Multiple Linear Regression with Categorical Independent Variables

1. Regression Equation

The regression equation remains the same as before except for the addition of an extra predictor. For example, suppose we have the following data with two predictors, student sex and teacher:

Table 1								
Math	Student	Teacher	Math	Student	Teacher	Math	Student	Teacher
Scores	Sex		Scores	Sex		Scores	Sex	
72	F	Gunther	74	F	Bryan	78	F	Marijke
73	F	Gunther	75	F	Bryan	79	F	Marijke
74	F	Gunther	76	F	Bryan	80	F	Marijke
76	Μ	Gunther	80	Μ	Bryan	83	М	Marijke
77	М	Gunther	81	М	Bryan	84	М	Marijke
78	М	Gunther	82	М	Bryan	85	М	Marijke

These data may be downloaded from the following link. Note the structure of the data in SPSS and JASP – all data in three columns, one for scores, one for sex, and one for teacher. Dummy variables are then created to identify group membership for sex or teacher.

http://www.bwgriffin.com/gsu/courses/edur8132/notes/math_scores.sav

The sample regression equation takes this form:

$$Y_{i} = b_{0} + b_{1}Male_{1i} + b_{2}Bryan_{2i} + b_{3}Marijke_{3i} + e_{i},$$
(1)

Regression coefficients maintain interpretations as learned previously, except now we adding the notion of statistical control since there is more than one predictor.

 b_1 = since Male will be the dummy variable, b_1 is mean difference in math scores between males and females controlling for teacher.

 b_2 = since the dummy variable Bryan identifies the teacher, b_2 is the mean difference in math scores between Bryan and Gunther (the omitted or reference teacher) controlling for student sex.

 b_3 = and the dummy variable for teacher Marijke, b_3 is the mean difference in math scores between Marijke and Gunther controlling for student sex.

 b_0 = predicted value of Y, Y', when IV equal zero; note that when dummy variables are in the equation, values of 0 for dummy represent the omitted group; literal interpretation for b_0 in this equation:

 $b_{0}\xspace$ is the predicted mean math score for females in Gunther's class.

2. Predicted Values and Errors

As before, predicted values are obtained using the equation:

 $Y' = b_0 + b_1 Male_{1i} + b_2 Bryan_{2i} + b_3 Marijke_{3i}$

Predicted means are often called **marginal means** or **adjusted means** in software such as SPSS and JASP and covariates or other categorical variables are usually set at their mean values when calculating marginal means.

(2)

Residuals are obtained by

 $e_i = Y - Y'$.

For the current data the following results are obtained:

Y' = 72.5 + 5.00 (Male) + 3.00 (Bryan) + 6.50 (Marijke)

(1) What is the predicted mean score for Females in Gunther's class?

Females in Gunther's class is the comparison, or omitted, group in the regression equation so the intercept, b_0 , represents their predicted mean score = 72.5

Y' = 72.5 + 5.00 (Male) + 3.00 (Bryan) + 6.50 (Marijke) Y' = 72.5 + 5.00 (0) + 3.00 (0) + 6.50 (0) Y' = 72.5

(2) What is the predicted mean score for Males in Gunther's class?

Y' = 72.5 + 5.00 (Male) + 3.00 (Bryan) + 6.50 (Marijke) Y' = 72.5 + 5.00 (1) + 3.00 (0) + 6.50 (0) Y' = 72.5 + 5.00 Y' = 77.5

(3) What is the predicted mean score for Females in Bryan's class?

Y' = 72.5 + 5.00 (Male) + 3.00 (Bryan) + 6.50 (Marijke) Y' = 72.5 + 5.00 (0) + 3.00 (1) + 6.50 (0) Y' = 72.5 + 3.00 Y' = 75.5

(4) What is the predicted mean score for Males in Bryan's class?

 $\begin{array}{l} Y' = 72.5 + 5.00 \mbox{ (Male)} + 3.00 \mbox{ (Bryan)} + 6.50 \mbox{ (Marijke)} \\ Y' = 72.5 + 5.00 \mbox{ (1)} & + 3.00 \mbox{ (1)} & + 6.50 \mbox{ (0)} \\ Y' = 72.5 + 5.00 & + 3.00 \\ Y' = 80.5 \end{array}$

(5) What is the predicted mean score for Females in Marijke's class?

