

# CHAPTER 5

## ELEMENTS OF MULTIPLE REGRESSION ANALYSIS: TWO INDEPENDENT VARIABLES



# Today's Class

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- Multiple Regression
  - ▣ Moving from one IV to multiple lvs
- Similar stuff you saw with a single IV...

# Moving from 1 Independent Variable to Multiple IV's

- The simple linear regression equation with one IV is as follows:

$$Y = a + bX + e$$

- Inferring this to multiple IV's is simple, just add more  $b$ 's!

$$Y = a + b_1X_1 + b_2X_2 + \dots + b_kX_k + e$$

# Multiple Regression

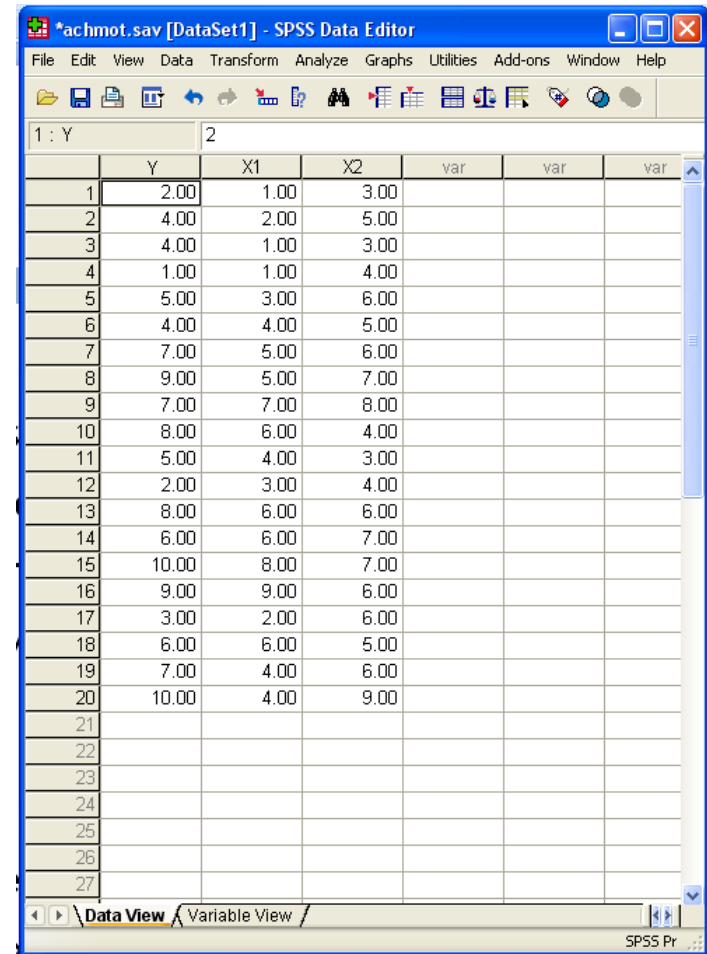
- Each independent variables has its own regression coefficient.
- Instead of the slope, we can think of each regression coefficient as follows:
  - ▣ The amount of change in the predicted value of  $Y$  as we increase 1 unit of  $X_i$ .

# Calculation of Multiple Regression

- If we have only two independent variables, calculation is tedious by not too difficult.
- Once we have more than 2 IV's we will have to rely on Matrix operations to perform calculations. We will learn more about that in the lectures to come. (YAY!)

# Data Set

- 20 students were given a test that measured Reading Achievement (Y), Verbal Aptitude ( $X_1$ ), and Achievement Motivation ( $X_2$ ).
- The data is given on Pg. 98 in the text.



SPSS Data Editor window showing a dataset with 20 rows and 7 columns. The columns are labeled Y, X1, X2, and three empty 'var' columns. The data is entered for the first 20 rows.

