

BOSTON COLLEGE
 Department of Economics
 EC 228 Econometrics, Prof. Baum, Ms. Yu, Fall 2003
Problem Set 6 Solutions

Problem sets should be your own work. You may work together with classmates, but if you're not figuring this out on your own, you will eventually regret it.

1. (7.13)

```
. use http://fmwww.bc.edu/ec-p/data/wooldridge/CEOSAL1
. gen rosneg=(ros<0)
. browse
. regress lsalary lsales roe rosneg
```

Source	SS	df	MS	Number of obs =	209
Model	19.7902034	3	6.59673446	F(3, 205) =	28.81
Residual	46.9319665	205	.228936422	Prob > F =	0.0000
				R-squared =	0.2966
				Adj R-squared =	0.2863
Total	66.7221699	208	.320779663	Root MSE =	.47847

lsalary	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lsales	.2883868	.0336172	8.58	0.000	.222107	.3546666
roe	.0166571	.0039681	4.20	0.000	.0088336	.0244806
rosneg	-.2256748	.109338	-2.06	0.040	-.441246	-.0101036
_cons	4.297602	.2932526	14.65	0.000	3.719424	4.87578

The estimated equation is

$$\log(\widehat{salary}) = 4.30 + .228 \log(sales) + .167 roe - .226 rosneg$$

$$n = 209, R^2 = .297, \bar{R}^2 = .286.$$

The coefficient on *rosneg* implies that if the CEO's firm had a negative return on its stock over the 1988 to 1990 period, the CEO salary was predicted to be about 22.6% lower, for given levels of *sales* and *roe*. The *t* statistic is about -2.07, which is significant at the 5% level against a two-sided alternative.

2. (7.14)

```
(i) . by male, sort: regress sleep totwrk educ age agesq yngkid
```

```
-----
-> male = 0
```

Source	SS	df	MS	Number of obs =	306
Model	6201576.18	5	1240315.24	F(5, 300) =	6.50
Residual	57288575.9	300	190961.92	Prob > F =	0.0000
				R-squared =	0.0977
				Adj R-squared =	0.0826
Total	63490152.1	305	208164.433	Root MSE =	436.99

sleep	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
totwrk	-.1399495	.0276594	-5.06	0.000	-.1943806	-.0855184
educ	-10.20514	9.588848	-1.06	0.288	-29.07506	8.664786
age	-30.35657	18.53091	-1.64	0.102	-66.82361	6.110463
agesq	.3679406	.2233398	1.65	0.101	-.0715705	.8074516
yngkid	-118.2826	93.18757	-1.27	0.205	-301.6666	65.10153
_cons	4238.729	384.8923	11.01	0.000	3481.299	4996.16

-> male = 1

Source	SS	df	MS	Number of obs =	400
Model	11806161.6	5	2361232.32	F(5, 394) =	14.59
Residual	63763979.0	394	161837.51	Prob > F =	0.0000
				R-squared =	0.1562
				Adj R-squared =	0.1455
Total	75570140.6	399	189398.849	Root MSE =	402.29

sleep	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
totwrk	-.1821232	.0244855	-7.44	0.000	-.2302618	-.1339846
educ	-13.05238	7.414218	-1.76	0.079	-27.62876	1.523995
age	7.156591	14.32037	0.50	0.618	-20.99731	35.31049
agesq	-.0447674	.1684053	-0.27	0.791	-.3758528	.286318
yngkid	60.38021	59.02278	1.02	0.307	-55.65877	176.4192
_cons	3648.208	310.0393	11.77	0.000	3038.67	4257.747

