



Migration and Labor Market

STATA II:

An Introduction into Panel Regression Models

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1 Generating dummy variables

- Gen command, replace command
- Advanced generation (forvalues 1/3 {...})

2 Organise your work with globals

- glo name varlist
- \$name

3 Summary statistics

- sum varlist



Making Graphs

- One Y-axis:
- `graph twoway line var1 var2 if ed == 1 & ex == 1`

- Two Y-axes:
- `graph twoway line (var1 var2) (var3 var2, yaxis(2)) if...`

- Scatter plots:
- `graph scatter var1 var2 if ...`

- Scatter plots with regression line:
- `graph scatter var1 var2 || lfit var1 var2 if ...`



- 1 Discussing your 'homework'
 - Generating the data sets and dummy variable
 - Descriptive statistics
 - Graphs
 - Problems?
- 2 Introduction into regressions commands in STATA
- 3 Discussing the next steps

$$y_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \dots + \beta_k x_{ki} + \varepsilon_i$$

The general econometric model:

- y_i indicates the dependent (or: endogenous) variable
- $x_{1i,ki}$ exogenous variable, explaining the independent variable
- β_0 constant or the y-axis intercept (if $x = 0$)
- $\beta_{1,2,k}$ regression coefficient or parameter of regression
- ε_i residual, disturbance term.



- The **ordinary least square (OLS)** estimator minimizes the squared deviations from the linear regression line
- Under the assumption that (i) the **error term is normally distributed** and has an **expected mean value of zero**, and (ii) the **variance of the error term is constant** and limited, the OLS estimator delivers **unbiased** results.
- Depending on the sample size, you can draw inference on the **total population**
- For the example, you can use the standard errors or t-statistics of the coefficients of the explanatory variables to test the **null-hypothesis** whether the **estimated coefficient** is zero



- **Endogeneity (simultaneous equation bias):** if your explanatory variables are not truly exogenous but correlated with the explanatory variable, the mean value of the error term isn't zero and you obtain biased estimates.
 - Solutions: Instrumental Variable (IV) estimation, natural experiments
- **Omitted variable bias:** If you have omitted relevant explanatory variables (i) the estimates of the remaining coefficients might be biased and (ii) the error term might be not zero.



- **Multicollinearity:** If your explanatory variables are correlated, you may have problems to identify the true correlation coefficient for the individual regressors, in particular when your sample is small. Is less serious.
- **Heteroscedasticity:** If the variance is not constant across groups in your sample (e.g. education groups) the standard errors are not properly estimated and the coefficients might be biased as well.
Solutions: Robust Standard Error estimates, Generalized Least Square (GLS) estimation, and others.



- **Contemporary correlation of the error term:** If your error term across the panels is contemporaneously correlated, e.g. due to simultaneous time shocks, you obtain biased estimates
- Solutions: GLS-, Generalized Methods of Moments estimators
- **And many, many others**



Regression Analysis with STATA



General Syntax

- The standard OLS regression syntax in STATA is:
 - **regress depvar [list of indepvar] [if], [options]**

Example:

- Regress log wage on migration share, controlling for education, working experience and time
 - **regress ln_wqjt m_qjt \$D_i \$D_j \$D_t**

$$y_{ijt} = \theta p_{ijt} + s_i + x_j + \pi_t + (s_i \times x_j) + (s_i \times \pi_t) + (x_j \times \pi_t) + \Psi_{ijt},$$

This model in STATA Syntax:

```
regress ln_wqjt m_qjt $D_i $D_j $D_t $D_ed_ex $D_ed_t $D_ex_t
```

where

- ln_wqjt: dependent variable (log wage)
- m_qjt: migration share in education-experience cell
- \$D_i: global for education dummies
- \$D_j: global for experience dummies
- \$D_t: global for time dummies
- \$D_ed_ex: global for interaction education-experience dummies
- \$D_ed_t : global for education-time interaction dummies
- \$D_ex_t: global for experience-time interaction dummies



```
*****  
***** Simple Regression *****  
*****  
regress ln_wqjt mqjt $D_i $D_j $D_t
```

Globals "\$"

Regression
command

Dependent
variable

Independent
variables

STATA: Output a regression model



```
. regress ln_wqjt mqjt $D_i $D_j $D_t
```

