

ECONOMETRICS USING STATA

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INTRODUCTION

1.1 AN OVERVIEW OF STATA'S DISTINCTIVE FEATURES

- You can easily learn STATA commands, even if you do not know the syntax
- You can use STATA's do-file Editor to save time developing your analysis
- A simple command performs all computations for all the desired observations
- Looping over variables saves time & effort
- STATA's by-groups reduce the need for programming
- STATA has many statistical features that make it uniquely powerful
- You can avoid problems by keeping STATA up to date
- STATA is infinitely extensible
- STATA's user community provides a wealth of useful additions to STATA
- STATA is cross-platform compatible
- STATA can be fun

WORKING WITH ECONOMIC AND FINANCIAL DATA IN STATA

2.1 THE BASICS

2.1.1 *The use Command*

```
. use census2c.(1980 Census data for NE and NC states)

. list, sep(0)
```

```
+-----+
| state      region      pop    popurb  medage  marr  divr |
+-----+
1. | Connecticut  NE        3107.6  2449.8   32.00   26.0  13.5 |
2. | Illinois     N Cntrl   11426.5  9518.0   29.90  109.8  51.0 |
3. | Indiana      N Cntrl   5490.2  3525.3   29.20   57.9  40.0 |
4. | Iowa         N Cntrl   2913.8  1708.2   30.00   27.5  11.9 |
5. | Kansas       N Cntrl   2363.7  1575.9   30.10   24.8  13.4 |
6. | Maine        NE       1124.7   534.1   30.40   12.0   6.2 |
7. | Massachusetts NE       5737.0  4808.3   31.20   46.3  17.9 |
8. | Michigan     N Cntrl   9262.1  6551.6   28.80   86.9  45.0 |
9. | Minnesota    N Cntrl   4076.0  2725.2   29.20   37.6  15.4 |
10. | Missouri     N Cntrl   4916.7  3349.6   30.90   54.6  27.6 |
11. | Nebraska     N Cntrl   1569.8   987.9   29.70   14.2   6.4 |
12. | New Hampshire NE        920.6   480.3   30.10    9.3   5.3 |
13. | New Jersey   NE       7364.8  6557.4   32.20   55.8  27.8 |
14. | New York     NE      17558.1 14858.1   31.90  144.5  62.0 |
15. | N. Dakota    N Cntrl    652.7   318.3   28.30    6.1   2.1 |
16. | Ohio         N Cntrl  10797.6  7918.3   29.90   99.8  58.8 |
17. | Pennsylvania NE      11863.9  8220.9   32.10   93.7  34.9 |
18. | Rhode Island NE       947.2   824.0   31.80    7.5   3.6 |
19. | S. Dakota    N Cntrl    690.8   320.8   28.90    8.8   2.8 |
20. | Vermont      NE       511.5   172.7   29.40    5.2   2.6 |
21. | Wisconsin    N Cntrl   4705.8  3020.7   29.40   41.1  17.5 |
+-----+
```

2.1.2 *Variable Types*

```
. use census2c.(1980 Census data for NE and NC states)

. describe
```

Contains data from census2c.dta

obs: 21

1980 Census data for NE and NC
states

```
vars:          7                               14 Jun 2006 08:48
size:         1,050
```

variable name	storage type	display format	value label	variable label
state	str13	%-13s		State
region	byte	%-8.0g	cenreg	Census region
pop	double	%8.1f		1980 Population, '000
popurb	double	%8.1f		1980 Urban population, '000
medage	float	%9.2f		Median age, years
marr	double	%8.1f		Marriages, '000
divr	double	%8.1f		Divorces, '000

Sorted by:

2.1.3 *Generate & Replace*

```
. use census2c.(1980 Census data for NE and NC states)

. generate urbanized = popurb/pop

. summarize urbanized
```

Variable	Obs	Mean	Std. Dev.	Min	Max
urbanized	21	.6667691	.1500842	.3377319	.8903645

```
. use census2c.(1980 Census data for NE and NC states)

. generate urbanized = popurb/pop

. replace urbanized = 100*urbanized
(21 real changes made)

. summarize urbanized
```

Variable	Obs	Mean	Std. Dev.	Min	Max
urbanized	21	66.67691	15.00843	33.77319	89.03645

2.1.4 *sort & gsort*

```
. use census2c.(1980 Census data for NE and NC states)

. list region state pop, sepby(region)
```

region	state	pop

```

1. | NE          Connecticut      3107.6 |
   |-----|
2. | N Cntrl     Illinois         11426.5 |
3. | N Cntrl     Indiana          5490.2 |
4. | N Cntrl     Iowa             2913.8 |
5. | N Cntrl     Kansas           2363.7 |
   |-----|
6. | NE          Maine            1124.7 |
7. | NE          Massachusetts    5737.0 |
   |-----|
8. | N Cntrl     Michigan         9262.1 |
9. | N Cntrl     Minnesota        4076.0 |
10. | N Cntrl     Missouri         4916.7 |
11. | N Cntrl     Nebraska         1569.8 |
   |-----|
12. | NE          New Hampshire    920.6 |
13. | NE          New Jersey       7364.8 |
14. | NE          New York        17558.1 |
   |-----|
15. | N Cntrl     N. Dakota         652.7 |
16. | N Cntrl     Ohio             10797.6 |
   |-----|
17. | NE          Pennsylvania     11863.9 |
18. | NE          Rhode Island     947.2 |
   |-----|
19. | N Cntrl     S. Dakota         690.8 |
   |-----|
20. | NE          Vermont          511.5 |
   |-----|
21. | N Cntrl     Wisconsin        4705.8 |
   +-----+

```

```

.
. sort region -pop
. list region state pop, sepby(region)

```

```

+-----+
| region   state                pop |
+-----+
1. | NE      Vermont              511.5 |
2. | NE      New Hampshire        920.6 |
3. | NE      Rhode Island         947.2 |
4. | NE      Maine                1124.7 |
5. | NE      Connecticut          3107.6 |
6. | NE      Massachusetts        5737.0 |
7. | NE      New Jersey           7364.8 |
8. | NE      Pennsylvania         11863.9 |
9. | NE      New York             17558.1 |
   |-----|
10. | N Cntrl N. Dakota            652.7 |
11. | N Cntrl S. Dakota            690.8 |
12. | N Cntrl Nebraska            1569.8 |
13. | N Cntrl Kansas              2363.7 |
14. | N Cntrl Iowa                2913.8 |
15. | N Cntrl Minnesota           4076.0 |
16. | N Cntrl Wisconsin           4705.8 |

```

```

17. | N Cntrl  Missouri      4916.7 |
18. | N Cntrl  Indiana      5490.2 |
19. | N Cntrl  Michigan     9262.1 |
20. | N Cntrl  Ohio        10797.6 |
21. | N Cntrl  Illinois     11426.5 |
+-----+

```

```

.
. gsort region -pop

. list region state pop, sepby(region)

```

```

+-----+
| region   state           pop |
+-----+
1. | NE      New York       17558.1 |
2. | NE      Pennsylvania   11863.9 |
3. | NE      New Jersey     7364.8 |
4. | NE      Massachusetts  5737.0 |
5. | NE      Connecticut    3107.6 |
6. | NE      Maine          1124.7 |
7. | NE      Rhode Island   947.2 |
8. | NE      New Hampshire  920.6 |
9. | NE      Vermont        511.5 |
+-----+
10. | N Cntrl Illinois      11426.5 |
11. | N Cntrl Ohio        10797.6 |
12. | N Cntrl Michigan     9262.1 |
13. | N Cntrl Indiana      5490.2 |
14. | N Cntrl Missouri     4916.7 |
15. | N Cntrl Wisconsin    4705.8 |
16. | N Cntrl Minnesota    4076.0 |
17. | N Cntrl Iowa         2913.8 |
18. | N Cntrl Kansas       2363.7 |
19. | N Cntrl Nebraska     1569.8 |
20. | N Cntrl S. Dakota     690.8 |
21. | N Cntrl N. Dakota     652.7 |
+-----+

```

ORGANIZING & HANDLING ECONOMIC DATA

3.1 CROSS-SECTIONAL DATA & IDENTIFIER VARIABLES

3.2 TIME SERIES DATA

3.3 POOLED CROSS-SECTIONAL TIME SERIES DATA

3.4 PANEL DATA

3.5 TOOLS FOR MANIPULATING PANEL DATA

3.6 COMBINING CROSS-SECTIONAL & TIME SERIES DATASETS

3.7 CREATING LONG-FORMAT DATASETS WITH APPEND

3.8 THE RESHAPE COMMAND

3.9 USING STATA FOR REPRODUCIBLE RESEARCH

3.9.1 *Using do-files*

3.9.2 *Data Validation: Assert & Duplicates*

LINEAR REGRESSION

4.1 LINEAR REGRESSION

```
. use hprice2a.(Housing price data for Boston-area communities)

. summarize price lprice lnox ldlist rooms stratio, sep(0)
```

Variable	Obs	Mean	Std. Dev.	Min	Max
price	506	22511.51	9208.856	5000	50001
lprice	506	9.941057	.409255	8.517193	10.8198
lnox	506	1.693091	.2014102	1.348073	2.164472
ldlist	506	1.188233	.539501	.1222176	2.495682
rooms	506	6.284051	.7025938	3.56	8.78
stratio	506	18.45929	2.16582	12.6	22

```
. use hprice2a.(Housing price data for Boston-area communities)

. regress lprice lnox ldlist rooms stratio
```

Source	SS	df	MS	Number of obs	=	506
Model	49.3987735	4	12.3496934	F(4, 501)	=	175.86
Residual	35.1834974	501	.070226542	Prob > F	=	0.0000
				R-squared	=	0.5840
				Adj R-squared	=	0.5807
Total	84.5822709	505	.167489645	Root MSE	=	.265

lprice	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnox	-.95354	.1167418	-8.17	0.000	-1.182904	-.7241762
ldlist	-.1343401	.0431032	-3.12	0.002	-.2190255	-.0496548
rooms	.2545271	.0185303	13.74	0.000	.2181203	.2909338
stratio	-.0524512	.0058971	-8.89	0.000	-.0640373	-.0408651
_cons	11.08387	.3181115	34.84	0.000	10.45887	11.70886

4.2 INFORMATION CRITERIA

```
. use hprice2a.(Housing price data for Boston-area communities)

. quietly regress lprice lnox ldlist rooms stratio

. estat ic
```

Akaike's information criterion and Bayesian information criterion

Model	Obs	ll(null)	ll(model)	df	AIC	BIC
.	506	-265.4135	-43.49514	5	96.99028	118.123

Note: N=Obs used in calculating BIC; see [R] BIC note.

4.3 THE COEFFICIENT ESTIMATES & BETA COEFFICIENTS

```
. use hprice2a.(Housing price data for Boston-area communities)
```

```
. quietly regress lprice lnox ldist rooms stratio
```

```
. regress, beta
```

Source	SS	df	MS	Number of obs	=	506
Model	49.3987735	4	12.3496934	F(4, 501)	=	175.86
Residual	35.1834974	501	.070226542	Prob > F	=	0.0000
Total	84.5822709	505	.167489645	R-squared	=	0.5840
				Adj R-squared	=	0.5807
				Root MSE	=	.265

lprice	Coef.	Std. Err.	t	P> t	Beta
lnox	-.95354	.1167418	-8.17	0.000	-.4692738
ldist	-.1343401	.0431032	-3.12	0.002	-.1770941
rooms	.2545271	.0185303	13.74	0.000	.4369626
stratio	-.0524512	.0058971	-8.89	0.000	-.2775771
_cons	11.08387	.3181115	34.84	0.000	.

4.4 RECOVERING ESTIMATION RESULTS

```
. use hprice2a.(Housing price data for Boston-area communities)
```

```
. quietly regress lprice lnox ldist rooms stratio
```

```
. ereturn list
```

scalars:

```

e(N) = 506
e(df_m) = 4
e(df_r) = 501
e(F) = 175.8550695227946
e(r2) = .5840322442976398
e(rmse) = .2650029089298266
e(mss) = 49.39877352102587

```

```

e(rss) = 35.18349741237627
e(r2_a) = .5807111444517128
e(ll) = -43.4951392092929
e(ll_0) = -265.4134648194153
e(rank) = 5

```

macros:

```

e(cmdline) : "regress lprice lnox ldist rooms stratio"
e(title) : "Linear regression"
e(marginsok) : "XB default"
e(vce) : "ols"
e(depvar) : "lprice"
e(cmd) : "regress"
e(properties) : "b V"
e(predict) : "regres_p"
e(model) : "ols"
e(estat_cmd) : "regress_estat"

```

matrices:

```

e(b) : 1 x 5
e(V) : 5 x 5

```

functions:

```

e(sample)

```

```

. use hprice2a.(Housing price data for Boston-area communities)

```

```

. quietly regress lprice lnox ldist rooms stratio

```

```

. matrix list e(b)

```

```

e(b)[1,5]

```

	lnox	ldist	rooms	stratio	_cons
y1	-.95354002	-.13434015	.25452706	-.05245119	11.083865

```

. use hprice2a.(Housing price data for Boston-area communities)

```

```

. quietly regress lprice lnox ldist rooms stratio

```

```

. estat vce

```

Covariance matrix of coefficients of regress model

e(V)	lnox	ldist	rooms	stratio	_cons
-----+-----					
lnox	.01362865				
ldist	.00426247	.00185789			
rooms	.00035279	.00003043	.00034337		
stratio	9.740e-07	.00002182	.00003374	.00003478	
_cons	-.03037429	-.01001835	-.00341397	-.00088151	.10119496

4.5 DETECTING COLLINEARLITY IN REGRESSION

```
. use hprice2a.(Housing price data for Boston-area communities)

. quietly regress lprice lnox ldist rooms stratio

. estat vif
```

Variable	VIF	1/VIF
-----+-----		
lnox	3.98	0.251533
ldist	3.89	0.257162
rooms	1.22	0.820417
stratio	1.17	0.852488
-----+-----		
Mean VIF	2.56	

4.6 PRESENTING REGRESSION ESTIMATES

```
. use hprice2a.(Housing price data for Boston-area communities)

. generate rooms2 = rooms^2

.

. quietly regress lprice rooms

. estimates store model1

.

. quietly regress lprice rooms rooms2 ldist

. estimates store model2

.

. quietly regress lprice ldist stratio lnox

. estimates store model3

.

. quietly regress lprice lnox ldist rooms stratio

. estimates store model4

.

. estimates table model1 model2 model3 model4, stat(r2_a, rmse) b(%7.3g) se(%6.
> 3g) p(%4.3f)
```

Variable	model1	model2	model3	model4
-----+-----				
rooms	.369	-.821		.255
	.0201	.183		.0185

		0.000	0.000		0.000
rooms2			.0889		
			.014		
			0.000		
ldist			.237	-.157	-.134
			.0255	.0505	.0431
			0.000	0.002	0.002
stratio				-.0775	-.0525
				.0066	.0059
				0.000	0.000
lnox				-1.22	-.954
				.135	.117
				0.000	0.000
_cons		7.62	11.3	13.6	11.1
		.127	.584	.304	.318
		0.000	0.000	0.000	0.000
-----+-----					
r2_a		.399	.5	.424	.581
rmse		.317	.289	.311	.265
-----+-----					

legend: b/se/p

```
. use hprice2a.(Housing price data for Boston-area communities)

. generate rooms2 = rooms^2

.
. quietly regress lprice rooms

. estimates store model1

.
. quietly regress lprice rooms rooms2 ldist

. estimates store model2

.
. quietly regress lprice ldist stratio lnox

. estimates store model3

.
. quietly regress lprice lnox ldist rooms stratio

. estimates store model4

.
. estimates table model1 model2 model3 model4, stat(r2_a rmse ll) b(%7.3g) star
> title("Models of Median Housing Price")
```

Models of Median Housing Price

Variable	model1	model2	model3	model4
rooms	.369***	-.821***		.255***

rooms2		.0889***		
ldist		.237***	-.157**	-.134**
stratio			-.0775***	-.0525***
lnox			-1.22***	-.954***
_cons	7.62***	11.3***	13.6***	11.1***
-----+				
r2_a	.399	.5	.424	.581
rmse	.317	.289	.311	.265
ll	-136	-88.6	-124	-43.5

legend: * p<0.05; ** p<0.01; *** p<0.001				

4.7 HYPOTHESIS TESTS, LINEAR RESTRICTIONS, & CONSTRAINED LEAST SQUARES

4.7.1 Wald Tests with Test

```
. use hprice2a.(Housing price data for Boston-area communities)
```

```
. regress lprice lnox ldist rooms stratio
```

Source	SS	df	MS	Number of obs	=	506
-----+				F(4, 501)	=	175.86
Model	49.3987735	4	12.3496934	Prob > F	=	0.0000
Residual	35.1834974	501	.070226542	R-squared	=	0.5840
-----+				Adj R-squared	=	0.5807
Total	84.5822709	505	.167489645	Root MSE	=	.265

lprice	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+						
lnox	-.95354	.1167418	-8.17	0.000	-1.182904	-.7241762
ldist	-.1343401	.0431032	-3.12	0.002	-.2190255	-.0496548
rooms	.2545271	.0185303	13.74	0.000	.2181203	.2909338
stratio	-.0524512	.0058971	-8.89	0.000	-.0640373	-.0408651
_cons	11.08387	.3181115	34.84	0.000	10.45887	11.70886

```
. test rooms
```

```
( 1) rooms = 0
```

```
F( 1, 501) = 188.67
Prob > F = 0.0000
```

```
. use hprice2a.(Housing price data for Boston-area communities)
```

```
. quietly regress lprice lnox ldist rooms stratio
```

```
. test rooms = 0.33
```

```
( 1) rooms = .33
```

```

F( 1, 501) = 16.59
Prob > F = 0.0001

```

4.7.2 Wald Tests involving Linear Combinations of Parameters

```

. use hprice2a.(Housing price data for Boston-area communities)

. quietly regress lprice lnox ldist rooms stratio

. lincom rooms + ldist + stratio

( 1) ldist + rooms + stratio = 0

```

	lprice	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
	(1)	.0677357	.0490714	1.38	0.168	-.0286753	.1641468

```

. test ldist = stratio

```

```

( 1) ldist - stratio = 0

```

```

F( 1, 501) = 3.63
Prob > F = 0.0574

```

```

. test lnox = 10*stratio

```

```

( 1) lnox - 10*stratio = 0

```

```

F( 1, 501) = 10.77
Prob > F = 0.0011

```

```

. use hprice2a.(Housing price data for Boston-area communities)

. constraint def 1 rooms + ldist + stratio = 0

. cnsreg lprice lnox ldist rooms stratio, constraint(1)

```

Constrained linear regression	Number of obs	=	506
	F(3, 502)	=	233.42
	Prob > F	=	0.0000
	Root MSE	=	0.2652

```

( 1) ldist + rooms + stratio = 0

```

	lprice	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
	lnox	-1.083392	.0691935	-15.66	0.000	-1.219337	-.9474478
	ldist	-.1880712	.0185284	-10.15	0.000	-.2244739	-.1516684
	rooms	.2430633	.01658	14.66	0.000	.2104886	.2756381
	stratio	-.0549922	.0056075	-9.81	0.000	-.0660092	-.0439752

_cons	11.48651	.1270377	90.42	0.000	11.23691	11.7361
-------	----------	----------	-------	-------	----------	---------

4.7.3 *Joint Hypothesis Tests*

```
. use hprice2a.(Housing price data for Boston-area communities)

. quietly regress lprice lnox ldlist rooms stratio

. test lnox ldlist

( 1)  lnox = 0
( 2)  ldlist = 0

      F( 2, 501) = 58.95
      Prob > F = 0.0000
```

```
. use hprice2a.(Housing price data for Boston-area communities)

. quietly regress lprice lnox ldlist rooms stratio

. test (lnox = 10*stratio) (ldlist = stratio)

( 1)  lnox - 10*stratio = 0
( 2)  ldlist - stratio = 0

      F( 2, 501) = 5.94
      Prob > F = 0.0028
```

4.7.4 *Testing Nonlinear Restrictions & Forming Nonlinear Combinations*

```
. use hprice2a.(Housing price data for Boston-area communities)

. quietly regress lprice lnox ldlist rooms stratio

. testnl _b[lnox] * _b[stratio] = 0.06

(1)  _b[lnox] * _b[stratio] = 0.06

      chi2(1) = 1.44
      Prob > chi2 = 0.2300

. use hprice2a.(Housing price data for Boston-area communities)

. quietly regress lprice lnox ldlist rooms stratio

. testnl (_b[lnox] * _b[stratio] = 0.06) (_b[rooms] * _b[ldlist] = 3 * _b[lnox])
>
```



```
(1) _b[lnox] * _b[stratio] = 0.06
(2) _b[rooms] * _b[lldist] = 3 * _b[lnox]
```

```
chi2(2) =      184.94
Prob > chi2 =      0.0000
```

4.7.5 *Testing Competing (Non-Nested) Models*

```
. use hprice2a.(Housing price data for Boston-area communities)

. nnest lprice lnox lldist rooms stratio (crime proptax lldist rooms stratio)
command nnest is unrecognized
r(199);

end of do-file
r(199);
```

4.8 COMPUTING RESIDUALS & PREDICTED VALUES

```
. use hprice2a.(Housing price data for Boston-area communities)

. quietly regress lprice lnox lldist rooms stratio

. predict double lpricehat, xb

. label var lpricehat "Predicted log price"

. twoway (scatter lpricehat lprice, msize(small) mcolor(black) msize(tiny)) ||
> (line lprice lprice if lprice <., clwidth(thin)), ytitle("Predicted log media
> n housing price") xtitle("Actual log median housing price") aspectratio(1) le
> gend(off)
```

4.8.1 *Computing Interval Predictions*

```
. use hprice2a.(Housing price data for Boston-area communities)

. quietly regress lprice lnox if _n <= 100

. predict double xb if e(sample)
(option xb assumed; fitted values)
(406 missing values generated)

. predict double stdpred if e(sample), stdp
(406 missing values generated)

. scalar tval = invttail(e(df_r), 0.975)
```

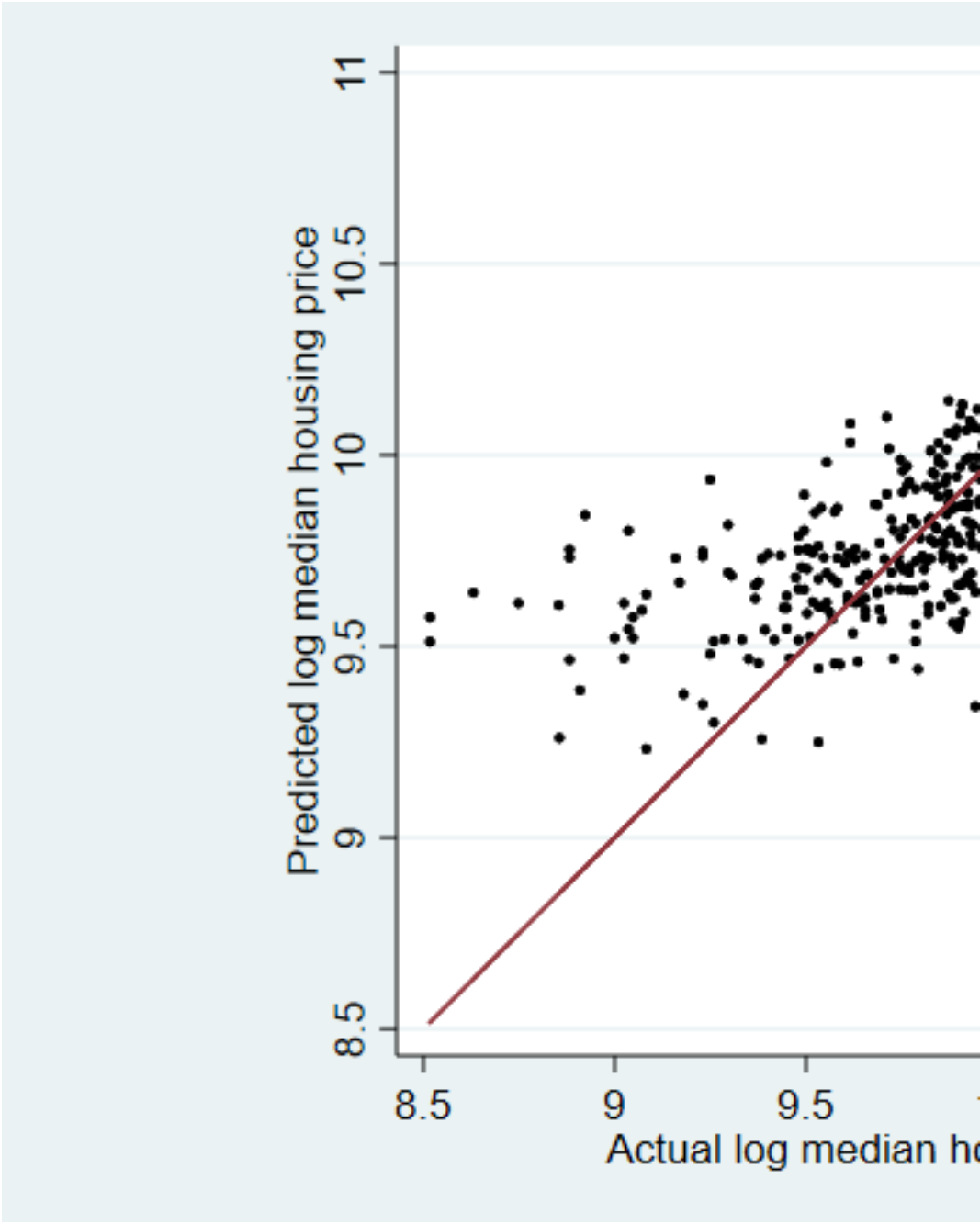


Figure 1: Actual versus predicted values from regression model

```

. generate double uplim = xb + tval * stdpred
(406 missing values generated)

. generate double lowlim = xb - tval * stdpred
(406 missing values generated)

.
. summarize lnox if e(sample), meanonly

. generate lnoxbar = r(mean)

.
. label var xb "Pred"

. label var uplim "95% prediction interval"

. label var lowlim "95% prediction interval"

.
. twoway (scatter lprice lnox if e(sample), sort ms(0h) xline('lnoxbar')) (conn
> ected xb lnox if e(sample), sort msize(small)) (rline uplim lowlim lnox if e(
> sample), sort), ytitle(Actual and predicted log price) legend(cols(3))

.

