

PROBLEM SET 3: SOLUTIONS

Point Distribution:

- 1), 2) : 5 points each
- 3), 4): 10 points each
- 5), 6): 15 points each

1) Denote the true value of  $\beta_1$  and  $\beta_2$  by  $\beta_{10}$  and  $\beta_{20}$ , respectively. Then.

$$c_i = \beta_{10} + \beta_{20}y_i^* + u_i^* = \beta_{10} + \beta_{20}y_i + (u_i^* - \beta_{20}\nu_i)$$

where we have used the equation  $y_i = y_i^* + \nu_i$ . Thus, if we run the regression of  $c_i$  on  $y_i$  and a constant, we have the error term in the regression,  $u_i = u_i^* - \beta_{20}\nu_i$ . Now, the covariance of  $y_i$  and  $u_i$  is calculated below:

$$\begin{aligned} \text{cov}(y_i, u_i) &= E[(y_i^* + \nu_i)(u_i^* - \beta_{20}\nu_i)] - E[(y_i^* + \nu_i)] E[(u_i^* - \beta_{20}\nu_i)] & (1) \\ &= -\beta_{20}\omega^2 & (2) \end{aligned}$$

since  $\text{cov}(\nu_i, y_i^*) = 0$ ,  $\text{cov}(\nu_i, u_i^*) = 0$ , and  $\text{cov}(u_i^*, y_i^*) = 0$ , and since  $u_i^*$  and  $\nu_i$  are mean-zero.

Since,  $\beta_{20} > 0, \omega^2 > 0$ , we have that  $\text{cov}(y_i, u_i) < 0$ , which implies that the correlation is negative.

2)

```
. ivreg2 M Y L.M L2.M (R = L.R L2.R)
```

IV (2SLS) estimation

-----

					Number of obs =	126
					F( 4, 121) =	11731.54
					Prob > F =	0.0000
Total (centered) SS	=	13.09922457			Centered R2 =	0.9974
Total (uncentered) SS	=	15967.60984			Uncentered R2 =	1.0000
Residual SS	=	.0336859252			Root MSE =	.01635

-----

M	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
R	-.0038475	.0007997	-4.81	0.000	-.0054149    -.0022802

-----

```

      Y |   .0674069   .0126278    5.34  0.000   .0426568   .0921569
      M |
    L1. |   1.404471   .0764595   18.37  0.000   1.254613   1.554329
    L2. |  -.4583713   .0719214   -6.37  0.000  -.5993346  -.3174081
  _cons |  -.2502209   .077605    -3.22  0.001  -.4023238  -.098118

```

```

-----
Anderson canon. corr. LR statistic (underidentification test):      178.178
                                                                Chi-sq(2) P-val =    0.0000
-----

```

```

Cragg-Donald F statistic (weak identification test):                186.771
Stock-Yogo weak ID test critical values: 10% maximal IV size      19.93
                                           15% maximal IV size      11.59
                                           20% maximal IV size       8.75
                                           25% maximal IV size       7.25

```

Source: Stock-Yogo (2005). Reproduced by permission.

```

-----
Sargan statistic (overidentification test of all instruments):      15.464
                                                                Chi-sq(1) P-val =    0.0001
-----

```

```

Instrumented:      R
Included instruments: Y L.M L2.M
Excluded instruments: L.R L2.R

```

```
. ivendog
```

Tests of endogeneity of: R

H0: Regressor is exogenous

Wu-Hausman F test: 0.08902 F(1,120) P-value = 0.76595

Durbin-Wu-Hausman chi-sq test: 0.09340 Chi-sq(1) P-value = 0.75990

The rejection of the null in the Sargan test indicates that the excluded instruments are not valid instruments. The Durbin-Wu-Hausman test of exogeneity of  $r_t$  i.e. of the appropriateness of OLS fails to reject this null hypothesis, suggesting that IV regression is not needed. Since one would expect that lagged values of a variable to be reasonable instruments, the results of these two tests suggests that the original model might have been misspecified, and that the lagged values of  $r_t$  that were used as instruments ought not to have been omitted from the regression model.