The estimated equation for men is

$$\widehat{sleep} = 3,648.2 - .182 \text{ totwrk} - 13.05 \text{ educ} + 7.16 \text{ age} - .0448 \text{ age}^2 + 60.38 \text{ yngkid}$$

$$n = 400, R^2 = .156.$$

The estimated equation for women is

$$\widehat{sleep} = 4,238.7 - .140 \text{ totwrk} - 10.21 \text{ educ} - 30.36 \text{ age} - .368 \text{ age}^2 - 118.28 \text{ yngkid}$$

$$n = 306, R^2 = .098.$$

There are certainly notable differences in the point estimates. For example, having a young child in the household leads to less sleep for women (about two hours a week) while men are estimated to sleep about an hour more. The quadratic in *age* is a hump-shape for men but a U-shape for women. The intercepts for men and women are also notably different.

```
(ii) . gen maletotwrk= male* totwrk

. gen maleeduc= male* educ

. gen maleage= male* age

. gen maleagesq=male*agesq

. gen maleyugkid=male*yngkid

. regress sleep totwrk educ age agesq yngkid male maletotwrk
maleeduc maleage maleagesq maleyugkid
```

Source	SS	df	MS	Number of obs =	706
Model	18187280.8	11	1653389.17	F(11, 694) =	9.48
Residual	121052555	694	174427.313	Prob > F =	0.0000
				R-squared =	0.1306
				Adj R-squared =	0.1168
Total	139239836	705	197503.313	Root MSE =	417.64

sleep	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
totwrk	-.1399495	.0264349	-5.29	0.000	-.1918514 -.0880476
educ	-10.20514	9.164321	-1.11	0.266	-28.19826 7.787983
age	-30.35657	17.71049	-1.71	0.087	-65.12914 4.415998
agesq	.3679406	.2134519	1.72	0.085	-.0511483 .7870294
yngkid	-118.2826	89.06187	-1.33	0.185	-293.1456 56.58046
male	-590.5211	488.7916	-1.21	0.227	-1550.209 369.1665
maletotwrk	-.0421737	.036674	-1.15	0.251	-.114179 .0298317
maleeduc	-2.847243	11.96795	-0.24	0.812	-26.34497 20.65048
maleage	37.51316	23.12332	1.62	0.105	-7.886887 82.91321
maleagesq	-.4127079	.2759136	-1.50	0.135	-.9544333 .1290175
maleyugkid	178.6628	108.1051	1.65	0.099	-33.5895 390.915
_cons	4238.729	367.8519	11.52	0.000	3516.493 4960.965

```
. test male maletotwrk maleeduc maleage maleagesq maleyugkid
```

- (1) male = 0.0
- (2) maletotwrk = 0.0
- (3) maleeduc = 0.0
- (4) maleage = 0.0
- (5) maleagesq = 0.0
- (6) maleyugkid = 0.0

```
F( 6, 694) = 2.12
Prob > F = 0.0495
```

The F statistic (with 6 and 694 df) is about 2.12 with p -value $\approx .05$, and so we reject the null that sleep equations are the same at the 5% level.

```
(iii) . test maletotwrk maleeduc maleage maleagesq maleyugkid
```

```
( 1) maletotwrk = 0.0
( 2) maleeduc   = 0.0
( 3) maleage    = 0.0
( 4) maleagesq  = 0.0
( 5) maleyugkid = 0.0
```

```
F( 5, 694) = 1.26
Prob > F = 0.2814
```

If we leave the coefficient on *male* unspecified under H_0 , and test only the five interaction terms, $male \cdot totwrk$, $male \cdot educ$, $male \cdot age$, $male \cdot age^2$, and $male \cdot yngkid$, the F statistic (with 5 and 694 df) is about 1.26 and p -value $\approx .28$.

(iv) The outcome of the test in part (iii) shows that, once an intercept difference is allowed, there is not strong evidence of slope differences between men and women. This is one of those cases where the practically important differences in estimates for women and men in part (i) do not translate into statistically significant differences. We apparently need a larger sample size to determine whether there are differences in slopes. For the purposes of studying the sleep-work tradeoff, the original model with *male* added as an explanatory variable seems sufficient.