Source	SS	df	MS			
Model	21.7438431	22	.988356504	Number of obs = 204		
Residual	.224021026	181	.001237685	F(22, 181) = 798.55		
				Prob > F = 0.0000		
				R-squared = 0.9898		
				Adj R-squared = 0.9886		
				Root MSE = .03518		
Total	21.9678641	203	.108216079			

ln_wqjt	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
mqjt	.3158438	.0501982	6.29	0.000	.2167948	.4148927
Ded_2	.2226291	.0133158	16.72	0.000	.1963548	.2489033
Ded_3	.7617929	.0141016	54.02	0.000	.7339682	.7896176
Dex_2	.1942821	.0070443	27.58	0.000	.1803825	.2081816
Dex_3	.2962792	.0074409	39.82	0.000	.2815972	.3109612
Dex_4	.3591109	.0072256	49.70	0.000	.3448537	.3733681
year2	.0144838	.0143728	1.01	0.315	-.0138761	.0428437
year3	.0063602	.0143947	0.44	0.659	-.0220429	.0347632
year4	.0179271	.0144253	1.24	0.216	-.0105362	.0463904
year5	.013435	.0144442	0.93	0.354	-.0150657	.0419358
year6	.0059516	.014444	0.41	0.681	-.0225488	.0344519
year7	.0022585	.0144567	0.16	0.876	-.0262668	.0307839
year8	.002628	.0144624	0.18	0.856	-.0259086	.0311646
year9	.0011193	.0144725	0.08	0.938	-.0274373	.0296759
year10	.0078542	.0144864	0.54	0.588	-.0207297	.036438
year11	.0215036	.0144755	1.49	0.139	-.0070589	.050066
year12	.0185784	.0144583	1.28	0.200	-.00995	.0471068
year13	.0131597	.0144415	0.91	0.363	-.0153357	.0416551
year14	-.0159386	.0144263	-1.10	0.271	-.044404	.0125268
year15	-.0394645	.0144193	-2.74	0.007	-.0679161	-.0110129
year16	-.0524586	.0144216	-3.64	0.000	-.0809146	-.0240025
year17	-.0586402	.0144174	-4.07	0.000	-.087088	-.0301924
_cons	3.941102	.0172898	227.94	0.000	3.906986	3.975217

STATA: How to interpret the output of a regression?



variance of model degrees of freedom Mean squared ANOVA table

```
. reg ln_wqkt mqkt
```

Source	SS	df	MS
Model	23.4146717	1	23.4146717
Residual	87.9145738	798	.110168639
Total	111.329246	799	.139335727

Number of obs = 800
 F(1, 798) = 212.53
 Prob > F = 0.0000
 R-squared = 0.2103
 Adj R-squared = 0.2093
 Root MSE = .33192

- Overall Model Fit**
1. Observations
 2. fit of the model
 3. F-Test
 4. R-squared
 5. adjusted R-squared
 6. Root Mean Standard Error

ln_wqkt	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
mqkt	-1.369118	.093913	-14.58	0.000	-1.553464 -1.184772
_cons	4.706176	.017403	270.42	0.000	4.672015 4.740337

β_1

β_0

analysis of significance levels

95% confidence interval



- Instrumental variables (IVs) are variables which are correlated with the (potentially) endogenous explanatory variable, but not the dependent variable and, hence, not the error term
- Syntax of a IV model in stata:
ivregress depvar indepvar
(endogenous variable = iv)
- Example:
ivregress ln_wqjt \$D_i \$D_j ... (mqjt = iv1 iv2 ...)



- Very often you use panel models, i.e. models which have a group and time series dimension
- There exist special estimators for this, e.g. fixed or random effects models
 - A **fixed effects** model is a model where you have a fixed (constant) effect for each individual/group. This is equivalent to a dummy variable for each group
 - A **random effects** model is a model where you have a random effect for each individual group, which is based on assumptions on the distribution of individual effects



Consider a simple linear model:

$$Y_{it} = \beta_0 + \beta_1 X_{it} + \alpha_i + u_{it}$$

unobserved heterogeneity



- **Assumption:** $\text{Cov}(\alpha_i, X_{it}) = 0$
- If assumption holds:
 - $\hat{\beta}_{RE}, \hat{\beta}_{FE}$ are consistent
 - $se(\hat{\beta}_{RE}) < se(\hat{\beta}_{FE})$
- If assumption does not hold
 - $\hat{\beta}_{FE}$ is solely consistent ($\hat{\beta}_{RE}$ no longer consistent)
- Hausman test – RE vs FE



Preparation for Panel Models:

- For running panel models STATA needs to identify the group(individual) and time series dimension
- Therefore you need an index for each group and an index for each time period
- Then use the **tsset** command to organize you dataset as a panel data set
- Syntax:
 - **tsset index year**
- where **index** is the group/individual index and **year** the time index