3)

```
. regress logQty X_2 X_3 logPrice
```

Source	SS	df	MS	Number of obs =	120
Model	9.1604655	3	3.0534885	F( 3, 116) =	116.03
Residual	3.05268076	116	.026316213	Prob > F =	0.0000
				R-squared =	0.7500
				Adj R-squared =	0.7436

Total | 12.2131463 119 .102631481 Root MSE = .16222

logQty	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
X_2	.3830345	.0226551	16.91	0.000	.3381632	.4279058
X_3	-.197489	.0343902	-5.74	0.000	-.265603	-.1293749
logPrice	-.3951405	.025181	-15.69	0.000	-.4450146	-.3452665
_cons	3.4532	.1234017	27.98	0.000	3.208788	3.697613

. ivreg2 logQty X\_2 X\_3 (logPrice = X\_4 X\_5)

IV (2SLS) estimation

Total (centered) SS	=	12.21314626	Number of obs =	120
Total (uncentered) SS	=	3303.294588	F( 3, 116) =	97.07
Residual SS	=	4.686889168	Prob > F	= 0.0000
			Centered R2	= 0.6162
			Uncentered R2	= 0.9986
			Root MSE	= .1976

logQty	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
logPrice	-.5935736	.0389166	-15.25	0.000	-.6698487	-.5172985
X_2	.491335	.0305378	16.09	0.000	.4314819	.551188
X_3	-.2252342	.0420297	-5.36	0.000	-.307611	-.1428575
_cons	3.061598	.157588	19.43	0.000	2.752731	3.370465

Anderson canon. corr. LR statistic (underidentification test): 116.545  
Chi-sq(2) P-val = 0.0000

Cragg-Donald F statistic (weak identification test): 94.366  
Stock-Yogo weak ID test critical values: 10% maximal IV size 19.93  
15% maximal IV size 11.59  
20% maximal IV size 8.75  
25% maximal IV size 7.25

Source: Stock-Yogo (2005). Reproduced by permission.

Sargan statistic (overidentification test of all instruments): 0.975  
Chi-sq(1) P-val = 0.3235

Instrumented: logPrice  
Included instruments: X\_2 X\_3

Excluded instruments: X\_4 X\_5

-----  
. ivendog

Tests of endogeneity of: logPrice

H0: Regressor is exogenous

Wu-Hausman F test: 831.99234 F(1,115) P-value = 0.00000

Durbin-Wu-Hausman chi-sq test: 105.42755 Chi-sq(1) P-value = 0.00000

The null of the Durbin-Wu-Hausman test of exogeneity of  $p_t$  is overwhelmingly rejected, which implies that the OLS estimation is not valid. This can be seen also by comparing the coefficients from the OLS regression with those from the IV regression. The differences between the two are much larger than the standard errors of the OLS coefficients, which should not be the case if the OLS estimates are consistent. The null of the Sargan test for validity of the instruments is not rejected at any reasonable level, which indicates that the instruments, and the IV regression, is valid.

. regress logPrice X\_2 X\_3 logQty

Source	SS	df	MS	Number of obs =	120
Model	52.6754551	3	17.558485	F( 3, 116) =	153.25
Residual	13.2904765	116	.114573073	Prob > F =	0.0000
-----				R-squared =	0.7985
-----				Adj R-squared =	0.7933
Total	65.9659316	119	.55433556	Root MSE =	.33849

logPrice	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
X_2	.8337184	.0418204	19.94	0.000	.7508878	.9165491
X_3	-.3845204	.0730633	-5.26	0.000	-.5292316	-.2398093
logQty	-1.720326	.1096305	-15.69	0.000	-1.937463	-1.503189
_cons	5.308666	.5204841	10.20	0.000	4.277782	6.33955

-----  
. ivreg2 logPrice X\_2 X\_3 (logQty = X\_4 X\_5)

IV (2SLS) estimation

	Number of obs =	120	
	F( 3, 116) =	145.70	
	Prob > F =	0.0000	
Total (centered) SS =	65.96593159	Centered R2 =	0.7983
Total (uncentered) SS =	294.8998539	Uncentered R2 =	0.9549

Residual SS = 13.30781268 Root MSE = .333

---

logPrice	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
logQty	-1.677681	.1102461	-15.22	0.000	-1.893759	-1.461603
X_2	.8265807	.0413213	20.00	0.000	.7455926	.9075689
X_3	-.3784546	.0719554	-5.26	0.000	-.5194846	-.2374247
_cons	5.12815	.5211023	9.84	0.000	4.106808	6.149492

---

Anderson canon. corr. LR statistic (underidentification test): 378.000  
Chi-sq(2) P-val = 0.0000

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Cragg-Donald F statistic (weak identification test): 1284.323  
Stock-Yogo weak ID test critical values: 10% maximal IV size 19.93  
15% maximal IV size 11.59  
20% maximal IV size 8.75  
25% maximal IV size 7.25

Source: Stock-Yogo (2005). Reproduced by permission.

