,  $mpg_i$  has no effect on  $price_i$ ; or car  $price_i$  is unrelated to fuel efficiency as measured by  $mpg_i$ .

• 
$$H_0$$
:  $\beta_4 = 0$  and  $\beta_5 = 0$  and  $\beta_6 = 0$   $\Rightarrow$   $\frac{\partial \, price_i}{\partial \, mpg_i} = \beta_4 + 2\beta_5 mpg_i + \beta_6 wgt_i = 0$ 

**Restricted** model corresponding to  $H_0$ : set  $\beta_4 = 0$  and  $\beta_5 = 0$  and  $\beta_6 = 0$  in PRE (4).

$$price_{_{i}}=\,\beta_{_{1}}+\beta_{_{2}}wgt_{_{i}}+\beta_{_{3}}wgt_{_{i}}^{^{2}}+u_{_{i}}\,.$$

$$k_0 = 3$$
;  $k_0 - 1 = 2$ .

• 
$$H_1$$
:  $\beta_4 \neq 0$  and/or  $\beta_5 \neq 0$  and/or  $\beta_6 \neq 0 \Rightarrow \frac{\partial \operatorname{price}_i}{\partial \operatorname{mpg}_i} = \beta_4 + 2\beta_5 \operatorname{mpg}_i + \beta_6 \operatorname{wgt}_i$ 

*Unrestricted* model corresponding to H<sub>1</sub>: is PRE (4).

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + \beta_{6}wgt_{i}mpg_{i} + u_{i}.$$
 (4)  

$$k = 6; \quad k - 1 = 5.$$

**4.** The *marginal* effect of  $mpg_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $mpg_i$  and/or  $wgt_i$ .

• 
$$H_0$$
:  $\beta_5 = 0$  and  $\beta_6 = 0$   $\Rightarrow \frac{\partial \, price_i}{\partial \, mpg_i} = \beta_4 + 2\beta_5 mpg_i + \beta_6 wgt_i = \beta_4$ 

**Restricted** model corresponding to  $H_0$ : set  $\beta_5 = 0$  and  $\beta_6 = 0$  in PRE (4).

$$price_{_{i}} = \beta_{_{1}} + \beta_{_{2}}wgt_{_{i}} + \beta_{_{3}}wgt_{_{i}}^{^{2}} + \beta_{_{4}}mpg_{_{i}} + u_{_{i}}.$$

$$k_0 = 4;$$
  $k_0 - 1 = 3.$ 

• 
$$H_1$$
:  $\beta_5 \neq 0$  and/or  $\beta_6 \neq 0$   $\Rightarrow \frac{\partial \operatorname{price}_i}{\partial \operatorname{mpg}_i} = \beta_4 + 2\beta_5 \operatorname{mpg}_i + \beta_6 \operatorname{wgt}_i$ 

Unrestricted model corresponding to H<sub>1</sub>: is PRE (4).

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + \beta_{6}wgt_{i}mpg_{i} + u_{i}.$$
 (4)

$$k = 6;$$
  $k - 1 = 5.$ 

## Model (4)

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + \beta_{6}wgt_{i}mpg_{i} + u_{i}$$
. (4)

## The OLS SRE for Model (4)

#### . regress price wgt wgtsq mpg mpgsq wgtmpg

Source	SS	df 	MS		Number of obs F( 5, 68)	
Model   Residual	308384833 326680563		76966.6 1125.93		Prob > F R-squared Adj R-squared	<b>= 0.0000 =</b> 0.4856
Total	635065396	73 8699	9525.97		Root MSE	= 2191.8
price	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
wgt   wgtsq mpg mpgsq wgtmpg _cons	-31.88985 .0034574 -3549.495 38.74472 .5421927 92690.55	9.148215 .0008629 1126.464 12.62339 .1971854 25520.53	-3.486 4.007 -3.151 3.069 2.750 3.632	0.001 0.000 0.002 0.003 0.008 0.001	-50.14483 .0017355 -5797.318 13.55514 .1487154 41765.12	-13.63487 .0051792 -1301.672 63.9343 .9356701 143616