```

4.9 COMPUTING MARGINAL EFFECTS

```

. use hprice2a.(Housing price data for Boston-area communities)

. regress price nox dist rooms stratio proptax

```

Source	SS	df	MS	Number of obs	=	506
-----+-----				F(5, 500)	=	165.85
Model	2.6717e+10	5	5.3434e+09	Prob > F	=	0.0000
Residual	1.6109e+10	500	32217368.7	R-squared	=	0.6239
-----+-----				Adj R-squared	=	0.6201
Total	4.2826e+10	505	84803032	Root MSE	=	5676

price	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+-----					
nox	-2570.162	407.371	-6.31	0.000	-3370.532 -1769.793
dist	-955.7175	190.7124	-5.01	0.000	-1330.414 -581.021
rooms	6828.264	399.7034	17.08	0.000	6042.959 7613.569
stratio	-1127.534	140.7653	-8.01	0.000	-1404.099 -850.9699
proptax	-52.24272	22.53714	-2.32	0.021	-96.52188 -7.963555
_cons	20440.08	5290.616	3.86	0.000	10045.5 30834.66
-----+-----					

```

. mfx, eyex

```

Elasticities after regress

y = Fitted values (predict)

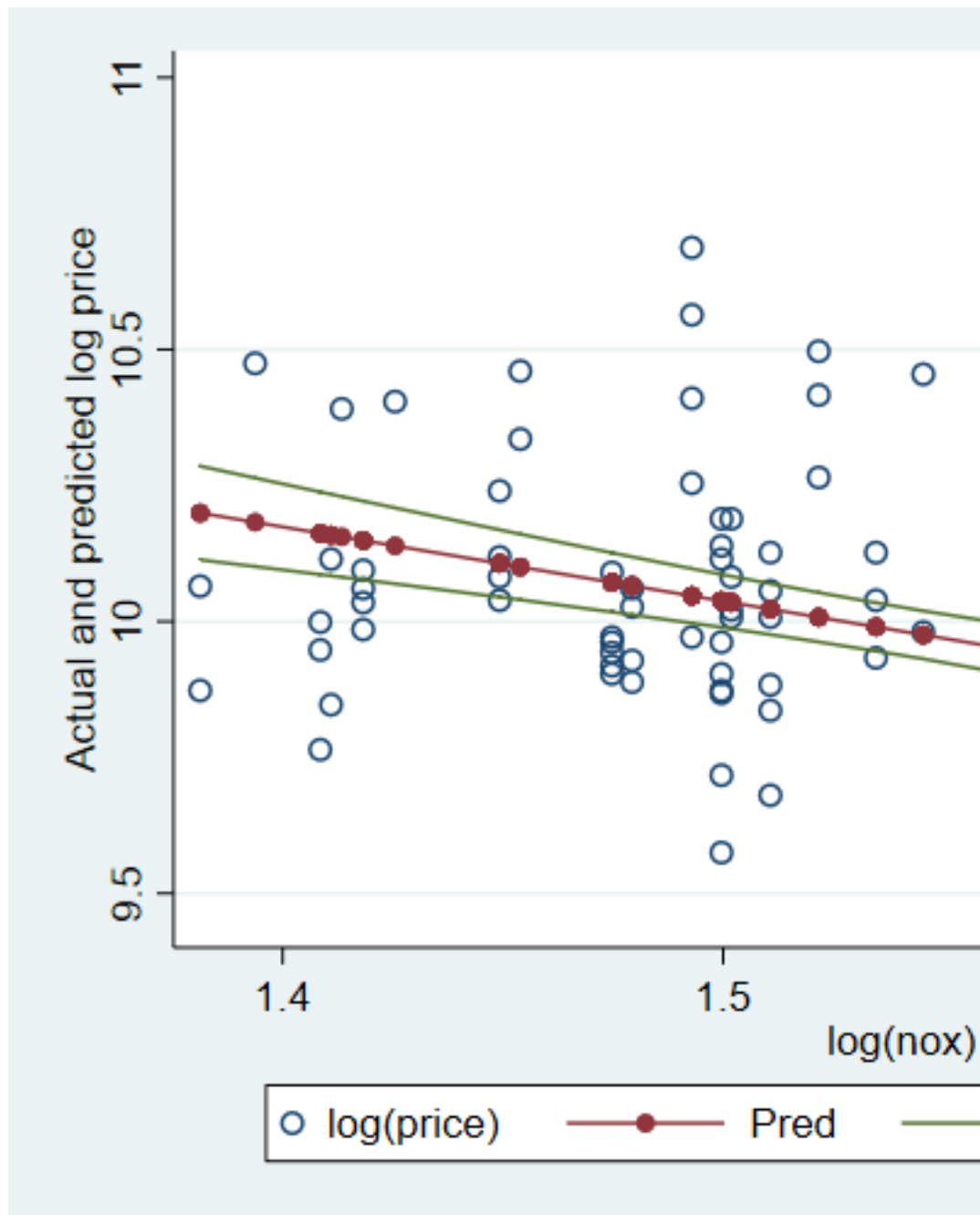


Figure 2: Point and interval predictions from bivariate regression

= 22511.51

variable	ey/ex	Std. Err.	z	P> z	[95% C.I.]	X
nox	-.6336244	.10068	-6.29	0.000	-.830954	-.436295		5.54978
dist	-.1611472	.03221	-5.00	0.000	-.224273	-.098022		3.79575
rooms	1.906099	.1136	16.78	0.000	1.68344	2.12876		6.28405
stratio	-.9245706	.11589	-7.98	0.000	-1.15171	-.697429		18.4593
proptax	-.0947401	.04088	-2.32	0.020	-.174871	-.014609		40.8237

SPECIFYING THE FUNCTIONAL FORM

5.1 INTRODUCTION

5.2 SPECIFICATION ERROR

5.2.1 *Omitting relevant Variables from the Model*

5.2.2 *Graphically Analyzing Regression data*

```
. use hprice2a.(Housing price data for Boston-area communities)

. graph matrix lprice lnox ldist rooms stratio, ms(0h) msize(tiny)
```

5.2.3 *Added-Variable Plots*

```
. use hprice2a.(Housing price data for Boston-area communities)

. generate rooms2 = rooms^2

. regress lprice lnox ldist rooms rooms2 stratio lproptax
```

Source		SS		df		MS		Number of obs	=	506
-----+										
Model		52.8357813		6		8.80596356		F(6, 499)	=	138.41
Residual		31.7464896		499		.06362022		Prob > F	=	0.0000
-----+										
								R-squared	=	0.6247
								Adj R-squared	=	0.6202
Total		84.5822709		505		.167489645		Root MSE	=	.25223

-----+										
lprice		Coef.		Std. Err.		t		P> t		[95% Conf. Interval]
-----+										
lnox		-.6615694		.1201606		-5.51		0.000		-.8976524 -.4254864
ldist		-.095087		.0421435		-2.26		0.024		-.1778875 -.0122864
rooms		-.5625662		.1610315		-3.49		0.001		-.8789496 -.2461829
rooms2		.0634347		.0124621		5.09		0.000		.0389501 .0879193
stratio		-.0362928		.0060699		-5.98		0.000		-.0482185 -.0243671
lproptax		-.2211125		.0410202		-5.39		0.000		-.301706 -.1405189
_cons		14.15454		.5693846		24.86		0.000		13.03585 15.27323

```
. avplots, ms(0h) msize(small) col(2)
```

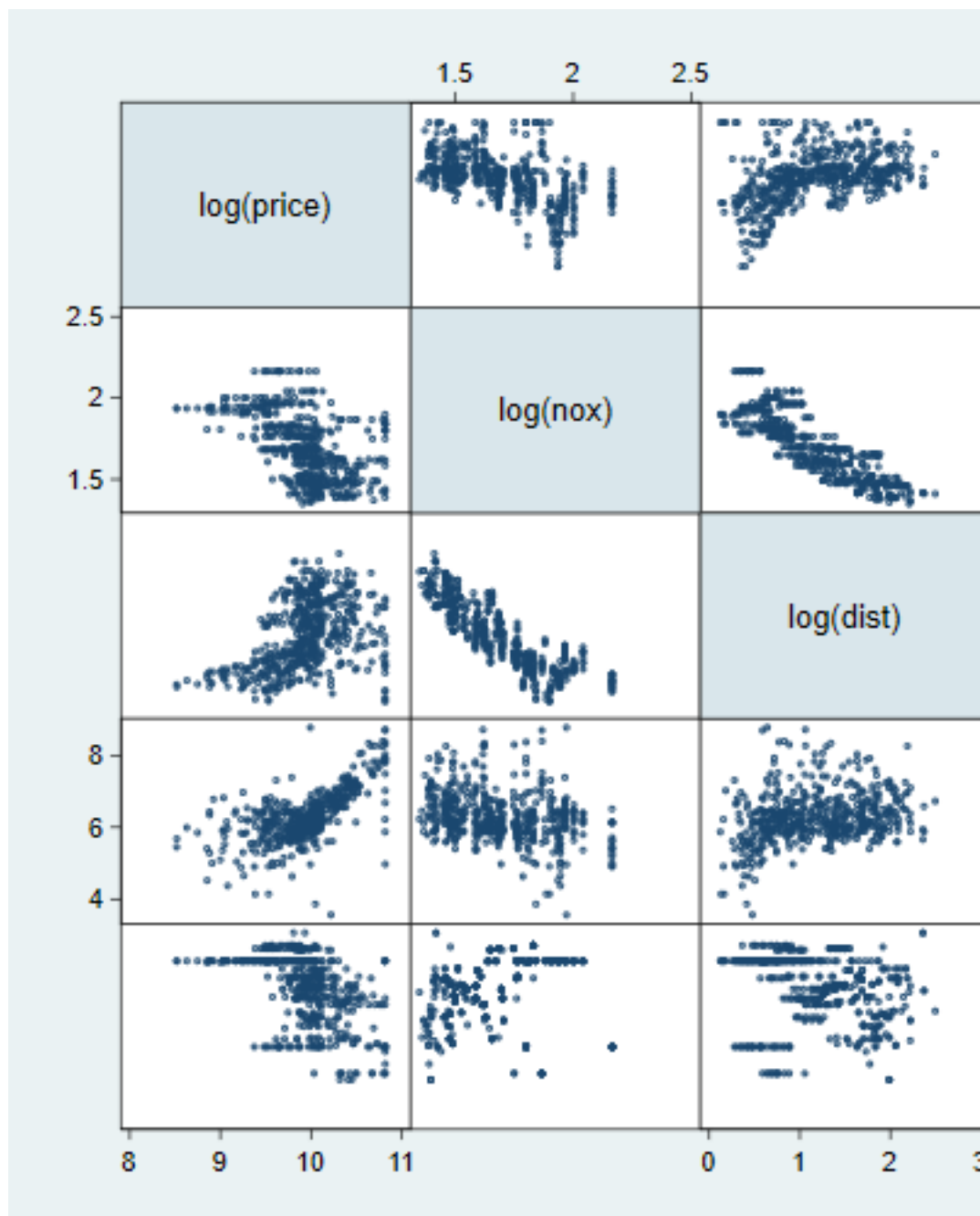


Figure 3: graph matrix of regression variables

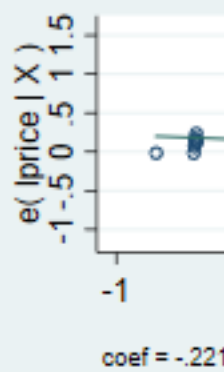
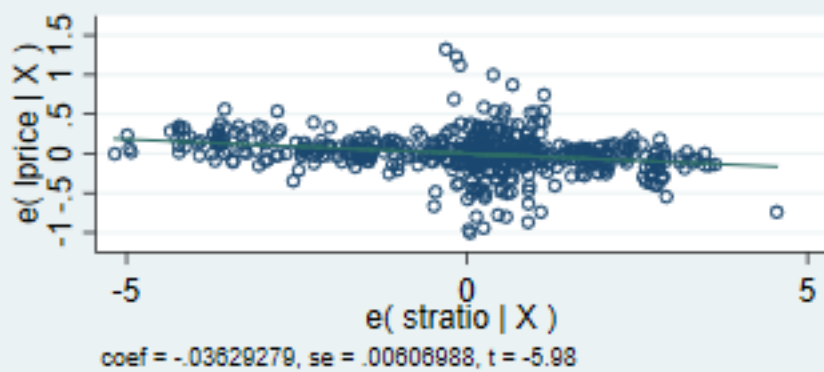
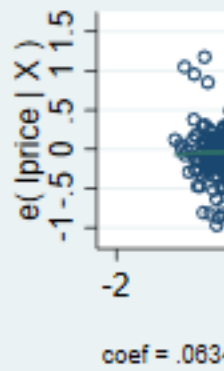
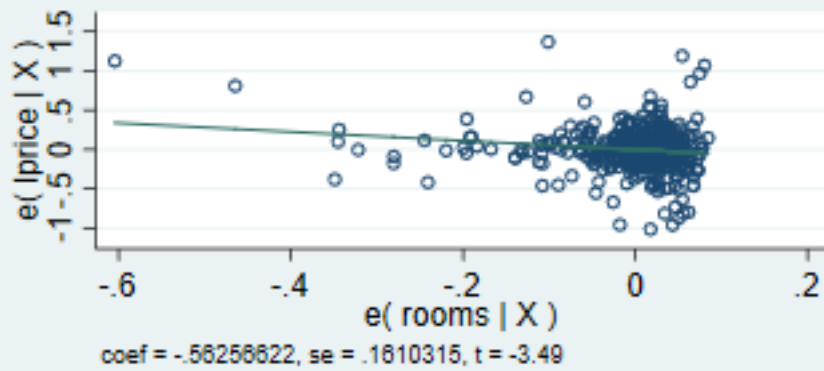
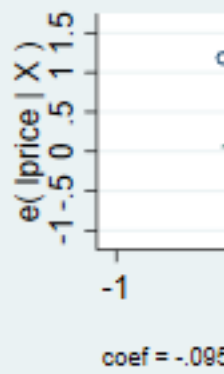
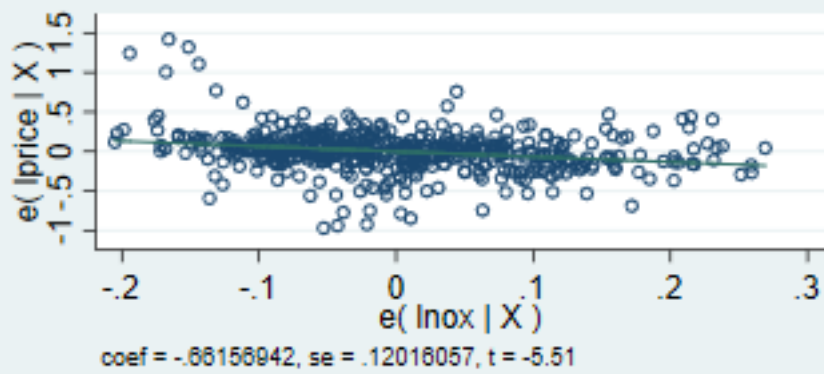


Figure 4: Added-variable plots

5.2.4 Including Irrelevant Variables in the Model

5.2.5 The Asymmetry of Specification Error

5.2.6 Misspecification of the Functional Form

5.2.7 Ramsey's RESET

```
. use hprice2a.(Housing price data for Boston-area communities)
```

```
. quietly regress lprice lnox ldist rooms stratio
```

```
. estat ovtest
```

Ramsey RESET test using powers of the fitted values of lprice

Ho: model has no omitted variables

F(3, 498) = 9.69

Prob > F = 0.0000

```
. estat ovtest, rhs
```

Ramsey RESET test using powers of the independent variables

Ho: model has no omitted variables

F(12, 489) = 11.79

Prob > F = 0.0000

```
. use hprice2a.(Housing price data for Boston-area communities)
```

```
. generate rooms2 = rooms^2
```

```
. regress lprice lnox ldist rooms rooms2 stratio lproptax
```

Source		SS	df	MS	Number of obs	=	506
-----+-----					F(6, 499)	=	138.41
Model		52.8357813	6	8.80596356	Prob > F	=	0.0000
Residual		31.7464896	499	.06362022	R-squared	=	0.6247
-----+-----					Adj R-squared	=	0.6202
Total		84.5822709	505	.167489645	Root MSE	=	.25223

-----+-----						
lprice		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+-----						
lnox		-.6615694	.1201606	-5.51	0.000	-.8976524 -.4254864
ldist		-.095087	.0421435	-2.26	0.024	-.1778875 -.0122864
rooms		-.5625662	.1610315	-3.49	0.001	-.8789496 -.2461829
rooms2		.0634347	.0124621	5.09	0.000	.0389501 .0879193
stratio		-.0362928	.0060699	-5.98	0.000	-.0482185 -.0243671
lproptax		-.2211125	.0410202	-5.39	0.000	-.301706 -.1405189
_cons		14.15454	.5693846	24.86	0.000	13.03585 15.27323
-----+-----						

```
. estat ovtest
```

Ramsey RESET test using powers of the fitted values of lprice

Ho: model has no omitted variables

F(3, 496) = 1.64
Prob > F = 0.1798

5.2.8 Specification Plots

```
. use hprice2a.(Housing price data for Boston-area communities)

. generate rooms2 = rooms^2

. quietly regress lprice lnox ldlist rooms rooms2 stratio lproptax

. rvppplot ldlist, ms(0h) yline(0)
```

5.2.9 Specification & Interaction Terms

```
. use hprice2a.(Housing price data for Boston-area communities)

. generate taxschl = lproptax * stratio

. regress lprice lnox ldlist lproptax stratio taxschl
```

Source		SS		df	MS	Number of obs	=	506
-----+								
Model		38.7301562		5	7.74603123	F(5, 500)	=	84.47
Residual		45.8521148		500	.09170423	Prob > F	=	0.0000
-----+								
						R-squared	=	0.4579
-----+								
						Adj R-squared	=	0.4525
Total		84.5822709		505	.167489645	Root MSE	=	.30283

lprice		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+						
lnox		-.9041103	.1441253	-6.27	0.000	-1.187276 - .6209444
ldlist		-.1430541	.0501831	-2.85	0.005	-.2416499 - .0444583
lproptax		-1.48103	.5163117	-2.87	0.004	-2.495438 - .4666219
stratio		-.4388722	.1538321	-2.85	0.005	-.7411093 - .1366351
taxschl		.0641648	.026406	2.43	0.015	.0122843 .1160452
_cons		21.47905	2.952307	7.28	0.000	15.6786 27.27951

5.2.10 Outlier Statistics & Measures of Leverage

```
. use hprice2a.(Housing price data for Boston-area communities)

. generate rooms2 = rooms^2
```

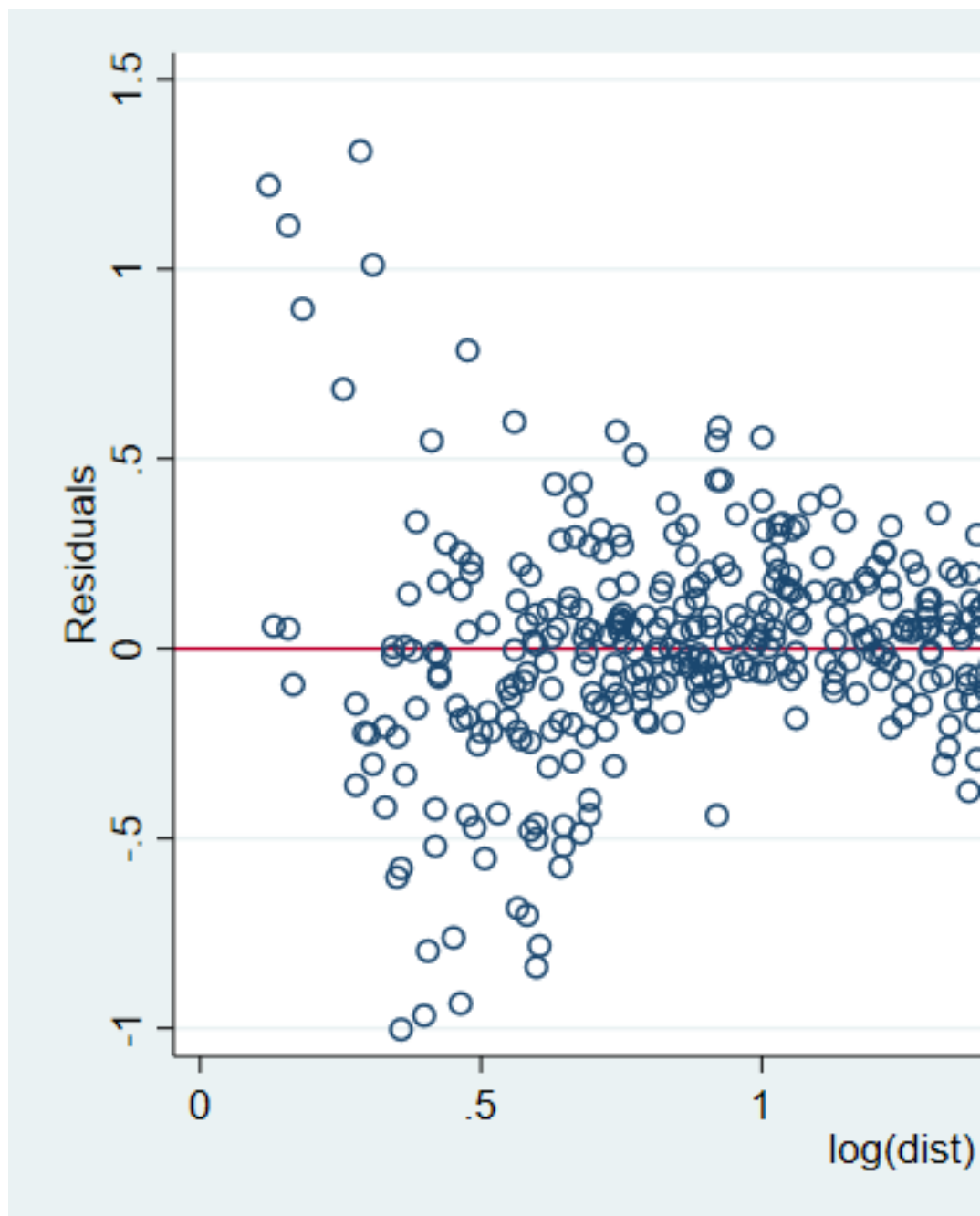


Figure 5: Residual-versus-predictor plot

```
. quietly regress lprice lnox ldist rooms rooms2 stratio lproptax

. generate town = _n

. predict double lev if e(sample), leverage

. predict double eps if e(sample), res

. generate eps2 = eps^2

. summarize price lprice
```