Y' = 72.5 + 5.00 (Male) + 3.00 (Bryan) + 6.50 (Marijke) Y' = 72.5 + 5.00 (0) + 3.00 (0) + 6.50 (1) Y' = 72.5 + 6.50 Y' = 79

(6) What is the predicted mean score for Males in Marijke's class?

 $\begin{array}{l} Y' = 72.5 + 5.00 \mbox{ (Male)} + 3.00 \mbox{ (Bryan)} + 6.50 \mbox{ (Marijke)} \\ Y' = 72.5 + 5.00 \mbox{ (1)} & + 3.00 \mbox{ (0)} & + 6.50 \mbox{ (1)} \\ Y' = 72.5 + 5.00 & + 6.50 \\ Y' = 84 \end{array}$

(7a) What is the estimated student sex difference in math holding constant teacher?

 $Y' = b_0 + b_1 Male_{1i} + b_2 Bryan_{2i} + b_3 Marijke_{3i}$ Y' = 72.5 + 5.00 (Male) + 3.00 (Bryan) + 6.50 (Marijke) $b_1 = 5.00$ (2)

(7b) Is this sex difference the same for all teachers? How does the average/estimated difference compare with actual?

Yes, the same, because there is no interaction term in the regression equation to model possible changes in sex differences across teacher. An interaction term would allow the difference of 5.00 to vary across teachers.

(8) What are the estimated teacher differences in math holding constant student sex?

Bryan vs. Gunther= b_2Bryan_{2i} Marijke vs. Gunther= $b_3Marijke_{3i}$ Bryan vs. Marijke= must set Marijke as the comparison teacher and include Gunther dummy to estimate
coefficient, but it should be the difference between their coefficient estimates, i.e.,
 $b_2 - b_3 = 3.00 - 6.50 = -3.50$

3. Predicted Values Holding Constant One IV

If one wished to obtain the **predicted means**, or **adjusted means** or **marginal means**, for each teacher controlling for sex – not predicting means separately for males and females, but instead holding constant sex—one must include sex in the regression equation but instead of using the scores 0, 1, one instead using the mean value of the dummy variable for sex. Thus, as noted above, predicted means are usually done by using mean scores for each covariate and grouping variable. This will be explained and illustrated in two future presentations: regression with both quantitative and categorical variables and in ANCOVA.

In this example, since sex is evenly divided, the mean for the sex dummy variable is M = 0.50, so **predicted means** (**marginal means**, or **adjusted means**) are obtained by using the male dummy variable mean as the multiplier for Male rather than the dummy code of 1.

 $\begin{array}{ll} Y' = b_0 & + b_1 Male_{1i} & + b_2 Bryan_{2i} & + b_3 Marijke_{3i} \\ Y' = 72.5 + 5.00 \ (Male) + 3.00 \ (Bryan) + 6.50 \ (Marijke) \\ Y' = 72.5 + 5.00 \ \textbf{(0.50)} & + 3.00 \ (Bryan) + 6.50 \ (Marijke) \end{array}$

(9) What is the predicted mean score for Gunther's class, holding constant sex?

Y' = 72.5 + 5.00 (Male) + 3.00 (Bryan) + 6.50 (Marijke) Y' = 72.5 + 5.00 (0.50) + 3.00 (0) + 6.50 (0) Y' = 72.5 + 5.00 (0.50) Y' = 75

(10) What is the predicted mean score for Bryan's class, holding constant sex?

Y' = 72.5 + 5.00 (Male) + 3.00 (Bryan) + 6.50 (Marijke) Y' = 72.5 + 5.00 (0.50) + 3.00 (1) + 6.50 (0) Y' = 72.5 + 5.00 (0.50) + 3.00 (1) Y' = 78 (11) What is the predicted mean score for Marijke's class, holding constant sex?