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Sargan statistic (overidentification test of all instruments): 0.970  
Chi-sq(1) P-val = 0.3246

---

Instrumented: logQty  
Included instruments: X\_2 X\_3  
Excluded instruments: X\_4 X\_5

---

. ivendog

Tests of endogeneity of: logQty

H0: Regressor is exogenous

Wu-Hausman F test: 3.45111 F(1,115) P-value = 0.06577

Durbin-Wu-Hausman chi-sq test: 3.49624 Chi-sq(1) P-value = 0.06151

The DWH test rejects the null hypothesis that the OLS regression is consistent at the 10% level, but not the 5% level. Therefore exogeneity of  $q_t$  is doubtful, but in contrast to the previous specification, the level of correlation between the suspect regressor and the error term is much lower. An examination of the difference in the parameter estimates between the OLS and the IV regressions supports the notion that the OLS estimates are only slightly biased. The Sargan test indicates that the instrument set used is valid, and so the IV regression is valid.

Rewriting the demand equation as an inverse demand equation, we obtain

$$p_t = -\frac{\beta_1}{\gamma} - \frac{\beta_2}{\gamma}x_{t2} - \frac{\beta_3}{\gamma}x_{t3} + \frac{1}{\gamma}q_t - \frac{1}{\gamma}u_t$$

Thus, estimating the inverse demand equation

$$p_t = \beta_1^* + \beta_2^* x_{t2} + \beta_3^* x_{t3} + \gamma^* q_t + v_t$$

we obtain the following relationships between the parameters of the two regression models:

$$\beta_i^* = -\frac{\beta_i}{\gamma}, \quad \gamma^* = \frac{1}{\gamma} \implies \beta_i = -\frac{\beta_i^*}{\gamma^*}$$

Thus, we have four estimates of  $\gamma$ :

$$\hat{\gamma}_{OLS} = -0.3951, \quad \hat{\gamma}_{2SLS} = -0.5936, \quad \frac{1}{\hat{\gamma}_{OLS}^*} = -\frac{1}{1.7203} = -0.5813, \quad \frac{1}{\hat{\gamma}_{2SLS}^*} = -\frac{1}{1.6777} = -0.5961$$

It is a little surprising that the estimate obtained from OLS estimation of the inverse demand function matches so closely the estimates obtained from the two IV(2SLS) regressions. This adds support to the earlier finding that the OLS estimation of the inverse demand equation is adequate. However, it is clear that OLS estimation of the demand equation yields biased parameter estimates.

4) a)

. regress lwage educ exper tenure married black south urban

Source	SS	df	MS	Number of obs =	935
Model	41.8377677	7	5.97682396	F( 7, 927) =	44.75
Residual	123.818527	927	.133569069	Prob > F =	0.0000
Total	165.656294	934	.177362199	R-squared =	0.2526
				Adj R-squared =	0.2469
				Root MSE =	.36547

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
educ	.0654307	.0062504	10.47	0.000	.0531642 .0776973
exper	.014043	.0031852	4.41	0.000	.007792 .020294
tenure	.0117473	.002453	4.79	0.000	.0069333 .0165613
married	.1994171	.0390502	5.11	0.000	.1227801 .2760541
black	-.1883499	.0376666	-5.00	0.000	-.2622717 -.1144282
south	-.0909036	.0262485	-3.46	0.001	-.142417 -.0393903
urban	.1839121	.0269583	6.82	0.000	.1310056 .2368185
_cons	5.395497	.113225	47.65	0.000	5.17329 5.617704

Ceteris paribus, the approximate difference in the log wage of blacks and nonblacks is -0.18835, where blacks receive a lower wage. The difference is statistically significant, as the p-value of the t-test for significance is basically 0 i.e. the null of the coefficient on black being zero is rejected at 0.1% level. Going from log wages to wages, we

obtain that ceteris paribus the ratio of wages of blacks to nonblacks is  $e^{-0.18835} = 0.8283$  i.e. about blacks earn about 17% lower than non-blacks, all other things being equal.

b)

```
. gen blackXeduc = black * educ
```

```
. regress lwage educ blackXeduc exper tenure married black south urban
```

Source	SS	df	MS	Number of obs =	935
Model	42.0055536	8	5.2506942	F( 8, 926) =	39.32
Residual	123.650741	926	.133532117	Prob > F =	0.0000
				R-squared =	0.2536
				Adj R-squared =	0.2471
Total	165.656294	934	.177362199	Root MSE =	.36542

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
educ	.0671153	.0064277	10.44	0.000	.0545008	.0797299
blackXeduc	-.0226237	.0201827	-1.12	0.263	-.0622327	.0169854
exper	.0138259	.0031906	4.33	0.000	.0075642	.0200876
tenure	.011787	.0024529	4.81	0.000	.0069732	.0166009
married	.1989077	.0390474	5.09	0.000	.1222761	.2755394
black	.0948094	.2553995	0.37	0.711	-.4064194	.5960383
south	-.0894495	.0262769	-3.40	0.001	-.1410187	-.0378803
urban	.1838523	.0269547	6.82	0.000	.130953	.2367516
_cons	5.374817	.1147027	46.86	0.000	5.149709	5.599924

```
. test black blackXeduc
```

```
( 1) black = 0
( 2) blackXeduc = 0
```

```
F( 2, 926) = 13.13
Prob > F = 0.0000
```