	Y	X1	X2	var	var	var
1	2.00	1.00	3.00			
2	4.00	2.00	5.00			
3	4.00	1.00	3.00			
4	1.00	1.00	4.00			
5	5.00	3.00	6.00			
6	4.00	4.00	5.00			
7	7.00	5.00	6.00			
8	9.00	5.00	7.00			
9	7.00	7.00	8.00			
10	8.00	6.00	4.00			
11	5.00	4.00	3.00			
12	2.00	3.00	4.00			
13	8.00	6.00	6.00			
14	6.00	6.00	7.00			
15	10.00	8.00	7.00			
16	9.00	9.00	6.00			
17	3.00	2.00	6.00			
18	6.00	6.00	5.00			
19	7.00	4.00	6.00			
20	10.00	4.00	9.00			
21						
22						
23						
24						
25						
26						
27						

# Knowing the Data



## MOTIVATION

IF A PRETTY POSTER AND A CUTE SAYING ARE ALL IT TAKES TO MOTIVATE YOU,  
YOU PROBABLY HAVE A VERY EASY JOB. THE KIND ROBOTS WILL BE DOING SOON.

# Preliminary Calculations

$$\sum Y = 117$$

$$\sum X_1 = 87$$

$$\sum X_2 = 110$$

$$\sum Y^2 = 825$$

$$\sum X_1^2 = 481$$

$$\sum X_2^2 = 658$$

$$\sum X_1 Y = 604$$

$$\sum X_2 Y = 702$$

$$\sum X_1 X_2 = 517$$

$$N = 20$$

$$\sum y^2 = \sum Y^2 - \frac{\sum Y^2}{N} = 825 - \frac{117^2}{20} = 825 - 684.45 = 140.55$$

$$\sum x_1^2 = \sum X_1^2 - \frac{\sum X_1^2}{N} = 481 - \frac{87^2}{20} = 481 - 378.45 = 102.55$$

$$\sum x_2^2 = \sum X_2^2 - \frac{\sum X_2^2}{N} = 658 - \frac{110^2}{20} = 658 - 605 = 53$$

$$\sum x_1 y = \sum X_1 Y - \frac{\sum X_1 \sum Y}{N} = 604 - \frac{87 \cdot 117}{20} = 604 - 508.95 = 95.05$$

$$\sum x_1 y = \sum X_2 Y - \frac{\sum X_2 \sum Y}{N} = 702 - \frac{110 \cdot 117}{20} = 702 - 643.5 = 58.5$$

$$\sum x_1 x_2 = \sum X_1 X_2 - \frac{(\sum X_1)(\sum X_2)}{N} = 517 - \frac{(87)(110)}{20} = 517 - 478.5 = 38.5$$

# Calculation of $b_1$

$$b_1 = \frac{\sum x_2^2 \sum x_1 y - \sum x_1 x_2 \sum x_2 y}{\sum x_1^2 \sum x_2^2 - \sum x_1 x_2^2}$$

$$b_1 = \frac{53 \quad 95.05 - 38.5 \quad 58.5}{102.55 \quad 53 - 38.5^2} = \frac{5037.65 - 2252.25}{5435.15 - 1482.25}$$

$$b_1 = \frac{2785.4}{3952.9} = .7046$$

# Calculation of $b_2$

$$b_2 = \frac{\sum x_1^2 \sum x_2 y - \sum x_1 x_2 \sum x_1 y}{\sum x_1^2 \sum x_2^2 - \sum x_1 x_2^2}$$

$$b_2 = \frac{102.55 \quad 58.5 - 38.5 \quad 95.05}{102.55 \quad 53 - 38.5^2} = \frac{5999.175 - 3659.425}{5435.15 - 1482.25}$$