 $\begin{array}{l} Y' = 72.5 + 5.00 \mbox{ (Male)} + 3.00 \mbox{ (Bryan)} + 6.50 \mbox{ (Marijke)} \\ Y' = 72.5 + 5.00 \mbox{ (0.50)} + 3.00 \mbox{ (0)} & + 6.50 \mbox{ (1)} \\ Y' = 72.5 + 5.00 \mbox{ (0.50)} & + 6.50 \mbox{ (1)} \\ Y' = 81.5 \end{array}$

The values calculated above are also produced in JASP and SPSS; see results below.

JASP marginal means

Marginal M	leans			
Marginal Mea	ans - teacher			
		95% CI for Me		
teacher	Marginal Mean	Lower	Upper	SE
Bryan	78.000	77.094	78.906	0.423
Gunther	75.000	74.094	75.906	0.423
Marijke	81.500	80.594	82.406	0.423

SPSS marginal means

E	stimate	ed Marg	inal Mea	ans	
te	eacher				
	Depender	it Variable: m	Estimat ath	es	
				95% Confide	ence Interval
	teacher	Mean	Std. Error	Lower Bound	Upper Bound
	Bryan	78.000	.408	77.111	78.889
	Gunther	75.000	.408	74.111	75.889
	Marijke	81.500	.408	80.611	82.389

4. Overall Mode Fit and Statistical Inference

The usual statistics apply for overall model fit (R^2 , adjusted R^2 , MSE, SEE) and the testing of model fit (F-value). If the F ratio for the model is significant (testing Ho: $R^2 = 0.00$) then that means more variance in the DV (math scores) is being predicted than expected by chance. It also implies that at least two group means are significantly different, either the sex means or two of the teachers' means. Use IV coefficient testing to identify which variables are significant.

JASP results for R² and model fit

Model	R	R²	Adjusted R ²		RMSE	_		
H.	0.000	0.000	0.000		3.869			
H ₁	0.706	0.499	0.432		2.915			
NOVA						-		
NOVA Model		Sum of S	Squares	df	Mea	an Square	F	p
	Regression		Squares 27.000	df 2	Mea	an Square 63.500	F 7.471	p 0.006
Model	Regression Residual	12			Mea			

5. Individual IV Statistical Inference

As before, each regression coefficient is tested with a t-ratio (b/se = t). However, coefficient t-ratios are insufficient to assess the contribution of a categorical variable with more than two categories. The **global effect** (overall statistical effect on the regression model) of a categorical IV with more than two categories is assessed by the F-test of the $\Delta R^2(X_k)$ contribution to the model. For categorical variables this null is Ho: $\Delta R^2(X_k) = 0.00$ or Ho: category group means are all equal.

Current Example: Teacher Global Effect:

Table 2			
Model	R ²	Regression df	Error df
$Y' = b_0 + b_1 Male_{1i}$.442	1	16
$Y' = b_0 + b_1 Male_{1i} + b_2 Bryan_{2i} + b_3 Marijke_{3i}$.941	3	14
ΔR^2 (Teacher) =	.941442 = .499	$\Delta R^2 df_1 = 3-1 = 2$	$\Delta R^2 df_2 = 14$ (smaller df)

In SPSS:

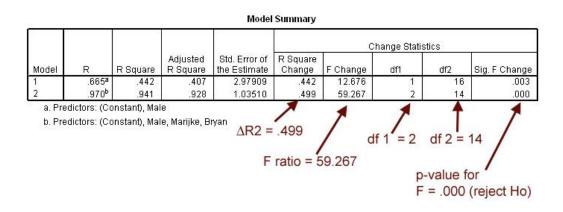
1. Choose Regression, enter Math in the Dependent box

2. Enter Male in Independents box, then click on Statistics->R-square Change->Continue

3. Click Next, then enter Bryan and Marijke dummy variables in IV box

4. Click Ok

The image below shows SPSS results testing the global effect ΔR^2 (Teacher).



Alternatively, one can use the TEST function in SPSS syntax, i.e., /TEST = (MALE) (Bryan Marijke). See the presentation notes and video on calculating ΔR^2 . Results presented below.

