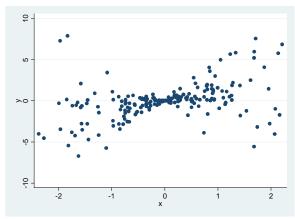
# Sociology 63993 Exam 1 February 22, 2007

- *I. True-False.* (20 points) Indicate whether the following statements are true or false. If false, briefly explain why.
- 1. An outlier on Y will have the most effect on the regression line when its value for X is equal to the mean of X.
- 2. Serial correlation will not affect the unbiasedness or consistency of OLS estimators, but it does affect their efficiency.
- 3. Increasing the number of items in a scale will always increase the value of Cronbach's Alpha.
- 4. Religion has four categories: Catholic, Protestant, Jewish and Other. The researcher wants to use Catholic as her reference category. Therefore, she could compute 3 dummy variables: Protestant, Jewish and Other. On each of these variables, Catholics should be coded as missing.
- 5. A researcher conducts the following analysis:
- . scatter y x



- . quietly reg y x
- . hettest

```
Breusch-Pagan / Cook-Weisberg test for heteroskedasticity
    Ho: Constant variance
    Variables: fitted values of y

chi2(1) = 0.18
    Prob > chi2 = 0.6721
```

Based on the above, she can conclude that heteroskedasticity is probably not a problem with her data.

*II.* Short answer. Discuss all three of the following five problems. (15 points each, 45 points total,) In each case, the researcher has used Stata to test for a possible problem, concluded that there is a problem, and then adopted a strategy to address that problem. Explain (a) what problem the researcher was testing for, and why she concluded that there was a problem, (b) the rationale behind the solution she chose, i.e. how does it try to address the problem, and (c) one alternative solution she could have tried, and why. (NOTE: a few sentences on each point will probably suffice – you don't have to repeat everything that was in the lecture notes.)

II-1.

#### . reg y x1 x2 x3

Source	SS	df		MS		Number of obs F( 3, 3971)		3975 24.14
Model   Residual	7240.36972 397049.865	3 3971		.45657 873746		Prob > F R-squared Adj R-squared	= =	0.0000 0.0179 0.0172
Total	404290.234	3974	101.	733828		Root MSE	=	9.9994
У	Coef.	Std.	Err.	t	P> t	[95% Conf.	In	terval]
x1   x2   x3   _cons	.4211835 .8435702 1.025274 .2881428	.8683 .8633 .7932 .2493	128	0.49 0.98 1.29 1.16	0.628 0.329 0.196 0.248	-1.281205 8490077 5298791 2006937	2	.123572 .536148 .580428 7769793

### . alpha x1 x2 x3, i gen(xscale)

Test scale = mean(unstandardized items)

Item	Obs	Sign	item-test correlation	item-rest correlation	average inter-item covariance	alpha
x1 x2	-+   3975   3975	+ +	0.8006 0.8079	0.6392 0.6466	.3240937	0.8219 0.8135
x3	3975 -+	+	0.9928	0.9702	.0760545	0.4777
Test scale					.2385009	0.8263

## . reg y xscale

Source	SS	df		MS		Number of obs F( 1, 3973)		3975 71.79
Model Residual	397114.534	1 3973	99.95			Prob > F R-squared Adj R-squared	= =	0.0000 0.0177 0.0175
Total	404290.234	3974	101.7	33828		Root MSE		9.9977
У	Coef.	Std.	Err.	t	P> t	[95% Conf.	In	terval]
xscale _cons	2.501135 .3140862	.2951		8.47 1.29	0.000	1.922395 1648963	_	.079876

# II-2.

# . reg y2 x11

Source	SS	df	MS		Number of obs F( 1, 3973)		3975 11.39
Model   Residual	21109.1326 7361172.91	1 3973	21109.132 1852.799		Prob > F R-squared Adj R-squared	=	0.0007 0.0029 0.0026
Total	7382282.04	3974	1857.64	52	Root MSE	=	43.044
y2	Coef.	Std.	Err.	t P> t	[95% Conf.	In	terval]
x11   _cons	2.790641 0881977	.8267 1.03		.38 0.001 .09 0.932	1.169714 -2.121672	_	.411567 .945276

## . dfbeta

DFx11: DFbeta(x11)

## . extremes DFx11 y2 x11

_					L
	obs:	DFx11	y2	x11	
	2100. 619. 3124. 3828. 3739.	0235065 0191516 0170745 0157568 0153048	-25.31333 33.0916 -23.67617 -15.25209 -20.67157	2.568708 3147684 2.227267 2.531287 2.223721	
+	2008. 3950. 906. 2546.	.0140591 .0141953 .0145061 .0156778 6.147187	35.20511 -30.203 -26.45573 -29.41599 2643.918	2.00463   1235355   3324702   286212   2.15582	