The returns to education for blacks is lower than that of whites, but the difference is not statistically significantly different from zero at any reasonable level. It is interesting to note that the negative effect of being black on the log wage is no longer negative or significant. Jointly testing to see if there is an effect of being black, we reject the hypothesis that the regression is stable over the category of race. Thus, even though neither `black` nor `blackXeduc` were significantly different from zero individually, they are jointly significantly different from zero.

c)

```

. gen MB = married * black
. gen mB = (1 - married) * black
. gen Mb = married * (1 - black)
. gen mb = (1 - married ) * (1 - black)

. regress lwage educ exper tenure MB Mb mB south urban

```

Source	SS	df	MS	Number of obs =	935
Model	41.8849419	8	5.23561773	F( 8, 926) =	39.17
Residual	123.771352	926	.133662368	Prob > F =	0.0000
Total	165.656294	934	.177362199	R-squared =	0.2528
				Adj R-squared =	0.2464
				Root MSE =	.3656

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
educ	.0654751	.006253	10.47	0.000	.0532034	.0777469
exper	.0141462	.003191	4.43	0.000	.0078837	.0204087
tenure	.0116628	.0024579	4.74	0.000	.006839	.0164866
MB	.0094485	.0560131	0.17	0.866	-.1004788	.1193757
Mb	.1889147	.0428777	4.41	0.000	.1047659	.2730635
mB	-.2408201	.0960229	-2.51	0.012	-.4292678	-.0523724
south	-.0919894	.0263212	-3.49	0.000	-.1436455	-.0403333
urban	.1843501	.0269778	6.83	0.000	.1314053	.2372948
_cons	5.403793	.1141222	47.35	0.000	5.179825	5.627761

```

. lincom MB -Mb

```

```

( 1) MB - Mb = 0

```

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
(1)	-.1794663	.0405386	-4.43	0.000	-.2590244	-.0999082

The log wage differential between married black and married nonblacks is -0.1795, and is significantly different from zero. In terms of wage levels, this difference translates to a 16.5% lower wage for married blacks than for married nonblacks ( $e^{-0.1795} = 0.8357$ ).



5) a) One would expect that higher SAT scores would imply higher college GPA. Similarly, lower academic percentile (measured as the percentage who performed better than the student in question) would suggest better college performance as measured by college GPA. It is unclear how size (or the square of the size) of the high-school class would affect college GPA. One might suppose that being an athlete would detract from studying, which would lower college GPA. It is unclear how gender would affect GPA.

b)

```
. regress colgpa hsize hsize^2 hspc sat female athlete
```

Source	SS	df	MS			
Model	524.819305	6	87.4698842	Number of obs =	4137	
Residual	1269.37637	4130	.307355053	F( 6, 4130) =	284.59	
Total	1794.19567	4136	.433799728	Prob > F =	0.0000	
				R-squared =	0.2925	
				Adj R-squared =	0.2915	
				Root MSE =	.5544	

  

colgpa	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
hsize	-.0568543	.0163513	-3.48	0.001	-.0889117	-.0247968
hsize^2	.0046754	.0022494	2.08	0.038	.0002654	.0090854
hspc	-.0132126	.0005728	-23.07	0.000	-.0143355	-.0120896
sat	.0016464	.0000668	24.64	0.000	.0015154	.0017774
female	.1548814	.0180047	8.60	0.000	.1195826	.1901802
athlete	.1693064	.0423492	4.00	0.000	.0862791	.2523336
_cons	1.241365	.0794923	15.62	0.000	1.085517	1.397212

The estimated GPA differential between athletes and nonathletes is 0.169, and is statistically significant at the 0.1% level.

c)

```
. regress colgpa hsize hsize^2 hspc female athlete
```

Source	SS	df	MS			
Model	338.217123	5	67.6434246	Number of obs =	4137	
Residual	1455.97855	4131	.35245184	F( 5, 4131) =	191.92	
Total	1794.19567	4136	.433799728	Prob > F =	0.0000	
				R-squared =	0.1885	
				Adj R-squared =	0.1875	
				Root MSE =	.59368	