				ANOVAC				
Model			Sum of Squares	df	Mean Square	F	Sig.	R Square Change
1	Subset	Male	112.500	1	112.500	105.000	.000a	.442
	Tests	Bryan, Marijke	127.000	2	63.500	59.267	.000ª	.499
Regression			239.500	3	79.833	74.511	.000 ^b	
	Residual		15.000	14	1.071			
	Total		254.500	17				
а. т	ested against th	e full model.						
b. P	redictors in the f	Full Model: (Const	ant), Marijke, I	Male, Bryan.				
C. D	ependent Variat	ole: math						

In JASP we can use the H_0 and H_1 models to create steps of entry for predictors. JASP results shown below match those produced by SPSS.

Aodel Summa	ary - math 🔻								
Model	R	R²	Adjusted R ²	RMSE	R ² Change	F Change	df1	df2	р
Ho	0.665	0.442	0.407	2.979	0.442	12.676	1	16	0.003
H ₁	0.970	0.941	0.928	1.035	0.499	59.267	2	14	< .001

6. Pairwise Comparisons Among IV Categories

Typically, with multinomial predictors, it is useful to provide a table of pairwise comparisons. Procedures to obtain pairwise comparisons were presented in the previous video on regression with a multinomial variable. Two approaches were used: (a) estimate separate regression equations with different groups serving the role of comparison groups, or (b) use ANOVA commands to obtain all pairwise comparisons automatically with post hoc procedures.

One must also control for familywise Type 1 error inflation when performing multiple comparisons. Pairwise comparisons in the context of regression, or ANOVA, when a second, or more, predictor is present means pairwise comparisons are performed on **adjusted means** (i.e., **predicted means** or **marginal means**). For this example, interest lies in determining whether math means differ among teachers after controlling for student sex.

6a. Regression Model Approach

1. Estimate teacher differences in regression. To obtain all mean differences you will have to change reference groups in the regression equation, for example:

 $Y' = b_0 + b_1 Male_{1i} + b_2 Bryan_{2i} + b_3 Marijke_{3i}$

 b_2 = Bryan vs. Gunther mean difference b_3 = Marijke vs. Gunther mean difference

How to get Bryan vs. Marijke mean difference? Rerun regression with Marijke as the omitted, referenced teacher:

 $Y' = b_0 + b_1 Male_{1i} + b_2 Bryan_{2i} + b_3 Gunther_{3i}$

Now, $b_2 =$ Bryan vs. Marijke mean difference.

2. Find standard errors (se) for each mean difference

3. Find appropriate Bonferroni or Scheffé critical t-value

4. Calculate CI for each mean difference, e.g,

Upper CI: b2 + se * critical t Lower CI: b2 – se * critical t

For the current example Bonferroni critical t = 2.709 (with comparisons = 3 and df = 14). For the Bryan vs. Gunther comparison (b2 = 3.00, se = 0.598), the 95% Bonferroni CI is

Upper CI: 3.00 + 0.598 * 2.709 = 4.62 Lower CI: 3.00 - 0.598 * 2.709 = 1.38

Below is a table showing complete results (only one adjusted CI needed; four are provided to show differences in CI among Tukey, Bonferroni, and Scheffé procedures.

Table 3

			Bonf. Adj. 95%	Tukey Adj. 95% Cl	Scheffé 95%
Difference	Difference	CI - hand cal.	CI -SPSS	- SPSS, JASP	CI - SPSS
3.00*	0.598	1.38, 4.62	1.395, 4.604	1.459, 4.54	1.39, 4.609
6.50*	0.598	4.88, 8.12	4.895, 8.104	4.959, 8.04	4.89, 8.109
-3.50*	0.598	-5.12, -1.88	-5.104, -1.895	-5.04, -1.959	-5.109, -1.89
	3.00* 6.50*	3.00* 0.598 6.50* 0.598	3.00* 0.598 1.38, 4.62 6.50* 0.598 4.88, 8.12	3.00* 0.598 1.38, 4.62 1.395, 4.604 6.50* 0.598 4.88, 8.12 4.895, 8.104	3.00* 0.598 1.38, 4.62 1.395, 4.604 1.459, 4.54 6.50* 0.598 4.88, 8.12 4.895, 8.104 4.959, 8.04

*p<.05, where p-values are adjusted using the Bonferroni method.

6b. ANOVA Approach

Since software typically provides multiple comparisons as an option in ANOVA, using the ANOVA option is the more efficient means to obtain multiple comparisons with adjusted confidence intervals.