Variable	Obs	Mean	Std. Dev.	Min	Max
price	506	22511.51	9208.856	5000	50001
lprice	506	9.941057	.409255	8.517193	10.8198

```
. gsort -lev

. list town price lprice lev eps2 in 1/5
```

	town	price	lprice	lev	eps2
1.	366	27499	10.2219	.17039262	.6181372
2.	368	23100	10.04759	.11272637	.3002205
3.	365	21900	9.994242	.10947853	.3308896
4.	258	50001	10.8198	.08036068	.0604706
5.	226	50001	10.8198	.0799096	.0338277

```
. gsort -eps2

. list town price lprice lev eps2 in 1/5
```

	town	price	lprice	lev	eps2
1.	369	50001	10.8198	.02250047	1.71812
2.	373	50001	10.8198	.01609848	1.489409
3.	372	50001	10.8198	.02056901	1.242105
4.	370	50001	10.8198	.0172083	1.022456
5.	406	5000	8.517193	.00854955	1.006366

```
. predict double dfits if e(sample), dfits

. gsort -dfits

. quietly generate cutoff = abs(dfits) > 2 * sqrt((e(df_m) + 1 )/e(N)) & e(sample) > e)

. list town price lprice dfits if cutoff
```

	town	price	lprice	dfits
--	------	-------	--------	-------

```

1. | 366 27499 10.2219 1.5679033 |
2. | 368 23100 10.04759 .82559867 |
3. | 369 50001 10.8198 .8196735 |
4. | 372 50001 10.8198 .65967704 |
5. | 373 50001 10.8198 .63873964 |
6. | 371 50001 10.8198 .55639311 |
7. | 370 50001 10.8198 .54354054 |
8. | 361 24999 10.12659 .32184327 |
9. | 359 22700 10.03012 .31516743 |
10. | 408 27901 10.23642 .31281326 |
11. | 367 21900 9.994242 .31060611 |
12. | 360 22600 10.02571 .28892457 |
13. | 363 20800 9.942708 .27393758 |
14. | 358 21700 9.985067 .24312885 |
490. | 386 7200 8.881836 -.23838749 |
491. | 388 7400 8.909235 -.25909393 |
492. | 491 8100 8.999619 -.26584795 |
493. | 400 6300 8.748305 -.28782824 |
494. | 416 7200 8.881836 -.29288953 |
495. | 402 7200 8.881836 -.29595696 |
496. | 381 10400 9.249561 -.29668364 |
497. | 258 50001 10.8198 -.30053391 |
498. | 385 8800 9.082507 -.302916 |
499. | 420 8400 9.035987 -.30843965 |
500. | 490 7000 8.853665 -.3142718 |
501. | 401 5600 8.630522 -.33273658 |
502. | 417 7500 8.922658 -.34950136 |
503. | 399 5000 8.517193 -.36618139 |
504. | 406 5000 8.517193 -.37661853 |
505. | 415 7012 8.855378 -.43879798 |
506. | 365 21900 9.994242 -.85150064 |
+-----+

```

```

.
. dfbeta lnox
      _dfbeta_1: dfbeta(lnox)

. quietly generate dfcut = abs(_dfbeta_1) > 2 * sqrt(e(N)) & e(sample)

. sort _dfbeta_1

. summarize lnox

      Variable |      Obs      Mean   Std. Dev.      Min      Max
-----+-----
      lnox |      506   1.693091   .2014102   1.348073   2.164472

. list town price lprice lnox _dfbeta_1

+-----+

```

	town	price	lprice	lnox	_dfbeta_1

1.	369	50001	10.8198	1.842136	-.4316933
2.	372	50001	10.8198	1.842136	-.4257791
3.	373	50001	10.8198	1.899118	-.3631822
4.	371	50001	10.8198	1.842136	-.2938702
5.	370	50001	10.8198	1.842136	-.2841335

6.	365	21900	9.994242	1.971299	-.2107066
7.	408	27901	10.23642	1.885553	-.1728729
8.	368	23100	10.04759	1.842136	-.1309522
9.	11	15000	9.615806	1.656321	-.1172723
10.	410	27499	10.2219	1.786747	-.1117743

11.	413	17900	9.792556	1.786747	-.0959273
12.	437	9600	9.169518	2.00148	-.0955826
13.	146	13800	9.532424	2.164472	-.0914387
14.	438	8700	9.071078	2.00148	-.0856147
15.	420	8400	9.035987	1.971299	-.085223

16.	145	11800	9.375854	2.164472	-.0816827
17.	439	8400	9.035987	2.00148	-.070508
18.	182	36199	10.49679	1.585145	-.0668001
19.	423	20800	9.942708	1.814825	-.064928
20.	157	13100	9.480368	2.164472	-.0622912

21.	258	50001	10.8198	1.867176	-.0570776
22.	158	41299	10.62859	1.800058	-.0563827
23.	409	17200	9.752665	1.786747	-.0549674
24.	343	16500	9.711116	1.644805	-.0542086
25.	414	16300	9.69892	1.786747	-.0522002

26.	143	13400	9.50301	2.164472	-.0499795
27.	446	11800	9.375854	2.00148	-.0498875
28.	31	12700	9.449357	1.682688	-.0480591
29.	156	15600	9.655026	2.164472	-.0474292
30.	493	20100	9.908475	1.806648	-.0472557

31.	480	21400	9.971147	1.814825	-.0469688
32.	481	23001	10.04329	1.671473	-.0465269
33.	451	13400	9.50301	1.964311	-.0461007
34.	9	16500	9.711116	1.656321	-.0449441
35.	33	13200	9.487972	1.682688	-.0427313

36.	28	14800	9.602383	1.682688	-.0383155
37.	32	14500	9.581903	1.682688	-.0374429
38.	188	32000	10.37349	1.474763	-.0370926
39.	435	11700	9.367344	1.964311	-.0370922
40.	454	17794	9.786616	1.964311	-.0369095

41.	436	13400	9.50301	2.00148	-.0368249
42.	35	13500	9.510445	1.682688	-.036522
43.	161	27000	10.20359	1.800058	-.035759
44.	441	10500	9.25913	2.00148	-.0354014
45.	26	13900	9.539644	1.682688	-.0336016

46.	445	10800	9.287301	2.00148	-.0330708

47.	10	18900	9.846917	1.656321	-.0328998
48.	101	27499	10.2219	1.648659	-.0317264
49.	189	29801	10.3023	1.474763	-.0315825
50.	12	18900	9.846917	1.656321	-.0311876

51.	472	19600	9.883285	1.671473	-.0303079
52.	200	34899	10.46021	1.393766	-.0302973
53.	34	13100	9.480368	1.682688	-.0300799
54.	470	20100	9.908475	1.757858	-.0299316
55.	482	23699	10.07319	1.671473	-.0295599

56.	162	50001	10.8198	1.800058	-.0294572
57.	263	48801	10.79551	1.867176	-.0288593
58.	23	15200	9.62905	1.682688	-.0284555
59.	468	19100	9.857444	1.764731	-.0278265
60.	448	12600	9.441452	2.00148	-.027274

61.	450	13000	9.472705	1.964311	-.0268602
62.	24	14500	9.581903	1.682688	-.0267329
63.	506	11900	9.384294	1.745715	-.0259339
64.	75	24101	10.09001	1.474763	-.0255723
65.	159	24299	10.09819	1.800058	-.0253549

66.	191	37001	10.5187	1.474763	-.0246281
67.	106	19500	9.87817	1.648659	-.0242416
68.	292	37298	10.5267	1.413423	-.0242084
69.	107	19500	9.87817	1.648659	-.0240593
70.	484	21800	9.989665	1.671473	-.0238902

71.	21	13600	9.517825	1.682688	-.0234366
72.	230	31499	10.35771	1.617406	-.0231037
73.	102	26500	10.1849	1.648659	-.0219822
74.	62	16000	9.680344	1.510722	-.0219783
75.	489	15200	9.62905	1.806648	-.0218215

76.	25	15600	9.655026	1.682688	-.0215176
77.	127	15700	9.661416	1.83098	-.0211289
78.	430	9500	9.159047	1.915451	-.0207799
79.	455	14900	9.609117	1.964311	-.0206806
80.	108	20400	9.92329	1.648659	-.020656

81.	173	23100	10.04759	1.629241	-.020537
82.	215	23699	10.07319	1.587192	-.0205038
83.	275	32400	10.38591	1.497388	-.0203821
84.	184	32500	10.389	1.585145	-.019082
85.	185	26399	10.18108	1.585145	-.0189528

86.	190	34899	10.46021	1.474763	-.0188717
87.	456	14100	9.55393	1.964311	-.0187349
88.	483	24999	10.12659	1.671473	-.0186684
89.	473	23200	10.05191	1.757858	-.0183966
90.	224	30101	10.31231	1.623341	-.0176759

91.	233	41702	10.6383	1.623341	-.0173393
92.	180	37201	10.52409	1.585145	-.0173209
93.	417	7500	8.922658	1.915451	-.0171022
94.	375	13800	9.532424	1.899118	-.0166074

95.	316	16200	9.692766	1.693779	-.0165045
96.	279	29100	10.27849	1.497388	-.0162037
97.	119	20400	9.92329	1.699279	-.0161968
98.	176	29401	10.28878	1.629241	-.0161278
99.	29	18400	9.820106	1.682688	-.0157718
100.	374	13800	9.532424	1.899118	-.0145677
101.	183	37900	10.54271	1.585145	-.0144407
102.	40	30801	10.3353	1.453953	-.014429
103.	495	24499	10.10639	1.766442	-.0138828
104.	264	30999	10.34171	1.867176	-.013838
105.	105	20100	9.908475	1.648659	-.013284
106.	457	12700	9.449357	1.964311	-.0129107
107.	341	18700	9.836279	1.638997	-.0129061
108.	193	36399	10.5023	1.474763	-.0128875
109.	96	28399	10.25411	1.492904	-.0127983
110.	201	32899	10.4012	1.393766	-.0126864
111.	192	30501	10.32551	1.474763	-.0125861
112.	257	43998	10.6919	1.371181	-.0124312
113.	276	32000	10.37349	1.497388	-.0120773
114.	186	29599	10.2955	1.585145	-.0116235
115.	338	18500	9.825526	1.638997	-.0116081
116.	494	21800	9.989665	1.766442	-.0114905
117.	166	24999	10.12659	1.800058	-.0114317
118.	99	43800	10.68739	1.492904	-.0108364
119.	41	34899	10.46021	1.453953	-.0108243
120.	471	19900	9.898475	1.757858	-.0104401
121.	278	33100	10.40729	1.497388	-.0101313
122.	94	24999	10.12659	1.534714	-.0101088
123.	111	21700	9.985067	1.699279	-.01002
124.	317	17800	9.786954	1.693779	-.0099336
125.	434	14300	9.568015	1.964311	-.0098488
126.	340	19000	9.852194	1.638997	-.0097685
127.	117	21200	9.961757	1.699279	-.0093754
128.	346	17500	9.769957	1.48614	-.0091785
129.	469	17052	9.744022	1.757858	-.0091705
130.	452	15200	9.62905	1.964311	-.0091184
131.	231	24299	10.09819	1.617406	-.0090826
132.	268	50001	10.8198	1.7492	-.008913
133.	179	29900	10.30561	1.629241	-.0087004
134.	280	35102	10.46601	1.4884	-.0086304
135.	68	22000	9.998797	1.408545	-.0085998
136.	347	17200	9.752665	1.48614	-.0085918
137.	53	22620	10.02659	1.479329	-.0084467
138.	429	11000	9.305651	1.915451	-.0083504
139.	150	15400	9.642123	2.164472	-.0080456
140.	349	24499	10.10639	1.470176	-.0077041
141.	213	22400	10.01682	1.587192	-.0074679

142.	163	50001	10.8198	1.800058	-.0072357
143.	203	42302	10.65259	1.423108	-.0072147
144.	214	28099	10.24349	1.587192	-.0071618
145.	488	20600	9.933046	1.763017	-.0069738

146.	169	23799	10.0774	1.800058	-.0067039
147.	168	23799	10.0774	1.800058	-.0066772
148.	426	8300	9.024011	1.915451	-.0066314
149.	393	9700	9.179881	1.94591	-.0066296
150.	81	28001	10.24	1.449269	-.0064669

151.	140	17800	9.786954	1.83098	-.0063847
152.	83	24800	10.1186	1.449269	-.0063639
153.	52	20500	9.92818	1.479329	-.0061827
154.	306	28399	10.25411	1.551809	-.0060002
155.	104	19300	9.867861	1.648659	-.0059841

156.	51	19700	9.888374	1.479329	-.0058348
157.	236	24000	10.08581	1.623341	-.0057875
158.	499	21200	9.961757	1.766442	-.0056033
159.	1	24000	10.08581	1.682688	-.0055244
160.	254	42800	10.66429	1.460938	-.0054843

161.	165	22700	10.03012	1.800058	-.0053979
162.	216	24999	10.12659	1.587192	-.0053433
163.	416	7200	8.881836	1.915451	-.0053274
164.	228	31600	10.36091	1.617406	-.0051245
165.	293	27901	10.23642	1.413423	-.0049782

166.	27	16600	9.717158	1.682688	-.0049226
167.	49	14400	9.574984	1.499623	-.0049053
168.	229	46700	10.7515	1.617406	-.0047393
169.	147	15600	9.655026	2.164472	-.0047173
170.	175	22600	10.02571	1.629241	-.0047105

171.	128	16200	9.692766	1.83098	-.0045839
172.	312	22099	10.00329	1.693779	-.0044402
173.	122	20300	9.918376	1.759581	-.0044247
174.	235	29001	10.27509	1.623341	-.0042688
175.	274	35200	10.4688	1.534714	-.0042283

176.	73	22800	10.03452	1.418277	-.0041709
177.	206	22600	10.02571	1.587192	-.004164
178.	125	18800	9.841612	1.759581	-.0037207
179.	2	21599	9.980402	1.545433	-.0036146
180.	396	13100	9.480368	1.93586	-.0034284

181.	50	19400	9.873029	1.499623	-.0033576
182.	39	24701	10.1146	1.607436	-.0031628
183.	178	24600	10.1105	1.629241	-.0030147
184.	110	19400	9.873029	1.648659	-.0029739
185.	61	18700	9.836279	1.510722	-.0029386

186.	177	23200	10.05191	1.629241	-.0028818
187.	210	20000	9.903487	1.587192	-.0027495
188.	120	19300	9.867861	1.759581	-.0027346
189.	124	17300	9.758462	1.759581	-.0025984

190.	148	14600	9.588777	2.164472	-.0025087
191.	123	20500	9.92818	1.759581	-.0022385
192.	245	17600	9.775654	1.460938	-.0022249
193.	167	50001	10.8198	1.800058	-.002166
194.	500	17500	9.769957	1.766442	-.0021589
195.	428	10900	9.296518	1.915451	-.0021474
196.	196	50001	10.8198	1.439835	-.0021369
197.	248	20500	9.92818	1.460938	-.002122
198.	18	17500	9.769957	1.682688	-.0020629
199.	42	26601	10.1887	1.499623	-.0019077
200.	342	32778	10.39751	1.48614	-.0018963
201.	48	16600	9.717158	1.499623	-.0018894
202.	60	19600	9.883285	1.510722	-.0018844
203.	226	50001	10.8198	1.617406	-.001832
204.	155	17000	9.740969	2.164472	-.0018229
205.	353	18600	9.830916	1.413423	-.0016996
206.	137	17400	9.764226	1.83098	-.0016503
207.	394	13800	9.532424	1.93586	-.0016382
208.	333	19400	9.873029	1.477049	-.0016011
209.	211	21700	9.985067	1.587192	-.0015952
210.	223	27499	10.2219	1.623341	-.0015933
211.	330	22600	10.02571	1.526056	-.0015749
212.	220	23001	10.04329	1.704748	-.0015483
213.	390	11500	9.350102	1.94591	-.0014979
214.	272	25200	10.1346	1.534714	-.0014252
215.	209	24399	10.1023	1.587192	-.0013174
216.	395	12700	9.449357	1.93586	-.0013078
217.	505	22000	9.998797	1.745715	-.0012906
218.	282	35401	10.47449	1.4884	-.0012697
219.	63	22199	10.0078	1.510722	-.0011577
220.	324	18500	9.825526	1.595339	-.0011329
221.	348	23100	10.04759	1.456287	-.0011269
222.	74	23400	10.06049	1.418277	-.0010706
223.	142	14400	9.574984	1.83098	-.001046
224.	45	21200	9.961757	1.499623	-.0009506
225.	296	28601	10.2612	1.474763	-.0009321
226.	129	18000	9.798127	1.83098	-.0008757
227.	246	18500	9.825526	1.460938	-.0008337
228.	309	22800	10.03452	1.693779	-.0007587
229.	118	19200	9.862665	1.699279	-.0005592
230.	112	22800	10.03452	1.699279	-.000556
231.	303	26399	10.18108	1.465567	-.0005297
232.	335	20700	9.937889	1.638997	-.0004579
233.	80	20300	9.918376	1.474763	-.0003887
234.	208	22501	10.02131	1.587192	-.00029
235.	126	21400	9.971147	1.759581	-.0002649
236.	504	23899	10.08159	1.745715	-.0001415

237.	397	12500	9.433484	1.93586	-.0001163
238.	477	16700	9.723164	1.814825	-.0000081
239.	86	26601	10.1887	1.501853	-.0000522
240.	232	31701	10.3641	1.617406	-.0000151

241.	310	20300	9.918376	1.693779	.0001054
242.	219	21500	9.975808	1.704748	.0001567
243.	283	46000	10.7364	1.4884	.0001949
244.	449	14100	9.55393	1.964311	.0002028
245.	277	33200	10.41031	1.497388	.0002292

246.	44	24701	10.1146	1.499623	.0002693
247.	305	36098	10.49399	1.551809	.000271
248.	344	23899	10.08159	1.576915	.0002758
249.	297	27100	10.20729	1.474763	.0003851
250.	84	22900	10.03889	1.449269	.0005833

251.	212	19300	9.867861	1.587192	.0005905
252.	43	25301	10.1386	1.499623	.0006433
253.	237	25099	10.13058	1.623341	.0006712
254.	37	20000	9.903487	1.607436	.0006828
255.	170	22299	10.0123	1.800058	.0007145

256.	130	14300	9.568015	1.83098	.0007622
257.	70	20900	9.947504	1.408545	.0008296
258.	207	24399	10.1023	1.587192	.000875
259.	251	24399	10.1023	1.460938	.00088
260.	314	21600	9.980449	1.693779	.0009465

261.	501	16800	9.729135	1.766442	.0010252
262.	238	31499	10.35771	1.623341	.0010367
263.	71	24200	10.09411	1.418277	.0010956
264.	57	24701	10.1146	1.410987	.0011089
265.	138	17100	9.746834	1.83098	.0011234

266.	252	24800	10.1186	1.460938	.0011321
267.	503	20600	9.933046	1.745715	.0011443
268.	323	20400	9.92329	1.595339	.0011816
269.	227	37602	10.53481	1.617406	.0013204
270.	30	21000	9.952278	1.682688	.0013535

271.	56	35401	10.47449	1.393766	.0014627
272.	440	12800	9.4572	2.00148	.0016096
273.	144	15600	9.655026	2.164472	.0016168
274.	313	19400	9.873029	1.693779	.0017158
275.	281	45401	10.72329	1.4884	.0017327

276.	217	23300	10.05621	1.704748	.0018928
277.	136	18100	9.803667	1.83098	.0019621
278.	15	18200	9.809176	1.682688	.0020112
279.	271	21100	9.957028	1.534714	.0020471
280.	172	19100	9.857444	1.800058	.0023016

281.	267	30699	10.33199	1.867176	.0023292
282.	113	18800	9.841612	1.699279	.0023302
283.	87	22501	10.02131	1.501853	.0023375
284.	475	13800	9.532424	1.764731	.0023762

285.	352	24101	10.09001	1.413423	.002526
286.	244	23699	10.07319	1.453953	.0025868
287.	85	23899	10.08159	1.501853	.0026809
288.	72	21700	9.985067	1.418277	.0027141
289.	225	44802	10.71001	1.617406	.0027266
290.	304	33100	10.40729	1.551809	.0027396
291.	90	28701	10.26469	1.587192	.0029018
292.	403	12100	9.400961	1.93586	.0029982
293.	135	15600	9.655026	1.83098	.0030177
294.	337	19500	9.87817	1.638997	.0030569
295.	384	12300	9.417355	1.94591	.0031938
296.	38	13668	9.522813	1.607436	.0032047
297.	266	22800	10.03452	1.867176	.0032496
298.	187	50001	10.8198	1.585145	.0033213
299.	291	28501	10.25769	1.413423	.0033408
300.	422	14200	9.560997	1.971299	.0033722
301.	351	22900	10.03889	1.456287	.0033961
302.	131	19200	9.862665	1.83098	.00347
303.	322	23100	10.04759	1.595339	.003478
304.	249	24499	10.10639	1.460938	.0034818
305.	204	48499	10.7893	1.425515	.0035265
306.	164	50001	10.8198	1.800058	.0035734
307.	13	21700	9.985067	1.656321	.0036516
308.	474	29801	10.3023	1.814825	.0039417
309.	250	26200	10.17351	1.460938	.0039509
310.	181	39799	10.5916	1.585145	.0041167
311.	79	21200	9.961757	1.474763	.0041254
312.	498	18300	9.814656	1.766442	.0041364
313.	502	22400	10.01682	1.745715	.0042514
314.	339	20548	9.930519	1.638997	.0043115
315.	234	48301	10.78521	1.623341	.0043629
316.	308	28201	10.24711	1.551809	.004465
317.	354	30101	10.31231	1.410987	.0044676
318.	321	23799	10.0774	1.595339	.0046011
319.	383	11300	9.332558	1.94591	.0047599
320.	389	10200	9.230143	1.94591	.0047679
321.	194	31101	10.345	1.388791	.0047706
322.	260	30101	10.31231	1.867176	.0047774
323.	294	23899	10.08159	1.474763	.0049389
324.	222	21700	9.985067	1.623341	.0049435
325.	121	22000	9.998797	1.759581	.0049438
326.	243	22199	10.0078	1.453953	.0050146
327.	171	17400	9.764226	1.800058	.0053822
328.	315	23799	10.0774	1.693779	.0055875
329.	259	36001	10.4913	1.867176	.0056653
330.	273	24399	10.1023	1.534714	.0056988
331.	331	19800	9.893437	1.526056	.0057538

332.	109	19800	9.893437	1.648659	.005793
333.	404	8300	9.024011	1.93586	.0060482
334.	319	23100	10.04759	1.693779	.006068
335.	444	15400	9.642123	2.00148	.0062065

336.	114	18700	9.836279	1.699279	.0064821
337.	47	20000	9.903487	1.499623	.0065764
338.	350	26601	10.1887	1.456287	.0065786
339.	3	34700	10.4545	1.545433	.006829
340.	195	29100	10.27849	1.388791	.0068445

341.	411	15000	9.615806	1.786747	.0069358
342.	247	24299	10.09819	1.460938	.0069601
343.	497	19700	9.888374	1.766442	.0070464
344.	355	18200	9.809176	1.418277	.0070865
345.	320	21000	9.952278	1.693779	.0070909

346.	307	33400	10.41631	1.551809	.007096
347.	479	14600	9.588777	1.814825	.0070969
348.	36	18900	9.846917	1.607436	.0071105
349.	261	33799	10.42819	1.867176	.0072586
350.	116	18300	9.814656	1.699279	.0072618

351.	20	18200	9.809176	1.682688	.0075003
352.	285	32199	10.37969	1.386294	.007564
353.	318	19800	9.893437	1.693779	.0076158
354.	329	19300	9.867861	1.526056	.0076533
355.	64	24999	10.12659	1.510722	.0076632

356.	46	19300	9.867861	1.499623	.0077691
357.	415	7012	8.855378	1.93586	.007824
358.	419	8800	9.082507	1.915451	.007874
359.	458	13500	9.510445	1.964311	.0080279
360.	205	50001	10.8198	1.425515	.0081019