$$b_2 = \frac{2339.75}{3952.9} = .5919$$

# Calculation of $a$

$$a = \bar{Y} - b_1 \bar{X}_1 - b_2 \bar{X}_2$$

$$a = 5.85 - .7046 \quad 4.35 - .5919 \quad 5.5$$

$$a = -.4705$$

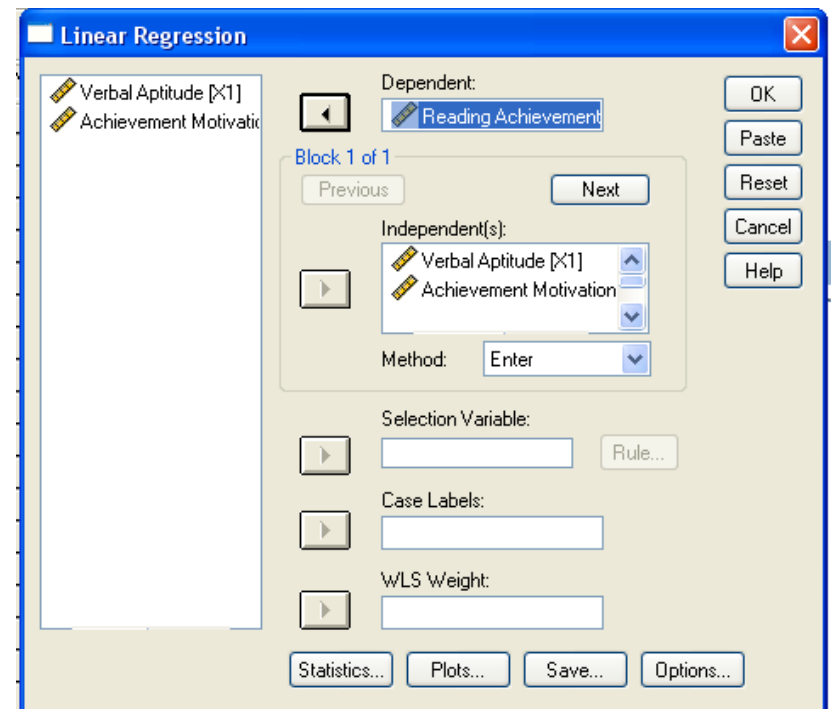
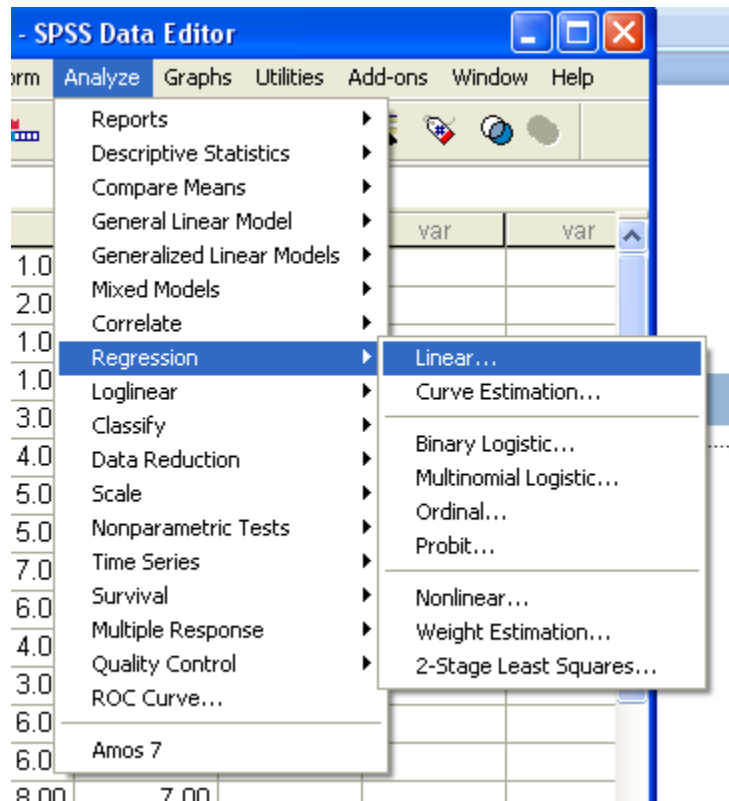
# The final regression equation