  

colgpa	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
hsize	-.0568543	.0163513	-3.48	0.001	-.0889117	-.0247968
hsize^2	.0046754	.0022494	2.08	0.038	.0002654	.0090854
hspc	-.0132126	.0005728	-23.07	0.000	-.0143355	-.0120896
sat	.0016464	.0000668	24.64	0.000	.0015154	.0017774
female	.1548814	.0180047	8.60	0.000	.1195826	.1901802
athlete	.1693064	.0423492	4.00	0.000	.0862791	.2523336
_cons	1.241365	.0794923	15.62	0.000	1.085517	1.397212

hsize		-.0534038	.0175092	-3.05	0.002	-.0877313	-.0190763
hsizesq		.0053228	.0024086	2.21	0.027	.0006007	.010045
hsperc		-.0171365	.0005892	-29.09	0.000	-.0182916	-.0159814
female		.0581231	.0188162	3.09	0.002	.0212333	.095013
athlete		.0054487	.0447871	0.12	0.903	-.0823582	.0932556
_cons		3.047698	.0329148	92.59	0.000	2.983167	3.112229

---

Yes, the effect is greatly lessened, though still positive. However, the estimate is not statistically significantly different from zero at any reasonable level.

Why would this happen? It suggests that `athlete` and `sat` are negatively correlated. What we have is omitted variable bias. Suppose the true specification of some process is

$$y = \beta_1 + \beta_2 x_2 + \beta_3 x_3 + \epsilon$$

Suppose instead we run the regression

$$y = \alpha_1 + \alpha_2 x_2 + \nu$$

Then,  $\nu = \beta_3 x_3 + \epsilon$ , and our estimate of  $\alpha_2$  will not be unbiased.

$$\begin{aligned} \hat{\alpha}_2 &= \frac{\text{cov}(y, x_2)}{\text{var}(x_2)} = \frac{\text{cov}(\alpha_1 + \alpha_2 x_2 + \nu, x_2)}{\text{var}(x_2)} \\ &= \alpha_2 + \beta_3 \frac{\text{cov}(x_3, x_2)}{\text{var}(x_2)} = \alpha_2 + \beta_3 \rho_{x_2, x_3} \sqrt{\frac{\text{var}(x_3)}{\text{var}(x_2)}} \end{aligned}$$

So, what must be happening to have `athlete` become insignificantly different from 0 is that the negative correlation between the `athlete` and `sat` is biasing the estimated effect of begin an athlete in this variant of the model. The following unconditional correlation matrix and conditional correlation (via the regression) both indicate the negative correlation necessary for such a biased estimate result.

```
. corr athlete sat
(obs=4137)
```

		athlete	sat
athlete		1.0000	
sat		-0.1851	1.0000

```
. regress sat athlete female white hsize hsizesq hsperc
```

Source		SS	df	MS	Number of obs =	4137
Model		14637941.7	6	2439656.96	F( 6, 4130) =	153.28
Residual		65735904.6	4130	15916.6839	Prob > F =	0.0000
					R-squared =	0.1821

```
-----+-----
Total | 80373846.3 4136 19432.7481
Adj R-squared = 0.1809
Root MSE = 126.16
```

```
-----+-----
sat | Coef. Std. Err. t P>|t| [95% Conf. Interval]
-----+-----
athlete | -74.9073 9.679523 -7.74 0.000 -93.88437 -55.93022
female | -57.84925 3.999139 -14.47 0.000 -65.68971 -50.00878
white | 106.3164 7.613348 13.96 0.000 91.39015 121.2427
hsize | 1.211836 3.7214 0.33 0.745 -6.084113 8.507785
hsizesq | .3737667 .5118486 0.73 0.465 -.6297322 1.377266
hsperc | -2.495847 .1254639 -19.89 0.000 -2.741823 -2.24987
_cons | 1002.015 9.763235 102.63 0.000 982.8739 1021.156
-----+-----
```

d)

```
. gen FA = female * athlete
. gen Fa = female * (1 - athlete)
. gen fA = (1 - female) * athlete
. gen fa = (1 - female) * (1 - athlete)
. regress colgpa hsize hsizesq hsperc sat FA Fa fa
```

```
-----+-----
Source | SS df MS
-----+-----
Model | 524.821272 7 74.9744674
Residual | 1269.3744 4129 .307429015
Total | 1794.19567 4136 .433799728
Number of obs = 4137
F( 7, 4129) = 243.88
Prob > F = 0.0000
R-squared = 0.2925
Adj R-squared = 0.2913
Root MSE = .55446
```

```
-----+-----
colgpa | Coef. Std. Err. t P>|t| [95% Conf. Interval]
-----+-----
hsize | -.0568006 .0163671 -3.47 0.001 -.0888889 -.0247124
hsizesq | .0046699 .0022507 2.07 0.038 .0002573 .0090825
hsperc | -.0132114 .000573 -23.06 0.000 -.0143349 -.012088
sat | .0016462 .0000669 24.62 0.000 .0015151 .0017773
FA | .3297256 .0840593 3.92 0.000 .1649242 .4945271
Fa | .1546151 .0183122 8.44 0.000 .1187133 .1905168
fA | .1674185 .0484877 3.45 0.001 .0723564 .2624806
_cons | 1.241575 .0795453 15.61 0.000 1.085623 1.397526
-----+-----
```

```
. test FA = Fa
```

```
( 1) FA - Fa = 0
```

```
F( 1, 4129) = 4.34  
Prob > F = 0.0372
```

```
. lincom FA - Fa
```

```
( 1) FA - Fa = 0
```

colgpa	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
(1)	.1751106	.0840258	2.08	0.037	.0103748	.3398464