361.	447	14900	9.609117	2.00148	.0081669
362.	298	20300	9.918376	1.474763	.0085907
363.	76	21400	9.971147	1.474763	.0091335
364.	82	23899	10.08159	1.449269	.0092037
365.	287	20100	9.908475	1.348073	.0093339

366.	270	20700	9.937889	1.534714	.0094668
367.	391	15100	9.62245	1.94591	.0096289
368.	387	10500	9.25913	1.94591	.0096712
369.	332	17100	9.746834	1.477049	.009686
370.	461	16400	9.705036	1.964311	.0097336

371.	103	18600	9.830916	1.648659	.0097767
372.	7	22900	10.03889	1.656321	.0099829
373.	255	21900	9.994242	1.366092	.0101005
374.	239	23699	10.07319	1.453953	.0102395
375.	132	19600	9.883285	1.83098	.0102551

376.	288	23200	10.05191	1.398717	.0103194
377.	284	50001	10.8198	1.388791	.0103747
378.	487	19100	9.857444	1.763017	.0104986
379.	311	16100	9.686575	1.693779	.0105595

380.	325	24999	10.12659	1.595339	.0105819
381.	433	16100	9.686575	1.764731	.0107606
382.	256	20900	9.947504	1.366092	.0108386
383.	290	24800	10.1186	1.398717	.0114381
384.	174	23600	10.069	1.629241	.0116044
385.	496	23100	10.04759	1.766442	.0117644
386.	65	33001	10.40429	1.425515	.0121224
387.	78	20800	9.942708	1.474763	.0121406
388.	265	36501	10.50509	1.867176	.0122671
389.	421	16700	9.723164	1.971299	.0123169
390.	218	28701	10.26469	1.704748	.012537
391.	54	23400	10.06049	1.479329	.0129076
392.	55	18900	9.846917	1.410987	.0129513
393.	221	26699	10.19238	1.623341	.0129747
394.	327	23001	10.04329	1.595339	.0132197
395.	100	33200	10.41031	1.492904	.01351
396.	58	32562	10.3909	1.413423	.0140252
397.	202	24101	10.09001	1.423108	.0143737
398.	240	23300	10.05621	1.453953	.0146344
399.	93	22900	10.03889	1.534714	.0146731
400.	88	22199	10.0078	1.501853	.0147033
401.	328	22199	10.0078	1.595339	.0147417
402.	295	21700	9.985067	1.474763	.0149248
403.	91	22600	10.02571	1.587192	.0149329
404.	418	10400	9.249561	1.915451	.0149713
405.	459	14900	9.609117	1.964311	.0150557
406.	22	19600	9.883285	1.682688	.0152086
407.	398	8500	9.047821	1.93586	.0153095
408.	269	43499	10.68049	1.7492	.0162008
409.	478	12000	9.392662	1.814825	.0163773
410.	300	29001	10.27509	1.386294	.0168745
411.	139	13300	9.49552	1.83098	.0171233
412.	89	23600	10.069	1.587192	.0173629
413.	242	20100	9.908475	1.453953	.0176759
414.	289	22299	10.0123	1.398717	.0176951
415.	356	20600	9.933046	1.418277	.0178045
416.	253	29599	10.2955	1.460938	.0179755
417.	6	28701	10.26469	1.521699	.0180267
418.	405	8500	9.047821	1.93586	.0180952
419.	199	34600	10.45161	1.396245	.0184836
420.	92	22000	9.998797	1.587192	.0186358
421.	326	24600	10.1105	1.595339	.0192508
422.	134	18400	9.820106	1.83098	.0196824
423.	241	22000	9.998797	1.453953	.0201607
424.	153	15300	9.635608	2.164472	.0202187
425.	4	33400	10.41631	1.521699	.0206138
426.	424	13400	9.50301	1.814825	.0206475

427.	141	14000	9.546813	1.83098	.0206522
428.	59	23300	10.05621	1.510722	.0209189
429.	385	8800	9.082507	1.94591	.0216851
430.	485	20600	9.933046	1.763017	.0229157

431.	492	13600	9.517825	1.806648	.0231727
432.	115	18500	9.825526	1.699279	.0234778
433.	19	20200	9.913438	1.682688	.0239359
434.	336	21100	9.957028	1.638997	.0240607
435.	66	23501	10.0648	1.381282	.0240661

436.	412	17200	9.752665	1.786747	.0244541
437.	133	23001	10.04329	1.83098	.0251679
438.	453	16100	9.686575	1.964311	.0252416
439.	197	33300	10.41331	1.396245	.0256373
440.	334	22199	10.0078	1.638997	.0258975

441.	388	7400	8.909235	1.94591	.0263166
442.	69	17400	9.764226	1.408545	.0263849
443.	5	36199	10.49679	1.521699	.0266658
444.	95	20600	9.933046	1.534714	.0274035
445.	407	11900	9.384294	1.885553	.0274155

446.	402	7200	8.881836	1.93586	.0280494
447.	299	22501	10.02131	1.386294	.0307287
448.	98	38700	10.56359	1.492904	.0308795
449.	16	19900	9.898475	1.682688	.0313705
450.	345	31201	10.34821	1.576915	.0318101

451.	262	43101	10.6713	1.867176	.032273
452.	379	13100	9.480368	1.903599	.0328231
453.	198	30300	10.3189	1.396245	.0330362
454.	486	21200	9.961757	1.763017	.0337355
455.	302	22000	9.998797	1.465567	.0341089

456.	431	14500	9.581903	1.764731	.0342425
457.	301	24800	10.1186	1.386294	.0343599
458.	77	20000	9.903487	1.474763	.0356156
459.	97	21400	9.971147	1.492904	.0357894
460.	476	13300	9.49552	1.764731	.0361187

461.	442	17100	9.746834	2.00148	.0364184
462.	14	20400	9.92329	1.682688	.0369498
463.	377	13900	9.539644	1.903599	.0369729
464.	401	5600	8.630522	1.93586	.0383786
465.	367	21900	9.994242	1.971299	.0430825

466.	8	27100	10.20729	1.656321	.043922
467.	382	10900	9.296518	1.903599	.0451832
468.	378	13300	9.49552	1.903599	.0464509
469.	467	19000	9.852194	1.879465	.0484201
470.	386	7200	8.881836	1.94591	.048621

471.	376	15000	9.615806	1.903599	.0488837
472.	67	19400	9.873029	1.381282	.0490895
473.	425	11700	9.367344	1.764731	.0514093
474.	462	17700	9.78132	1.964311	.0535513

475.	443	18400	9.820106	2.00148	.0566888
476.	286	22000	9.998797	1.358409	.0573509
477.	17	23100	10.04759	1.682688	.0589057
478.	400	6300	8.748305	1.93586	.0589789
479.	432	14100	9.55393	1.764731	.0600489
480.	465	21400	9.971147	1.879465	.0677176
481.	380	10200	9.230143	1.903599	.0691514
482.	364	16800	9.729135	2.04122	.0709101
483.	392	23200	10.05191	1.94591	.0715526
484.	381	10400	9.249561	1.903599	.0725951
485.	149	17794	9.786616	2.164472	.0743867
486.	466	19900	9.898475	1.879465	.0744588
487.	399	5000	8.517193	1.93586	.0797595
488.	366	27499	10.2219	1.971299	.0797842
489.	357	17794	9.786616	2.04122	.0857706
490.	154	19400	9.873029	2.164472	.0910494
491.	463	19500	9.87817	1.964311	.0941472
492.	464	20200	9.913438	1.964311	.0974507
493.	427	10200	9.230143	1.764731	.1007114
494.	406	5000	8.517193	1.93586	.1024767
495.	151	21500	9.975808	2.164472	.1047597
496.	152	19600	9.883285	2.164472	.1120427
497.	460	20000	9.903487	1.964311	.1142668
498.	160	23300	10.05621	2.164472	.1165014
499.	491	8100	8.999619	1.806648	.1222368
500.	362	19900	9.898475	2.04122	.1376445
501.	363	20800	9.942708	2.04122	.1707894
502.	490	7000	8.853665	1.806648	.1791869
503.	358	21700	9.985067	2.04122	.1827834
504.	360	22600	10.02571	2.04122	.2209745
505.	361	24999	10.12659	2.04122	.2422512
506.	359	22700	10.03012	2.04122	.2483543

5.3 ENDOGENEITY & MEASUREMENT ERROR

REGRESSION WITH NON-IID ERRORS

6.1 THE GENERALIZED LINEAR REGRESSION MODEL

6.1.1 *Types of Deviations from i.i.d. Errors*

6.1.2 *The Robust Estimator of VCE*

```
. use fertil2.. describe
```

Contains data from fertil2.dta

obs: 4,361

vars: 30

2 Dec 2004 00:16

size: 484,071

variable name	storage type	display format	value label	variable label
mnthborn	float	%9.0g		
yearborn	float	%9.0g		
age	float	%9.0g		
electric	float	%9.0g		
radio	float	%9.0g		
tv	float	%9.0g		
bicycle	float	%9.0g		
educ	float	%9.0g		
ceb	float	%9.0g		
agefbrth	float	%9.0g		
children	float	%9.0g		
knowmeth	float	%9.0g		
usemeth	float	%9.0g		
monthfm	float	%9.0g		
yearfm	float	%9.0g		
agefm	float	%9.0g		
idlnchld	float	%9.0g		
heduc	float	%9.0g		
agesq	float	%9.0g		
urban	float	%9.0g		
urbeduc	float	%9.0g		
spirit	float	%9.0g		
protest	float	%9.0g		
catholic	float	%9.0g		
frsthalf	float	%9.0g		
educ0	float	%9.0g		
evermarr	float	%9.0g		
_est_OLS	byte	%8.0g		esample() from estimates store
_est_robust	byte	%8.0g		esample() from estimates store
_est_cluster	byte	%8.0g		esample() from estimates store

Sorted by:

. summarize

Variable	Obs	Mean	Std. Dev.	Min	Max
-----+					
mnthborn	4,361	6.331346	3.323333	1	12
yearborn	4,361	60.43362	8.682723	38	73
age	4,361	27.40518	8.685233	15	49
electric	4,358	.1402019	.3472363	0	1
radio	4,359	.7017665	.457535	0	1
-----+					
tv	4,359	.0929112	.2903413	0	1
bicycle	4,358	.2758146	.4469751	0	1
educ	4,361	5.855996	3.927075	0	20
ceb	4,361	2.441642	2.406861	0	13
agefbrth	3,273	19.0113	3.092333	10	38
-----+					
children	4,361	2.267828	2.222032	0	13
knowmeth	4,354	.9632522	.1881636	0	1
usemeth	4,290	.5776224	.4939956	0	1
monthfm	2,079	6.270322	3.619943	1	12
yearfm	2,079	76.91246	7.760183	50	88
-----+					
agefm	2,079	20.68639	5.002383	10	46
idlnchld	4,241	4.615892	2.219303	0	20
heduc	1,956	5.144683	4.803028	0	20
agesq	4,361	826.46	526.9232	225	2401
urban	4,361	.5166246	.4997808	0	1
-----+					
urbeduc	4,361	3.469158	4.294228	0	20
spirit	4,361	.4221509	.493959	0	1
protest	4,361	.2277001	.4193961	0	1
catholic	4,361	.1024994	.3033387	0	1
frsthalf	4,361	.5404724	.4984164	0	1
-----+					
educ0	4,361	.2077505	.4057437	0	1
evermarr	4,361	.4767255	.4995153	0	1
_est_OLS	4,361	.7367576	.4404433	0	1
_est_robust	4,361	.7367576	.4404433	0	1
_est_cluster	4,361	.7367576	.4404433	0	1

. use fertil2.. regress ceb age agefbrth usemeth

Source	SS	df	MS	Number of obs	=	3,213
-----+				F(3, 3209)	=	1433.16
Model	9202.53439	3	3067.51146	Prob > F	=	0.0000
Residual	6868.49331	3,209	2.14038433	R-squared	=	0.5726
-----+				Adj R-squared	=	0.5722
Total	16071.0277	3,212	5.00343328	Root MSE	=	1.463
-----+						
ceb	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+						
age	.2237368	.003448	64.89	0.000	.2169763	.2304974
agefbrth	-.2606634	.0087954	-29.64	0.000	-.2779085	-.2434184

```

      usemeth |   .1873702   .0554298   3.38   0.001   .0786888   .2960516
      _cons   |   1.358134   .1737828   7.82   0.000   1.017397   1.69887
-----

```

```

. estimates store nonRobust

```

```

. summarize ceb age agefbrth usemeth children if e(sample)

```

```

      Variable |      Obs      Mean   Std. Dev.      Min      Max
-----+-----
      ceb |      3,213   3.230003   2.236836         1       13
      age |      3,213   29.93931   7.920432        15       49
  agefbrth |      3,213   19.00498   3.098121        10       38
      usemeth |      3,213   .6791161   .4668889         0         1
  children |      3,213   2.999378   2.055579         0       13

```

```

. regress ceb age agefbrth usemeth, robust

```

```

Linear regression                                Number of obs   =      3,213
                                                F(3, 3209)          =      874.06
                                                Prob > F             =      0.0000
                                                R-squared            =      0.5726
                                                Root MSE            =      1.463

```

```

-----
      |
      ceb |      Coef.   Robust Std. Err.      t    P>|t|    [95% Conf. Interval]
-----+-----
      age |   .2237368   .0046619   47.99   0.000   .2145962   .2328775
  agefbrth |  -.2606634   .0095616  -27.26   0.000  -.2794109  -.2419159
      usemeth |   .1873702   .0606446    3.09   0.002   .0684642   .3062762
      _cons |   1.358134   .1675624    8.11   0.000   1.029593   1.686674
-----

```

```

. estimates store Robust

```

```

. estimates table nonRobust Robust, b(%9.4f) se(%5.3f) t(%5.2f) p(%4.3f) title(
> Estimates of CEB with OLS and Robust Standard Errors)

```

Estimates of CEB with OLS and Robust Standard Errors

```

-----
      Variable | nonRobust   Robust
-----+-----
      age |      0.2237      0.2237
      |      0.003      0.005
      |      64.89      47.99
      |      0.000      0.000
  agefbrth |     -0.2607     -0.2607
      |      0.009      0.010
      |     -29.64     -27.26
      |      0.000      0.000
      usemeth |     0.1874     0.1874
      |      0.055      0.061
      |      3.38      3.09
      |      0.001      0.002
      _cons |     1.3581     1.3581

```

	0.174	0.168
	7.82	8.11
	0.000	0.000

 legend: b/se/t/p

6.1.3 The Cluster Estimator of VCE

```
. use fertil2.. regress ceb age agefbrth usemeth, cluster(children)
```

Linear regression	Number of obs	=	3,213
	F(3, 13)	=	20.91
	Prob > F	=	0.0000
	R-squared	=	0.5726
	Root MSE	=	1.463

(Std. Err. adjusted for 14 clusters in children)

		Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
ceb							
age		.2237368	.0315086	7.10	0.000	.1556665	.2918071
agefbrth		-.2606634	.0354296	-7.36	0.000	-.3372045	-.1841224
usemeth		.1873702	.0943553	1.99	0.069	-.016472	.3912125
_cons		1.358134	.4248589	3.20	0.007	.4402818	2.275985

6.1.4 The Newey–West Estimator of VCE

```
. use ukrates.. describe
```

Contains data from ukrates.dta

obs:	526	
vars:	3	2 Dec 2004 10:43
size:	6,312	

variable name	storage type	display format	value label	variable label
rs	float	%9.0g		
month	float	%tm		
r20	float	%9.0g		

Sorted by: month

```
. summarize
```

Variable	Obs	Mean	Std. Dev.	Min	Max
rs	526	7.651513	3.553109	1.561667	16.18
month	526	168.5	151.9874	-94	431

```

r20 |      526      8.863726      3.224372      3.35      17.18

. quietly regress D.rs LD.r20

. estimates store nonHAC

. newey D.rs LD.r20, lag(5)

Regression with Newey-West standard errors      Number of obs      =      524
maximum lag: 5                                F( 1,      522) =      36.00
                                              Prob > F          =      0.0000

```

		Newey-West				
D.rs		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]

r20						
LD.		.4882883	.0813867	6.00	0.000	.3284026 .648174
_cons		.0040183	.0254102	0.16	0.874	-.0459004 .0539371

```

. estimates store NeweyWest

. estimates table nonHAC NeweyWest, b(%9.4f) se(%5.3f) t(%5.2f) p(%4.3f) title(
> Estimates of D.rs with OLS and Newey-West Standard Errors)

```

Estimates of D.rs with OLS and Newey-West Standard Errors

Variable	nonHAC	NeweyWest

r20		
LD.		0.4883
		0.067
		7.27
		0.000
_cons		0.0040
		0.022
		0.18
		0.858

legend: b/se/t/p

6.1.5 The Generalized Least Squares Estimator

6.2 HETEROSKEDASTICITY IN THE ERROR DISTRIBUTION

6.2.1 Heteroskedasticity Related to Scale

6.2.1.1 Testing for Heteroskedasticity Related to Scale

```
. use hprice2a.(Housing price data for Boston-area communities)
```

```
. regress lprice rooms crime ldist
```

Source	SS	df	MS	Number of obs	=	506
Model	47.9496883	3	15.9832294	F(3, 502)	=	219.03
Residual	36.6325827	502	.072973272	Prob > F	=	0.0000
				R-squared	=	0.5669
				Adj R-squared	=	0.5643
Total	84.5822709	505	.167489645	Root MSE	=	.27014

lprice	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
rooms	.3072343	.0178231	17.24	0.000	.2722172	.3422514
crime	-.0174486	.001591	-10.97	0.000	-.0205744	-.0143228
ldist	.074858	.0255746	2.93	0.004	.0246115	.1251045
_cons	7.984449	.1128067	70.78	0.000	7.762817	8.20608

```
. estat hettest, iid
```

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

Ho: Constant variance

Variables: fitted values of lprice

chi2(1) = 44.67

Prob > chi2 = 0.0000

```
. estat hettest rooms crime ldist, iid
```

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

Ho: Constant variance

Variables: rooms crime ldist

chi2(3) = 80.11

Prob > chi2 = 0.0000

```
. whitetst
```

command whitetst is unrecognized

```
r(199);
```

end of do-file

```
r(199);
```

6.2.1.2 Feasible Generalized Least Squares Estimation

```
. use hprice2a.(Housing price data for Boston-area communities)
```

```
. generate rooms2 = rooms^2
```

```
. regress lprice rooms crime ldist [aw=1/rooms2]
```

(sum of wgt is 13.31716591697057)

Source	SS	df	MS	Number of obs	=	506
--------	----	----	----	---------------	---	-----

-----+-----					F(3, 502)	=	159.98
Model		39.6051883	3	13.2017294	Prob > F	=	0.0000
Residual		41.426616	502	.082523139	R-squared	=	0.4888
-----+-----					Adj R-squared	=	0.4857
Total		81.0318042	505	.160459018	Root MSE	=	.28727

lprice		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+-----							
rooms		.2345368	.0194432	12.06	0.000	.1963367	.272737
crime		-.0175759	.0016248	-10.82	0.000	-.0207682	-.0143837
ldist		.0650916	.027514	2.37	0.018	.0110349	.1191483
_cons		8.450081	.1172977	72.04	0.000	8.219626	8.680536

6.2.2 Heteroskedasticity Between Groups of Observations

6.2.2.1 Testing for Heteroskedasticity Between Groups of Observations

```
. use NEdata.. describe
```

Contains data from NEdata.dta

```
obs:      120
vars:      6                      24 Oct 2004 13:28
size:     2,640
```

variable name	storage type	display format	value label	variable label

state	long	%8.0g	state	
year	int	%8.0g		
pop	float	%9.0g		
dpi	float	%9.0g		
dpipc	float	%9.0g		
ldpipc	float	%9.0g		

Sorted by:

```
. summarize
```

Variable	Obs	Mean	Std. Dev.	Min	Max
-----+-----					
state	120	3.5	1.714986	1	6
year	120	1990.5	5.790459	1981	2000
pop	120	2196276	1931629	515594	6362076
dpi	120	4.33e+07	4.46e+07	4385134	1.93e+08
dpipc	120	18.15802	5.662848	8.153382	33.38758
-----+-----					
ldpipc	120	2.848302	.3265395	2.098433	3.508184

```
. regress dpipc year
```

Source		SS	df	MS	Number of obs	=	120
-----+-----					F(1, 118)	=	440.17

Model		3009.33617	1	3009.33617	Prob > F	=	0.0000
Residual		806.737449	118	6.83675804	R-squared	=	0.7886
-----+					Adj R-squared	=	0.7868
Total		3816.07362	119	32.0678456	Root MSE	=	2.6147

dpipc		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+						
year		.8684582	.0413941	20.98	0.000	.7864865 .9504298
_cons		-1710.508	82.39534	-20.76	0.000	-1873.673 -1547.343

```
. predict double eps, residual
```

```
. robvar eps, by(state)
```

		Summary of Residuals		
state		Mean	Std. Dev.	Freq.
-----+				
CT		4.167853	1.3596266	20
MA		1.618796	.86550138	20
ME		-2.9841056	.93797625	20
NH		.51033312	.61139299	20
RI		-.8927223	.63408722	20
VT		-2.4201543	.71470977	20
-----+				
Total		-6.063e-14	2.6037101	120

```
W0 = 4.3882072 df(5, 114) Pr > F = 0.00108562
```

```
W50 = 3.2989851 df(5, 114) Pr > F = 0.00806751
```

```
W10 = 4.2536245 df(5, 114) Pr > F = 0.00139064
```

6.2.2.2 Feasible Generalized Least Squares Estimation

```
. use NEdata.. regress dpipc year
```

Source		SS	df	MS	Number of obs	=	120
-----+					F(1, 118)	=	440.17
Model		3009.33617	1	3009.33617	Prob > F	=	0.0000
Residual		806.737449	118	6.83675804	R-squared	=	0.7886
-----+					Adj R-squared	=	0.7868
Total		3816.07362	119	32.0678456	Root MSE	=	2.6147

dpipc		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+						
year		.8684582	.0413941	20.98	0.000	.7864865 .9504298
_cons		-1710.508	82.39534	-20.76	0.000	-1873.673 -1547.343

```
. predict double eps, residual
```

```
. by state, sort: egen sd_eps = sd(eps)
```

```
. generate double gw_wt = 1/sd_eps^2
```

```
. tabstat sd_eps gw_wt, by(state)
```

Summary statistics: mean
by categories of: state

state	sd_eps	gw_wt
-----+-----		
CT	1.359627	.5409545
MA	.8655014	1.334948
ME	.9379762	1.136623
NH	.611393	2.675218
RI	.6340872	2.48715
VT	.7147098	1.957675
-----+-----		
Total	.8538824	1.688761

```
. regress dpipc year [aw = gw_wt]
(sum of wgt is 202.6513649171444)
```

Source	SS	df	MS	Number of obs	=	120
-----+-----						
Model	2845.55409	1	2845.55409	F(1, 118)	=	698.19
Residual	480.921278	118	4.07560405	Prob > F	=	0.0000
-----+-----						
Total	3326.47537	119	27.9535745	R-squared	=	0.8554
				Adj R-squared	=	0.8542
				Root MSE	=	2.0188

dpipc	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+-----						
year	.8444948	.0319602	26.42	0.000	.7812049	.9077847
_cons	-1663.26	63.61705	-26.14	0.000	-1789.239	-1537.281

6.2.3 Heteroskedasticity in Grouped Data

6.2.3.1 Feasible Generalized Least Squares Estimation

```
. use pubschl.. describe
```

Contains data from pubschl.dta

obs: 420

vars: 18

2 Dec 2004 12:36

size: 58,380

variable name	storage type	display format	value label	variable label

observation_n~r	float	%9.0g		
dist_cod	float	%9.0g		
county	str18	%18s		

```

district      str53  %53s
gr_span       str8   %8s
enrl_tot      float  %9.0g
teachers       float  %9.0g
calw_pct      float  %9.0g
meal_pct      float  %9.0g
computer      float  %9.0g
testscr       float  %9.0g
comp_stu      float  %9.0g
expn_stu      float  %9.0g
str           float  %9.0g
avginc       float  %9.0g
el_pct       float  %9.0g
read_scr     float  %9.0g
math_scr     float  %9.0g

