

STATA Commands for Unobserved Effects Panel Data

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1 Introduction

Panel data or cross-sectional timeseries are observations on a panel of i units or cases over t time periods. Most panel data commands start with **xt**. For an overview type **help xt**. These notes present the annotated log of a STATA session demonstrating the use of many of these commands. The data sets used are those used in the STATA cross-sectional time series reference manual. This note should be regarded as an introduction to that manual and to the STATA on-line help files which give comprehensive descriptions of the facilities in STATA for cross-sectional time series analysis.

To obtain the optimum benefit from these notes I would recommend that one should work through the STATA session with a copy of Wooldridge available for reference. The emphasis here is on the implementation of the methods described in Chapter 10 of Wooldridge and no attempt is made to explain the theory set out there. Note the different fonts used for comments (this font), instructions in these comments (**help** **xt**) and for computer input/output (**help xt**).

. help xt

help for xt, iis, tis

manual: [XT] xt
dialogs: iis tsset

Cross-sectional time-series analysis

```
xt ... [, i(varname) t(varname) ... ]  
iis [varname] [, clear]  
tis [varname] [, clear]
```

Description

The `xt` series of commands provide tools for analyzing cross-sectional time-series (panel) datasets:

<code>help xtdes</code>	Describe pattern of <code>xt</code> data
<code>help xtsum</code>	Summarize <code>xt</code> data
<code>help xtab</code>	Tabulate <code>xt</code> data
<code>help xtdata</code>	Faster specification searches with <code>xt</code> data
<code>help xtline</code>	Line plots with <code>xt</code> data
<code>help xtreg</code>	Fixed-, between- and random-effects, and population-averaged linear models
<code>help xtregar</code>	Fixed- and random-effects linear models with an AR(1) disturbance
<code>help xtgls</code>	Panel-data models using GLS
<code>help xtpcse</code>	OLS or Prais-Winsten models with panel-corrected standard errors
<code>help xtrchh</code>	Hildreth-Houck random coefficients models
<code>help xtivreg</code>	Instrumental variables and two-stage least squares for panel-data models
<code>help xtabond</code>	Arellano-Bond linear, dynamic panel data estimator
<code>help xttobit</code>	Random-effects tobit models
<code>help xtintreg</code>	Random-effects interval data regression models
<code>help xtlogit</code>	Fixed-effects, random-effects, & population-averaged logit models
<code>help xtprobit</code>	Random-effects and population-averaged probit models
<code>help xtclglog</code>	Random-effects and population-averaged cloglog models
<code>help xtpoisson</code>	Fixed-effects, random-effects, & population-averaged Poisson models
<code>help xtnbreg</code>	Fixed-effects, random-effects, & population-averaged negative binomial models
<code>help xtgee</code>	Population-averaged panel-data models using GEE

Each observation in a cross-sectional time-series (`xt`) dataset is an observation on `x` for unit `i` at time `t`.

`iis` is related to the `i()` option of the other `xt` commands. Command `iis` or option `i()` sets the name of the variable corresponding to index `i`.

`tis` is similarly related to the `t()` option. Command `tis` or option `t()` sets the name of the variable corresponding to index `t`.

Some `xt` commands use time-series operators in their internal calculations and thus require that your data be `tsset`; see help `tsset`. For instance, since `xtabond` uses time-series operators in its internal calculations, you must `tsset` your data before using it. The particular help file will indicate if `tsset` is required for the command.

Options

`i(varname)` specifies the variable name corresponding to index `i`. This must be a single, numeric variable, although whether it takes on the values 1, 2, 3 or 1, 7, 9, etc., is irrelevant. (If the identifying variable is a string, use `egen`'s `group()` function to make a numeric variable; see help `egen`.)

`t(varname)` specifies the variable name corresponding to index `t`. This must be a single, numeric variable.

`clear` removes the definition of `i` or `t`. For instance, typing "tis, clear" makes Stata forget the identity of the `t()` variable.

Remarks

Once `i()` and `t()` have been specified, either by option or by the `iis` and `tis` commands, they need not be specified again except to change the variable's identity.

`iis` and `tis`, without arguments, list the current name of the index variable.

Example

An `xt` dataset:

pid	yr_visit	fev	age	sex	height	smokes
1071	1991	1.21	25	1	69	0
1071	1992	1.52	26	1	69	0
1071	1993	1.32	28	1	68	0
1072	1991	1.33	18	1	71	1
1072	1992	1.18	20	1	71	1

```
1072      1993    1.19    21     1      71      0
```

The other xt commands need to know the identities of the variables identifying patient and time. You could type

```
. iis pid  
. tis yr_visit
```

Also see

Manual: [XT] intro,
[XT] xt

Online: help for xtabond, xtcloglog, xtdta, xtdes, xtgee, xtgls,
xtintreg, xtivreg, xtline, xtlogit, xtnbreg, xtpcse, xtpoisson,
xtprobit, xtrchh, xtreg, xtregar, xtsum, xttab, xtobit; tsset