- After we finish all the calculations, we can put it all together

$$Y' = a + b_1X_1 + b_2X_2$$

$$Y' = -.4705 + .7046 X_1 + .5919 X_2$$

# Multiple Regression in SPSS



# SPSS Output

**Model Summary**

Model	R	R Square	Adjusted R Square	Std. Error of the Estimate
1	.850 <sup>a</sup>	.723	.690	1.51360

a. Predictors: (Constant), Achievement Motivation, Verbal Aptitude

Proportion of Variance in Y accounted for by X1 and X2

**ANOVA<sup>b</sup>**

Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	101.603	2	50.802	22.175	.000 <sup>a</sup>
	Residual	38.947	17	2.291		
	Total	140.550	19			

a. Predictors: (Constant), Achievement Motivation, Verbal Aptitude

b. Dependent Variable: Reading Achievement

Test of hypothesis that either X1 or X2 (or both) have some relationship with Y (a non-zero regression coefficient)

**Coefficients<sup>a</sup>**

Model		Unstandardized Coefficients		Standardized Coefficients	t	Sig.
		B	Std. Error	Beta		
1	(Constant)	-.471	1.194		-.394	.698
	Verbal Aptitude	.705	.175	.602	4.021	.001
	Achievement Motivation	.592	.244	.363	2.428	.027

a. Dependent Variable: Reading Achievement

Test of hypothesis the coefficient for X1 is zero.

Test of hypothesis the coefficient for X2 is zero.

# Predicted Values

□ Once we have the final equation, we can use it for prediction.

□ Let's try to predict 2 subjects:

▣ Subject 1:  $X_1=1$  and  $X_2=3$

▣ Subject 20:  $X_1=4$  and  $X_2=9$

$$Y'(\text{Subject 1}) = -.4705 + .7046 \cdot 1 + .5919 \cdot 3 = 2.0098$$

$$Y'(\text{Subject 20}) = -.4705 + .7046 \cdot 4 + .5919 \cdot 9 = 7.6750$$

# Residuals

- Once we have the predicted values, we can calculate residuals for these two subjects.

Residual Subject 1  $e_1 = Y - Y' = 2 - 2.0098 = -.0098$

Residual Subject 20  $e_{20} = Y - Y' = 10 - 7.765 = 2.325$

# Sum of Squares

$$SS_{reg} = b_1 \sum x_1 y + b_2 \sum x_2 y$$

$$SS_{reg} = .7046 \quad 95.05 + .5919 \quad 58.5 = 101.6$$

$$SS_{res} = \sum y^2 - SS_{reg}$$

$$SS_{res} = 140.55 - 101.6 = 38.95$$

# Squared Multiple Regression Coefficient ( $R^2$ )

- ▶ Remember from CH. 2,  $R^2$  was the amount of variance accounted for by the independent variable

$$R^2 = \frac{SS_{reg}}{\sum y^2}$$

From our previous example this would be:

$$R^2 = \frac{101.60}{140.55} = .723$$

- ▶ 72% of the total variance of the dependent variable (Y) was accounted for by the two independent variables ( $X_1$  &  $X_2$ )

# Alternative methods of Calculations

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- We can calculate this by hand in terms of the correlation coefficients...(example shown in book).
- We can also perform the calculations with the click of the mouse in SPSS.