#### . test wgt wgtsq wgtmpg

#### F-test of hypothesis 1

- (1) wgt = 0.0
- (2) wgtsq = 0.0
- (3) wgtmpg = 0.0

$$F(3, 68) = 6.42$$
  
 $Prob > F = 0.0007$ 

#### . test wgtsq wgtmpg

### F-test of hypothesis 2

- (1) wgtsq = 0.0
- (2) wgtmpg = 0.0

$$F(2, 68) = 8.80$$
  
 $Prob > F = 0.0004$ 

#### . test mpg mpgsq wgtmpg

- (1) mpg = 0.0
- (2) mpgsq = 0.0
- (3) wgtmpg = 0.0

$$F(3, 68) = 4.03$$
  
 $Prob > F = 0.0106$ 

#### . test mpgsq wgtmpg

## F-test of hypothesis 4

- (1) mpgsq = 0.0
- (2) wgtmpg = 0.0

$$F(2, 68) = 4.75$$
  
 $Prob > F = 0.0117$ 

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## **Multiple Linear Regression Model (5)**

### The PRE is:

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + \beta_{6}wgt_{i}mpg_{i} + \beta_{7}foreign_{i} + u_{i}$$
... (5)

$$k = 7$$
;  $k - 1 = 6$ .

## Marginal or partial effect of wgti

$$\frac{\partial \operatorname{price}_{i}}{\partial \operatorname{wgt}_{i}} = \frac{\partial \operatorname{E}(\operatorname{price}_{i} \big| \operatorname{wgt}_{i}, \operatorname{mpg}_{i}, \operatorname{foreign}_{i})}{\partial \operatorname{wgt}_{i}} = \frac{\partial \operatorname{E}(\operatorname{price}_{i} \big| \bullet)}{\partial \operatorname{wgt}_{i}} = \beta_{2} + 2\beta_{3} \operatorname{wgt}_{i} + \beta_{6} \operatorname{mpg}_{i}$$

• Marginal effect of  $wgt_i$  on  $price_i$  is a linear function of  $wgt_i$  and  $mpg_i$ ; it is not a constant.

# Marginal or partial effect of mpgi

$$\frac{\partial \, price_{_{i}}}{\partial \, mpg_{_{i}}} = \frac{\partial \, E(price_{_{i}} \big| \, wgt_{_{i}}, \, mpg_{_{i}}, \, foreign_{_{i}})}{\partial \, mpg_{_{i}}} = \frac{\partial \, E(price_{_{i}} \big| \, \bullet \,)}{\partial \, mpg_{_{i}}} = \beta_{_{4}} + 2\, \beta_{_{5}} mpg_{_{i}} + \, \beta_{_{6}} wgt_{_{i}}$$

• Marginal effect of  $mpg_i$  on  $price_i$  is a linear function of  $mpg_i$  and  $wgt_i$ ; it is not a constant.

# Hypotheses of interest

- 1. The *marginal* effect of  $wgt_i$  on  $price_i$  is zero: i.e.,  $wgt_i$  has no effect on  $price_i$ ; or  $car\ price_i$  is unrelated to  $car\ wgt_i$ .
- 2. The *marginal* effect of  $wgt_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $wgt_i$  and/or  $mpg_i$ .
- 3. The *marginal* effect of  $mpg_i$  on  $price_i$  is zero: i.e.,  $mpg_i$  has no effect on  $price_i$ ; or car  $price_i$  is unrelated to fuel efficiency as measured by  $mpg_i$ .
- **4.** The *marginal* effect of  $mpg_i$  on  $price_i$  is *constant*: i.e., it does not depend on  $mpg_i$  and/or  $wgt_i$ .

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# Interpretation of the slope coefficient $\beta_7$ on the foreign<sub>i</sub> dummy variable regressor

$$price_{i} = \beta_{1} + \beta_{2}wgt_{i} + \beta_{3}wgt_{i}^{2} + \beta_{4}mpg_{i} + \beta_{5}mpg_{i}^{2} + \beta_{6}wgt_{i}mpg_{i} + \beta_{7}foreign_{i} + u_{i}$$
... (5)

Compare the expressions implied by regression equation (5) for the conditional mean prices of foreign and domestic cars that have the same weight and fuel efficiency.