Load the data set nlswork.dta

```
. use nlswork, clear  
. describe
```

Contains data

National Longitudinal Survey.
Young Women 14-26 years of age
in 1968

18 Feb 2005 22:17

obs: 28,534
vars: 21
size: 1,055,758

variable	name	storage	display	value	variable label
		type	format	label	
idcode		int	%8.0g		NLS id
year		byte	%8.0g		interview year
birth_yr		byte	%8.0g		birth year
age		byte	%8.0g		age in current year
race		byte	%8.0g		1=white, 2=black, 3=other
msp		byte	%8.0g		1 if married, spouse present
nev_mar		byte	%8.0g		1 if never yet married
grade		byte	%8.0g		current grade completed
collgrad		byte	%8.0g		1 if college graduate
not_smsa		byte	%8.0g		1 if not SMSA
c_city		byte	%8.0g		1 if central city
south		byte	%8.0g		1 if south
ind_code		byte	%8.0g		industry of employment
occ_code		byte	%8.0g		occupation
union		byte	%8.0g		1 if union
wks_ue		byte	%8.0g		weeks unemployed last year

```

ttl_exp      float %9.0g          total work experience
tenure       float %9.0g          job tenure, in years
hours        int   %8.0g           usual hours worked
wks_work     int   %8.0g           weeks worked last year
ln_wage      float %9.0g          ln(wage/GNP deflator)
-----
```

Sorted by: idcode year

To start one must set the indices i (units) and t (time). As already described this can be done using the **iis tis** commands, **i()** **t()** options or the **tsset** command. Examples of the commands follow.

```

. iis idcode

. tis year

.

. iis
i() is idcode

. tis
t() is year

. iis, clear

. iis
(i() has not been defined)

.

. tsset idcode year
    panel variable: idcode, 1 to 5159
    time variable: year, 68 to 88, but with gaps

. tsset
    panel variable: idcode, 1 to 5159
    time variable: year, 68 to 88, but with gaps
```

xtdes describes the participation pattern of panel data. We have 4711 women in the survey. The maximum number of years over which any women is observed is 15. the most common pattern is participation in only the first year (136 or 2.89% are observed in this pattern). The bottom line of the table give the totals for participation patterns not observed.

```

. xtdes

idcode: 1, 2, ..., 5159                      n =      4711
year: 68, 69, ..., 88                          T =      15
                                                Delta(year) = 1; (88-68)+1 = 21
                                                (idcode*year uniquely identifies each observation)
```

Distribution of T_i:	min	5%	25%	50%	75%	95%	max
	1	1	3	5	9	13	15

Freq.	Percent	Cum.		Pattern
136	2.89	2.89		1.....
114	2.42	5.31	1
89	1.89	7.20	1.11
87	1.85	9.04	11
86	1.83	10.87		111111.1.11.1.11.1.11
61	1.29	12.16	11.1.11
56	1.19	13.35		11.....
54	1.15	14.50	1.1.11
54	1.15	15.64	1.11.1.11.1.11
3974	84.36	100.00		(other patterns)
4711	100.00			XXXXXX.X.XX.X.XX.X.XX

. xtdes, pattern(20)

```

idcode: 1, 2, ..., 5159                               n =      4711
year: 68, 69, ..., 88                                T =       15
Delta(year) = 1; (88-68)+1 = 21
(idcode*year uniquely identifies each observation)

```

Distribution of T_i:	min	5%	25%	50%	75%	95%	max
	1	1	3	5	9	13	15

Freq.	Percent	Cum.		Pattern
136	2.89	2.89		1.....
114	2.42	5.31	1
89	1.89	7.20	1.11
87	1.85	9.04	11
86	1.83	10.87		111111.1.11.1.11.1.11
61	1.29	12.16	11.1.11
56	1.19	13.35		11.....
54	1.15	14.50	1.1.11
54	1.15	15.64	1.11.1.11.1.11
49	1.04	16.68	11.1.11.1.11
45	0.96	17.64	1.11.1.11.1.11
43	0.91	18.55		1111.....
42	0.89	19.44		...1.....
40	0.85	20.29	1.1.11.1.11.1.11
38	0.81	21.10	11.1.11.1.11.1.11
38	0.81	21.91		111.....
34	0.72	22.63		..1111.1.11.1.11.1.11
31	0.66	23.29	1...
30	0.64	23.92	1.1.11.1.11
29	0.62	24.54		...111.1.11.1.11.1.11

```

3555      75.46  100.00 | (other patterns)
-----+-----
4711      100.00          | XXXXXX.X.XX.X.XX.X.XX

```

xtsum generalizes summarize by reporting means and standard for panel data. It differs from summarize in that it decomposes the standard deviation into between and within components.

```
. summ hours
```

Variable	Obs	Mean	Std. Dev.	Min	Max
hours	28467	36.55956	9.869623	1	168

```
. xtsum hours
```

Variable	Mean	Std. Dev.	Min	Max	Observations
hours overall	36.55956	9.869623	1	168	N = 28467
between		7.846585	1	83.5	n = 4710
within		7.520712	-2.154726	130.0596	T-bar = 6.04395

```
. xtsum birth_yr /* Time invariant variable */
```

Variable	Mean	Std. Dev.	Min	Max	Observations
birth_yr overall	48.08509	3.012837	41	54	N = 28534
between		3.051795	41	54	n = 4711
within		0	48.08509	48.08509	T-bar = 6.05689

xttab generalizes tabulate by performing one-way tabulations and by decomposing counts into between and within components in panel data.

```
. summ msp
```

Variable	Obs	Mean	Std. Dev.	Min	Max
msp	28518	.6029175	.4893019	0	1

```
. tab msp
```