Both approaches are illustrated in the video. Below are results from both SPSS and JASP.

SPSS Multiple Comparisons from Univariate ANOVA command.

			Multiple C	omparisons			
Dependent V	/ariable: math						
	(I) teacher	(J) teacher	Mean Difference (I-J)	Std. Error	Sig.	95% Confide	ence Interval Upper Bound
Tukey HSD	Brvan	Gunther	3.0000*	.57735	.001	1,4597	4,5403
rukey nob	Dijan	Marijke	-3.5000*	.57735	.000	-5.0403	-1.9597
	Gunther	Bryan	-3.0000*	.57735	.000	-4.5403	-1.4597
		Marijke	-6.5000*	.57735	.000	-8.0403	-4.9597
	Marijke	Bryan	3.5000*	.57735	.000	1.9597	5.0403
	-	Gunther	6.5000*	.57735	.000	4.9597	8.0403
Scheffe	Bryan	Gunther	3.0000*	.57735	.001	1.3906	4.6094
		Marijke	-3.5000*	.57735	.000	-5.1094	-1.8906
	Gunther	Bryan	-3.0000*	.57735	.001	-4.6094	-1.3906
		Marijke	-6.5000*	.57735	.000	-8.1094	-4.8906
	Marijke	Bryan	3.5000*	.57735	.000	1.8906	5.1094
		Gunther	6.5000*	.57735	.000	4.8906	8.1094
Bonferroni	Bryan	Gunther	3.0000*	.57735	.001	1.3953	4.6047
		Marijke	-3.5000*	.57735	.000	-5.1047	-1.8953
	Gunther	Bryan	-3.0000*	.57735	.001	-4.6047	-1.3953
		Marijke	-6.5000*	.57735	.000	-8.1047	-4.8953
	Marijke	Bryan	3.5000*	.57735	.000	1.8953	5.1047
		Gunther	6.5000*	.57735	.000	4.8953	8.1047

JASP Multiple Comparisons from ANOVA command.

			95% CI for Mean Difference)				
		Mean Difference	Lower	Upper	SE	t	p _{tukey}	p _{scheffe}	p _{bonf}
Bryan	Gunther	3.000	1.436	4.564	0.598	5.020	< .001	< .001	< .001
	Marijke	-3.500	-5.064	-1.936	0.598	-5.857	< .001	< .001	< .001
Gunther	Marijke	-6.500	-8.064	-4.936	0.598	-10.877	< .001	< .001	< .001

7. APA Style Results

Variable		Correla	ations	
	Math Scores	Male	Bryan	Marijke
Math Scores				
Male	.67*			
Bryan	03	.00		
Marijke	.63*	.00	50*	
Mean	78.17	0.50	0.33	0.33
SD	3.87	0.51	0.49	0.49

Table 4: Descriptive Statistics and Correlations Among Math Scores, Student Sex, and Teachers

Note: Male (male = 1, female = 0), Bryan (=1, others = 0) and Marijke (=1, others = 0) are dummy variables; n = 18. *p < .05.

(Optional: provide table of math score means by teacher)

Table 5: Regression of Math Scores on Student Sex and Teachers

Variable	b	se	ΔR^2	95% CI	F	t
Male	5.00	0.49	.44	3.95, 6.05		10.25*
Teacher			.50		59.27*	
Bryan	3.00	0.60		1.72, 4.28		5.02*
Marijke	6.50	0.60		5.22, 7.78		10.88*
Intercept	72.50	0.49		71.45, 73.55		148.58*

Note: $R^2 = .94$, adj. $R^2 = .93$, $F_{3,14} = 74.51^*$, MSE = 1.071, n = 18. ΔR^2 represents the semi-partial correlation or the increment in R^2 due to adding the respective variable. Male (male = 1, female = 0), Bryan (=1, others = 0) and Marijke (=1, others = 0) are dummy variables.

*p < .05.