3. (7.15)

(i) When $educ = 12.5$, the approximate proportionate difference in estimated *wage* between women and men is $-.227 - .0056(12.5) = -.297$. When $educ = 0$, the difference is $-.227$. So the differential at 12.5 years of education is about 7 percentage points greater.

```
(ii) . use http://fmwww.bc.edu/ec-p/data/wooldridge/WAGE1
```

```
. gen femaleeduc1=female*(educ-12.5)
```

```
. regress lwage female educ femaleeduc1 exper expersq tenure
tenursq
```

Source	SS	df	MS	Number of obs =	526
Model	65.4081526	7	9.3440218	F(7, 518) =	58.37
Residual	82.9216091	518	.160080326	Prob > F =	0.0000
				R-squared =	0.4410
				Adj R-squared =	0.4334
Total	148.329762	525	.28253288	Root MSE =	.4001

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
female	-.296345	.0358358	-8.27	0.000	-.3667465	-.2259436
educ	.0823692	.0084699	9.72	0.000	.0657296	.0990088
femaleeduc1	-.0055645	.0130618	-0.43	0.670	-.0312252	.0200962
exper	.0293366	.0049842	5.89	0.000	.019545	.0391283
expersq	-.0005804	.0001075	-5.40	0.000	-.0007916	-.0003691
tenure	.0318967	.006864	4.65	0.000	.018412	.0453814
tenursq	-.00059	.0002352	-2.51	0.012	-.001052	-.000128
_cons	.388806	.1186871	3.28	0.001	.1556388	.6219733

We can write the model underlying (7.18) as

$$\begin{aligned}
 \log(\text{wage}) &= \beta_0 + \delta_0 \text{female} + \beta_1 \text{educ} + \delta_1 \text{female} \cdot \text{educ} + \text{other factors} \\
 &= \beta_0 + (\delta_0 + 12.5\delta_1) \text{female} + \beta_1 \text{educ} + \delta_1 \text{female} \cdot (\text{educ} - 12.5) + \text{other factors} \\
 &= \beta_0 + \theta_0 \text{female} + \beta_1 \text{educ} + \delta_1 \text{female} \cdot (\text{educ} - 12.5) + \text{other factors},
 \end{aligned}$$

where $\theta_0 = \delta_0 + 12.5\delta_1$ is the gender differential at 12.5 years of education. When we run this regression we obtain about $-.294$ as the coefficient on *female* (which differs from $-.297$ due to rounding error). Its standard error is about $.036$.

- (iii) The t statistic on *female* from part (ii) is about -8.17 , which is very significant. This is because we are estimating the gender differential at a reasonable number of years of education, 12.5 which is close to the average. In equation (7.18), the coefficient on *female* is the gender differential when $\text{educ} = 0$. There are no people of either gender with close to zero years of education, and so we cannot hope – nor do we want to – to estimate the gender differential at $\text{educ} = 0$.

4. (7.19)

- (i) . use <http://fmwww.bc.edu/ec-p/data/wooldridge2k/401KSUBS-10>

. summ nettf

Variable	Obs	Mean	Std. Dev.	Min	Max
nettf	928	21.18766	74.44089	-121.472	1462.115

The average is 21.188 , the minimum is -121.472 , the maximum is 1462.115 .

- (ii)

. regress nettf e401k

Source	SS	df	MS	Number of obs =	928
Model	155419.609	1	155419.609	F(1, 926) =	28.89
				Prob > F =	0.0000

Residual		4981501.04	926	5379.59076	R-squared	=	0.0303

Total		5136920.65	927	5541.44622	Adj R-squared	=	0.0292
					Root MSE	=	73.346

nettf		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
e401k		26.21824	4.877813	5.37	0.000	16.64538 35.79109
_cons		10.16922	3.162157	3.22	0.001	3.963395 16.37505

This can be easily done by regressing *nettf* on *e401k* and doing a *t* test on $\hat{\beta}_{ec401k}$; the estimate is the average difference in *nettf* for those eligible for a 401(k) and those not eligible. Using the 928 observation gives $\hat{\beta}_{ec401k} = 26.218$ and $t_{e401k} = 4.878$. Therefore, we strongly reject the null hypothesis that there is no difference in the average. The coefficient implies that, on average, a family eligible for a 401(k) plan has 26,218 more on net total financial assets.