1 if married, spouse present	Freq.	Percent	Cum.
0	11,324	39.71	39.71
1	17,194	60.29	100.00

```
Total | 28,518 100.00
```

```
. xttab msp
```

msp	Overall		Between		Within
	Freq.	Percent	Freq.	Percent	Percent
0	11324	39.71	3113	66.08	55.06
1	17194	60.29	3643	77.33	71.90
Total	28518	100.00	6756	143.41	64.14
		(n = 4711)			

xttrans is another generalization of tabulate. It reports changes in a single categorical variable over time.

```
. xttrans msp
```

1 if				
married,	1 if married, spouse			
spouse	present			
present	0	1	Total	
0	80.49	19.51	100.00	
1	7.96	92.04	100.00	
Total	37.11	62.89	100.00	

```
. xttrans msp, freq /* Does not normalize for missing time periods */
```

1 if				
married,	1 if married, spouse			
spouse	present			
present	0	1	Total	
0	7,697	1,866	9,563	
	80.49	19.51	100.00	
1	1,133	13,100	14,233	
	7.96	92.04	100.00	
Total	8,830	14,966	23,796	
	37.11	62.89	100.00	

```
. * Rectangularize the data  
. fillin idcode year
```

```
. xttrans msp, freq

    1 if |
married, | 1 if married, spouse
    spouse |      present
  present |        0          1 |      Total
-----+-----+-----+
    0 |     6,792     1,446 |   8,238
      |     82.45     17.55 | 100.00
-----+-----+-----+
    1 |     813     10,954 | 11,767
      |     6.91     93.09 | 100.00
-----+-----+-----+
  Total |    7,605    12,400 | 20,005
      |    38.02    61.98 | 100.00
```

xtline draws line plots for panel data.

2 Estimation using **xtreg**

The basic linear unobserved effects panel data model may is

$$y_{it} = \mathbf{X}_{it}\beta + c_i + u_{it} \quad (1)$$

(For a full explanation of the symbols see Wooldridge page 251, etc.). In equation 1 u_i is the unit specific residual and differs *between* units but not across time *within* units. Averaging equation 1 over time we get

$$\bar{y}_i = \bar{\mathbf{X}}_i\beta + c_i + \bar{u}_i \quad (2)$$

Subtracting equation 2 from equation 1 gives equation 3

$$(y_{it} - \bar{y}_i) = (\mathbf{X}_{it} - \bar{\mathbf{X}}_i)\beta + (u_{it} - \bar{u}_i) \quad (3)$$

These three equations form the basis for the various ways of estimating β .

xtreg ..., fe gives the fixed effects or within estimator of β and is derived from equation 3. It is equivalent to performing *OLS* on equation pd3.

xtreg ..., be gives the between effects and corresponds to *OLS* estimation of equation 2.

xtreg ..., re gives the random effects estimator and is a weighted average of the within and between effects estimator. The random effects estimator is equivalent to estimating

$$(y_{it} - \theta\bar{y}_i) = (\mathbf{X}_{it} - \theta\bar{\mathbf{X}}_i)\beta + (1 - \theta)c_i + (u_{it} - \theta\bar{u}_i) \quad (4)$$

where θ is a function of σ_c^2 and σ_u^2 .

xtreg ..., mle produces maximum likelihood estimates of the random effects estimator.

For other options available with the **xtreg** command see the on-line help files or the STATA manuals.

```

. tsset idcode year
    panel variable: idcode, 1 to 5159
    time variable: year, 68 to 88, but with gaps

. qui gen age2 = age^2

. qui gen ttl_exp2 = ttl_exp^2

. qui gen tenure2 = tenure^2

. gen byte black = race==2

.

. * OLS
. regress ln_w grade age* ttl_exp* tenure* black not_smsa south

Source |      SS          df          MS
-----+-----
Model |  2402.22796     10   240.222796
Residual |  4011.63592 28080   .142864527
-----+-----
Total |  6413.86388 28090   .228332641

Number of obs = 28091
F( 10, 28080) = 1681.47
Prob > F      = 0.0000
R-squared      = 0.3745
Adj R-squared = 0.3743
Root MSE       = .37797

-----+-----
ln_wage |      Coef.      Std. Err.          t      P>|t| [95% Conf. Interval]
-----+-----
grade |   .0629238   .0010313     61.01  0.000   .0609024   .0649452
age |    .038598   .003467     11.13  0.000   .0318025   .0453935
age2 |   -.0007082   .0000563    -12.57  0.000  -.0008186  -.0005978
ttl_exp |   .0211279   .002335      9.05  0.000   .0165511   .0257046
ttl_exp2 |   .0004473   .0001246      3.59  0.000   .0002031   .0006916
tenure |   .0473687   .0019626     24.14  0.000   .0435219   .0512156
tenure2 |   -.002027   .0001338    -15.15  0.000  -.0022893  -.0017648
black |   -.0699386   .0053207    -13.14  0.000  -.0803673  -.0595098
not_smsa |  -.1720455   .0051675    -33.29  0.000  -.182174   -.161917
south |  -.1003387   .0048938    -20.50  0.000  -.1099308  -.0907467
_cons |   .2472833   .0493319      5.01  0.000   .1505903   .3439762

-----+-----

. * Fixed-effects model (within-group estimator)
. xtreg ln_w grade age* ttl_exp* tenure* black not_smsa south, fe

Fixed-effects (within) regression
Group variable (i): idcode
Number of obs      = 28091
Number of groups  = 4697

R-sq: within  = 0.1727
      between = 0.3505
Obs per group: min = 1
                           avg = 6.0