• The **conditional mean price of** *foreign* **cars** is obtained from equation (5) by **setting**  $foreign_i = 1$  and taking the conditional expectation of  $price_i$ :

$$\begin{split} E(\text{price}_i | \text{ wgt}_i, \text{mpg}_i, \text{foreign}_i &= 1) \\ &= \beta_1 + \beta_2 \text{wgt}_i + \beta_3 \text{wgt}_i^2 + \beta_4 \text{mpg}_i + \beta_5 \text{mpg}_i^2 + \beta_6 \text{wgt}_i \text{mpg}_i + \beta_7 \end{split}$$

• The **conditional mean price of** *domestic* **cars** is obtained from equation (5) by **setting**  $foreign_i = 0$  and taking the conditional expectation of  $price_i$ :

$$\begin{split} E & \left( price_i \middle| \ wgt_i, mpg_i, foreign_i = 0 \right) \\ & = \beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + \beta_4 mpg_i + \beta_5 mpg_i^2 + \beta_6 wgt_i mpg_i \end{split}$$

• Take the difference between the conditional mean price of foreign cars and the conditional mean price of "similar" domestic cars, where "similar" means foreign and domestic cars with **the** *same* **values of** *wgt<sub>i</sub>* **and** *mpg<sub>i</sub>*.

$$\begin{split} E \big( price_i \big| \ wgt_i, mpg_i, foreign_i = 1 \big) - \ E \big( price_i \big| \ wgt_i, mpg_i, foreign_i = 0 \big) \\ &= \beta_1 + \beta_2 wgt_i + \beta_3 wgt_i^2 + \beta_4 mpg_i + \beta_5 mpg_i^2 + \beta_6 wgt_i mpg_i + \beta_7 \\ &- \beta_1 - \beta_2 wgt_i - \beta_3 wgt_i^2 - \beta_4 mpg_i - \beta_5 mpg_i^2 - \beta_6 wgt_i mpg_i \\ &= \beta_7 \end{split}$$

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**Result:** The coefficient  $\beta_7$  on the foreign<sub>i</sub> dummy variable in equation (5) is

$$\beta_7 = E(price_i | wgt_i, mpg_i, foreign_i = 1) - E(price_i | wgt_i, mpg_i, foreign_i = 0)$$

- = the mean price of *foreign* cars with given values of  $wgt_i$  and  $mpg_i$  minus
  - the mean price of *domestic* cars with the same values of  $wgt_i$  and  $mpg_i$
- = the adjusted mean price difference between foreign and domestic cars of the same weight and fuel efficiency, that have the same values of wgt<sub>i</sub> and mpg<sub>i</sub>
- Compare  $\beta_7$  in equation (5) with  $\beta_2$  in the following simple regression equation:

$$price_i = \beta_1 + \beta_2 foreign_i + u_i$$

$$\beta_2 = E(price_i | foreign_i = 1) - E(price_i | foreign_i = 0)$$

- = the mean price of all *foreign* cars *minus* 
  - the mean price of all *domestic* cars
- = the *unadjusted* mean price difference between *all foreign* and *all* domestic cars regardless of their weight and fuel efficiency

# Hypothesis of interest

- 5. There is no difference between the mean price of foreign and domestic cars that have the same weight and fuel efficiency.
- $H_0$ :  $\beta_7 = 0 \implies$   $E(price_i | wgt_i, mpg_i, foreign_i = 1) E(price_i | wgt_i, mpg_i, foreign_i = 0) = 0$
- $H_1: \beta_7 \neq 0 \Rightarrow$  $E(price_i | wgt_i, mpg_i, foreign_i = 1) - E(price_i | wgt_i, mpg_i, foreign_i = 0) \neq 0$