# . replace y2 = y2/100 in 10 (1 real change made)

# . reg y2 x11

Source	SS	df		MS		Number of obs F( 1, 3973)		3975 71.23
Model   Residual	7120.96787 397169.267	1 3973		0.96787 9670945		Prob > F R-squared Adj R-squared	= =	0.0000 0.0176 0.0174
Total	404290.234	3974	101.	733828		Root MSE		9.9984
y2	Coef.	Std.	Err.	t	P> t	[95% Conf.	In	terval]
x11   _cons	1.620833 .3580865	.1920		8.44 1.49	0.000	1.244322 1142516	_	.997344 8304245

# II-3.

# . reg income skills scale01

Source	SS	df	MS		Number of obs	
Model   Residual   Total	444.0924 162.827596 	7 23.	22.0462 2610852  4355552		F( 2, 7) Prob > F R-squared Adj R-squared Root MSE	= 0.0100 = 0.7317
income	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
skills   scale01   _cons	.6777237 3.271439 -8.71115	.3230182 .8299957 6.793964	2.10 3.94 -1.28	0.074 0.006 0.241	086093 1.308811 -24.77632	1.44154 5.234067 7.354021

#### . list

-	+   income	black	skills	scale01	scale02
1. 2. 3. 4.	5 9.7 28.4 8.8	black black black black black	9 18 21 12 14	1.741101 3.031433 5.278032 5.278032 5.278032	· · · ·
6. 7. 8. 9.	26.6 25.4 23.1 22.5 19.5	black black black black black	16 16 9 18 5	6.309574 7.300372 7.300372 7.300372 7.300372	·   .   .   .   .   .
11. 12. 13. 14. 15.	21.7 24.8 30.1 24.8 28.5	white white white white white	7 9 12 17 19	: : : :	7.300372 7.783137 8.258524 8.258524 8.727161
16. 17. 18. 19.	26   38.9   22.1   33.1   48.3	white white white white white	6 17 1 10 17	· · · ·	8.727161 9.189587 9.189587 9.646264 11.42288

## . sum

Variable	Obs	Mean	Std. Dev.	Min	Max
income	20	24.415	9.788354	5	48.3
black	20	.5	.5129892	0	1
skills	20	12.65	5.460625	1	21
scale01	10	5.611769	1.93984	1.741101	7.300372
scale02	10	8.850319	1.142792	7.300372	11.42288

## . gen xscale01 = scale01

(10 missing values generated)

# . replace xscale01 = 5.611769 if missing(scale01)

(10 real changes made)

. gen md = 0

# . replace md = 1 if missing(scale01)

(10 real changes made)

## . reg income skills xscale01 md

Source	ss	df	MS		Number of obs F( 3, 16)	
Model Residual	1242.10592 578.319556		.035306		Prob > F R-squared	= 0.0003 = 0.6823
Total	1820.42548	19 95.	8118671	Adj R-squared Root MSE		= 6.0121
income	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
skills xscale01 md _cons	.7629064 3.283386 12.58468 -9.953716	.258854 1.033725 2.753807 7.175127	2.95 3.18 4.57 -1.39	0.009 0.006 0.000 0.184	.2141605 1.091988 6.746875 -25.1643	1.311652 5.474785 18.42249 5.256874

# III. Computation and interpretation. (35 points total)

Both Sociologists and Public Health researchers are interested in the determinants of self-reported health. The NHANES2F data, available from Stata's web site, includes information on the following.

Variable	Description
health	Self-reported health. Values range from 1 (poor health) to 5 (excellent health)
age	Age in years
female	Coded 1 if female, 0 otherwise
black	Coded 1 if black, 0 otherwise
rural	Coded 1 if respondent lives in a rural area, 0 otherwise

[NOTE: These data are weighted and proper analysis should take that into account. In addition, the dependent variable is ordinal and would probably be better analyzed by ordinal regression methods that we will talk about later. For simplicity, we will ignore such details for now.]

An analysis of the data yields the following results.