```

```

overall = 0.2625                         max =      15
                                                F(8, 23386)      =  610.12
corr(u_i, Xb)  = 0.1936                  Prob > F      = 0.0000
-----
ln_wage |      Coef.    Std. Err.      t    P>|t|   [95% Conf. Interval]
-----+
grade | (dropped)
      age | .0359987  .0033864   10.63  0.000   .0293611  .0426362
      age2 | -.000723  .0000533  -13.58  0.000  -.0008274  -.0006186
      ttl_exp | .0334668  .0029653   11.29  0.000   .0276545  .039279
      ttl_exp2 | .0002163  .0001277    1.69  0.090  -.0000341  .0004666
      tenure | .0357539  .0018487   19.34  0.000   .0321303  .0393775
      tenure2 | -.0019701  .000125  -15.76  0.000  -.0022151  -.0017251
      black | (dropped)
      not_smsa | -.0890108  .0095316   -9.34  0.000  -.1076933  -.0703282
      south | -.0606309  .0109319   -5.55  0.000  -.0820582  -.0392036
      _cons | 1.03732  .0485546   21.36  0.000   .9421497  1.13249
-----+
sigma_u | .35562203
sigma_e | .29068923
rho | .59946283 (fraction of variance due to u_i)
-----+
F test that all u_i=0:   F(4696, 23386) =      5.13      Prob > F = 0.0000

```

Version 8 of STATA does not have a robust option on the `xtreg` command. (This option is available in version 9. In version 8 use the `areg` command.

```
. areg ln_w grade age* ttl_exp* tenure* black not_smsa south, ///
absorb(idcode) cluster(idcode)
```

Regression with robust standard errors	Number of obs = 28091
	F(9, 4696) = 202.74
	Prob > F = 0.0000
	R-squared = 0.6919
	Adj R-squared = 0.6299
	Root MSE = .2907

(standard errors adjusted for clustering on idcode)

ln_wage		Robust				
		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
grade	-.0147384	237927.9	-0.00	1.000	-466450.3	466450.3
age	.0359987	.005743	6.27	0.000	.0247397	.0472576
age2	-.000723	.0000926	-7.81	0.000	-.0009045	-.0005414
ttl_exp	.0334668	.0044589	7.51	0.000	.0247251	.0422084
ttl_exp2	.0002163	.0001932	1.12	0.263	-.0001625	.0005951
tenure	.0357539	.0027049	13.22	0.000	.0304511	.0410567

tenure2	-.0019701	.0001859	-10.60	0.000	-.0023345	-.0016057
black	(dropped)					
not_smsa	-.0890108	.0150819	-5.90	0.000	-.1185784	-.0594432
south	-.0606309	.0179022	-3.39	0.001	-.0957276	-.0255341
_cons	1.222086	2982750	0.00	1.000	-5847589	5847592
<hr/>						
idcode	absorbed					(4697 categories)

Note that STATA has no direct command for two way fixed effects. If you wish to also introduce a second set of fixed effects for, say, time periods create a set of appropriate dummy variables for inclusion in your regressions and use a one way estimator.

```
. * Between-group estimator
. xtreg ln_w grade age* ttl_exp* tenure* black not_smsa south, be
```

Between regression (regression on group means)	Number of obs	=	28091
Group variable (i): idcode	Number of groups	=	4697
R-sq: within = 0.1591	Obs per group: min	=	1
between = 0.4900	avg	=	6.0
overall = 0.3695	max	=	15
sd(u_i + avg(e_i.))= .3036114	F(10, 4686)	=	450.23
	Prob > F	=	0.0000

ln_wage	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
grade	.0607602	.0020006	30.37	0.000	.0568382 .0646822
age	.0323158	.0087251	3.70	0.000	.0152105 .0494211
age2	-.0005997	.0001429	-4.20	0.000	-.0008799 -.0003194
ttl_exp	.0138853	.0056749	2.45	0.014	.0027598 .0250108
ttl_exp2	.0007342	.0003267	2.25	0.025	.0000936 .0013747
tenure	.0698419	.0060729	11.50	0.000	.0579361 .0817476
tenure2	-.0028756	.0004098	-7.02	0.000	-.0036789 -.0020722
black	-.0564167	.0105131	-5.37	0.000	-.0770272 -.0358061
not_smsa	-.1860406	.0112495	-16.54	0.000	-.2080949 -.1639862
south	-.0993378	.010136	-9.80	0.000	-.1192091 -.0794665
_cons	.3339113	.1210434	2.76	0.006	.0966093 .5712133

```
. * Random-effects model (GLS estimator)
. xtreg ln_w grade age* ttl_exp* tenure* black not_smsa south, re
```