# Tests of Significance

- Once we have calculated the parameter values it is important to determine if they are significant.
- The following are tests used for all the parameters

# Test of $R^2$

$$F = \frac{\frac{R^2}{k}}{\frac{1 - R^2}{N - k - 1}} \quad \begin{array}{l} df = k, N - k - 1 \\ Df = 2, 17 \end{array}$$

$$F = \frac{\frac{.723}{2}}{\frac{1 - .723}{20 - 2 - 1}} = \frac{.3615}{.0163} = 22.18$$

Or alternatively,

$$F = \frac{\frac{SS_{\text{reg}}}{df_{\text{reg}}}}{\frac{SS_{\text{res}}}{df_{\text{res}}}} = \frac{\frac{101.6}{2}}{\frac{38.95}{17}} = \frac{50.8}{2.29} = 22.18$$

# Test of b's

- Want to test if b is significantly different from 0.
- This is done in the same way as it was done with one variable....take the b value and divide it by its standard error.

In our example with 2 IV's, the standard error for  $b_1$  is:

$$s_{b_{y1.2}} = \sqrt{\frac{s_{y.12}^2}{\sum x_1^2 1 - r_{12}^2}}$$

the standard error for  $b_2$  is:

$$s_{b_{y2.1}} = \sqrt{\frac{s_{y.12}^2}{\sum x_2^2 1 - r_{12}^2}}$$

# Test of b's

- Using our same example from before

$$s_{y.12}^2 = \frac{SS_{res}}{N - k - 1}$$

$$SS_{res} = 38.95$$

$$s_{y.12}^2 = \frac{38.95}{20 - 2 - 1} = 2.29$$

$$\sum x_1^2 = 102.55$$

$$\sum x_2^2 = 53$$

$$b_1 = .7046$$

$$b_2 = .5919$$

$$r_{12} = .522$$

$$s_{b_1} = \sqrt{\frac{2.29}{102.55 \cdot 1 - .522^2}} = .1752$$

$$t_{b_1} = \frac{b_1}{s_{b_1}} = \frac{.7046}{.1752} = 4.02$$

$$s_{b_2} = \sqrt{\frac{2.29}{53 \cdot 1 - .522^2}} = .2437$$

$$t_{b_2} = \frac{b_2}{s_{b_2}} = \frac{.5919}{.2437} = 2.43$$

# Test of $R^2$ vs. Test of $b$

- The test of  $R^2$  is the same as testing all the  $b$ 's simultaneously.
- When we test each  $b$  individually, we are testing the given  $b$  while controlling for all other independent variables.

# Confidence Intervals

- We can calculate confidence intervals in multiple regression similar to the way we did in simple regression:

$$b \pm t_{\alpha/2, df} s_b$$

Using our same example:

*CI* for  $b_1$

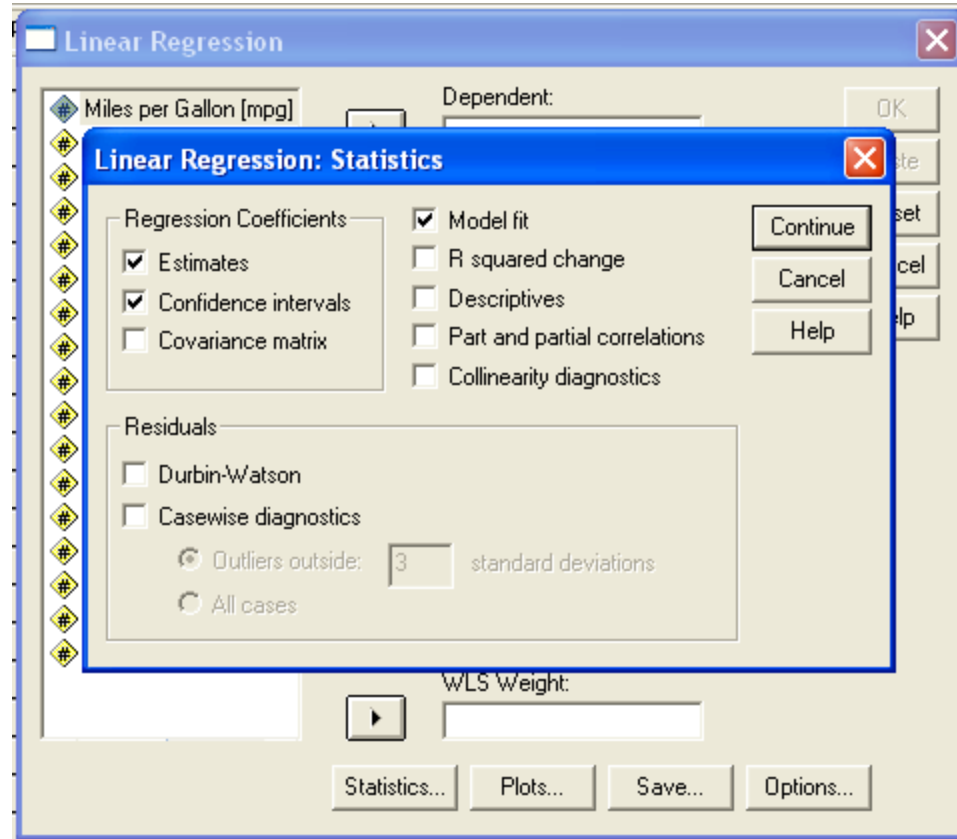
$$.7046 \pm 2.11 \cdot .1752 = .3349, 1.0743$$