(iii) . regress nettf e401k inc incsq age agesq male

Source		SS	df	MS	Number of obs =	928

Model		1211139.92	6	201856.653	F(6, 921) =	47.36
Residual		3925780.74	921	4262.5198	Prob > F	= 0.0000

Total		5136920.65	927	5541.44622	R-squared	= 0.2358
					Adj R-squared	= 0.2308
					Root MSE	= 65.288

nettf		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
e401k		14.21904	4.590444	3.10	0.002	5.2101 23.22799
inc		-.5482641	.253173	-2.17	0.031	-1.045127 -.0514011
incsq		.0140768	.0019759	7.12	0.000	.0101989 .0179546
age		-2.567236	1.818878	-1.41	0.158	-6.136862 1.002391
agesq		.0428191	.0209215	2.05	0.041	.0017597 .0838786
male		.201791	5.470784	0.04	0.971	-10.53486 10.93844
_cons		34.81393	37.44084	0.93	0.353	-38.66533 108.2932

The equation estimated by OLS is

$$\widehat{nettf} = 34.814 + 14.219 e401k - .548 inc + .014 inc^2 - 2.567 age + .0428 age^2 + .202 male$$

(37.44)
(4.59)
(.253)
(.0020)
(1.819)

(.021)
(5.47)

$n = 928, R^2 = .236.$

Now holding income and age fixed, a 401(k)-eligible family is estimated to have \$14,219 more in wealth than a non-eligible family.

(iv)

```
. gen e401kage1= e401k*(age-41)

. gen e401kage2= e401k*(age-41)^2

. regress nettf a e401k inc incsq age agesq male e401kage1
e401kage2
```

Source	SS	df	MS	Number of obs =	928
Model	1257734.26	8	157216.782	F(8, 919) =	37.25
Residual	3879186.39	919	4221.0951	Prob > F =	0.0000
				R-squared =	0.2448
				Adj R-squared =	0.2383
				Root MSE =	64.97
Total	5136920.65	927	5541.44622		

nettf a	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
e401k	8.357268	6.187644	1.35	0.177	-3.786285 20.50082
inc	-.4700326	.2532375	-1.86	0.064	-.9670235 .0269583
incsq	.0133709	.0019785	6.76	0.000	.009488 .0172538
age	-1.791962	2.264044	-0.79	0.429	-6.235259 2.651334
agesq	.028537	.0258394	1.10	0.270	-.0221741 .0792481
male	.4487733	5.445848	0.08	0.934	-10.23897 11.13651
e401kage1	1.14543	.4725547	2.42	0.016	.218019 2.072842
e401kage2	.0595252	.0434693	1.37	0.171	-.0257854 .1448358
_cons	27.12249	47.16079	0.58	0.565	-65.43285 119.6778

Only the interaction $e401k \cdot (age - 41)$ is significant. Its coefficient is 1.145 ($t = 2.42$). It shows that the effect of 401(k) eligibility on financial wealth increases with age. The coefficient on $e401k \cdot (age - 41)^2$ is .060 (t statistic = 1.37), so it is not significant.

(v) The effect of e401k in part (iii) is the same for all ages, 14.219. For the regression in part (iv), the coefficient on e401k from part (iv) is about 8.357, which is the effect at the average age, $age = 41$.

(vi) . tab fsize, gen(fsize)

family size	Freq.	Percent	Cum.
1	203	21.88	21.88
2	217	23.38	45.26
3	198	21.34	66.59
4	188	20.26	86.85
5	74	7.97	94.83
6	31	3.34	98.17
7	11	1.19	99.35
8	5	0.54	99.89