Random-effects GLS regression	Number of obs	=	28091
Group variable (i): idcode	Number of groups	=	4697
R-sq: within = 0.1715	Obs per group: min	=	1
between = 0.4784	avg	=	6.0
overall = 0.3708	max	=	15

```

Random effects u_i ~ Gaussian
corr(u_i, X)      = 0 (assumed)          Wald chi2(10)      = 9244.87
                                                               Prob > chi2      = 0.0000

-----+
ln_wage |   Coef.   Std. Err.      z   P>|z|   [95% Conf. Interval]
-----+
grade |  .0646499  .0017811  36.30  0.000   .0611589  .0681408
age |   .036806  .0031195  11.80  0.000   .0306918  .0429201
age2 |  -.0007133  .00005  -14.27  0.000  -.0008113  -.0006153
ttl_exp |  .0290207  .0024219  11.98  0.000   .0242737  .0337676
ttl_exp2 |  .0003049  .0001162   2.62  0.009   .000077  .0005327
tenure |   .039252  .0017555  22.36  0.000   .0358114  .0426927
tenure2 |  -.0020035  .0001193  -16.80  0.000  -.0022373  -.0017697
black |  -.0530532  .0099924  -5.31  0.000  -.0726379  -.0334685
not_smsa |  -.1308263  .0071751  -18.23  0.000  -.1448891  -.1167634
south |  -.0868927  .0073031  -11.90  0.000  -.1012066  -.0725788
_cons |   .2387209  .0494688    4.83  0.000   .1417639  .335678
-----+
sigma_u |  .25790313
sigma_e |  .29069544
rho |  .44043812 (fraction of variance due to u_i)
-----+
. estimates hold re
.
. * Random-effects model (Gaussian ML or fully iterated GLS estimator)
. xtreg ln_w grade age* ttl_exp* tenure* black not_smsa south, mle

Fitting constant-only model:
Iteration 0: log likelihood = -13690.161
Iteration 1: log likelihood = -12819.317
Iteration 2: log likelihood = -12662.039
Iteration 3: log likelihood = -12649.744
Iteration 4: log likelihood = -12649.614

Fitting full model:
Iteration 0: log likelihood = -8922.145
Iteration 1: log likelihood = -8853.6409
Iteration 2: log likelihood = -8853.4255
Iteration 3: log likelihood = -8853.4254

Random-effects ML regression                               Number of obs      =     28091
Group variable (i): idcode                            Number of groups   =      4697
                                                               Obs per group: min =         1
                                                               avg =       6.0
                                                               max =       15

```

					LR chi2(10) =	7592.38
					Prob > chi2 =	0.0000
Log likelihood = -8853.4254						

ln_wage	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
grade .0646093	.0017372	37.19	0.000	.0612044	.0680142	
age .0368531	.0031226	11.80	0.000	.030733	.0429732	
age2 -.0007132	.0000501	-14.24	0.000	-.0008113	-.000615	
ttl_exp .0288196	.0024143	11.94	0.000	.0240877	.0335515	
ttl_exp2 .000309	.0001163	2.66	0.008	.0000811	.0005369	
tenure .0394371	.0017604	22.40	0.000	.0359868	.0428875	
tenure2 -.0020052	.0001195	-16.77	0.000	-.0022395	-.0017709	
black -.0533394	.0097338	-5.48	0.000	-.0724172	-.0342615	
not_smsa -.1323433	.0071322	-18.56	0.000	-.1463221	-.1183644	
south -.0875599	.0072143	-12.14	0.000	-.1016998	-.0734201	
_cons .2390837	.0491902	4.86	0.000	.1426727	.3354947	
/sigma_u .2485556	.0035017	70.98	0.000	.2416925	.2554187	
/sigma_e .2918458	.001352	215.87	0.000	.289196	.2944956	

rho .4204033	.0074828			.4057959	.4351212	

Likelihood-ratio test of sigma_u=0: chibar2(01)= 7339.84 Prob>=chibar2 = 0.000

3 Testing after xtreg

```

*
. /* After xtreg, re */
.
. estimates unhold re
.
. * Breusch & Pagan score test for random effects
. xttest0

```

Breusch and Pagan Lagrangian multiplier test for random effects:

```
ln_wage[idcode,t] = Xb + u[idcode] + e[idcode,t]
```

Estimated results:

	Var	sd = sqrt(Var)
ln_wage	.2283326	.4778416
e	.0845038	.2906954
u	.066514	.2579031

```

Test:  Var(u) = 0
chi2(1) = 14779.98
Prob > chi2 = 0.0000

. * Hausman specification test (compares fe and re)

.qui xtreg ln_wage grade age age2 ttl_exp ttl_exp2 tenure tenure2 not_smsa south,fe
F(4696, 23386) = 5.19          Prob > F = 0.0000

.estimates store fe

.qui xtreg ln_wage grade age age2 ttl_exp ttl_exp2 tenure tenure2 not_smsa south, re
.estimates store re

.hausman fe re

      ---- Coefficients ----
      |      (b)          (B)          (b-B)      sqrt(diag(V_b-V_B))
      |      fe            re           Difference   S.E.
-----+-----
age | .0359987       .0363062     -.0003075    .0013183
age2 | -.000723       -.000705     -.000018    .0000184
ttl_exp | .0334668       .0292321     .0042347    .0017085
ttl_exp2 | .0002163       .0002946     -.0000783    .0000529
tenure | .0357539       .0390983     -.0033444    .0005789
tenure2 | -.0019701      -.0020014     .0000313    .0000372
not_smsa | -.0890108      -.1268961     .0378853    .0063038
south | -.0606309      -.094716     .0340851    .008259
-----+
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(8) = (b-B)'[(V_b-V_B)^(-1)](b-B)
= 142.53
Prob>chi2 = 0.0000

```

4 Prediction after xtreg

```
. * Syntax:
```

```
. * predict [type] newvarname [if exp] [in range] [, statistic nooffset]
. * where statistic is:
. *     xb      fitted values (the default)
. *     stdp   standard error of the fitted values
. *     ue     the combined residuals
. *     xbu    prediction, including effect
. *     u      the fixed effect component
. *     e      the random error component
.
. predict xb /* computes the linear predictor (the default) */
(option xb assumed; fitted values)
(443 missing values generated)

. predict stdp, stdp
(443 missing values generated)
```