# Model (5)

$$price_{_{i}} = \beta_{_{1}} + \beta_{_{2}}wgt_{_{i}} + \beta_{_{3}}wgt_{_{i}}^{^{2}} + \beta_{_{4}}mpg_{_{i}} + \beta_{_{5}}mpg_{_{i}}^{^{2}} + \beta_{_{6}}wgt_{_{i}}mpg_{_{i}} + \beta_{_{7}}foreign_{_{i}} + u_{_{i}}.$$

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## The OLS SRE for Model (5)

#### . regress price wgt wgtsq mpg mpgsq wgtmpg foreign

Source	SS	df	MS		Number of obs	
Model   Residual	358503838 276561558		0639.7 784.45		Prob > F R-squared Adj R-squared	= 0.0000 = 0.5645
Total	635065396	73 8699	525.97		Root MSE	= 2031.7
price	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
wgt   wgtsq   mpg   mpgsq	-15.3063 .0020985 -1407.999 14.23503	9.724082 .0008898 1211.602 13.65253	-1.574 2.358 -1.162 1.043	0.120 0.021 0.249 0.301	-34.71565 .0003224 -3826.366 -13.01554	4.103051 .0038747 1010.368 41.4856

 wgtmpg
 .2373812
 .2026331
 1.171
 0.246
 -.1670761
 .6418385

 foreign
 2749.963
 789.1946
 3.485
 0.001
 1174.724
 4325.202

 \_cons
 39826.01
 28102.9
 1.417
 0.161
 -16267.6
 95919.63

#### . test wgt wgtsq wgtmpg

### F-test of hypothesis 1

```
(1) wgt = 0.0 (2) wgtsq = 0.0
```

(3) wgtmpg = 0.0

$$F(3, 67) = 8.46$$
  
 $Prob > F = 0.0001$ 

### . test wgtsq wgtmpg

## F-test of hypothesis 2

$$(1)$$
 wgtsq = 0.0  $(2)$  wgtmpg = 0.0

$$F(2, 67) = 4.55$$
  
 $Prob > F = 0.0140$ 

## . test mpg mpgsq wgtmpg

#### F-test of hypothesis 3

$$(1) mpg = 0.0$$

$$(2)$$
 mpgsq =  $0.0$ 

$$(3)$$
 wgtmpg =  $0.0$ 

$$F(3, 67) = 0.56$$
  
 $Prob > F = 0.6419$ 

#### . test mpgsq wgtmpg

### F-test of hypothesis 4

$$(1)$$
 mpgsq = 0.0

$$(2)$$
 wgtmpg = 0.0

$$F(2, 67) = 0.69$$
  
Prob > F = 0.5068

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## The OLS SRE for Model (5)

#### . regress price wgt wgtsq mpg mpgsq wgtmpg foreign

Source	SS	df	MS		Number of obs	
Model   Residual	358503838 276561558		0639.7 784.45		Prob > F R-squared Adi R-squared	= 0.0000 = 0.5645
Total	635065396	73 8699	525.97		Root MSE	= 2031.7
price	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
wgt wgtsq mpg mpgsq wgtmpg foreign _cons	-15.3063 .0020985 -1407.999 14.23503 .2373812 2749.963 39826.01	9.724082 .0008898 1211.602 13.65253 .2026331 789.1946 28102.9	-1.574 2.358 -1.162 1.043 1.171 3.485 1.417	0.120 0.021 0.249 0.301 0.246 0.001 0.161	-34.71565 .0003224 -3826.366 -13.01554 1670761 1174.724 -16267.6	4.103051 .0038747 1010.368 41.4856 .6418385 4325.202 95919.63

. test foreign = 0

## F-test of hypothesis 5

(1) foreign = 0.0

$$F( 1, 67) = 12.14$$
  
Prob >  $F = 0.0009$ 

. test foreign

## F-test of hypothesis 5 (again)

(1) foreign = 0.0

. lincom foreign

## t-test of hypothesis 5

(1) foreign = 0.0

- '	Coef.		 [95% Conf.	Interval]
:	2749.963	3.485	1174.724	4325.202

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