Table 6: Comparisons of Adjusted Mean Math Scores Among Teachers

• • • • • • • • • • • • • • • • • • •	2	8	
Comparison	Estimated Mean	Standard Error of	Bonferroni Adjusted
	Difference	Difference	95% CI
Bryan vs. Gunther	3.00*	0.598	1.38, 4.62
Marijke vs. Gunther	6.50*	0.598	4.88, 8.12
Bryan vs. Marijke	-3.50*	0.598	-5.12, -1.88
* 05 1		c , , , , ,	

*p<.05, where p-values are adjusted using the Bonferroni method.

Regression results show that both student sex and teachers are statistically related to students' math scores at the .05 level of significance. Males score about 5 points higher than females, and students in Marijke's class tend to score higher than students in either of Bryan's or Gunther's class. Students in Gunther's class score lower than in either Bryan's or Marijke's class. Note that all teacher comparisons are statistically different.

8. Exercises

(1) According to the leadership literature, there are a number of different leadership styles. Listed below are scores obtained from an instrument designed to measure a particular leadership style, which will be referred to as style X. Of interest is whether X differs by school district type in terms of urbanity, and by sex. A stratified random sample of school principals were selected from three district types (mostly urban, mostly suburban, and mostly rural).

The scores on style X range from 100 to 0. The closely the score to 100, the more the respondent conforms to style X, while the closer the score to 0, the less the respondent conforms to style X.

Sex	District Type	Style X
m	urban	85
m	urban	98
m	urban	75
f	urban	63
m	urban	91
f	urban	49
f	urban	62
f	suburban	49
f	suburban	48
m	suburban	56
m	suburban	78
f	suburban	35
m	suburban	50
m	rural	33
m	rural	95
f	rural	26
f	rural	11
f	rural	33
m	rural	25
m	rural	65

Is there any evidence that X differs among the three district types, or by sex?

(2) A researcher is interested in learning whether frequency of reading at home to elementary-aged children produces differential effects on reading achievement. After obtaining information from a randomly selected sample of parents about this behavior, the following classifications and standardized achievement scores were recorded. (Note: frequency classifications as follows: a = less than once per month, b = once to three times per month, c = more than three times per month.) In addition to reading frequency, information regarding the family's status concerning whether or not the family's child receives either free or reduced lunch is recorded as a proxy for SES.

SES	Freq. of Reading	Achievement
	rieq. of Reading	
fr	а	48
fr	а	37
no	а	47
no	а	65
no	b	57
fr	b	39
fr	b	49
no	b	45
no	С	61
no	С	55
fr	С	51
fr	С	30

Note. FR inc	licates free or	reduced lunch	received.	NO indicates otherwise.
<u>noter</u> i it inte	nearco nee or	i caacca ianon		

Is frequency of reading at home related to student reading achievement once SES is taken into account?

(3) An administrator wishes to know whether student behavioral problems can be linked to student performance. If students were suspended or reprimanded more than once, they are classified as having behavioral problems. In addition, each student's SES is known, and should be taken into account. The administrator randomly selects 13 students and collects the appropriate data.

Student	GPA	Student SES	Behavioral Problems
Bill	3.33	h	n
Bob	1.79	I	У
Stewart	2.21	m	n
Linda	3.54	h	У
Lisa	2.89	m	n
Ann	2.54	m	n
Fred	2.66	h	У
Carter	1.10	I	У
Bill	3.10	h	n
Sue	2.10	I	У
Kara	2.07	I	У
Loser	2.31	m	n
Kathy	3.67	h	n

9. Exercise Answers

(1) Results for leadership style analysis.

Table 1a

Descriptive Statistics for Lead	dership Style, District Type, and Sex
Variable	Correlations

variable	correlations					
	Style	Urban	Suburban	Male		
Style						
Urban	.55*					
Suburban	10	48*				
Male	.54*	.03	07			
Mean	56.35	.350	.300	.550		
SD	25.07	.489	.470	.510		

Note: Male is a dummy variable (male = 1, female = 0), as are Urban (1, 0 = other) and Suburban (1, 0 = other); n = 20.