5 Faster estimation of alternative models using `xtdata`

xtdata *varlist* ... produces a converted data set of the variables specified or, if *varlist* is not specified, all the variables in the data. Once converted, Stata's ordinary regress command may be used to perform various panel data regressions more quickly than use **xtreg**. Before using **xtdata** you must eliminate any variables that you do not intend to use and that have missing values. After converting the data, with **xtdata** you may form linear transformations of the regressors. All nonlinear transformations of the data must be done before conversion. .

```
. xtdat ln_w grade age* ttl_exp* tenure* black not_smsa south, fe clear
```

```
. regress ln_w grade age ttl_exp tenure black not_smsa south
```

Source	SS	df	MS	Number of obs = 28091		
Model	356.233455	6	59.3722424	F(6, 28084) = 820.44		
Residual	2032.33275	28084	.072366214	Prob > F = 0.0000		
Total	2388.5662	28090	.085032617	R-squared = 0.1491		
				Adj R-squared = 0.1490		
				Root MSE = .26901		
ln_wage	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
grade	.0375399	5.04e+08	0.00	1.000	-9.88e+08	9.88e+08
age	-.0026787	.0007876	-3.40	0.001	-.0042224	-.0011349
ttl_exp	.0287709	.0013209	21.78	0.000	.0261819	.0313599
tenure	.0114355	.0008422	13.58	0.000	.0097847	.0130863
black	(dropped)					
not_smsa	-.0921689	.0088194	-10.45	0.000	-.1094553	-.0748825
south	-.0633396	.0101132	-6.26	0.000	-.083162	-.0435172

```

_cons | 1.121064 6.32e+09 0.00 1.000 -1.24e+10 1.24e+10
-----
. regress ln_w grade age* ttl_exp* tenure* black not_smsa south

      Source |       SS        df        MS
-----+-----+
      Model | 412.443881    9 45.8270979
      Residual | 1976.12232 28081 .07037222
-----+-----+
      Total | 2388.5662 28090 .085032617

      Number of obs = 28091
      F( 9, 28081) = 651.21
      Prob > F = 0.0000
      R-squared = 0.1727
      Adj R-squared = 0.1724
      Root MSE = .26528

-----+
      ln_wage |     Coef.   Std. Err.      t    P>|t|    [95% Conf. Interval]
-----+
      grade | -.0147384 4.97e+08 -0.00 1.000 -9.75e+08 9.75e+08
      age | .0359987 .0030904 11.65 0.000 .0299414 .0420559
      age2 | -.000723 .0000486 -14.88 0.000 -.0008183 -.0006277
      ttl_exp | .0334668 .0027061 12.37 0.000 .0281626 .0387709
      ttl_exp2 | .0002163 .0001166 1.86 0.064 -.0000122 .0004448
      tenure | .0357539 .0016871 21.19 0.000 .0324471 .0390607
      tenure2 | -.0019701 .0001141 -17.27 0.000 -.0021937 -.0017465
      black | (dropped)
      not_smsa | -.0890108 .0086984 -10.23 0.000 -.10606 -.0719616
      south | -.0606309 .0099763 -6.08 0.000 -.0801849 -.0410769
      _cons | 1.222086 6.23e+09 0.00 1.000 -1.22e+10 1.22e+10
-----+
.

. xtdta ln_w grade age* ttl_exp* tenure* black not_smsa south, re ratio(.95) clear

----- theta -----
      min      5%      median      95%      max
0.2750  0.2750    0.5741    0.7198  0.7377

. /* (ratio is the ratio of the std. dev. of the individual effect and the
. * random error)
. regress ln_w constant grade age ttl_exp tenure black not_smsa south, noconstant

      Source |       SS        df        MS
-----+-----+
      Model | 11775.6413    6 1962.60688
      Residual | 2032.33275 28085 .072363637
-----+-----+
      Total | 13807.974 28091 .491544411

      Number of obs = 28091
      F( 6, 28085) = 27121.45
      Prob > F = 0.0000
      R-squared = 0.8528
      Adj R-squared = 0.8528
      Root MSE = .269

-----+
      ln_wage |     Coef.   Std. Err.      t    P>|t|    [95% Conf. Interval]
-----+
      constant | (dropped)

```

```

grade |   .1269649   .0013954    90.99    0.000     .1242299    .1296999
      age |  -.0026787   .0007876   -3.40    0.001     -.0042224   -.001135
      ttl_exp |   .0287709   .0013209   21.78    0.000     .026182    .0313599
      tenure |   .0114355   .0008422   13.58    0.000     .0097847    .0130863
      black | (dropped)
      not_smsa |  -.0921689   .0088192   -10.45    0.000     -.109455   -.0748828
      south |  -.0633396   .010113    -6.26    0.000     -.0831616   -.0435175
-----
. regress ln_w constant grade age* ttl_exp* tenure* black not_smsa south, noconstant

      Source |      SS       df      MS
-----+-----+
      Model | 11831.8517      9 1314.65019
      Residual | 1976.12232 28082  .070369714
-----+-----+
      Total | 13807.974 28091  .491544411

      Number of obs = 28091
      F( 9, 28082) = 18682.05
      Prob > F    = 0.0000
      R-squared    = 0.8569
      Adj R-squared = 0.8568
      Root MSE     = .26527

-----+
      ln_wage |      Coef.    Std. Err.      t    P>|t| [95% Conf. Interval]
-----+-----+
      constant | (dropped)
      grade |   .0827449   .0035478    23.32    0.000     .0757911    .0896987
      age |   .0359987   .0030903    11.65    0.000     .0299415    .0420558
      age2 |  -.000723   .0000486   -14.88    0.000     -.0008183   -.0006277
      ttl_exp |   .0334668   .0027061    12.37    0.000     .0281627    .0387708
      ttl_exp2 |  .0002163   .0001166     1.86    0.064     -.0000122    .0004447
      tenure |   .0357539   .0016871    21.19    0.000     .0324472    .0390606
      tenure2 |  -.0019701   .0001141   -17.27    0.000     -.0021937   -.0017465
      black | (dropped)
      not_smsa |  -.0890108   .0086982   -10.23    0.000     -.1060597   -.0719619
      south |  -.0606309   .0099761    -6.08    0.000     -.0801845   -.0410772
-----+