- . webuse nhanes2f, clear
- . keep health age female black rural

# . corr , means (obs=10335)

Variable	Mean	Std. Dev.	Min	Max
health age	3.413836 47.56584	1.206196 17.21752	1 20	5 74
female	.5250121	.4993982	0	1
black	.1050798	.3066711	0	1
rural	.3672956	.4820913	0	1

	health	age	female	black	rural
health	1.0000				
age	-0.3686	1.0000			
female	-0.0320	0.0090	1.0000		
black	-0.1286	-0.0321	0.0100	1.0000	
rural	-0.0827	0.0565	-0.0341	-0.1838	1.0000

# . alpha female black rural, i gen(demscale)

Test scale = mean(unstandardized items)

Item	Obs	Sign	item-test correlation	item-rest correlation	average inter-item covariance	alpha
female black rural	10337   10337   10337	+ + -	0.6438 0.4966 0.6892	0.0316 0.1326 0.1247	.0271779 .0082063 .0015225	0.2855 0.0659 0.0176
Test scale					.0123022	0.1704

### . drop demscale

# . pcorr2 health age female black rural

(obs=10335)

Partial and Semipartial correlations of health with

Variable	Partial	SemiP	Partial^2	SemiP^2	Sig.
age	-0.3730	-0.3675	0.1391	0.1350	0.000
female	-0.0331	-0.0303	0.0011	0.0009	0.001
black	-0.1664	-0.1542	0.0277	0.0238	0.000
rural	-0.0981	-0.0901	0.0096	0.0081	0.000

#### . reg health age female black rural

#### . test rural

- a) (10 pts) Fill in the missing quantities [1] [5].
- b) (25 points) Answer the following questions about the analysis and the results, explaining how the printout supports your conclusions.
- 1. Briefly interpret the results, i.e. explain how race, gender, age and geographic location affect self-reported health. What types of people report the highest levels of health and which types of people report the lowest?
- 2. Suppose that, 20 years from now, your friend decides to finally move away from the big city and live in a small farm out in the country. His mother thinks it is a terrible idea but at least she manages to talk him out of having a sex change operation too. According to the above model, how much higher/lower can your friend expect his health score to be then than it is now?
- 3. The researcher created a variable called demscale but then immediately deleted it. Why did he do this?
- 4. Suppose the researcher now ran backwards stepwise regression using the .05 level of significance, i.e. gave the command
- . sw, pr(.05): reg health age female black rural

How would the results differ from the regression reported above?

5. Suppose that in previous studies it has been found that, after controlling for other variables, on average women score a tenth of a point lower on health than do men. The researcher therefore decides to test

$$H_0$$
:  $\beta_{female} = -.10$   
 $H_A$ :  $\beta_{female} \neq -.10$ 

Based on the results presented above and using the .05 level of significance, should the researcher reject or not reject the null hypothesis?

# IV. Extra Credit. (Up to 10 points.)

Y is coded 0 if failed, 1 if succeeded. X1, X2 and X3 are explanatory variables. An analysis of these data reveals the following:

## . reg y x1 x2 x3

Source	SS	df	MS		Number of obs	
Model Residual Total	3.1589065   4.0598435 +	28 .144	296883 994411  862903		F( 3, 28) Prob > F R-squared Adj R-squared Root MSE	= 0.0009 = 0.4376
У	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
x1 x2 x3 _cons	.0348776 .0088082 .3690164 6336808	.011257 .0191632 .1365854 .3923563	3.10 0.46 2.70 -1.62	0.004 0.649 0.012 0.118	.0118187 0304459 .089234 -1.437386	.0579365 .0480622 .6487988 .1700247

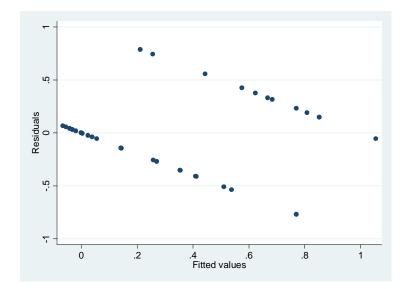
### . predict yhat

(option xb assumed; fitted values)

- . predict e, resid
- . extremes yhat e y

+			
obs:	yhat	е	У
3.   6.   5.   8.   4.	0662836 0550311 0425547 0328912 0198185	.0662836 .0550311 .0425547 .0328912 .0198185	0 0 0 0 0
+   25.   27.   26.   29.   31.	.7690787 .769383 .8080336 .851562 1.054008	.2309213 769383 .1919664 .148438 0540077	1 0 1 1 1

## . rvfplot



- a) Why does the plot of residuals versus fitted values (i.e. yhat versus e) look the way it does? [HINT: Any OLS regression using a binary dependent variable is going to look more or less like the above. Think about why that has to be.]
- b) Does anything in the above analysis suggest any problems with the use of OLS regression with binary dependent variables, e.g. are any OLS assumptions being violated, are the estimates sensible? [HINT: the yhat values can be interpreted as the predicted probability of success given the values of the x variables, e.g. a yhat value of .73 implies a 73% chance of succes.]