```

Sorted by:

```
. summarize
```

Variable	Obs	Mean	Std. Dev.	Min	Max
observatio~r	420	210.5	121.3878	1	420
dist_cod	420	67472.81	3466.995	61382	75440
county	0				
district	0				
gr_span	0				
enrl_tot	420	2628.793	3913.105	81	27176
teachers	420	129.0674	187.9127	4.85	1429
calw_pct	420	13.24604	11.45482	0	78.9942
meal_pct	420	44.70524	27.12338	0	100
computer	420	303.3833	441.3413	0	3324
testscr	420	654.1565	19.05335	605.55	706.75
comp_stu	420	.1359266	.0649558	0	.4208333
expn_stu	420	5312.408	633.9371	3926.07	7711.507
str	420	19.64043	1.891812	14	25.8
avginc	420	15.31659	7.22589	5.335	55.328
el_pct	420	15.76816	18.28593	0	85.53972
read_scr	420	654.9705	20.10798	604.5	704
math_scr	420	653.3426	18.7542	605.4	709.5

```
. regress read_scr expn_stu comp_stu meal_pct
```

Source	SS	df	MS	Number of obs	=	420
Model	136046.267	3	45348.7558	F(3, 416)	=	565.36
Residual	33368.3632	416	80.2124115	Prob > F	=	0.0000
				R-squared	=	0.8030
				Adj R-squared	=	0.8016
Total	169414.631	419	404.330861	Root MSE	=	8.9561

read_scr	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
expn_stu	.0046699	.0007204	6.48	0.000	.0032538 .006086

```

      comp_stu |   19.88584   7.168347    2.77   0.006   5.795143   33.97654
      meal_pct |   -.635131   .0164777  -38.54   0.000   -.667521   -.602741
      _cons |   655.8528   3.812206  172.04   0.000   648.3592   663.3464
-----+-----
. regress read_scr expn_stu comp_stu meal_pct [aw = enr_l_tot]
(sum of wgt is 1,104,093)

      Source |           SS           df           MS       Number of obs   =         420
-----+-----+-----+-----+-----+-----+-----
      Model |   123692.671            3   41230.8903       Prob > F           =         0.0000
      Residual |   18915.9815          416   45.4711093       R-squared            =         0.8674
-----+-----+-----+-----+-----+-----
      Total |   142608.652          419   340.354779       Adj R-squared       =         0.8664
                                         Root MSE           =         6.7432
-----+-----

      read_scr |           Coef.      Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----+-----+-----+-----+-----
      expn_stu |    .0055534    .0008322     6.67   0.000    .0039176    .0071892
      comp_stu |   27.26378    8.197228     3.33   0.001   11.15063   43.37693
      meal_pct |  -.6352229    .013149   -48.31   0.000   -.6610696   -.6093762
      _cons |   648.988    4.163875   155.86   0.000   640.8031   657.1728
-----+-----

```

6.3 SERIAL CORRELATION IN THE ERROR DISTRIBUTION

6.3.1 *Testing for Serial Correlation*

```
. use ukrates.. describe
```

Contains data from ukrates.dta

```

obs:           526
vars:           3                      2 Dec 2004 10:43
size:          6,312

```

```

-----+-----
      storage  display  value
variable name  type    format  label      variable label
-----+-----+-----+-----+-----
rs             float    %9.0g
month          float    %tm
r20            float    %9.0g
-----+-----

```

Sorted by: month

```
. summarize
```

```

      Variable |           Obs           Mean        Std. Dev.        Min        Max
-----+-----+-----+-----+-----+-----
           rs |           526       7.651513       3.553109       1.561667       16.18
        month |           526        168.5       151.9874         -94         431
          r20 |           526       8.863726       3.224372         3.35       17.18

```

```
. regress D.rs LD.r20
```

Source	SS	df	MS	Number of obs	=	524
				F(1, 522)	=	52.88
Model	13.8769739	1	13.8769739	Prob > F	=	0.0000
Residual	136.988471	522	.262430021	R-squared	=	0.0920
				Adj R-squared	=	0.0902
Total	150.865445	523	.288461654	Root MSE	=	.51228

D.rs	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
r20						
LD.	.4882883	.0671484	7.27	0.000	.356374	.6202027
_cons	.0040183	.022384	0.18	0.858	-.0399555	.0479921

. predict double eps, residual
(2 missing values generated)

. estat bgodfrey, lags(6)

Breusch-Godfrey LM test for autocorrelation

lags(p)	chi2	df	Prob > chi2
6	17.237	6	0.0084

H0: no serial correlation

. wntestq eps

Portmanteau test for white noise

Portmanteau (Q) statistic = 82.3882
Prob > chi2(40) = 0.0001

. ac eps

6.3.2 FGLS Estimation with Serial Correlation

. use ukrates.. regress D.rs LD.r20

Source	SS	df	MS	Number of obs	=	524
				F(1, 522)	=	52.88
Model	13.8769739	1	13.8769739	Prob > F	=	0.0000
Residual	136.988471	522	.262430021	R-squared	=	0.0920
				Adj R-squared	=	0.0902
Total	150.865445	523	.288461654	Root MSE	=	.51228

D.rs	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
r20						
LD.	.4882883	.0671484	7.27	0.000	.356374	.6202027

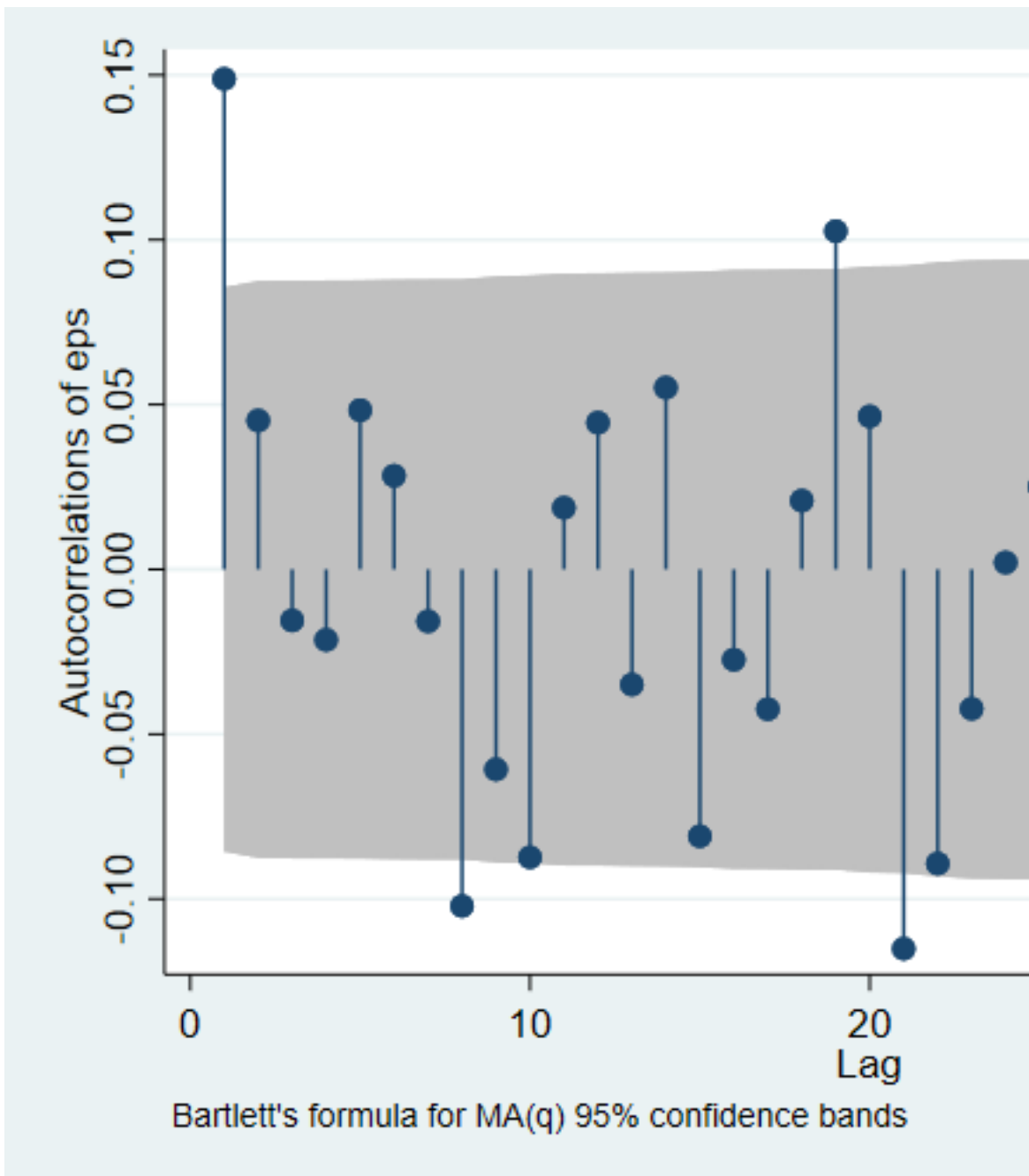


Figure 6: Autocorrelogram of regression residuals

_cons		.0040183	.022384	0.18	0.858	-.0399555	.0479921

. prais D.rs LD.r20, nolog

Prais-Winsten AR(1) regression -- iterated estimates

Source		SS	df	MS	Number of obs	=	524
					F(1, 522)	=	25.73
Model		6.56420242	1	6.56420242	Prob > F	=	0.0000
Residual		133.146932	522	.25507075	R-squared	=	0.0470
					Adj R-squared	=	0.0452
Total		139.711134	523	.2671341	Root MSE	=	.50505

D.rs		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
<hr/>							
r20							
LD.		.3495857	.068912	5.07	0.000	.2142067	.4849647
<hr/>							
_cons		.0049985	.0272145	0.18	0.854	-.0484649	.0584619
<hr/>							
rho		.1895324					

Durbin-Watson statistic (original) 1.702273

Durbin-Watson statistic (transformed) 2.007414

REGRESSION WITH INDICATOR VARIABLES

7.1 TESTING FOR SIGNIFICANCE OF A QUALITATIVE FACTOR

7.1.1 *Regression with One Qualitative Measure*

```
. use NEdata.. describe
```

Contains data from NEdata.dta

```
obs:      120
vars:      6                      24 Oct 2004 13:28
size:     2,640
```

```
-----+-----
      storage   display   value
variable name  type    format   label   variable label
-----+-----
state          long    %8.0g     state
year           int     %8.0g
pop            float   %9.0g
dpi            float   %9.0g
dpipc          float   %9.0g
ldpipc         float   %9.0g
-----+-----
```

Sorted by:

```
. summarize
```

```
-----+-----
Variable |      Obs      Mean   Std. Dev.   Min      Max
-----+-----
state |      120        3.5   1.714986        1         6
year |      120     1990.5   5.790459     1981     2000
pop |      120    2196276  1931629    515594    6362076
dpi |      120    4.33e+07  4.46e+07   4385134    1.93e+08
dpipc |      120    18.15802   5.662848   8.153382    33.38758
-----+-----
ldpipc |      120    2.848302   .3265395   2.098433    3.508184
-----+-----
```

```
. mean dpipc, over(state)
```

```
Mean estimation      Number of obs   =      120
```

```
CT: state = CT
MA: state = MA
ME: state = ME
NH: state = NH
RI: state = RI
VT: state = VT
```

	Over	Mean	Std. Err.	[95% Conf. Interval]	
-----+-----					
dpipc					
	CT	22.32587	1.413766	19.52647	25.12527
	MA	19.77681	1.298507	17.20564	22.34798
	ME	15.17391	.9571251	13.27871	17.06911
	NH	18.66835	1.193137	16.30582	21.03088
	RI	17.26529	1.045117	15.19586	19.33473
	VT	15.73786	1.020159	13.71784	17.75788

. tabulate state, generate(NE)

state	Freq.	Percent	Cum.
-----+-----			
CT	20	16.67	16.67
MA	20	16.67	33.33
ME	20	16.67	50.00
NH	20	16.67	66.67
RI	20	16.67	83.33
VT	20	16.67	100.00
-----+-----			
Total	120	100.00	

. regress dpipc NE2-NE6

Source	SS	df	MS	Number of obs	=	120
-----+-----						
Model	716.218512	5	143.243702	F(5, 114)	=	5.27
Residual	3099.85511	114	27.1917115	Prob > F	=	0.0002
-----+-----						
				R-squared	=	0.1877
				Adj R-squared	=	0.1521
Total	3816.07362	119	32.0678456	Root MSE	=	5.2146

dpipc	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+-----						
NE2	-2.549057	1.648991	-1.55	0.125	-5.815695	.7175814
NE3	-7.151959	1.648991	-4.34	0.000	-10.4186	-3.88532
NE4	-3.65752	1.648991	-2.22	0.029	-6.924158	-.3908815
NE5	-5.060575	1.648991	-3.07	0.003	-8.327214	-1.793937
NE6	-6.588007	1.648991	-4.00	0.000	-9.854646	-3.321369
_cons	22.32587	1.166013	19.15	0.000	20.01601	24.63573

. use NEdata.. tabulate state, generate(NE)

state	Freq.	Percent	Cum.
-----+-----			
CT	20	16.67	16.67
MA	20	16.67	33.33
ME	20	16.67	50.00
NH	20	16.67	66.67
RI	20	16.67	83.33
VT	20	16.67	100.00
-----+-----			
Total	120	100.00	

```
. forvalues i = 1/5 {
  2. generate NE_`i' = NE_`i' - NE6
  3. }

. regress dpipc NE_*
```

Source		SS	df	MS	Number of obs	=	120
-----+					F(5, 114)	=	5.27
Model		716.218512	5	143.243702	Prob > F	=	0.0002
Residual		3099.85511	114	27.1917115	R-squared	=	0.1877
-----+					Adj R-squared	=	0.1521
Total		3816.07362	119	32.0678456	Root MSE	=	5.2146

dpipc		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+							
NE_1		4.167853	1.064419	3.92	0.000	2.059247	6.276459
NE_2		1.618796	1.064419	1.52	0.131	-.48981	3.727402
NE_3		-2.984106	1.064419	-2.80	0.006	-5.092712	-.8754996
NE_4		.5103331	1.064419	0.48	0.633	-1.598273	2.618939
NE_5		-.8927223	1.064419	-0.84	0.403	-3.001328	1.215884
_cons		18.15802	.4760227	38.15	0.000	17.21502	19.10101

```
.
. lincom -(NE_1 + NE_2 + NE_3 + NE_4 + NE_5)

( 1) - NE_1 - NE_2 - NE_3 - NE_4 - NE_5 = 0
```

dpipc		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+							
(1)		-2.420154	1.064419	-2.27	0.025	-4.52876	-.3115483

7.1.2 Regression with Two Qualitative Measures

```
. use nlsw88.(NLSW, 1988 extract)
```

```
. describe
```

Contains data from nlsw88.dta

obs:	2,246	NLSW, 1988 extract
vars:	17	21 Jun 2006 11:33
size:	60,642	(_dta has notes)

variable name	storage type	display format	value label	variable label

idcode	int	%8.0g		NLS id
age	byte	%8.0g		age in current year
race	byte	%8.0g	racelbl	race
married	byte	%8.0g	marlbl	married

never_married	byte	%8.0g		never married
grade	byte	%8.0g		current grade completed
collgrad	byte	%16.0g	gradlbl	college graduate
south	byte	%8.0g		lives in south
smsa	byte	%9.0g	smsalbl	lives in SMSA
c_city	byte	%8.0g		lives in central city
industry	byte	%23.0g	indlbl	industry
occupation	byte	%22.0g	occlbl	occupation
union	byte	%8.0g	unionlbl	union worker
wage	float	%9.0g		hourly wage
hours	byte	%8.0g		usual hours worked
ttl_exp	float	%9.0g		total work experience
tenure	float	%9.0g		job tenure (years)

Sorted by: idcode

. summarize

Variable	Obs	Mean	Std. Dev.	Min	Max
idcode	2,246	2612.654	1480.864	1	5159
age	2,246	39.15316	3.060002	34	46
race	2,246	1.282725	.4754413	1	3
married	2,246	.6420303	.4795099	0	1
never_married	2,246	.1041852	.3055687	0	1
grade	2,244	13.09893	2.521246	0	18
collgrad	2,246	.2368655	.4252538	0	1
south	2,246	.4194123	.4935728	0	1
smsa	2,246	.7039181	.4566292	0	1
c_city	2,246	.2916296	.4546139	0	1
industry	2,232	8.189516	3.010875	1	12
occupation	2,237	4.642825	3.408897	1	13
union	1,878	.2454739	.4304825	0	1
wage	2,246	7.766949	5.755523	1.004952	40.74659
hours	2,242	37.21811	10.50914	1	80
ttl_exp	2,246	12.53498	4.610208	.1153846	28.88461
tenure	2,231	5.97785	5.510331	0	25.91667

. keep if !missing(wage + race + union)
(368 observations deleted)

. generate lwage = log(wage)

. tabulate race, generate(R)

race	Freq.	Percent	Cum.
white	1,353	72.04	72.04
black	501	26.68	98.72
other	24	1.28	100.00
Total	1,878	100.00	

. regress lwage R1 R2 union

Source	SS	df	MS	Number of obs	=	1,878
				F(3, 1874)	=	38.73
Model	29.3349228	3	9.77830761	Prob > F	=	0.0000
Residual	473.119209	1,874	.252464893	R-squared	=	0.0584
				Adj R-squared	=	0.0569
Total	502.454132	1,877	.267690001	Root MSE	=	.50246

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
R1	-.0349326	.1035125	-0.34	0.736	-.2379444	.1680793
R2	-.2133924	.1049954	-2.03	0.042	-.4193126	-.0074721
union	.239083	.0270353	8.84	0.000	.1860606	.2921054
_cons	1.913178	.1029591	18.58	0.000	1.711252	2.115105

```
. test R1 R2
```

```
( 1) R1 = 0
```

```
( 2) R2 = 0
```

```

F( 2, 1874) = 23.25
Prob > F = 0.0000

```

7.1.2.1 Interaction Effects

```
. use nlsw88.(NLSW, 1988 extract)
```

```
. keep if !missing(wage + race + union)
(368 observations deleted)
```

```
. generate lwage = log(wage)
```

```
. tabulate race, generate(R)
```

race	Freq.	Percent	Cum.
white	1,353	72.04	72.04
black	501	26.68	98.72
other	24	1.28	100.00
Total	1,878	100.00	

```
. generate R1u = R1 * union
```

```
. generate R2u = R2 * union
```

```
. regress lwage R1 R2 union R1u R2u
```

Source	SS	df	MS	Number of obs	=	1,878
				F(5, 1872)	=	26.63
Model	33.3636017	5	6.67272035	Prob > F	=	0.0000
Residual	469.09053	1,872	.250582548	R-squared	=	0.0664
				Adj R-squared	=	0.0639

Total | 502.454132 1,877 .267690001 Root MSE = .50058

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+-----						
R1	-.1818955	.1260945	-1.44	0.149	-.4291962	.0654051
R2	-.4152863	.1279741	-3.25	0.001	-.6662731	-.1642995
union	-.2375316	.2167585	-1.10	0.273	-.6626452	.187582
R1u	.4232627	.2192086	1.93	0.054	-.0066561	.8531816
R2u	.6193578	.2221704	2.79	0.005	.1836302	1.055085
_cons	2.07205	.1251456	16.56	0.000	1.82661	2.317489
-----+-----						

. test R1u R2u

(1) R1u = 0

(2) R2u = 0

F(2, 1872) = 8.04

Prob > F = 0.0003

7.2 REGRESSION WITH QUALITATIVE & QUANTITATIVE FACTORS

. use nlsw88.(NLSW, 1988 extract)

. keep if !missing(wage + race + union)

(368 observations deleted)

. generate lwage = log(wage)

. tabulate race, generate(R)

race	Freq.	Percent	Cum.
-----+-----			
white	1,353	72.04	72.04
black	501	26.68	98.72
other	24	1.28	100.00
-----+-----			
Total	1,878	100.00	

. regress lwage R1 R2 union tenure

Source	SS	df	MS	Number of obs	=	1,868
-----+-----				F(4, 1863)	=	85.88
Model	77.1526731	4	19.2881683	Prob > F	=	0.0000
Residual	418.434693	1,863	.224602626	R-squared	=	0.1557
-----+-----				Adj R-squared	=	0.1539
Total	495.587366	1,867	.265445831	Root MSE	=	.47392

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+-----						
R1	-.070349	.0976711	-0.72	0.471	-.2619053	.1212073
R2	-.2612185	.0991154	-2.64	0.008	-.4556074	-.0668297

```

      union |   .1871116   .0257654    7.26   0.000   .1365794   .2376438
    tenure |   .0289352   .0019646   14.73   0.000   .0250823   .0327882
      _cons |   1.777386   .0975549   18.22   0.000   1.586058   1.968715
-----
. test R1 R2

( 1)  R1 = 0
( 2)  R2 = 0

      F( 2, 1863) =   29.98
      Prob > F =   0.0000

```

7.2.1 Testing for Slope Differences

```

. use nls88.(NLSW, 1988 extract)

. keep if !missing(wage + race + union)
(368 observations deleted)

. generate lwage = log(wage)

. tabulate race, generate(R)

```

race	Freq.	Percent	Cum.
white	1,353	72.04	72.04
black	501	26.68	98.72
other	24	1.28	100.00
Total	1,878	100.00	

```

. generate uTen = union * tenure
(10 missing values generated)

. regress lwage R1 R2 union tenure uTen

```

Source	SS	df	MS	Number of obs	=	1,868
Model	77.726069	5	15.5452138	F(5, 1862)	=	69.27
Residual	417.861297	1,862	.224415304	Prob > F	=	0.0000
				R-squared	=	0.1568
				Adj R-squared	=	0.1546
Total	495.587366	1,867	.265445831	Root MSE	=	.47372

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
R1	-.0715443	.0976332	-0.73	0.464	-.2630264 .1199377
R2	-.2638742	.0990879	-2.66	0.008	-.4582093 -.0695391
union	.2380442	.0409706	5.81	0.000	.157691 .3183975
tenure	.0309616	.0023374	13.25	0.000	.0263774 .0355458
uTen	-.0068913	.0043112	-1.60	0.110	-.0153467 .001564
_cons	1.766484	.0977525	18.07	0.000	1.574768 1.9582

```
. generate R1ten = R1 * tenure
(10 missing values generated)
```

```
. generate R2ten = R2 * tenure
(10 missing values generated)
```

```
. regress lwage R1 R2 union tenure R1ten R2ten
```

Source		SS	df	MS	Number of obs	=	1,868
-----+					F(6, 1861)	=	57.26
Model		77.2369283	6	12.8728214	Prob > F	=	0.0000
Residual		418.350438	1,861	.224798731	R-squared	=	0.1558
-----+					Adj R-squared	=	0.1531
Total		495.587366	1,867	.265445831	Root MSE	=	.47413

lwage		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+						
R1		-.082753	.1395	-0.59	0.553	-.3563459 .1908398
R2		-.291495	.1422361	-2.05	0.041	-.570454 -.012536
union		.1876079	.0257915	7.27	0.000	.1370246 .2381912
tenure		.0257611	.0186309	1.38	0.167	-.0107785 .0623007
R1ten		.0024973	.0187646	0.13	0.894	-.0343045 .0392991
R2ten		.0050825	.018999	0.27	0.789	-.032179 .0423441
_cons		1.794018	.1382089	12.98	0.000	1.522957 2.065078

```
. test R1ten R2ten
```

```
( 1) R1ten = 0
```

```
( 2) R2ten = 0
```

```

F( 2, 1861) = 0.19
Prob > F = 0.8291
```

```
. regress lwage R1 R2 union tenure uTen R1ten R2ten
```

Source		SS	df	MS	Number of obs	=	1,868
-----+					F(7, 1860)	=	49.48
Model		77.8008722	7	11.1144103	Prob > F	=	0.0000
Residual		417.786494	1,860	.224616394	R-squared	=	0.1570
-----+					Adj R-squared	=	0.1538
Total		495.587366	1,867	.265445831	Root MSE	=	.47394

lwage		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+						
R1		-.0697096	.1396861	-0.50	0.618	-.3436676 .2042485
R2		-.2795277	.1423788	-1.96	0.050	-.5587668 -.0002886
union		.238244	.0410597	5.80	0.000	.1577161 .3187718
tenure		.0304528	.0188572	1.61	0.106	-.0065308 .0674364
uTen		-.0068628	.0043311	-1.58	0.113	-.0153572 .0016316
R1ten		-.0001912	.0188335	-0.01	0.992	-.0371283 .0367459
R2ten		.0023429	.0190698	0.12	0.902	-.0350576 .0397433