```

6 More general error structures

xtpcse and **xtgls** estimate linear panel data models using feasible GLS. **xtpcse** computes OLS estimates with panel-corrected standard errors, while **xtgls** computes feasible GLS estimates. These commands allow estimation in the presence of AR(1) autocorrelation within panels, as well as heteroscedasticity or cross-sectional correlation across panels. In the case of cross-sectional correlation, **xtgls** requires $T > n$.

```

. use invest2, clear

. tsset company time
      panel variable: company, 1 to 5
      time variable: time, 1 to 20

```

```

. * OLS with panel-corrected standard errors
. xtpcse invest market stock /* Heterosk. and contemp. correlation (the default) */

```

Linear regression, correlated panels corrected standard errors (PCSEs)

Group variable:	company	Number of obs	=	100
Time variable:	time	Number of groups	=	5
Panels:	correlated (balanced)	Obs per group: min	=	20
Autocorrelation:	no autocorrelation	avg	=	20
		max	=	20
Estimated covariances	= 15	R-squared	=	0.7789
Estimated autocorrelations	= 0	Wald chi2(2)	=	755.43
Estimated coefficients	= 3	Prob > chi2	=	0.0000

Panel-corrected						
	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
market	.1050854	.0083183	12.63	0.000	.0887818	.1213891
stock	.3053655	.0330427	9.24	0.000	.240603	.3701281
_cons	-48.02974	10.81437	-4.44	0.000	-69.2255	-26.83397

```

. xtpcse invest market stock, corr(ar1) /* Heterosk., contemp. correlation and AR(1) autocorrelation
(note: estimates of rho outside [-1,1] bounded to be in the range [-1,1])

```

Prais-Winsten regression, correlated panels corrected standard errors (PCSEs)

Group variable:	company	Number of obs	=	100
Time variable:	time	Number of groups	=	5
Panels:	correlated (balanced)	Obs per group: min	=	20
Autocorrelation:	common AR(1)	avg	=	20
		max	=	20
Estimated covariances	= 15	R-squared	=	0.5909
Estimated autocorrelations	= 1	Wald chi2(2)	=	124.32
Estimated coefficients	= 3	Prob > chi2	=	0.0000

Panel-corrected						
	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
market	.093367	.0125705	7.43	0.000	.0687294	.1180046
stock	.354706	.0571221	6.21	0.000	.2427486	.4666633
_cons	-39.39866	40.22722	-0.98	0.327	-118.2426	39.44524
rho	.8530976					

```

. xtpcse invest market stock, corr(psar1) rhotype(tscorr)

```

Prais-Winsten regression, correlated panels corrected standard errors (PCSEs)

Group variable:	company	Number of obs	=	100
Time variable:	time	Number of groups	=	5
Panels:	correlated (balanced)	Obs per group: min	=	20
Autocorrelation:	panel-specific AR(1)	avg	=	20
		max	=	20
Estimated covariances	= 15	R-squared	=	0.8734
Estimated autocorrelations	= 5	Wald chi2(2)	=	483.87
Estimated coefficients	= 3	Prob > chi2	=	0.0000
<hr/>				
	Panel-corrected			
	Coef.	Std. Err.	z	P> z
				[95% Conf. Interval]
<hr/>				
market	.0976686	.009442	10.34	0.000
stock	.3726526	.0384121	9.70	0.000
_cons	-46.95183	16.78803	-2.80	0.005
<hr/>				
rhos =	.4735903	.704354	.8977688	.5249498
				.8558518
<hr/>				

. xtpcse invest market stock, hetonly /* Heterosk., no contemp. correlation */

Linear regression, heteroskedastic panels corrected standard errors

Group variable:	company	Number of obs	=	100
Time variable:	time	Number of groups	=	5
Panels:	heteroskedastic (balanced)	Obs per group: min	=	20
Autocorrelation:	no autocorrelation	avg	=	20
		max	=	20
Estimated covariances	= 5	R-squared	=	0.7789
Estimated autocorrelations	= 0	Wald chi2(2)	=	720.01
Estimated coefficients	= 3	Prob > chi2	=	0.0000
<hr/>				
	Het-corrected			
	Coef.	Std. Err.	z	P> z
				[95% Conf. Interval]
<hr/>				
market	.1050854	.0090625	11.60	0.000
stock	.3053655	.0409468	7.46	0.000
_cons	-48.02974	14.20367	-3.38	0.001
<hr/>				

. xtpcse invest market stock, hetonly corr(ar1) /* Heterosk. and AR(1) autocorr., no cont
(note: estimates of rho outside [-1,1] bounded to be in the range [-1,1])