13	1	0.11	100.00

Total	928	100.00	

. drop fsize5 fsize6 fsize7 fsize8 fsize9

. regress nettf a e401k inc incsq age agesq male fsize1 fsize2
fsize3 fsize4

Source	SS	df	MS	Number of obs =	928
Model	1249291.04	10	124929.104	F(10, 917) =	29.47
Residual	3887629.61	917	4239.50884	Prob > F =	0.0000
-----				R-squared =	0.2432
-----				Adj R-squared =	0.2349
Total	5136920.65	927	5541.44622	Root MSE =	65.112

nettf a	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
e401k	13.42462	4.595985	2.92	0.004	4.404754	22.44449
inc	-.5637908	.2564669	-2.20	0.028	-1.067121	-.0604606
incsq	.0142597	.001986	7.18	0.000	.0103621	.0181573
age	-1.732811	1.869153	-0.93	0.354	-5.401126	1.935504
agesq	.0321586	.0216034	1.49	0.137	-.0102393	.0745564
male	-1.783906	6.270077	-0.28	0.776	-14.08927	10.52146
fsize1	9.1958	8.194099	1.12	0.262	-6.885564	25.27716
fsize2	17.87712	7.54224	2.37	0.018	3.075066	32.67918
fsize3	.5817076	7.547443	0.08	0.939	-14.23056	15.39397
fsize4	6.537835	7.612689	0.86	0.391	-8.402482	21.47815
_cons	12.91241	39.44122	0.33	0.743	-64.49313	90.31795

. test fsize1 fsize2 fsize3 fsize4

- (1) fsize1 = 0
- (2) fsize2 = 0
- (3) fsize3 = 0
- (4) fsize4 = 0

F(4, 917) = 2.25
Prob > F = 0.0620

I chose *fsize5* as the base group. The estimated equation is

$$\widehat{nettf a} = 12.912 + 13.425 e401k - .564 inc + .014 inc^2 - 1.733 age + .032 age^2$$

$$\begin{matrix} (39.44) & (4.60) & (.256) & (.0020) & (1.869) & (.022) \\ - 1.784 male + 9.196 fsize1 + 17.877 fsize2 + .582 fsize3 + 6.538 fsize4 \\ (6.27) & (8.19) & (7.54) & (7.55) & (7.61) \end{matrix}$$

$$n = 928, R^2 = .243.$$

The *F* statistic for joint significance of the four family size dummies is about 2.25. With 4 and 917 *df*, this gives *p*-value = .062, so they are not jointly significant.

5. (8.9)

(i) . use <http://fmwww.bc.edu/ec-p/data/wooldridge/VOTE1>

. regress voteA prtystra democA lexpendA lexpendB

Source	SS	df	MS	Number of obs =	173
Model	38822.1768	4	9705.5442	F(4, 168) =	169.23
Residual	9635.07174	168	57.3516175	Prob > F =	0.0000
				R-squared =	0.8012
				Adj R-squared =	0.7964
Total	48457.2486	172	281.728189	Root MSE =	7.5731

voteA	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
prtystra	.2519175	.0712925	3.53	0.001	.1111729	.3926622
democA	3.792944	1.40652	2.70	0.008	1.016213	6.569674
lexpendA	5.779294	.3918197	14.75	0.000	5.00577	6.552819
lexpendB	-6.237836	.3974596	-15.69	0.000	-7.022495	-5.453178
_cons	37.66142	4.736036	7.95	0.000	28.3116	47.01123

. predict e if e(sample),resid

. regress e prtystra democA lexpendA lexpendB

Source	SS	df	MS	Number of obs =	173
Model	0	4	0	F(4, 168) =	0.00
Residual	9635.07169	168	57.3516172	Prob > F =	1.0000
				R-squared =	0.0000
				Adj R-squared =	-0.0238
Total	9635.07169	172	56.0178587	Root MSE =	7.5731

e	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
prtystra	7.22e-11	.0712925	0.00	1.000	-.1407447	.1407447
democA	-2.62e-08	1.40652	-0.00	1.000	-2.776731	2.77673
lexpendA	2.44e-08	.3918197	0.00	1.000	-.7735247	.7735247
lexpendB	1.00e-09	.3974596	0.00	1.000	-.7846588	.7846589
_cons	-1.27e-07	4.736036	-0.00	1.000	-9.349812	9.349811