_cons	1.76904	.1390492	12.72	0.000	1.496331	2.041749
-------	---------	----------	-------	-------	----------	----------

```
. test uTen R1ten R2ten
```

```
( 1) uTen = 0
( 2) R1ten = 0
( 3) R2ten = 0
```

```
F( 3, 1860) = 0.96
Prob > F = 0.4098
```

```
. regress lwage union tenure uTen
```

Source	SS	df	MS	Number of obs =	1,868
				F(3, 1864) =	92.25
Model	64.0664855	3	21.3554952	Prob > F =	0.0000
Residual	431.52088	1,864	.231502618	R-squared =	0.1293
				Adj R-squared =	0.1279
Total	495.587366	1,867	.265445831	Root MSE =	.48115

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
union	.2144586	.0414898	5.17	0.000	.1330872 .29583
tenure	.0298926	.0023694	12.62	0.000	.0252456 .0345395
uTen	-.0056219	.0043756	-1.28	0.199	-.0142035 .0029597
_cons	1.655054	.0193938	85.34	0.000	1.617018 1.69309

```
. regress lwage tenure if !union
```

Source	SS	df	MS	Number of obs =	1,408
				F(1, 1406) =	148.43
Model	36.8472972	1	36.8472972	Prob > F =	0.0000
Residual	349.032053	1,406	.248244703	R-squared =	0.0955
				Adj R-squared =	0.0948
Total	385.87935	1,407	.274256823	Root MSE =	.49824

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
tenure	.0298926	.0024536	12.18	0.000	.0250795 .0347056
_cons	1.655054	.0200828	82.41	0.000	1.615659 1.69445

```
. predict double unw if e(sample), res
(470 missing values generated)
```

```
. regress lwage tenure if union
```

Source	SS	df	MS	Number of obs =	460
				F(1, 458) =	55.95
Model	10.0775663	1	10.0775663	Prob > F =	0.0000

Residual		82.4888278	458	.180106611	R-squared	=	0.1089
-----+							
Total		92.5663941	459	.201669704	Adj R-squared	=	0.1069
					Root MSE	=	.42439

lwage		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+						
tenure		.0242707	.0032447	7.48	0.000	.0178944 .0306469
_cons		1.869513	.0323515	57.79	0.000	1.805937 1.933088

```
. predict double nunw if e(sample), res
(1,418 missing values generated)
```

```
.
. generate double allres = nunw
(1,418 missing values generated)
```

```
. replace allres = unw if unw <.
(1,408 real changes made)
```

```
. sdtest allres, by(union)
```

Variance ratio test

Group		Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]
-----+						
nonunion		1,408	7.51e-17	.0132735	.4980645	-.0260379 .0260379
union		460	3.33e-17	.0197657	.4239271	-.0388425 .0388425
-----+						
combined		1,868	6.48e-17	.0111235	.4807605	-.0218157 .0218157

ratio = sd(nonunion) / sd(union)	f =	1.3803
Ho: ratio = 1	degrees of freedom =	1407, 459

Ha: ratio < 1	Ha: ratio != 1	Ha: ratio > 1
Pr(F < f) = 1.0000	2*Pr(F > f) = 0.0000	Pr(F > f) = 0.0000

```
.
. regress lwage union tenure uTen, robust
```

Linear regression	Number of obs	=	1,868
	F(3, 1864)	=	109.84
	Prob > F	=	0.0000
	R-squared	=	0.1293
	Root MSE	=	.48115

		Robust				
lwage		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+						
union		.2144586	.0407254	5.27	0.000	.1345864 .2943308
tenure		.0298926	.0023964	12.47	0.000	.0251928 .0345924
uTen		-.0056219	.0038631	-1.46	0.146	-.0131984 .0019546
_cons		1.655054	.0210893	78.48	0.000	1.613693 1.696415

7.3 SEASONAL ADJUSTMENT WITH INDICATOR VARIABLES

```
. use turksales.. describe
```

Contains data from turksales.dta

```
obs:      40
vars:      2                      21 Jun 2006 11:33
size:     320                      (_dta has notes)
```

```
-----
      storage   display   value
variable name  type      format   label      variable label
-----
t              float     %tq
sales          float     %9.0g
-----
```

Sorted by: t

```
. summarize
```

```
-----+-----
      Variable |      Obs      Mean    Std. Dev.      Min      Max
-----+-----
          t |         40      139.5    11.69045       120      159
       sales |         40    105.6178    4.056961    97.84603    112.9617
-----+-----
```

```
. summarize sales, meanonly
```

```
. generate mu = r(mean)
```

```
. forvalues i=1/3 {
```

```
2.     generate qseas'i' = (quarter(dofq(t)) == 'i')
```

```
3. }
```

```
. regress sales qseas*
```

```
-----+-----
      Source |      SS      df      MS      Number of obs      =      40
-----+-----
      Model | 161.370376      3    53.7901254      F(3, 36)      =      4.03
      Residual | 480.52796      36    13.3479989      Prob > F      =      0.0143
-----+-----
      Total | 641.898336      39    16.4589317      R-squared      =      0.2514
      Adj R-squared      =      0.1890
      Root MSE      =      3.6535
-----+-----
```

```
-----+-----
      sales |      Coef.   Std. Err.      t    P>|t|     [95% Conf. Interval]
-----+-----
      qseas1 | -5.232047    1.633891    -3.20   0.003    -8.545731   -1.918362
      qseas2 | -2.842753    1.633891    -1.74   0.090    -6.156437    .4709317
      qseas3 | -.8969368    1.633891    -0.55   0.586    -4.210621    2.416748
      _cons | 107.8608     1.155335    93.36   0.000     105.5177    110.2039
-----+-----
```

```
. predict double salesSA, residual
```

```
. replace salesSA = salesSA + mu
```

(40 real changes made)

```
. summarize sales salesSA
```

Variable	Obs	Mean	Std. Dev.	Min	Max
-----+-----					
sales	40	105.6178	4.056961	97.84603	112.9617
salesSA	40	105.6178	3.510161	97.49429	111.9563

```
.
. label var salesSA "sales, seasonally adjusted"
```

```
. tsline sales salesSA, lpattern(solid dash)
```

```
.
. regress sales qseas* t
```

Source	SS	df	MS	Number of obs	=	40
-----+-----				F(4, 35)	=	54.23
Model	552.710487	4	138.177622	Prob > F	=	0.0000
Residual	89.1878487	35	2.54822425	R-squared	=	0.8611
-----+-----				Adj R-squared	=	0.8452
Total	641.898336	39	16.4589317	Root MSE	=	1.5963

sales	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----+-----					
qseas1	-4.415311	.7169299	-6.16	0.000	-5.870756 -2.959866
qseas2	-2.298262	.7152449	-3.21	0.003	-3.750287 -.846238
qseas3	-.6246916	.7142321	-0.87	0.388	-2.07466 .8252766
t	.2722452	.0219686	12.39	0.000	.2276466 .3168438
_cons	69.47421	3.138432	22.14	0.000	63.10285 75.84556
-----+-----					

```
. test qseas1 qseas2 qseas3
```

```
( 1) qseas1 = 0
```

```
( 2) qseas2 = 0
```

```
( 3) qseas3 = 0
```

```

F( 3, 35) = 15.17
Prob > F = 0.0000
```

```
. predict double salesSADT, residual
```

```
. replace salesSADT = salesSADT + mu
(40 real changes made)
```

```
.
. label var salesSADT "sales, detrended and SA"
```

```
. tsline sales salesSADT, lpattern(solid dash) yline('mu')
```

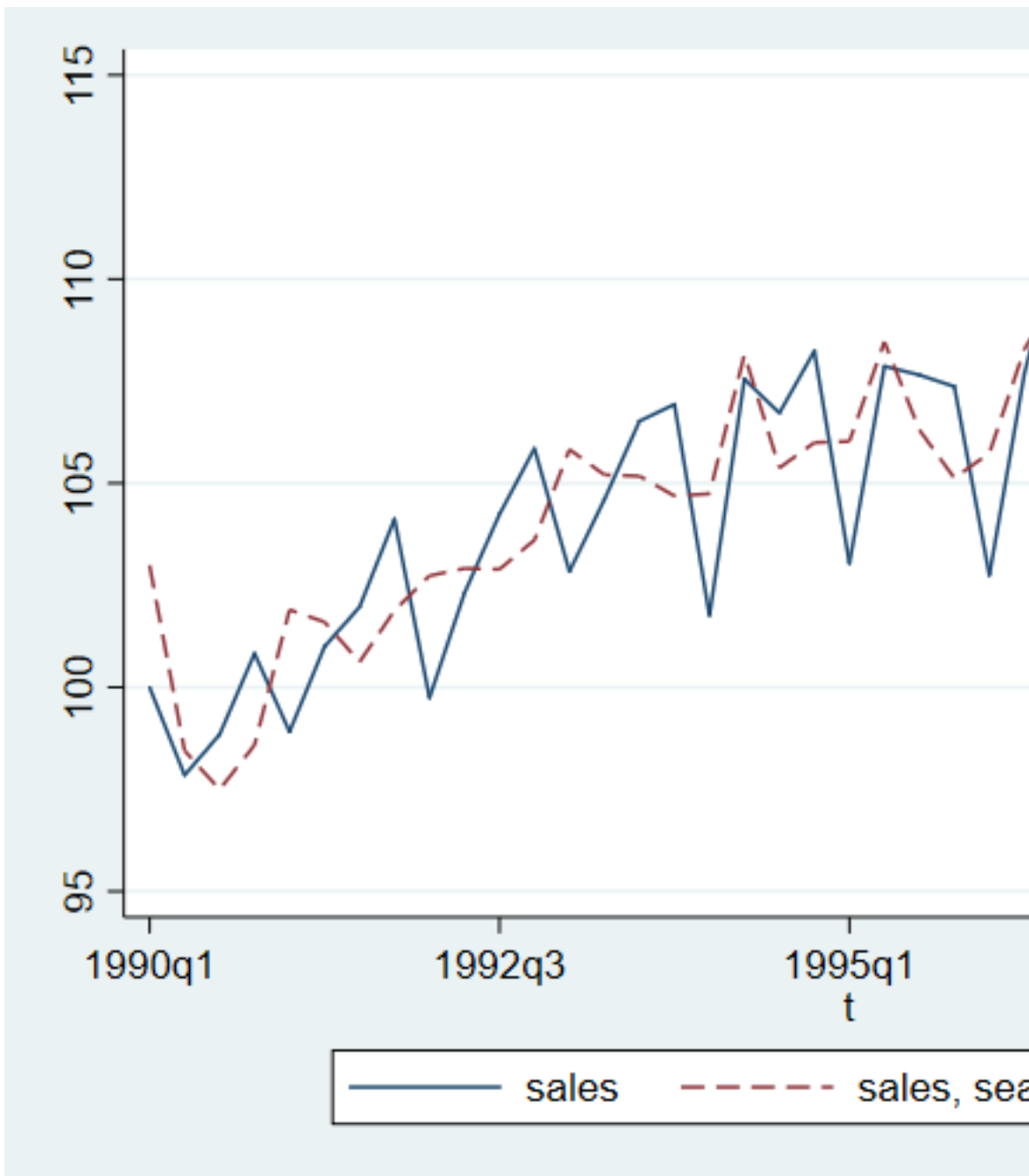


Figure 7: Seasonal adjustment of time series



Figure 8: Seasonal adjustment and detrending of time series

7.4 TESTING FOR STRUCTURAL STABILITY & STRUCTURAL CHANGE

7.4.1 Constraints of Continuity & Differentiability

```
. use nls88.(NLSW, 1988 extract)
```

```
. describe
```

Contains data from nls88.dta

```
obs:      2,246      NLSW, 1988 extract
vars:      17        21 Jun 2006 11:33
size:     60,642     (_dta has notes)
```

variable name	storage type	display format	value label	variable label
idcode	int	%8.0g		NLS id
age	byte	%8.0g		age in current year
race	byte	%8.0g	racelbl	race
married	byte	%8.0g	marlbl	married
never_married	byte	%8.0g		never married
grade	byte	%8.0g		current grade completed
collgrad	byte	%16.0g	gradlbl	college graduate
south	byte	%8.0g		lives in south
smsa	byte	%9.0g	smsalbl	lives in SMSA
c_city	byte	%8.0g		lives in central city
industry	byte	%23.0g	indlbl	industry
occupation	byte	%22.0g	occlbl	occupation
union	byte	%8.0g	unionlbl	union worker
wage	float	%9.0g		hourly wage
hours	byte	%8.0g		usual hours worked
ttl_exp	float	%9.0g		total work experience
tenure	float	%9.0g		job tenure (years)

Sorted by: idcode

```
. summarize
```

Variable	Obs	Mean	Std. Dev.	Min	Max
idcode	2,246	2612.654	1480.864	1	5159
age	2,246	39.15316	3.060002	34	46
race	2,246	1.282725	.4754413	1	3
married	2,246	.6420303	.4795099	0	1
never_marr~d	2,246	.1041852	.3055687	0	1
grade	2,244	13.09893	2.521246	0	18
collgrad	2,246	.2368655	.4252538	0	1
south	2,246	.4194123	.4935728	0	1
smsa	2,246	.7039181	.4566292	0	1
c_city	2,246	.2916296	.4546139	0	1
industry	2,232	8.189516	3.010875	1	12
occupation	2,237	4.642825	3.408897	1	13
union	1,878	.2454739	.4304825	0	1

wage	2,246	7.766949	5.755523	1.004952	40.74659
hours	2,242	37.21811	10.50914	1	80
-----+					
t1l_exp	2,246	12.53498	4.610208	.1153846	28.88461
tenure	2,231	5.97785	5.510331	0	25.91667

```
. generate lwage = log(wage)

. generate Ten2 = tenure <= 2

. generate Ten7 = !Ten2 & tenure <= 7

. generate Ten12 = !Ten2 & !Ten7 & tenure <= 12

. generate Ten25 = !Ten2 & !Ten7 & !Ten12 & tenure < .
```

```
.
. generate tTen2 = tenure * Ten2
(15 missing values generated)

. generate tTen7 = tenure * Ten7
(15 missing values generated)

. generate tTen12 = tenure * Ten12
(15 missing values generated)

. generate tTen25 = tenure * Ten25
(15 missing values generated)
```

```
.
. regress lwage Ten* tTen*, nocons hascons
```

Source	SS	df	MS	Number of obs	=	2,231
-----+				F(7, 2223)	=	37.12
Model	76.6387069	7	10.9483867	Prob > F	=	0.0000
Residual	655.578361	2,223	.294907045	R-squared	=	0.1047
-----+				Adj R-squared	=	0.1018
Total	732.217068	2,230	.328348461	Root MSE	=	.54305

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+						
Ten2	1.55662	.0383259	40.62	0.000	1.481462	1.631778
Ten7	1.708728	.060084	28.44	0.000	1.590901	1.826554
Ten12	1.870808	.1877798	9.96	0.000	1.502566	2.23905
Ten25	1.751961	.1691799	10.36	0.000	1.420194	2.083728
tTen2	.0897426	.0331563	2.71	0.007	.0247221	.1547631
tTen7	.0434089	.0140739	3.08	0.002	.0158095	.0710083
tTen12	.0154208	.019786	0.78	0.436	-.0233801	.0542218
tTen25	.0238014	.0102917	2.31	0.021	.0036191	.0439837

```
. predict double lwagehat
(option xb assumed; fitted values)
(15 missing values generated)
```



```

. label var lwagehat "Predicted log(wage)"

. sort tenure

.
. twoway (line lwagehat tenure if tenure <= 2) (line lwagehat tenure if tenure
> > 2 & tenure <= 7) (line lwagehat tenure if tenure > 7 & tenure <= 12) (lin
> e lwagehat tenure if tenure > 12 & tenure < .), legend(off)

.
. mkspline sTen2 2 sTen7 7 sTen12 12 sTen25 = tenure

. regress lwage sTen*

```

Source	SS	df	MS	Number of obs	=	2,231
-----+				F(4, 2226)	=	64.55
Model	76.1035947	4	19.0258987	Prob > F	=	0.0000
Residual	656.113473	2,226	.294749988	R-squared	=	0.1039
-----+				Adj R-squared	=	0.1023
Total	732.217068	2,230	.328348461	Root MSE	=	.54291

lwage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+						
sTen2	.1173168	.0248619	4.72	0.000	.0685619	.1660716
sTen7	.0471177	.009448	4.99	0.000	.02859	.0656455
sTen12	.0055041	.0111226	0.49	0.621	-.0163076	.0273158
sTen25	.0237767	.0083618	2.84	0.005	.007379	.0401744
_cons	1.539985	.0359605	42.82	0.000	1.469465	1.610505

```

.
. predict double lwageSpline
(option xb assumed; fitted values)
(15 missing values generated)

.
. label var lwageSpline "Predicted log(wage), splined"

. twoway line lwageSpline tenure

.

```

7.4.2 Structural Change in a Time Series Model

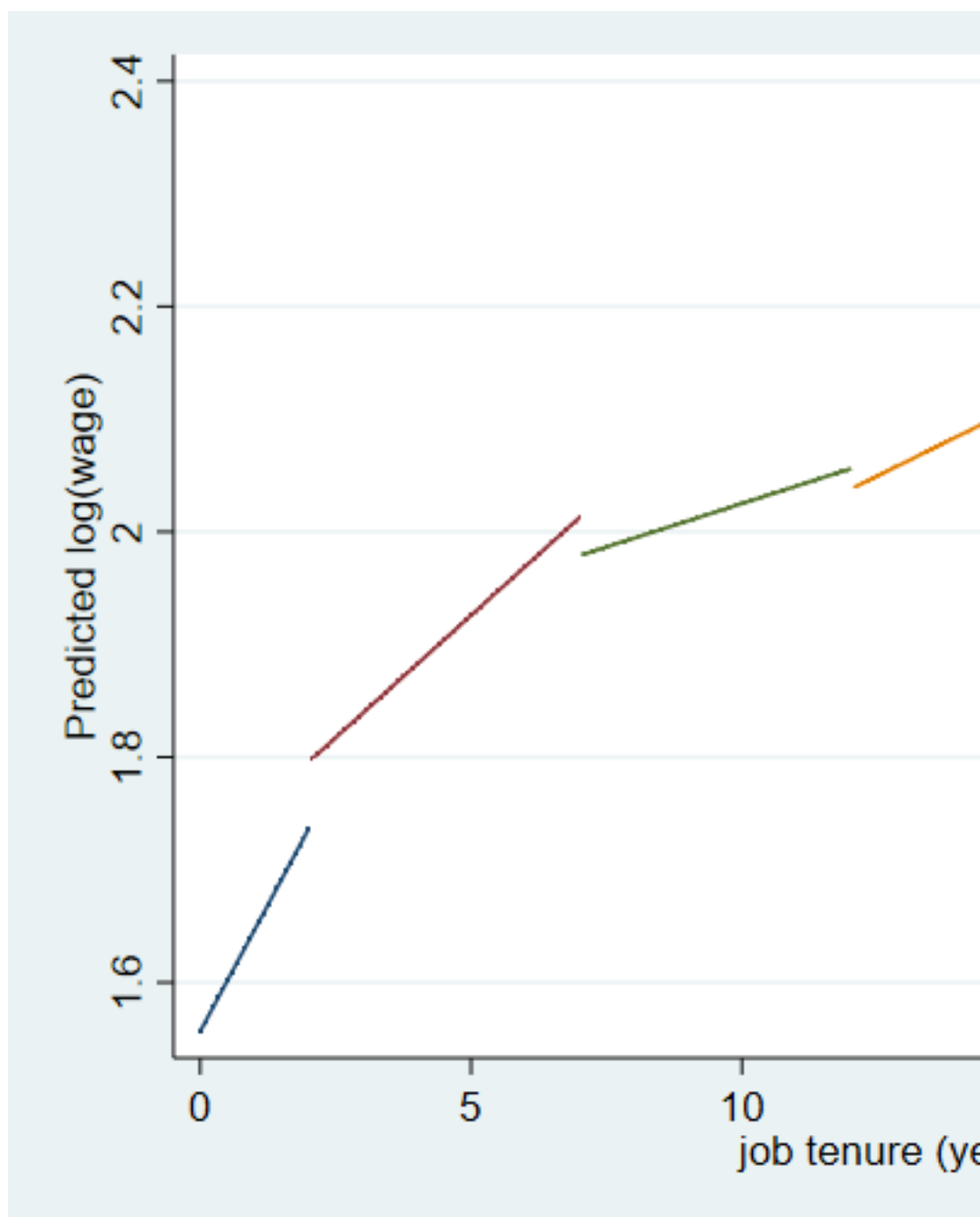


Figure 9: Piecewise wage-tenure profile

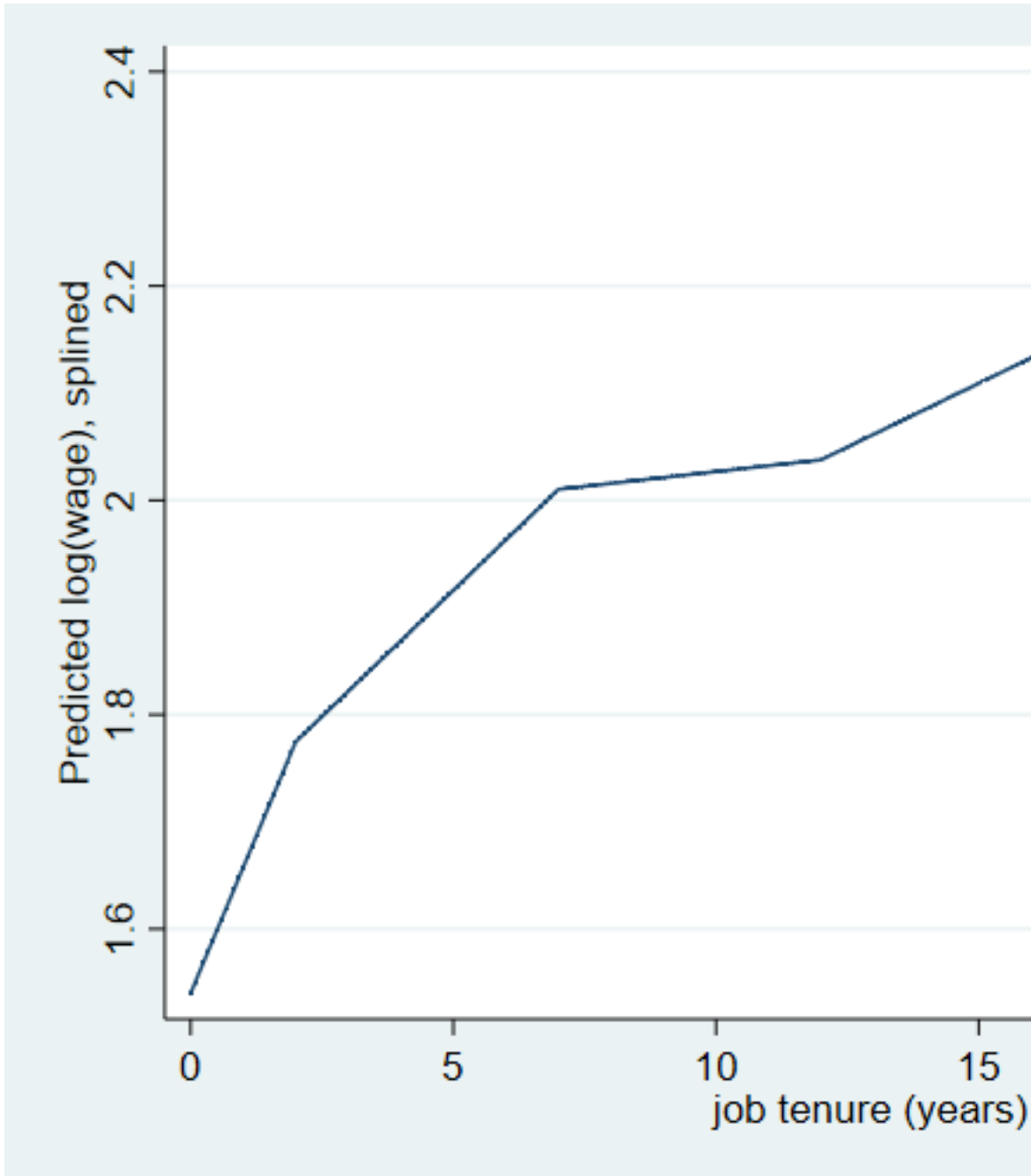


Figure 10: Piecewise linear wage-tenure profile

INSTRUMENTAL VARIABLES ESTIMATORS

8.1 INTRODUCTION

8.2 ENDOGENEITY IN ECONOMIC RELATIONSHIPS

8.3 2SLS

8.4 THE IVREG COMMAND

8.5 IDENTIFICATION & TESTS OF OVERIDENTIFYING RESTRICTIONS

8.6 COMPUTING IV ESTIMATES

```
. use griliches.(Wages of Very Young Men, Zvi Griliches, J.Pol.Ec. 1976)
```

```
. describe
```