Table 1b

Regression of Style on Sex and District Type

Variable	b	se	ΔR^2	95%CI	F	t
Male	26.29	7.53	.29	10.32, 42.26		3.49*
District Type			.33		7.14*	
Urban	33.57*	8.94		14.62, 52.52		3.76*
Suburban	13.40	9.32		-6.36, 33.16		1.44
Intercept	26.12	7.65		9.91, 42.33		3.42*
-	-					

Note: $R^2 = .625$, adj. $R^2 = .555$, $F_{3,16} = 8.90$, MSE = 279.70, n = 20. ΔR^2 represents the semi-partial correlation or the increment in R^2 due to adding the respective variable. Male is a dummy variable (male = 1, female = 0), as are Urban (1, 0 = other) and Suburban (1, 0 = other).

*p < .05.

Table 1c

Comparisons of Style Scores Among Urban, Suburban, and Rural Principals

Contrast	Estimated Mean	Standard Error of	Bonferroni Adjusted	
	Difference	Difference	95% CI	
Urban vs. Rural	33.57*	8.94	9.74, 57.40	
Suburban vs. Rural	13.40	9.32	-11.44, 38.24	
Urban vs. Suburban	20.17	9.32	-4.67, 45.01	

*p<.05, where p-values are adjusted using the Bonferroni method.

[Note, Bonferroni CI taken from Excel Spreadsheet is incorrect so must calculate CI using tabled values for Bonferroni comparisons. Use male = .55 in regression equation to obtain estimated means for each district.]

Both sex and district type are statistically related to leadership style. Once district type is taken into account, males average about 26 points higher than females. Among the three district types considered, principals in urban settings have a statistically higher score on style than do principals in rural districts, but not statistically higher than principals in suburban districts.

(2) Results for reading frequency.

Table 2a

Variable	Correlations				
	Achievement	В	С	SES	
Achievement					
B = 1 to 3 per month	09				
C = more than 3 per month	.04	50			
SES	65*	.00	.00		
Mean	48.66	.333	.333	.500	
SD	10.129	.492	.492	.522	

Note: SES is a dummy variable (free/reduced lunch = 1, otherwise = 0), as are B (1, 0 = other) and C (1, 0 = other); n = 12.

Table 2b

Regression of Achievement on Reading Frequency and SES

Variable	b	se	ΔR^2	95%CI	F	t
SES	-12.66	5.16	.426	-24.57, -0.77		-2.45*
Reading Freq.			.007		0.05	
В	-1.75	6.32		-16.32, 12.82		-0.28
С	-0.00	6.32		-14.57, 14.57		0.00
Intercept	55.58	5.16		43.68, 67.48		10.77*

Note: $R^2 = .43$, adj. $R^2 = .22$, $F_{3,8} = 2.04$, MSE = 79.89, n = 12. ΔR^2 represents the semi-partial correlation or the increment in R^2 due to adding the respective variable. SES is a dummy variable (free/reduced lunch = 1, otherwise = 0), as are B (1, 0 = other) and C (1, 0 = other).

<u>*p</u> < .05.

Table 2c

	3 3 1	,	
Contrast	Estimated Mean	Standard Error of	.95Cl
	Difference	Difference	
B vs. A	-1.75	6.32	-16.32, 12.82
C vs. A	-0.00	6.32	-14.57, 14.57
B vs. C	-1.75	6.32	-16.32, 12.82
* 05			

*p < .05.

[Note – the above comparison represents the unadjusted comparisons (no Bonferroni corrections); these numbers obtained from regression output. Bonferroni adjusted comparisons reported below in 2d.]

Table 2d

Comparison	Estimated Mean	Standard Error of	Bonferroni Adjusted
	Difference	Difference	95% CI
A vs. c	0.00	6.32	-18.99, 18.99
B vs. c	-1.75	6.32	-20.74, 17.24
A vs b	1.75	6.32	-17.24, 20.74

*p<.05, where p-values are adjusted using the Bonferroni method.

Only SES was statistically related to achievement scores, with those receiving free for reduced lunch scoring about 12 to 13 points lower than those not receiving free/reduced lunch, on average. There were no statistical differences observed among the three levels of reading frequency.

Bonferroni and Scheffe adjusted confidence intervals are reported below.