The estimated equation is

$$\widehat{voteA} = 37.66 + .252 prtystra + 3.793 democA + 5.779 \log(expendA) - 6.238 \log(expendB) + \hat{u}$$

(4.74)
(.071)
(1.407)
(.392)

(0.397)

$$n = 173, R^2 = .801, \bar{R}^2 = .796.$$

You can convince yourself that regressing the \hat{u}_i on all of the explanatory variables yields an R -squared of zero, although it might not be exactly zero in your computer output due to rounding error. Remember, this is how OLS works: the estimates $\hat{\beta}_j$ are chosen to make the residuals be uncorrelated in the sample with each independent variable (as well as have zero sample average).

(ii) Use the F statistic version

```
. gen esq=e^2
. regress esq prtystra democA lexpendA lexpendB
```

Source	SS	df	MS	Number of obs =	173
Model	61537.0938	4	15384.2735	F(4, 168) =	2.33
Residual	1109198.47	168	6602.37183	Prob > F =	0.0581
Total	1170735.56	172	6806.6021	R-squared =	0.0526
				Adj R-squared =	0.0300
				Root MSE =	81.255

esq	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
prtystra	-.2992641	.7649293	-0.39	0.696	-1.809376 1.210848
democA	15.61921	15.09117	1.03	0.302	-14.17356 45.41198
lexpendA	-10.30573	4.204007	-2.45	0.015	-18.60522 -2.006238
lexpendB	-.0514033	4.26452	-0.01	0.990	-8.470355 8.367549
_cons	113.9635	50.81502	2.24	0.026	13.6452 214.2817

```
. test prtystra democA lexpendA lexpendB
```

```
( 1) prtystra = 0
( 2) democA = 0
( 3) lexpendA = 0
( 4) lexpendB = 0
```

```
F( 4, 168) = 2.33
Prob > F = 0.0581
```

use the bpagan test

```
. regress voteA prtystra democA lexpendA lexpendB
```

Source	SS	df	MS	Number of obs =	173
Model	38822.1768	4	9705.5442	F(4, 168) =	169.23
Residual	9635.07174	168	57.3516175	Prob > F =	0.0000
Total	48457.2486	172	281.728189	R-squared =	0.8012
				Adj R-squared =	0.7964
				Root MSE =	7.5731

voteA	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]

prtystrA		.2519175	.0712925	3.53	0.001	.1111729	.3926622
democA		3.792944	1.40652	2.70	0.008	1.016213	6.569674
lexpendA		5.779294	.3918197	14.75	0.000	5.00577	6.552819
lexpendB		-6.237836	.3974596	-15.69	0.000	-7.022495	-5.453178
_cons		37.66142	4.736036	7.95	0.000	28.3116	47.01123

```
. bpagan prtystrA democA lexpendA lexpendB
```

```
Breusch-Pagan LM statistic: 9.919488 Chi-sq( 4) P-value =
.0418
```

The B-P test entails regressing the \hat{u}_i^2 on the independent variables in part (i). The F statistic for joint significant (with 4 and 168 df) is about 2.33 with p -value $\approx .058$. Therefore, there is some evidence of heteroskedasticity, but not quite at the 5% level.

(iii) use white test

```
. whitetst, fitted
```

```
White's special test statistic : 5.490049 Chi-sq( 2) P-value = .0642
```

use the F-statistic version

```
. predict voteA1 (option xb assumed; fitted values)
```

```
. gen voteA1sq= voteA1^2
```

```
. regress esq voteA1 voteA1sq
```

Source	SS	df	MS	Number of obs =	173
Model	37152.5749	2	18576.2875	F(2, 170) =	2.79
Residual	1133582.99	170	6668.13521	Prob > F =	0.0645
Total	1170735.56	172	6806.6021	R-squared =	0.0317
				Adj R-squared =	0.0203
				Root MSE =	81.659

esq	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
voteA1	-4.263682	2.166534	-1.97	0.051	-8.540455 .0130912
voteA1sq	.0357354	.0212419	1.68	0.094	-.0061964 .0776672
_cons	171.8584	53.14213	3.23	0.001	66.95499 276.7619

```
. test voteA1 voteA1sq
```

```
( 1) voteA1 = 0
```

```
( 2) voteA1sq = 0
```

```
F( 2, 170) = 2.79
```