The hypothesis that there is no difference between female athletes and female nonathletes is rejected. Female athletes have higher GPAs than female nonathletes.

e)

```
. regress colgpa hsize hsizesq hsperc sat femXsat female athlete
```

Source	SS	df	MS	Number of obs = 4137		
Model	524.867644	7	74.981092	F( 7, 4129) =	243.91	
Residual	1269.32803	4129	.307417784	Prob > F =	0.0000	
				R-squared =	0.2925	
				Adj R-squared =	0.2913	
				Root MSE =	.55445	
Total	1794.19567	4136	.433799728			

colgpa	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
hsize	-.0569121	.0163537	-3.48	0.001	-.0889741	-.0248501
hsizesq	.0046864	.0022498	2.08	0.037	.0002757	.0090972
hsperc	-.013225	.0005737	-23.05	0.000	-.0143497	-.0121003
sat	.0016255	.0000852	19.09	0.000	.0014585	.0017924
femXsat	.0000512	.0001291	0.40	0.692	-.000202	.0003044
female	.1023066	.1338023	0.76	0.445	-.1600179	.3646311
athlete	.1677568	.0425334	3.94	0.000	.0843684	.2511452
_cons	1.263743	.0974952	12.96	0.000	1.0726	1.454887

The effect of SAT scores do not differ by gender, since the coefficient on femXsat is statistically insignificantly

different from 0.

6) a)

```
. regress nettfa e401k
```

Source	SS	df	MS			
Model	786249.663	1	786249.663	Number of obs =	9275	
Residual	37157139.8	9273	4007.02468	F( 1, 9273) =	196.22	
Total	37943389.5	9274	4091.3726	Prob > F =	0.0000	
				R-squared =	0.0207	
				Adj R-squared =	0.0206	
				Root MSE =	63.301	

  

nettfa	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
e401k	18.85832	1.346275	14.01	0.000	16.21933	21.49732
_cons	11.67677	.8430406	13.85	0.000	10.02423	13.32932

The average net total financial assets, which is measured in thousands of dollars, does differ by 401k eligibility, and the estimated difference is \$18,858.

b)

```
. regress nettfa inc incsq age agesq male e401k
```

Source	SS	df	MS			
Model	7673992.51	6	1278998.75	Number of obs =	9275	
Residual	30269397	9268	3266.01176	F( 6, 9268) =	391.61	
Total	37943389.5	9274	4091.3726	Prob > F =	0.0000	
				R-squared =	0.2022	
				Adj R-squared =	0.2017	
				Root MSE =	57.149	

  

nettfa	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
inc	-.2702243	.0746105	-3.62	0.000	-.4164772	-.1239713
incsq	.010216	.0005871	17.40	0.000	.0090651	.0113669
age	-1.939771	.4834769	-4.01	0.000	-2.887492	-.9920497
agesq	.0345662	.0055482	6.23	0.000	.0236906	.0454418
male	3.369048	1.485813	2.27	0.023	.4565283	6.281569
e401k	9.713482	1.277127	7.61	0.000	7.210032	12.21693
_cons	21.19779	9.992211	2.12	0.034	1.610861	40.78472

Yes, both the quadratic terms included are statistically (and economically) significant. The estimated dollar effect of 401k eligibility is \$9,713, and is statistically significant.

c)

```
. gen e401kXage41 = e401k * (age - 41)
. gen e401kXage41sq = e401k * (age - 41) * (age - 41)
. regress nettfa inc incsq age agesq e401kXage41 e401kXage41sq male e401k
```

Source	SS	df	MS	Number of obs = 9275		
Model	7763594.46	8	970449.308	F( 8, 9266)	=	297.95
Residual	30179795	9266	3257.04673	Prob > F	=	0.0000
-----				R-squared	=	0.2046
-----				Adj R-squared	=	0.2039
Total	37943389.5	9274	4091.3726	Root MSE	=	57.071

  

nettfa	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
inc	-.2705924	.0745119	-3.63	0.000	-.4166522	-.1245326
incsq	.0101878	.0005864	17.37	0.000	.0090383	.0113373
age	-2.287514	.5908919	-3.87	0.000	-3.445792	-1.129235
agesq	.0360854	.0067801	5.32	0.000	.0227948	.0493759
e401kXage41	.6524833	.1313038	4.97	0.000	.395099	.9098676
e401kXage4~q	-.0038891	.0116248	-0.33	0.738	-.0266762	.0188981
male	3.310739	1.483828	2.23	0.026	.4021098	6.219369
e401k	9.978824	1.718176	5.81	0.000	6.610821	13.34683
_cons	32.75766	12.21115	2.68	0.007	8.821123	56.6942