Prais-Winsten regression, heteroskedastic panels corrected standard errors

```

Group variable: company Number of obs      = 100
Time variable: time   Number of groups    = 5
Panels:        heteroskedastic (balanced) Obs per group: min = 20
Autocorrelation: common AR(1)           avg = 20
                                         max = 20
Estimated covariances     = 5          R-squared       = 0.5909
Estimated autocorrelations = 1          Wald chi2(2)    = 120.57
Estimated coefficients    = 3          Prob > chi2    = 0.0000

```

	Het-corrected					
	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
market	.093367	.0128727	7.25	0.000	.0681369	.1185971
stock	.354706	.0587917	6.03	0.000	.2394763	.4699357
_cons	-39.39866	37.19875	-1.06	0.290	-112.3069	33.50954
rho	.8530976					

```

. * Feasible GLS
. xtgls invest market stock, panel(iid) corr(indep) nmk
```

Cross-sectional time-series FGLS regression

```

Coefficients: generalized least squares
Panels:        homoskedastic
Correlation:   no autocorrelation
```

```

Estimated covariances     = 1          Number of obs      = 100
Estimated autocorrelations = 0          Number of groups    = 5
Estimated coefficients    = 3          Time periods      = 20
Log likelihood            = -624.9928 Wald chi2(2)      = 341.63
                                         Prob > chi2    = 0.0000
```

	invest	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
market		.1050854	.0113778	9.24	0.000	.0827853 .1273855
stock		.3053655	.0435078	7.02	0.000	.2200918 .3906393
_cons		-48.02974	21.48016	-2.24	0.025	-90.13009 -5.929387

```
. * (same as regress ..., nmk uses n - k to normalize the RSS)
```

```
. * xtgls invest market stock, i(company) panel(hetero) /* Heterosk., no contemp. correlation
```

Cross-sectional time-series FGLS regression

Coefficients: generalized least squares
Panels: heteroskedastic
Correlation: no autocorrelation

Estimated covariances	=	5	Number of obs	=	100
Estimated autocorrelations	=	0	Number of groups	=	5
Estimated coefficients	=	3	Time periods	=	20
			Wald chi2(2)	=	865.38
Log likelihood	=	-570.1305	Prob > chi2	=	0.0000

invest	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
market	.0949905	.007409	12.82	0.000	.0804692 .1095118
stock	.3378129	.0302254	11.18	0.000	.2785722 .3970535
_cons	-36.2537	6.124363	-5.92	0.000	-48.25723 -24.25017

. xtgls invest market stock, panel(corr) corr(ar1) /* Heterosk., contemp. correlation and

Cross-sectional time-series FGLS regression

Coefficients: generalized least squares
Panels: heteroskedastic with cross-sectional correlation
Correlation: common AR(1) coefficient for all panels (.8651)

Estimated covariances	=	15	Number of obs	=	100
Estimated autocorrelations	=	1	Number of groups	=	5
Estimated coefficients	=	3	Time periods	=	20
			Wald chi2(2)	=	153.66
Log likelihood	=	-491.3974	Prob > chi2	=	0.0000

invest	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
market	.0745101	.0091391	8.15	0.000	.0565978 .0924225
stock	.3150971	.0447361	7.04	0.000	.2274158 .4027783
_cons	-2.770019	13.78308	-0.20	0.841	-29.78435 24.24431

. matrix list e(Sigma)

symmetric e(Sigma)[5,5]					
	_ee	_ee2	_ee3	_ee4	_ee5
_ee	5223.2164				
_ee2	-101.18031	302.56293			
_ee3	37.474924	146.75692	2578.9016		
_ee4	-173.62446	57.848228	619.37254	262.40269	
_ee5	-1093.8519	111.5931	537.76577	704.40596	8835.32

```
. xtgls invest market stock, panel(corr) corr(ar1) rhotype(dw)
```

Cross-sectional time-series FGLS regression

Coefficients: generalized least squares

Panels: heteroskedastic with cross-sectional correlation

Correlation: common AR(1) coefficient for all panels (0.8179)

Estimated covariances	=	15	Number of obs	=	100
Estimated autocorrelations	=	1	Number of groups	=	5
Estimated coefficients	=	3	Time periods	=	20
			Wald chi2(2)	=	203.26
Log likelihood	=	-495.6259	Prob > chi2	=	0.0000

	invest	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----	market	.0758752	.0090285	8.40	0.000	.0581796 .0935708
	stock	.3289528	.0409971	8.02	0.000	.2485999 .4093056
	_cons	-10.08235	11.9502	-0.84	0.399	-33.50432 13.33961
-----+-----						

```
. xtgls invest market stock, panel(corr) corr(psar1)
```

Cross-sectional time-series FGLS regression

Coefficients: generalized least squares

Panels: heteroskedastic with cross-sectional correlation

Correlation: panel-specific AR(1)

Estimated covariances	=	15	Number of obs	=	100
Estimated autocorrelations	=	5	Number of groups	=	5
Estimated coefficients	=	3	Time periods	=	20
			Wald chi2(2)	=	331.55
Log likelihood	=	-484.6178	Prob > chi2	=	0.0000

	invest	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----	market	.0820264	.0081381	10.08	0.000	.066076 .0979767
	stock	.3800689	.0313874	12.11	0.000	.3185508 .441587
	_cons	-11.51848	12.69055	-0.91	0.364	-36.39151 13.35455
-----+-----						

.

.

end of do-file

```
. exit, clear
```