*CI* for  $b_2$

$$.5919 \pm 2.11 \cdot .2437 = .0777, 1.1061$$

- Our CI's do not include 0, again confirming that the regression coefficients significantly differ from 0.

# Confidence Intervals in SPSS



# SPSS Output

**Coefficients<sup>a</sup>**

Model	Unstandardized Coefficients		Standardized Coefficients	t	Sig.	95% Confidence Interval for B	
	B	Std. Error	Beta			Lower Bound	Upper Bound
1 (Constant)	-.471	1.194		-.394	.698	-2.990	2.049
Verbal Aptitude	.705	.175	.602	4.021	.001	.335	1.074
Achievement Motivation	.592	.244	.363	2.428	.027	.078	1.106

a. Dependent Variable: Reading Achievement

# Test for increment in proportion of variance accounted for

- This tests the amount of variance accounted for as increased due to the adding of another independent variable.
- In our example, we can test the increment due to adding  $X_2$  on top of what information we already have from  $X_1$ ....along with testing the increment due to adding  $X_1$  on top of  $X_2$ .
- This test is actually equivalent to testing the individual b coefficient.

# Test for increment in proportion of variance accounted for

The test for  $X_2$

$$F = \frac{R_{y.12}^2 - R_{y.1}^2 / (2-1)}{1 - R_{y.12}^2 / N - 2 - 1}$$

$$\begin{aligned} \text{df} &= 1, N-2-1 \\ \text{df} &= 1, 17 \end{aligned}$$

$$F = \frac{.723 - .6273 / (2-1)}{1 - .723 / 20 - 2 - 1} = \frac{.0957}{.0163} = 5.87$$

The test for  $X_1$

$$F = \frac{R_{y.12}^2 - R_{y.2}^2 / (2-1)}{1 - R_{y.12}^2 / N - 2 - 1}$$

$$F = \frac{.723 - .4597 / (2-1)}{1 - .723 / 20 - 2 - 1} = \frac{.2633}{.0163} = 16.15$$

# Relative Importance of Variables

- The magnitude of  $b$  is in part affected by the scale of measurement.
- For example, if you measure objects in inches instead of feet, the nature of the regression and the tests of significance will not change.
- What will change is the magnitude of the  $b$ .
- Therefore remember, it isn't the size of the  $b$  that is important, it is its significance.

# Relative Importance of Variables

- Thinking back to our example:
  - ▣  $b_1 = .602$  and  $b_2 = .364$
- This does not mean that  $b_1$  is twice as important as  $b_2$ .
- They simply represent different variables measured on different scales.

# Next Time

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- Lab session: Regression in practice.
  - ▣ Meet in room 228.