The linear interaction term between e401k and age - 41 is significant, but not the quadratic interaction term. These interaction terms allow the effect of 401k eligibility to differ with age, centering around the age of 41. The difference in the effect of 401k eligibility between this interacted model with the previous model is not statistically significant, which is easy to see by noticing that the estimates have overlapping 95% confidence intervals.

d)

```
. replace fs1 = (fsize == 1)
(7258 real changes made)

. replace fs2 = (fsize == 2)
(7076 real changes made)

. replace fs3 = (fsize == 3)
```

(7446 real changes made)

```
. replace fs4 = (fsize == 4)
(7285 real changes made)
```

```
. replace fs5 = (fsize >= 5)
(424 real changes made)
```

```
. regress nettfa inc incsq age agesq male e401k fs2 fs3 fs4 fs5
```

Source	SS	df	MS	Number of obs =	9275
Model	7730274.7	10	773027.47	F( 10, 9264) =	237.03
Residual	30213114.8	9264	3261.34659	Prob > F =	0.0000
Total	37943389.5	9274	4091.3726	R-squared =	0.2037
				Adj R-squared =	0.2029
				Root MSE =	57.108

nettfa	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
inc	-.2412311	.0755141	-3.19	0.001	-.3892554 -.0932069
incsq	.0100501	.0005895	17.05	0.000	.0088946 .0112056
age	-1.504553	.4947967	-3.04	0.002	-2.474464 -.5346428
agesq	.029165	.0057032	5.11	0.000	.0179855 .0403444
male	1.323946	1.652795	0.80	0.423	-1.915896 4.563789
e401k	9.481517	1.278268	7.42	0.000	6.97583 11.9872
fs2	-.3536808	1.924384	-0.18	0.854	-4.125898 3.418536
fs3	-4.081595	2.013317	-2.03	0.043	-8.02814 -.1350509
fs4	-5.696103	2.021384	-2.82	0.005	-9.65846 -1.733746
fs5	-6.748335	2.235513	-3.02	0.003	-11.13043 -2.366237
_cons	15.74294	10.143	1.55	0.121	-4.13958 35.62545

```
. test fs2 = fs3 = fs4 = fs5 = 0
```

- ( 1) fs2 - fs3 = 0
- ( 2) fs2 - fs4 = 0
- ( 3) fs2 - fs5 = 0
- ( 4) fs2 = 0

```
F( 4, 9264) = 4.31
Prob > F = 0.0017
```

Yes, they are jointly significantly different from the zero vector. This implies that family size affects net total financial assets.

e)

```
. regress netffa inc incsq age agesq male e401k _INfs_2 _INfs_3 _INfs_4 _INfs_5 _INfsXinc_2
> _INfsXinc_3 _INfsXinc_4 _INfsXinc_5 _IN2fsXincsq_2 _IN2fsXincsq_3 _IN2fsXincsq_4
> _IN2fsXincsq_5 _AfsXage_2 _AfsXage_3 _AfsXage_4 _AfsXage_5 _A2f sXagesq_2
> _A2fsXagesq_3 _A2fsXagesq_4 _A2fsXagesq_5 _MfsXmal_2_1 _MfsXmal_3_1 _MfsXmal_4_1
> _MfsXmal_5_1 _EfsXe40_2_1 _EfsXe40_3_1 _EfsXe40_4_1 _EfsXe40_5_1
```

Source	SS	df	MS	Number of obs =	9275
Model	7962430.63	34	234189.136	F( 34, 9240) =	72.18
Residual	29980958.9	9240	3244.69252	Prob > F =	0.0000
				R-squared =	0.2099
				Adj R-squared =	0.2069
Total	37943389.5	9274	4091.3726	Root MSE =	56.962