Contains data from griliches.dta

```
obs:          758                Wages of Very Young Men, Zvi
                                Griliches, J.Pol.Ec. 1976
vars:          26                31 Oct 2004 14:12
size:         65,188
```

variable name	storage type	display format	value label	variable label
rns	float	%9.0g		residency in South
rns80	float	%9.0g		
mrt	float	%9.0g		marital status = 1 if married
mrt80	float	%9.0g		
smsa	float	%9.0g		reside metro area = 1 if urban
smsa80	float	%9.0g		
med	float	%9.0g		mother's education, years
iq	float	%9.0g		iq score
kww	float	%9.0g		score on knowledge in world of work test
year	float	%9.0g		
age	float	%9.0g		
age80	float	%9.0g		
s	float	%9.0g		completed years of schooling
s80	float	%9.0g		
expr	float	%9.0g		experience, years
expr80	float	%9.0g		
tenure	float	%9.0g		tenure, years
tenure80	float	%9.0g		

```

lw          float    %9.0g          log wage
lw80        float    %9.0g
_Iyear_67   byte     %8.0g          year==67
_Iyear_68   byte     %8.0g          year==68
_Iyear_69   byte     %8.0g          year==69
_Iyear_70   byte     %8.0g          year==70
_Iyear_71   byte     %8.0g          year==71
_Iyear_73   byte     %8.0g          year==73

```

Sorted by:

```
. summarize
```

Variable	Obs	Mean	Std. Dev.	Min	Max
-----+					
rns	758	.2691293	.4438001	0	1
rns80	758	.292876	.4553825	0	1
mrt	758	.5145119	.5001194	0	1
mrt80	758	.8984169	.3022988	0	1
smsa	758	.7044855	.456575	0	1
-----+					
smsa80	758	.7124011	.452942	0	1
med	758	10.91029	2.74112	0	18
iq	758	103.8562	13.61867	54	145
kww	758	36.57388	7.302247	12	56
year	758	69.03166	2.631794	66	73
-----+					
age	758	21.83509	2.981756	16	30
age80	758	33.01187	3.085504	28	38
s	758	13.40501	2.231828	9	18
s80	758	13.70712	2.214693	9	18
expr	758	1.735429	2.105542	0	11.444
-----+					
expr80	758	11.39426	4.210745	.692	22.045
tenure	758	1.831135	1.67363	0	10
tenure80	758	7.362797	5.05024	0	22
lw	758	5.686739	.4289494	4.605	7.051
lw80	758	6.826555	.4099268	4.749	8.032
-----+					
_Iyear_67	758	.0831135	.2762359	0	1
_Iyear_68	758	.1042216	.3057496	0	1
_Iyear_69	758	.1121372	.3157435	0	1
_Iyear_70	758	.0844327	.2782193	0	1
_Iyear_71	758	.121372	.3267747	0	1
-----+					
_Iyear_73	758	.2084433	.4064636	0	1

```
. use griliches.(Wages of Very Young Men, Zvi Griliches, J.Pol.Ec. 1976)
```

```
. ivreg lw s expr tenure rns smsa _I* (iq=med kww age mrt), first
```

First-stage regressions

Source	SS	df	MS	Number of obs	=	758
-----+				F(15, 742)	=	25.03

Model		47176.4676	15	3145.09784	Prob > F	=	0.0000
Residual		93222.8583	742	125.637275	R-squared	=	0.3360
-----+					Adj R-squared	=	0.3226
Total		140399.326	757	185.468066	Root MSE	=	11.209

iq	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+						
s	2.497742	.2858159	8.74	0.000	1.936638	3.058846
expr	-.033548	.2534458	-0.13	0.895	-.5311042	.4640082
tenure	.6158215	.2731146	2.25	0.024	.0796522	1.151991
rns	-2.610221	.9499731	-2.75	0.006	-4.475177	-.7452663
smsa	.0260481	.9222585	0.03	0.977	-1.784499	1.836595
_Iyear_67	.9254935	1.655969	0.56	0.576	-2.325449	4.176436
_Iyear_68	.4706951	1.574561	0.30	0.765	-2.620429	3.56182
_Iyear_69	2.164635	1.521387	1.42	0.155	-.8221007	5.15137
_Iyear_70	5.734786	1.696033	3.38	0.001	2.405191	9.064381
_Iyear_71	5.180639	1.562156	3.32	0.001	2.113866	8.247411
_Iyear_73	4.526686	1.48294	3.05	0.002	1.615429	7.437943
med	.2877745	.1622338	1.77	0.077	-.0307176	.6062665
kww	.4581116	.0699323	6.55	0.000	.3208229	.5954003
age	-.8809144	.2232535	-3.95	0.000	-1.319198	-.4426307
mrt	-.584791	.946056	-0.62	0.537	-2.442056	1.272474
_cons	67.20449	4.107281	16.36	0.000	59.14121	75.26776

Instrumental variables (2SLS) regression

Source		SS	df	MS	Number of obs	=	758
-----+					F(12, 745)	=	45.91
Model		59.2679161	12	4.93899301	Prob > F	=	0.0000
Residual		80.0182337	745	.107407025	R-squared	=	0.4255
-----+					Adj R-squared	=	0.4163
Total		139.28615	757	.183997556	Root MSE	=	.32773

lw	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+						
iq	.0001747	.0039374	0.04	0.965	-.0075551	.0079044
s	.0691759	.013049	5.30	0.000	.0435587	.0947931
expr	.029866	.006697	4.46	0.000	.0167189	.0430132
tenure	.0432738	.0076934	5.62	0.000	.0281705	.058377
rns	-.1035897	.0297371	-3.48	0.001	-.1619682	-.0452111
smsa	.1351148	.0268889	5.02	0.000	.0823277	.1879019
_Iyear_67	-.052598	.0481067	-1.09	0.275	-.1470388	.0418428
_Iyear_68	.0794686	.0451078	1.76	0.079	-.009085	.1680222
_Iyear_69	.2108962	.0443153	4.76	0.000	.1238984	.2978939
_Iyear_70	.2386338	.0514161	4.64	0.000	.1376962	.3395714
_Iyear_71	.2284609	.0441236	5.18	0.000	.1418396	.3150823
_Iyear_73	.3258944	.0410718	7.93	0.000	.2452642	.4065247
_cons	4.39955	.2708771	16.24	0.000	3.867777	4.931323

Instrumented: iq

Instruments: s expr tenure rns smsa _Iyear_67 _Iyear_68 _Iyear_69
_Iyear_70 _Iyear_71 _Iyear_73 med kww age mrt

8.7 IVREG2 & GMM ESTIMATION

8.7.1 *The GMM Estimator*

8.7.2 *GMM in a Homoskedastic Context*

8.7.3 *GMM & Heteroskedasticity-Consistent Standard Errors*

```
. use griliches.(Wages of Very Young Men, Zvi Griliches, J.Pol.Ec. 1976)
```

```
. ivreg2 lw s expr tenure rns smsa _I* (iq=med kww age mrt), gmm  
command ivreg2 is unrecognized  
r(199);
```

```
end of do-file  
r(199);
```

8.7.4 *GMM & Clustering*

8.7.5 *GMM & HAC Standard Errors*

8.8 TESTING & OVERIDENTIFYING RESTRICTIONS IN GMM

8.8.1 *Testing a subset of the Overidentifying Restrictions in GMM*

8.9 TESTING FOR HETEROSKEDASTICITY IN THE IV CONTEXT

8.10 TESTING THE RELEVANCE OF INSTRUMENTS

8.11 DURBIN–WU–HAUSMAN TESTS FOR ENDOGENEITY IN IV ESTIMATION

PANEL DATA MODELS

9.1 FIXED EFFECTS & RANDOM EFFECTS MODELS

9.1.1 *One-Way Fixed Effects Models*

```
. use traffic.. describe
```

Contains data from traffic.dta

```
obs:      336
vars:      54                      30 Nov 2004 10:23
size:     61,152
```

variable name	storage type	display format	value label	variable label
state	float	%9.0g	sid	State ID (FIPS) Code
year	int	%9.0g		Year
spircons	float	%9.0g		Spirits Consumption
unrate	float	%9.0g		Unemployment Rate
perinc	float	%9.0g		Per Capita Personal Income
emppop	float	%9.0g		Employment/Population Ratio
beertax	float	%9.0g		Tax on Case of Beer
sobapt	float	%9.0g		% Southern Baptist
mormon	float	%9.0g		% Mormon
mlda	float	%9.0g		Minimum Legal Drinking Age
dry	float	%9.0g		% Residing in Dry Counties
yngdrv	float	%9.0g		% of Drivers Aged 15-24
vmiles	float	%9.0g		Ave. Mile per Driver
breath	byte	%9.0g		Prelim. Breath Test Law
jaild	byte	%9.0g		Mandatory Jail Sentence
comserd	byte	%9.0g		Mandatory Community Service
allmort	int	%9.0g		# of Vehicle Fatalities (#VF)
mrall	float	%9.0g		Vehicle Fatality Rate (VFR)
allnite	int	%9.0g		# of Night-time VF (#NVF)
mralln	float	%9.0g		Night-time VFR (NVFR)
allsvn	int	%9.0g		# of Single VF (#SVN)
a1517	int	%9.0g		#VF, 15-17 year olds
mra1517	float	%9.0g		VFR, 15-17 year olds
a1517n	byte	%9.0g		#NVF, 15-17 year olds
mra1517n	float	%9.0g		NVFR, 15-17 year olds
a1820	int	%9.0g		#VF, 18-20 year olds
a1820n	int	%9.0g		#NVF, 18-20 year olds
mra1820	float	%9.0g		VFR, 18-20 year olds
mra1820n	float	%9.0g		NVFR, 18-20 year olds
a2124	int	%9.0g		#VF, 21-24 year olds
mra2124	float	%9.0g		VFR, 21-24 year olds
a2124n	int	%9.0g		#NVF, 21-24 year olds
mra2124n	float	%9.0g		NVFR, 21-24 year olds

aidall	float	%9.0g		# of alcohol-involved VF
mraidall	float	%9.0g		Alcohol-Involved VFR
pop	float	%9.0g		Population
pop1517	float	%9.0g		Population, 15-17 year olds
pop1820	float	%9.0g		Population, 18-20 year olds
pop2124	float	%9.0g		Population, 21-24 year olds
miles	float	%9.0g		total vehicle miles (millions
unus	float	%9.0g		U.S. unemployment rate
epopus	float	%9.0g		U.S. Emp/Pop Ratio
gspch	float	%9.0g		GSP Rate of Change
stabrv	str2	%9s		
region	str2	%9s		
reg	long	%8.0g	reg	
fatal	float	%9.0g		
fatal1517	float	%9.0g		
fatal1820	float	%9.0g		
fatal2124	float	%9.0g		
fatal1517n	float	%9.0g		
fatal1820n	float	%9.0g		
fatal2124n	float	%9.0g		
perincK	float	%9.0g		

Sorted by:

. summarize

Variable	Obs	Mean	Std. Dev.	Min	Max
state	336	30.1875	15.30985	1	56
year	336	1985	2.002983	1982	1988
spircons	336	1.75369	.6835745	.79	4.9
unrate	336	7.346726	2.533405	2.4	18
perinc	336	13880.18	2253.046	9513.762	22193.46
emppop	336	60.80568	4.721656	42.9932	71.26865
beertax	336	.513256	.4778442	.0433109	2.720764
sobapt	336	7.156925	9.762621	0	30.3557
mormon	336	2.801933	9.665279	.1	65.9165
mlda	336	20.45563	.8990255	18	21
dry	336	4.267074	9.500901	0	45.7921
yngdrv	336	.1859299	.0248736	.073137	.281625
vmiles	336	7890.754	1475.659	4576.346	26148.27
breath	336	.4613095	.4992443	0	1
jaild	335	.280597	.449963	0	1
comserd	335	.1850746	.388939	0	1
allmort	336	928.6637	934.0515	79	5504
mrall	336	.000204	.000057	.0000821	.0004218
allnite	336	182.5833	188.4311	13	1049
mralln	336	.0000388	.000011	.0000172	.0000944
allsvn	336	109.9494	108.5397	8	603
a1517	336	62.61012	55.72909	3	318
mra1517	336	.0003034	.0000937	.0001163	.0006735
a1517n	336	12.2619	12.25341	0	76
mra1517n	336	.0000598	.000033	0	.0002571

-----+-----					
a1820		336	106.6607	104.2236	7 601
a1820n		336	33.52679	33.23834	0 196
mra1820		336	.0004728	.0001522	.0001855 .0010952
mra1820n		336	.0001436	.0000613	0 .0005238
a2124		336	126.872	131.7886	12 770
-----+-----					
mra2124		336	.0004091	.0001225	.0002 .0008922
a2124n		336	41.37798	42.93031	1 249
mra2124n		336	.0001284	.0000422	.0000222 .0003143
aidall		336	293.3332	303.5807	24.6 2094.9
mrmaidall		336	.0000659	.000026	.0000234 .0001772
-----+-----					
pop		336	4930272	5073704	478999.7 2.83e+07
pop1517		336	230815.5	229896.3	21000.02 1172000
pop1820		336	249090.4	249345.6	20999.96 1321004
pop2124		336	336389.9	345304.4	30000.16 1892998
miles		336	37101.49	37454.37	3993 241575
-----+-----					
unus		336	7.528571	1.479376	5.5 9.7
epopus		336	59.97143	1.585048	57.8 62.3
gspch		336	.0253135	.0431732	-.1236415 .1423609
stabrv		0			
region		0			
-----+-----					
reg		336	2.395833	1.16971	1 4
fatal		336	2.040444	.5701938	.82121 4.21784
fatal1517		336	3.033664	.9374229	1.16279 6.73469
fatal1820		336	4.727677	1.521962	1.854922 10.9524
fatal2124		336	4.091315	1.224812	2 8.92157
-----+-----					
fatal1517n		336	.597681	.3297287	0 2.57143
fatal1820n		336	1.436218	.6128804	0 5.2381
fatal2124n		336	1.284335	.4224502	.22222 3.14286
perinck		336	13.88018	2.253046	9.513762 22.19345

. xtsum fatal beertax spircons unrte perinck state year

Variable		Mean	Std. Dev.	Min	Max	Observations	
fatal	overall	2.040444	.5701938	.82121	4.21784	N =	336
	between		.5461407	1.110077	3.653197	n =	48
	within		.1794253	1.45556	2.962664	T =	7
beertax	overall	.513256	.4778442	.0433109	2.720764	N =	336
	between		.4789513	.0481679	2.440507	n =	48
	within		.0552203	.1415352	.7935126	T =	7
spircons	overall	1.75369	.6835745	.79	4.9	N =	336
	between		.6734649	.8614286	4.388572	n =	48
	within		.147792	1.255119	2.265119	T =	7
unrate	overall	7.346726	2.533405	2.4	18	N =	336
	between		1.953377	4.1	13.2	n =	48
	within		1.634257	4.046726	12.14673	T =	7
perinck	overall	13.88018	2.253046	9.513762	22.19345	N =	336

	between		2.122712	9.95087	19.51582		n =	48
	within		.8068546	11.43261	16.55782		T =	7
state	overall		30.1875	15.30985	1	56	N =	336
	between			15.44883	1	56	n =	48
	within			0	30.1875	30.1875	T =	7
year	overall		1985	2.002983	1982	1988	N =	336
	between			0	1985	1985	n =	48
	within			2.002983	1982	1988	T =	7

```
. use traffic.. xtreg fatal beertax spircons unrte perinck, fe
```

```
Fixed-effects (within) regression                Number of obs   =       336
Group variable: state                          Number of groups =       48
```

```
R-sq:                                           Obs per group:
within  = 0.3526                               min =       7
between = 0.1146                               avg  =      7.0
overall = 0.0863                               max  =       7
```

```
corr(u_i, Xb) = -0.8804                        F(4,284)         =      38.68
                                           Prob > F         =      0.0000
```

	fatal		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
	beertax		-.4840728	.1625106	-2.98	0.003	-.8039508 -.1641948
	spircons		.8169652	.0792118	10.31	0.000	.6610484 .9728819
	unrate		-.0290499	.0090274	-3.22	0.001	-.0468191 -.0112808
	perinck		.1047103	.0205986	5.08	0.000	.064165 .1452555
	_cons		-.383783	.4201781	-0.91	0.362	-1.210841 .4432754
	sigma_u		1.1181913				
	sigma_e		.15678965				
	rho		.98071823	(fraction of variance due to u_i)			

```
F test that all u_i=0: F(47, 284) = 59.77                Prob > F = 0.0000
```

9.1.2 Time Effects & Two-Way Fixed Effects Models

```
. use traffic.. tabulate year, generate(yr)
```

	Year		Freq.	Percent	Cum.
	1982		48	14.29	14.29
	1983		48	14.29	28.57
	1984		48	14.29	42.86
	1985		48	14.29	57.14
	1986		48	14.29	71.43
	1987		48	14.29	85.71
	1988		48	14.29	100.00

Total | 336 100.00

```
.  
. local j 0  
  
. forvalues i=82/87 {  
2. local ++j  
3. rename yr'j' yr'i'  
4. quietly replace yr'i' = yr'i' - yr7  
5. }  
  
.   
  
. drop yr7  
  
.   
  
. xtreg fatal beertax spircons unrte perinck yr*, fe
```

Fixed-effects (within) regression	Number of obs	=	336
Group variable: state	Number of groups	=	48

R-sq:	Obs per group:
within = 0.4528	min = 7
between = 0.1090	avg = 7.0
overall = 0.0770	max = 7

	F(10,278)	=	23.00
corr(u_i, Xb) = -0.8728	Prob > F	=	0.0000

	fatal	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+-----							
beertax	-.4347195	.1539564	-2.82	0.005	-.7377878	-.1316511	
spircons	.805857	.1126425	7.15	0.000	.5841163	1.027598	
unrate	-.0549084	.0103418	-5.31	0.000	-.0752666	-.0345502	
perinck	.0882636	.0199988	4.41	0.000	.0488953	.1276319	
yr82	.1004321	.0355629	2.82	0.005	.0304253	.170439	
yr83	.0470609	.0321574	1.46	0.144	-.0162421	.1103638	
yr84	-.0645507	.0224667	-2.87	0.004	-.1087771	-.0203243	
yr85	-.0993055	.0198667	-5.00	0.000	-.1384139	-.0601971	
yr86	.0496288	.0232525	2.13	0.034	.0038554	.0954021	
yr87	.0003593	.0289315	0.01	0.990	-.0565933	.0573119	
_cons	.0286246	.4183346	0.07	0.945	-.7948812	.8521305	
-----+-----							
sigma_u	1.0987683						
sigma_e	.14570531						
rho	.98271904 (fraction of variance due to u_i)						

F test that all u_i=0: F(47, 278) = 64.52	Prob > F = 0.0000
---	-------------------

```
. test yr82 yr83 yr84 yr85 yr86 yr87
```

- (1) yr82 = 0
- (2) yr83 = 0
- (3) yr84 = 0
- (4) yr85 = 0
- (5) yr86 = 0
- (6) yr87 = 0

```

F( 6, 278) = 8.48
Prob > F = 0.0000

```

9.1.3 *The Between Estimator*

```
. use traffic.. xtreg fatal beertax spircons unrte perinck, be
```

```

Between regression (regression on group means) Number of obs   =      336
Group variable: state                        Number of groups  =      48

```

```

R-sq:                                     Obs per group:
    within = 0.0479                               min =      7
    between = 0.4565                               avg =     7.0
    overall = 0.2583                               max =      7

```

```

F(4,43) = 9.03
sd(u_i + avg(e_i.))= .4209489                Prob > F      = 0.0000

```

	fatal	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
	+						
beertax		.0740362	.1456333	0.51	0.614	-.2196614	.3677338
spircons		.2997517	.1128135	2.66	0.011	.0722417	.5272618
unrate		.0322333	.038005	0.85	0.401	-.0444111	.1088776
perinck		-.1841747	.0422241	-4.36	0.000	-.2693277	-.0990218
_cons		3.796343	.7502025	5.06	0.000	2.283415	5.309271

9.1.4 *One-Way Random Effects Models*

```
. use traffic.. xtreg fatal beertax spircons unrte perinck, re
```

```

Random-effects GLS regression                Number of obs   =      336
Group variable: state                        Number of groups  =      48

```

```

R-sq:                                     Obs per group:
    within = 0.2263                               min =      7
    between = 0.0123                               avg =     7.0
    overall = 0.0042                               max =      7

```

```

Wald chi2(4) = 49.90
corr(u_i, X) = 0 (assumed)                  Prob > chi2      = 0.0000

```

	fatal	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
	+						
beertax		.0442768	.1204613	0.37	0.713	-.191823	.2803765
spircons		.3024711	.0642954	4.70	0.000	.1764546	.4284877
unrate		-.0491381	.0098197	-5.00	0.000	-.0683843	-.0298919
perinck		-.0110727	.0194746	-0.57	0.570	-.0492423	.0270968

_cons		2.001973	.3811247	5.25	0.000	1.254983	2.748964
-----+							
sigma_u		.41675665					
sigma_e		.15678965					
rho		.87601197	(fraction of variance due to u_i)				

9.1.5 Testing the appropriateness of Random Effects Model

```
. use traffic.. quietly xtreg fatal beertax spircons unrte perinck, fe

. estimates store fix

.

. quietly xtreg fatal beertax spircons unrte perinck, re

. estimates store ran

.

. hausman fix ran
```

---- Coefficients ----				
		(b)	(B)	
		fix	ran	(b-B) Difference sqrt(diag(V_b-V_B)) S.E.
-----+				
beertax		-.4840728	.0442768	-.5283495 .1090815
spircons		.8169652	.3024711	.514494 .0462668
unrate		-.0290499	-.0491381	.0200882 .
perinck		.1047103	-.0110727	.115783 .0067112

b = consistent under Ho and Ha; obtained from xtreg
B = inconsistent under Ha, efficient under Ho; obtained from xtreg

Test: Ho: difference in coefficients not systematic

```
chi2(4) = (b-B)'[(V_b-V_B)^(-1)](b-B)
        = 130.93
Prob>chi2 = 0.0000
(V_b-V_B is not positive definite)
```

.