. regr achievement i.read_freq_num i.ses_num
. pwcompare read_freq_num, bonf

Pairwise comparisons of marginal linear predictions

	Contrast	Std. Err.	Bonfe [95% Conf.	-
+-				
read freq num				
	-1.75	6.320436	-20.81093	17.31093
3 vs 1	7.07e-15	6.320436	-19.06093	19.06093
			-17.31093	
	±•, •			
	<i>c</i>	,		
. pwcompare read	_freq_num, s	sch		
. pwcompare read		sch		
. pwcompare read	freq_num, s	sch	Sche	 ffe
. pwcompare read 			Sche [95% Conf.	
. pwcompare read 				
read_freq_num	Contrast	Std. Err.	[95% Conf.	Interval]
read_freq_num 2 vs 1	Contrast 	Std. Err. 6.320436	[95% Conf. -20.62467	Interval] 17.12467
read_freq_num 2 vs 1 3 vs 1	Contrast -1.75 7.07e-15	Std. Err. 6.320436 6.320436	[95% Conf. -20.62467 -18.87467	Interval] 17.12467 18.87467
read_freq_num 2 vs 1 3 vs 1	Contrast -1.75 7.07e-15	Std. Err. 6.320436	[95% Conf. -20.62467 -18.87467	Interval] 17.12467 18.87467

(3) Results for GPA analysis.

Table 3a

Descriptive Statistics for GPA, SES, and Behavioral Problems

Variable	Correlations			
	GPA	High	Mid	Behavior
GPA				
High SES	.78*			
Mid. SES	07	53		
Behavior	46	10	62*	
Mean	2.56	0.39	0.31	0.46
SD	0.74	0.51	0.48	0.52

Note: High (1, 0 = otherwise) and Mid. SES (1, 0 = otherwise) are dummy variables, as is behavior (1 for problems, 0 = otherwise); n = 13.

Table 3b

Regression of GPA on Behavioral Problems and SES

Variable	b	se	ΔR^2	95%CI	F	t
Behavioral Prob.	27	.37	.01	-1.10, 0.57		-0.72
SES			.56		11.31*	
High	1.34*	.35		0.54, 2.13		3.81*
Mid	.46	.47		-0.60, 1.51		0.98
Intercept	2.03	.42		1.08, 2.98		4.83*

Note: $R^2 = .78$, adj. $R^2 = .70$, $F_{3,9} = 10.35^*$, MSE = 0.164, n = 13. ΔR^2 represents the semi-partial correlation or the increment in R^2 due to adding the respective variable. *p < .05.

Table 3c

Comparisons of	of Achievement	among Reading	Frequency

Estimated Mean	Standard Error of	95% CI of Mean
Difference	Difference	Difference
1.34*	0.35	0.54, 2.21
0.46	0.47	60, 1.51
0.88	0.31	.18, 1.58
	Difference 1.34* 0.46	Difference Difference 1.34* 0.35 0.46 0.47

*p < .05.

Only SES was statistically related to GPA, with those in the high SES group showing statistically higher GPAs than either the middle or low SES groups. There was no statistical difference between the middle and low SES groups, nor was behavioral problem associated with GPA.

[Table 3c above are the unadjusted comparisons, Table 3d below contains the Bonferroni adjusted comparisons using the estimated means with behavioral problems mean used as 0.46 to obtained predicted means for each of the three SES groups.]

[Again note that the Excel spreadsheet se are too small and erroneous, so use tabled Bonferroni critical t and calculate CI using regression se.]

Table 3cComparisons of Achievement among Reading Frequency

Contrast	Estimated Mean	Standard Error of	Bonferroni Adjusted
	Difference	Difference	95% CI
High vs. Low	1.34*	0.35	0.31, 2.36
Mid vs. Low	0.46	0.47	-0.91, 1.83
High vs. Mid	0.88	0.31	-0.03, 1.79

*p<.05, where p-values are adjusted using the Bonferroni method.

[Bonferroni critical t = 2.923 (3 comparions, 9 df)]

Scheffé confidence intervals are reported below.

Table 3c

Comparisons of Achievement among Reading Frequency

Contrast	Estimated Mean	Standard Error of	Bonferroni Adjusted
	Difference	Difference	95% CI
High vs. Low	1.34*	0.35	0.31, 2.36
Mid vs. Low	0.46	0.47	-0.91, 1.82
High vs. Mid	0.88	0.31	-0.02, 1.78

*p<.05, where p-values are adjusted using the Scheffé method.