netffa	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
inc	.7324251	.2465294	2.97	0.003	.249173	1.215677
incsq	.0004576	.0025611	0.18	0.858	-.0045627	.0054779
age	-1.593533	.9856013	-1.62	0.106	-3.525529	.3384627
agesq	.0289718	.0115668	2.50	0.012	.0062982	.0516454
male	2.468105	2.622154	0.94	0.347	-2.671897	7.608106
e401k	7.060432	2.76795	2.55	0.011	1.63464	12.48622
_INfs_2	8.472165	27.89939	0.30	0.761	-46.2168	63.16113
_INfs_3	7.819918	30.6623	0.26	0.799	-52.28495	67.92479
_INfs_4	-15.20342	32.39481	-0.47	0.639	-78.7044	48.29755
_INfs_5	-2.197631	39.65744	-0.06	0.956	-79.93497	75.53971
_INfsXinc_2	-1.168416	.2829879	-4.13	0.000	-1.723134	-.6136971
_INfsXinc_3	-.9147828	.2906207	-3.15	0.002	-1.484463	-.3451022
_INfsXinc_4	-1.044873	.3030632	-3.45	0.001	-1.638944	-.4508028
_INfsXinc_5	-1.380207	.338948	-4.07	0.000	-2.044619	-.7157936
_IN2fsXinc~2	.0118186	.0027563	4.29	0.000	.0064157	.0172214
_IN2fsXinc~3	.0081092	.0027966	2.90	0.004	.0026272	.0135912
_IN2fsXinc~4	.0098486	.0029118	3.38	0.001	.0041407	.0155564
_IN2fsXinc~5	.0131491	.0032414	4.06	0.000	.0067952	.0195029
_AfsXage_2	-.0562306	1.349162	-0.04	0.967	-2.700886	2.588425
_AfsXage_3	.3185739	1.506122	0.21	0.832	-2.633759	3.270906
_AfsXage_4	1.54704	1.600657	0.97	0.334	-1.590602	4.684681
_AfsXage_5	1.542528	1.92175	0.80	0.422	-2.224526	5.309583
_A2fsXages~2	.0062072	.0155267	0.40	0.689	-.0242286	.036643
_A2fsXages~3	-.0032207	.0175261	-0.18	0.854	-.0375758	.0311343
_A2fsXages~4	-.0190828	.0188308	-1.01	0.311	-.0559953	.0178298
_A2fsXages~5	-.0242004	.0223179	-1.08	0.278	-.0679484	.0195475
_MfsXmal_2_1	-3.4121	4.331588	-0.79	0.431	-11.90297	5.078768
_MfsXmal_3_1	-1.427265	5.08321	-0.28	0.779	-11.39148	8.536948



_MfsXmal_4_1		-.0454875	5.213619	-0.01	0.993	-10.26533	10.17436
_MfsXmal_5_1		-3.886421	6.284615	-0.62	0.536	-16.20565	8.432812
_EfsXe40_2_1		6.348744	3.796426	1.67	0.095	-1.09309	13.79058
_EfsXe40_3_1		.9601566	3.992241	0.24	0.810	-6.865516	8.785829
_EfsXe40_4_1		.8992764	3.88614	0.23	0.817	-6.718416	8.516969
_EfsXe40_5_1		4.071537	4.510328	0.90	0.367	-4.769702	12.91278
_cons		2.123567	19.91196	0.11	0.915	-36.90827	41.15541

```

-----
. test _Ifs_2 _Ifs_3 _Ifs_4 _Ifs_5 _IfsXinc_2 _IfsXinc_3 _IfsXinc_4 _IfsXinc_5 _IfsXincsq_2 _IfsXincsq_3
> _3 _IfsXage_4 _IfsXage_5 _IfsXagesq_2 _IfsXagesq_3 _IfsXagesq_4 _IfsXagesq_5 _IfsXmale_2 _IfsXmale_3
> _3 _IfsXe401k_4 _IfsXe401k_5

```

```

( 1) _Ifs_2 = 0
( 2) _Ifs_3 = 0
( 3) _Ifs_4 = 0
( 4) _Ifs_5 = 0
( 5) _IfsXinc_2 = 0
( 6) _IfsXinc_3 = 0
( 7) _IfsXinc_4 = 0
( 8) _IfsXinc_5 = 0
( 9) _IfsXincsq_2 = 0
(10) _IfsXincsq_3 = 0
(11) _IfsXincsq_4 = 0
(12) _IfsXincsq_5 = 0
(13) _IfsXage_2 = 0
(14) _IfsXage_3 = 0
(15) _IfsXage_4 = 0
(16) _IfsXage_5 = 0
(17) _IfsXagesq_2 = 0
(18) _IfsXagesq_3 = 0
(19) _IfsXagesq_4 = 0
(20) _IfsXagesq_5 = 0
(21) _IfsXmale_2 = 0
(22) _IfsXmale_3 = 0
(23) _IfsXmale_4 = 0
(24) _IfsXmale_5 = 0
(25) _IfsXe401k_2 = 0
(26) _IfsXe401k_3 = 0
(27) _IfsXe401k_4 = 0
(28) _IfsXe401k_5 = 0

```

```

F( 28, 9240) = 3.17
Prob > F = 0.0000

```

The null hypothesis of the Chow test, which is distributed as an F-statistic, is that the categories don't matter. However, the null is rejected at 5% (and even the 0.1%!) level, which implies that the regression is not stable over family size categories.