9.1.6 Prediction from One-Way Fixed Effects Model & Random Effects Model

9.2 IV MODELS FOR PANEL DATA

9.3 DYNAMIC PANEL DATA MODELS

```
. use traffic.. tsset
```

```
panel variable:  state (strongly balanced)
time variable:  year, 1982 to 1988
delta:  1 unit
```

```
. xtabond2 fatal L.fatal spircons year, gmmstyle(beertax spircons unrte perinc
> K) ivstyle(year) twostep robust nolevelq
command xtabond2 is unrecognized
r(199);

end of do-file
r(199);
```

9.4 SEEMINGLY UNRELATED REGRESSION MODELS

9.4.1 *SUR with Identical Regressors*

9.5 MOVING-WINDOW REGRESSION ESTIMATES

MODELS OF DISCRETE & LIMITED DEPENDENT VARIABLES

10.1 BINOMIAL LOGIT & PROBIT MODELS

10.1.1 *The Latent Variable Approach*

10.1.2 *Marginal Effects & Predictions*

10.1.2.1 *Binomial Probit*

```
. use womenwk.. describe
```

Contains data from womenwk.dta

obs: 2,000

vars: 15

9 Nov 2004 20:23

size: 134,000

variable name	storage type	display format	value label	variable label
c1	double	%10.0g		
c2	double	%10.0g		
u	double	%10.0g		
v	double	%10.0g		
county	float	%9.0g		
age	int	%8.0g		
education	int	%8.0g		
married	byte	%8.0g		
children	int	%8.0g		
select	float	%9.0g		
wagefull	float	%9.0g		
wage	float	%9.0g		
lw	float	%9.0g		
work	float	%9.0g		
lwf	float	%9.0g		

Sorted by:

```
. summarize
```

Variable	Obs	Mean	Std. Dev.	Min	Max
c1	2,000	-.0023069	.9880571	-3.500514	3.614182
c2	2,000	-.0077596	1.006025	-3.410111	3.423961
u	2,000	-.0011535	.4940286	-1.750257	1.807091
v	2,000	-.0071367	1	-3.954782	3.229851
county	2,000	4.5	2.873	0	9

```

-----+-----
      age |      2,000      36.208      8.28656      20      59
education |      2,000      13.084      3.045912      10      20
    married |      2,000      .6705      .4701492      0      1
   children |      2,000      1.6445      1.398963      0      5
      select |      2,000     35.78556     14.98163    -14.45688     89.63869
-----+-----
    wagefull |      2,000     21.31176     7.012038    -1.680425     45.80979
      wage |      1,343     23.69217     6.305374     5.88497     45.80979
        lw |      1,343     3.126703     .2865111     1.772402     3.824498
      work |      2,000      .6715     .4697852      0      1
      lwf |      2,000     2.099581     1.487519      0     3.824498

```

```

.
. probit work age married children education, nolog

```

```

Probit regression                                Number of obs   =      2,000
                                                LR chi2(4)       =      478.32
                                                Prob > chi2      =      0.0000
Log likelihood = -1027.0616                    Pseudo R2       =      0.1889

```

```

-----+-----
      work |      Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
      age |   .0347211   .0042293     8.21   0.000   .0264318   .0430105
    married |   .4308575   .074208     5.81   0.000   .2854125   .5763025
   children |   .4473249   .0287417    15.56   0.000   .3909922   .5036576
 education |   .0583645   .0109742     5.32   0.000   .0368555   .0798735
      _cons |  -2.467365   .1925635    -12.81   0.000   -2.844782  -2.089948
-----+-----

```

```

. mfx compute

```

```

Marginal effects after probit
      y  = Pr(work) (predict)
      = .71835948

```

```

-----+-----
variable |      dy/dx   Std. Err.      z    P>|z|     [ 95% C.I.   ]      X
-----+-----
      age |   .011721    .00142     8.25   0.000   .008935   .014507     36.208
    married*|   .150478    .02641     5.70   0.000   .098716   .20224     .6705
   children |   .1510059    .00922    16.38   0.000   .132939   .169073     1.6445
educat~n |   .0197024    .0037     5.32   0.000   .012442   .026963    13.084
-----+-----

```

```

(*) dy/dx is for discrete change of dummy variable from 0 to 1

```

```

.
. margeff, dummies(married) count
command margeff is unrecognized
r(199);

```

```

end of do-file
r(199);

```

10.1.3 Evaluating Specification & Goodness of Fit

```
. use womenwk..
. logit work age married children education, nolog
```

```
Logistic regression               Number of obs   =      2,000
                                LR chi2(4)        =      476.62
                                Prob > chi2       =      0.0000
Log likelihood = -1027.9144       Pseudo R2    =      0.1882
```

work	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
age	.0579303	.007221	8.02	0.000	.0437773	.0720833
married	.7417775	.1264705	5.87	0.000	.4938998	.9896552
children	.7644882	.0515289	14.84	0.000	.6634935	.865483
education	.0982513	.0186522	5.27	0.000	.0616936	.134809
_cons	-4.159247	.3320401	-12.53	0.000	-4.810034	-3.508461

```
. mfx compute
```

```
Marginal effects after logit
      y = Pr(work) (predict)
      = .72678588
```

variable	dy/dx	Std. Err.	z	P> z	[95% C.I.]		X
age	.0115031	.00142	8.08	0.000	.008713	.014293	36.208
married*	.1545671	.02703	5.72	0.000	.101592	.207542	.6705
children	.151803	.00938	16.19	0.000	.133425	.170181	1.6445
educat~n	.0195096	.0037	5.27	0.000	.01226	.02676	13.084

(*) dy/dx is for discrete change of dummy variable from 0 to 1

```
. mfx compute, at(children = 0)
```

```
warning: no value assigned in at() for variables age married education;
means used for age married education
```

```
Marginal effects after logit
      y = Pr(work) (predict)
      = .43074191
```

variable	dy/dx	Std. Err.	z	P> z	[95% C.I.]		X
age	.0142047	.00178	7.97	0.000	.01071	.0177	36.208
married*	.1762562	.02825	6.24	0.000	.120897	.231616	.6705
children	.187455	.01115	16.82	0.000	.165609	.209301	0
educat~n	.0240915	.00458	5.26	0.000	.015115	.033068	13.084

(*) dy/dx is for discrete change of dummy variable from 0 to 1

10.2 ORDERED LOGIT & PROBIT MODELS

```
. use panel84extract.. describe
```

Contains data from panel84extract.dta

```
obs:          98
vars:          38                      1 Dec 2004 23:14
size:         14,896
```

variable name	storage type	display format	value label	variable label
s77	float	%9.0g		
s78	float	%9.0g		
s79	float	%9.0g		
s80	float	%9.0g		
s81	float	%9.0g		
s82	float	%9.0g		
s83	float	%9.0g		
b77	float	%9.0g		
b78	float	%9.0g		
b79	float	%9.0g		
b80	float	%9.0g		
b81	float	%9.0g		
b82	float	%9.0g		
b83	float	%9.0g		
is77	float	%9.0g		
is78	float	%9.0g		
is79	float	%9.0g		
is80	float	%9.0g		
is81	float	%9.0g		
is82	float	%9.0g		
is83	float	%9.0g		
ia77	float	%9.0g		
ia78	float	%9.0g		
ia79	float	%9.0g		
ia80	float	%9.0g		
ia81	float	%9.0g		
ia82	float	%9.0g		
ia83	float	%9.0g		
r77	float	%9.0g		
r78	float	%9.0g		
r79	float	%9.0g		
r80	float	%9.0g		
r81	float	%9.0g		
r82	float	%9.0g		
r83	float	%9.0g		
rating83	float	%9.0g		
rating83c	float	%9.0g	bondrating	Bond rating, 1983
dia	float	%9.0g		

Sorted by:

```
. summarize
```

Variable	Obs	Mean	Std. Dev.	Min	Max
-----+					
s77	98	2399226	6835850	28188	5.50e+07
s78	98	2757617	7812873	48432	6.32e+07
s79	98	3084304	8191207	103004	6.63e+07
s80	98	3081560	7295521	114562	5.77e+07
s81	98	3328268	7815398	167239	6.27e+07
-----+					
s82	98	3126689	7479808	115376	6.00e+07
s83	98	3409835	9008834	108397	7.46e+07
b77	98	1.019388	.2675541	.29	2.27
b78	98	1.035	.2937677	.29	2.27
b79	98	1.008265	.2601666	.29	2.27
-----+					
b80	98	1.032245	.2380095	.29	2
b81	98	1.015612	.2452773	.29	2.08
b82	98	.9689796	.2139394	.29	1.65
b83	98	.9932653	.2011348	.46	1.65
is77	98	11.24923	5.615914	-16.54562	26.56383
-----+					
is78	98	11.65585	4.528279	.5114942	26.84296
is79	98	11.36975	4.527863	.045454	23.66122
is80	98	11.04448	9.335074	-4.169284	88.73264
is81	98	10.34562	4.951754	-2.06479	25.77528
is82	98	7.204422	6.152187	-12.69882	24.89959
-----+					
is83	98	7.846085	6.495875	-16.40544	26.55174
ia77	98	14.95777	6.958497	-18.13673	37.67006
ia78	98	15.53974	5.104227	1.188427	29.87578
ia79	98	15.48429	5.302939	.0836498	28.42293
ia80	98	13.8006	6.085422	-6.350523	29.75574
-----+					
ia81	98	13.82875	5.952011	-2.480706	30.5371
ia82	98	9.407202	7.306611	-12.54386	32.30307
ia83	98	10.11473	7.441946	-13.08016	30.74564
r77	98	2.34113	1.874804	0	6
r78	98	2.419951	1.911573	0	6
-----+					
r79	98	2.373979	1.912279	0	6
r80	98	2.315973	1.855837	0	6
r81	98	2.399914	1.841927	0	6
r82	98	2.342588	1.851536	0	6
r83	98	2.391481	1.850125	0	6
-----+					
rating83	98	2.336735	1.787378	0	6
rating83c	98	3.479592	1.17736	2	5
dia	98	.7075242	4.711211	-10.79014	20.05367

. tabulate rating83c

Bond			
rating,			
1983	Freq.	Percent	Cum.
-----+			
BA_B_C	26	26.53	26.53
BAA	28	28.57	55.10

AA_A	15	15.31	70.41
AAA	29	29.59	100.00
-----+-----			
Total	98	100.00	

```
.
. ologit rating83c ia83 dia, nolog
```

Ordered logistic regression	Number of obs	=	98
	LR chi2(2)	=	11.54
	Prob > chi2	=	0.0031
Log likelihood = -127.27146	Pseudo R2	=	0.0434

rating83c	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
-----+-----						
ia83	.0939166	.0296196	3.17	0.002	.0358633	.1519699
dia	-.0866925	.0449789	-1.93	0.054	-.1748496	.0014646
-----+-----						
/cut1	-.1853053	.3571432			-.8852932	.5146826
/cut2	1.185726	.3882099			.4248488	1.946604
/cut3	1.908412	.4164896			1.092108	2.724717
-----+-----						

```
.
. predict spBA_B_C spBAA spAA_A spAAA
(option pr assumed; predicted probabilities)
```

```
. summarize spAAA, mean
```

```
. list sp* rating83c if spAAA == r(max)
```

+-----+-----+-----+-----+-----+-----+					
	spBA_B_C	spBAA	spAA_A	spAAA	rati~83c

31.	.0388714	.0985567	.1096733	.7528986	AAA
	+-----+-----+-----+-----+-----+				

```
.
. summarize spBA_B_C, mean
```

```
. list sp* rating83c if spBA_B_C == r(max)
```

+-----+-----+-----+-----+-----+-----+					
	spBA_B_C	spBAA	spAA_A	spAAA	rati~83c

67.	.7158453	.1926148	.0449056	.0466343	AAA
	+-----+-----+-----+-----+-----+				

10.3 TRUNCATED REGRESSION & TOBIT MODELS

10.3.1 *Truncation*

```
. use laborsub.. describe
```

Contains data from laborsub.dta

obs: 250

vars: 6

size: 1,750

25 Sep 2004 18:36

variable name	storage type	display format	value label	variable label
lfp	byte	%9.0g		1 if woman worked in 1975
whrs	int	%9.0g		Wife's hours of work
kl6	byte	%9.0g		# of children younger than 6
k618	byte	%9.0g		# of children between 6 and 18
wa	byte	%9.0g		Wife's age
we	byte	%9.0g		Wife's educational attainment

Sorted by:

. summarize

Variable	Obs	Mean	Std. Dev.	Min	Max
lfp	250	.6	.4908807	0	1
whrs	250	799.84	915.6035	0	4950
kl6	250	.236	.5112234	0	3
k618	250	1.364	1.370774	0	8
wa	250	42.92	8.426483	30	60
we	250	12.352	2.164912	5	17

.

. regress whrs kl6 k618 wa we if whrs > 0

Source	SS	df	MS	Number of obs	=	150
Model	7326995.15	4	1831748.79	F(4, 145)	=	2.80
Residual	94793104.2	145	653745.546	Prob > F	=	0.0281
Total	102120099	149	685369.794	R-squared	=	0.0717
				Adj R-squared	=	0.0461
				Root MSE	=	808.55

whrs	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
kl6	-421.4822	167.9734	-2.51	0.013	-753.4748 -89.48953
k618	-104.4571	54.18616	-1.93	0.056	-211.5538 2.639668
wa	-4.784917	9.690502	-0.49	0.622	-23.9378 14.36797
we	9.353195	31.23793	0.30	0.765	-52.38731 71.0937
_cons	1629.817	615.1301	2.65	0.009	414.0371 2845.597

.

. truncreg whrs kl6 k618 wa we, ll(0) nolog

(note: 100 obs. truncated)

Truncated regression

Limit: lower = 0

upper = +inf

Number of obs = 150

Wald chi2(4) = 10.05

Log likelihood = -1200.9157 Prob > chi2 = 0.0395

whrs	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
kl6	-803.0042	321.3614	-2.50	0.012	-1432.861	-173.1474
k618	-172.875	88.72898	-1.95	0.051	-346.7806	1.030579
wa	-8.821123	14.36848	-0.61	0.539	-36.98283	19.34059
we	16.52873	46.50375	0.36	0.722	-74.61695	107.6744
_cons	1586.26	912.355	1.74	0.082	-201.9233	3374.442
/sigma	983.7262	94.44303	10.42	0.000	798.6213	1168.831

10.3.2 Censoring

. use womenwk.. describe

Contains data from womenwk.dta

obs: 2,000
vars: 15 9 Nov 2004 20:23
size: 134,000

variable name	storage type	display format	value label	variable label
c1	double	%10.0g		
c2	double	%10.0g		
u	double	%10.0g		
v	double	%10.0g		
county	float	%9.0g		
age	int	%8.0g		
education	int	%8.0g		
married	byte	%8.0g		
children	int	%8.0g		
select	float	%9.0g		
wagefull	float	%9.0g		
wage	float	%9.0g		
lw	float	%9.0g		
work	float	%9.0g		
lwf	float	%9.0g		

Sorted by:

. summarize

Variable	Obs	Mean	Std. Dev.	Min	Max
c1	2,000	-.0023069	.9880571	-3.500514	3.614182
c2	2,000	-.0077596	1.006025	-3.410111	3.423961
u	2,000	-.0011535	.4940286	-1.750257	1.807091
v	2,000	-.0071367	1	-3.954782	3.229851
county	2,000	4.5	2.873	0	9

age	2,000	36.208	8.28656	20	59
education	2,000	13.084	3.045912	10	20
married	2,000	.6705	.4701492	0	1
children	2,000	1.6445	1.398963	0	5
select	2,000	35.78556	14.98163	-14.45688	89.63869
-----+-----					
wagefull	2,000	21.31176	7.012038	-1.680425	45.80979
wage	1,343	23.69217	6.305374	5.88497	45.80979
lw	1,343	3.126703	.2865111	1.772402	3.824498
work	2,000	.6715	.4697852	0	1
lwf	2,000	2.099581	1.487519	0	3.824498

```
.
. regress lwf age married children education
```

Source	SS	df	MS	Number of obs	=	2,000
-----+-----				F(4, 1995)	=	134.21
Model	937.873188	4	234.468297	Prob > F	=	0.0000
Residual	3485.34135	1,995	1.74703827	R-squared	=	0.2120
-----+-----				Adj R-squared	=	0.2105
Total	4423.21454	1,999	2.21271363	Root MSE	=	1.3218

lwf	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+-----						
age	.0363624	.003862	9.42	0.000	.0287885	.0439362
married	.3188214	.0690834	4.62	0.000	.1833381	.4543046
children	.3305009	.0213143	15.51	0.000	.2887004	.3723015
education	.0843345	.0102295	8.24	0.000	.0642729	.1043961
_cons	-1.077738	.1703218	-6.33	0.000	-1.411765	-.7437105
-----+-----						

```
.
. tobit lwf age married children education, ll(0)
```

Refining starting values:

Grid node 0: log likelihood = -3563.7251

Fitting full model:

```
Iteration 0: log likelihood = -3563.7251
Iteration 1: log likelihood = -3368.4259
Iteration 2: log likelihood = -3350.1512
Iteration 3: log likelihood = -3349.9689
Iteration 4: log likelihood = -3349.9685
Iteration 5: log likelihood = -3349.9685
```

Tobit regression	Number of obs	=	2,000
	Uncensored	=	1,343
Limits: lower = 0	Left-censored	=	657
upper = +inf	Right-censored	=	0
	LR chi2(4)	=	461.85
	Prob > chi2	=	0.0000
Log likelihood = -3349.9685	Pseudo R2	=	0.0645

lwf	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
age	.052157	.0057457	9.08	0.000	.0408888	.0634252
married	.4841801	.1035191	4.68	0.000	.2811633	.687197
children	.4860021	.0317055	15.33	0.000	.4238228	.5481814
education	.1149492	.0150913	7.62	0.000	.0853528	.1445455
_cons	-2.807696	.2632573	-10.67	0.000	-3.323984	-2.291408
var(e.lwf)	3.507421	.1498785			3.225466	3.814024

```
. mfx compute, predict(pr(0,.))
```

Marginal effects after tobit

```
y = Pr(lwf>0) (predict, pr(0,.))
= .81920975
```

variable	dy/dx	Std. Err.	z	P> z	[95% C.I.]		X
age	.0073278	.00083	8.84	0.000	.005703	.008952	36.208
married*	.0706994	.01576	4.48	0.000	.039803	.101596	.6705
children	.0682813	.00479	14.26	0.000	.058899	.077663	1.6445
educat~n	.0161499	.00216	7.48	0.000	.011918	.020382	13.084

(*) dy/dx is for discrete change of dummy variable from 0 to 1

```
. mfx compute, predict(e(0,.))
```

Marginal effects after tobit

```
y = E(lwf|lwf>0) (predict, e(0,.))
= 2.3102021
```

variable	dy/dx	Std. Err.	z	P> z	[95% C.I.]		X
age	.0314922	.00347	9.08	0.000	.024695	.03829	36.208
married*	.2861047	.05982	4.78	0.000	.168855	.403355	.6705
children	.2934463	.01908	15.38	0.000	.256041	.330852	1.6445
educat~n	.0694059	.00912	7.61	0.000	.051531	.087281	13.084

(*) dy/dx is for discrete change of dummy variable from 0 to 1

10.4 INCIDENTAL TRUNCATION & SAMPLE SELECTION MODELS

```
. use womenwk..
```

```
. heckman lw education age children, select(age married children education) no
> log
```

Heckman selection model	Number of obs	=	2,000
(regression model with sample selection)	Selected	=	1,343
	Nonselected	=	657

Log likelihood = -1052.857	Wald chi2(3)	=	454.78
	Prob > chi2	=	0.0000

	lw	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
lw							
	education	.0397189	.0024525	16.20	0.000	.0349121	.0445256
	age	.0075872	.0009748	7.78	0.000	.0056767	.0094977
	children	-.0180477	.0064544	-2.80	0.005	-.0306981	-.0053973
	_cons	2.305499	.0653024	35.30	0.000	2.177509	2.43349
select							
	age	.0350233	.0042344	8.27	0.000	.0267241	.0433225
	married	.4547724	.0735876	6.18	0.000	.3105434	.5990014
	children	.4538372	.0288398	15.74	0.000	.3973122	.5103621
	education	.0565136	.0110025	5.14	0.000	.0349492	.0780781
	_cons	-2.478055	.1927823	-12.85	0.000	-2.855901	-2.100208
	/athrho	.3377674	.1152251	2.93	0.003	.1119304	.5636045
	/lnsigma	-1.375543	.0246873	-55.72	0.000	-1.423929	-1.327156
	rho	.3254828	.1030183			.1114653	.5106469
	sigma	.2527024	.0062385			.2407662	.2652304
	lambda	.0822503	.0273475			.0286501	.1358505

10.5 BIVARIATE PROBIT & PROBIT WITH SELECTION

```

. use hmda..
. replace fanfred = . if deny
(285 real changes made, 285 to missing)

. rename s6 loanamt

. rename vr vacancy

. rename mi med_income

. rename s50 appr_value

. rename s17 appl_income

. replace appl_income = appl_income/1000
(2,379 real changes made)

. rename s46 debt_inc_r

.
. summarize

```

Variable	Obs	Mean	Std. Dev.	Min	Max
seq	2,380	2328.934	1293.337	2	4509
s3	2,380	1	0	1	1
s4	2,380	1	0	1	1
s5	2,380	1.036555	.2027758	1	3
loanamt	2,380	139.1353	83.42097	2	980
s7	2,380	1.268908	.6605115	1	3
s9	2,380	1120	0	1120	1120
s11	2,380	.1743697	.3795069	0	1
s13	2,380	4.715126	.6991424	3	5
s14	2,379	5.702816	1.580592	1	8
s15	2,380	1.223109	.4342082	1	3
s16	2,379	2.513241	.9831064	1	8
appl_income	2,380	13.9406	116.9485	0	999.9994
s18	2,265	1.954967	3.044967	0	9
s19a	0				
s19b	0				
s19c	0				
s19d	0				
s20	2,380	1261.521	35488.59	1	999999.4
s23a	0				
s24a	2,380	.7605042	1.104747	0	8

s25a		2,380	19758.32	139160.4	0	999999.4
s26a		2,380	22275.29	147586.7	0	999999.4
s27a		2,380	.1163866	.3207553	0	1
s30a		2,380	4332.712	4663.801	0	81000
-----+						
s30c		2,380	1360.359	2130.56	0	41667
s31a		2,380	4914.016	5162.458	0	81000
s31c		2,380	1471.81	2358.868	0	41667
s32		2,380	1457.293	854.9721	0	10798
s33		2,380	1870.239	40962.55	0	999999.4
-----+						
s34		2,380	4.507271	87.8796	0	3908
s35		2,380	5134.612	70836.87	0	999999.4
s39		2,380	3782.993	61390.39	0	999999.4
s40		2,380	1.75	23.60479	0	666
s41		2,380	433.6915	20497.71	0	999999.4
-----+						
s42		2,380	1.721008	.5372816	1	4
s43		2,380	2.116387	1.666721	1	6
s44		2,380	.0735294	.2610584	0	1
s45		2,380	25.53461	9.665561	0	300
debt_inc_r		2,380	33.08136	10.72573	0	300
-----+						
s47		2,380	1.670588	.4736616	1	3
s48		2,380	2022.587	40956.14	6	999999.4
s49		0				
appr_value		2,380	198.5426	152.9863	25	4316
s51		2,380	1.711765	.4530364	1	2
-----+						
s52		2,380	.2256303	.4180845	0	1
s53		2,380	.0201681	.1406045	0	1
s54		0				
s55		2,380	.0294118	.1689932	0	1
s56		2,380	.047479	.2127058	0	1
-----+						
s57		2,380	102102.5	302843.7	0	999999.4
netw		2,380	253.0412	1072.576	-7919	28023
uria		2,380	3.774496	2.027062	1.8	10.6
rt dum		2,380	.0726891	.25968	0	1
bd		2,380	.444958	.4970656	0	1
-----+						
med_income		2,380	.8294118	.3762278	0	1
old		2,380	.4630252	.5054334	0	2
vacancy		2,380	.4365546	.4960626	0	1
school		2,380	9679.206	97847.79	5	999999.4
chval		2,380	6993.166	81710.9	-8.333333	999999.4
-----+						
dnotown		2,380	.0336134	.1802699	0	1
dprop		2,380	0	0	0	0
deny		2,380	.1197479	.3247347	0	1
fanfred		2,095	.3331742	.4714608	0	1
approve		2,380	.8802521	.3247347	0	1
-----+						
black		2,380	.142437	.3495712	0	1

. heckprob fanfred loanamt vacancy med_income appr_value, select(approve = blac

```
> k appl_income debt_inc_r) nolog
```

```

Probit model with sample selection          Number of obs   =      2,380
                                           Selected        =      2,095
                                           Nonselected     =       285

                                           Wald chi2(4)      =       80.69
Log likelihood = -2063.066                  Prob > chi2       =       0.0000

```

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
-----+-----						
fanfred						
loanamt	-.0026434	.0008029	-3.29	0.001	-.0042169	-.0010698
vacancy	-.2163307	.0609798	-3.55	0.000	-.3358489	-.0968125
med_income	.2671341	.0893349	2.99	0.003	.0920409	.4422273
appr_value	-.0014358	.0005099	-2.82	0.005	-.0024351	-.0004364
_cons	.1684824	.1182055	1.43	0.154	-.0631961	.4001608
-----+-----						
approve						
black	-.7343534	.081858	-8.97	0.000	-.8947921	-.5739148
appl_income	-.0006596	.000236	-2.80	0.005	-.0011221	-.0001971
debt_inc_r	-.0262367	.0036441	-7.20	0.000	-.0333791	-.0190944
_cons	2.236424	.1319308	16.95	0.000	1.977845	2.495004
-----+-----						
/athrho	-.6006599	.2712535	-2.21	0.027	-1.132307	-.0690128
-----+-----						
rho	-.537519	.1928812			-.8118074	-.0689034
-----+-----						
LR test of indep. eqns. (rho = 0):			chi2(1) =	4.99	Prob > chi2 = 0.0255	

10.5.1 Binomial Probit with Selection

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