Prob > F = 0.0645

Now we regress \hat{u}_i^2 on \widehat{voteA}_i and $(\widehat{voteA}_i)^2$, where the \widehat{voteA}_i are the OLS fitted values from part (i). The F test, with 2 and 170 df , is about 2.79 with p -value $\approx .065$. This is slightly less evidence of heteroskedasticity than provided by the B-P test, but the conclusion is very similar.

6. (9.7)

(i) . use <http://fmwww.bc.edu/ec-p/data/wooldridge/WAGE2>

. regress lwage educ exper tenure married south urban black KWW

Source	SS	df	MS	Number of obs =	935
Model	42.8510762	8	5.35638452	F(8, 926) =	40.39
Residual	122.805218	926	.132619026	Prob > F =	0.0000
				R-squared =	0.2587
				Adj R-squared =	0.2523
Total	165.656294	934	.177362199	Root MSE =	.36417

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
educ	.0576277	.006838	8.43	0.000	.0442079 .0710475
exper	.0122284	.003241	3.77	0.000	.0058678 .018589
tenure	.011072	.0024564	4.51	0.000	.0062512 .0158927
married	.1894612	.0390774	4.85	0.000	.1127707 .2661517
south	-.0916006	.0261562	-3.50	0.000	-.142933 -.0402683
urban	.1755452	.0270323	6.49	0.000	.1224936 .2285969
black	-.1642666	.0385304	-4.26	0.000	-.2398837 -.0886495
KWW	.0050275	.0018188	2.76	0.006	.0014581 .008597
_cons	5.358797	.1136002	47.17	0.000	5.135853 5.581741

We estimate the model from column (2) but with KWW in place of IQ . The coefficient on $educ$ becomes about .058 (se $\approx .006$), so this is similar to the estimate obtained with IQ , although slightly larger and more precisely estimated.

(ii) . regress lwage educ exper tenure married south urban black KWW
IQ

Source	SS	df	MS	Number of obs =	935
Model	44.0968017	9	4.89964463	F(9, 925) =	37.28
Residual	121.559493	925	.131415668	Prob > F =	0.0000
				R-squared =	0.2662
				Adj R-squared =	0.2591
Total	165.656294	934	.177362199	Root MSE =	.36251

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
educ	.0498375	.007262	6.86	0.000	.0355856	.0640893
exper	.0127522	.0032308	3.95	0.000	.0064117	.0190927
tenure	.0109248	.0024457	4.47	0.000	.006125	.0157246
married	.1921449	.0389094	4.94	0.000	.1157839	.2685059
south	-.0820295	.0262222	-3.13	0.002	-.1334913	-.0305676
urban	.1758226	.0269095	6.53	0.000	.1230118	.2286334
black	-.1303995	.0399014	-3.27	0.001	-.2087073	-.0520917
KWW	.003826	.0018521	2.07	0.039	.0001911	.0074608
IQ	.0031183	.0010128	3.08	0.002	.0011306	.0051059
_cons	5.175643	.127776	40.51	0.000	4.924879	5.426408

When KWW and IQ are both used as proxies, the coefficient on $educ$ becomes about .049 (se \approx .007). Compared with the estimate when only KWW is used as a proxy, the return to education has fallen by almost a full percentage point.

(iii) `. test KWW IQ`

(1) $KWW = 0.0$

(2) $IQ = 0.0$

F(2, 925) = 8.59
 Prob > F = 0.0002

The t statistic on IQ is about 3.08 while that on KWW is about 2.07, so each is significant at the 5% level against a two-sided alternative. They are jointly very significant, with $F_{2,925} \approx 8.59$ and p -value $\approx .0002$.