Preparation: Running the *tsset* command

```
273  
274  
275 *****  
276 ***** Panel Regression *****  
277 *****  
278  
279 tsset index year  
280
```

```
. do "c:\Users\  
. tsset index year  
    panel variable:  index (strongly balanced)  
    time variable:   year, 1992 to 2008  
                   delta: 1 unit  
.  
end of do-file
```



- Then you can use panel estimators, e.g. the **xtreg** estimator
- Syntax
 - **xtreg depvar [list of indepvar] [if], [options]**
 - **xtreg ln_wqjt m_qjt, fe**
- i.e. in the example we run a simple fixed effects panel regression model which is equivalent to include a dummy variable for each group (in this case education-experience group),
but: no time or interaction dummies



Running a Panel Regression: command

```
*****  
***** Panel Regression *****  
*****
```

```
tsset index year
```

```
xtreg ln_wqjt mqjt, fe
```

STATA: Panel Regression Output



```
sers\ . xtreg ln_wqjt mqjt, fe
sers\ Fixed-effects (within) regression
sers\ Group variable: index
sers\ R-sq: within = 0.1569
sers\          between = 0.2948
sers\          overall = 0.2471
sers\ corr(u_i, Xb) = -0.6051
sers\ F(1,191) = 35.54
sers\ Prob > F = 0.0000
```

	ln_wqjt	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
	mqjt	.4030928	.0676135	5.96	0.000	.2697277	.536458
	_cons	4.466355	.010334	432.20	0.000	4.445971	4.486738
	sigma_u	.37037033					
	sigma_e	.03825941					
	rho	.98944169	(fraction of variance due to u_i)				

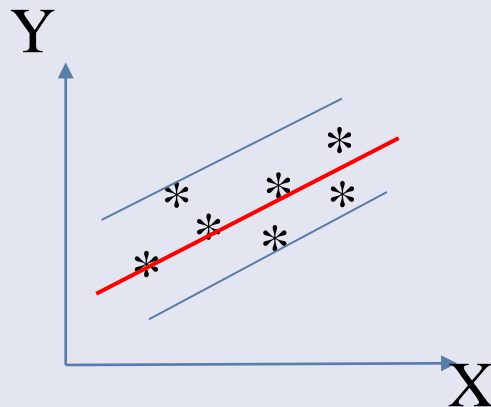
```
sers\ F test that all u_i=0: F(11, 191) = 1009.83 Prob > F = 0.0000
sers\ .
sers\ end of do-file
```



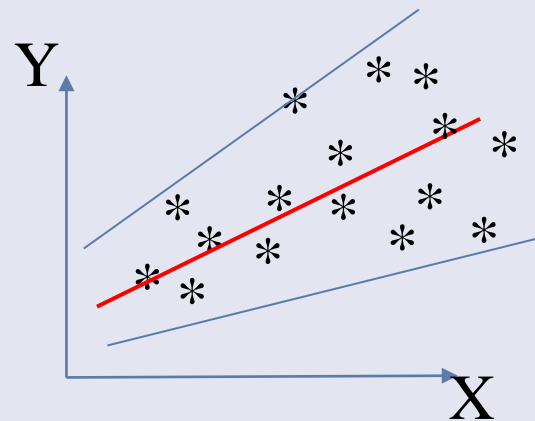
- There are other features of panel estimators which are helpful
- **Heteroscedasticity:**
 - Heteroscedasticity: the variance is not constant, but varies across groups
 - **xtpcse** , **p(h)** corrects for heteroscedastic standard errors
 - **xtgls** , **p(h)** corrects coefficient and standard errors for panel heteroscedasticity, but may produce biased results depending on the group and time dimension of the panel
 - *Note:* p(h) after the comma is a so-called option in the STATA syntax

- Variance of errors given independent variable is constant:
 $Var(u_i|X_i) = \sigma$
- Variance of errors given independent variable varies:
 $Var(u_i|X_i) = f(X_i)$

Homoscedasticity



Heteroscedasticity





- **Contemporary correlation across cross-sections**
 - Contemporary correlation: the error terms are contemporarily correlated across cross-sections, e.g. due to macroeconomic disturbances
 - **xtgls** , **p(c)** corrects for contemporary correlation and panel heteroscedasticity, but may produce biased results depending on the group and time dimension of the panel.



- **Introduction into your country and outline of research question**
- **[Very brief sketch of state of research]**
- **Theoretical hypothesis/hypotheses**
- **Description of data**
 - Data sources
 - Descriptive Statistics
 - Graphical presentation of descriptive evidence
- **Econometric findings**
 - Regression model
 - Regression results
 - Interpretation
- **Conclusions**

Next Meeting: June 29



- Begin: 12:00 – 14:00
- Topic: Descriptive data analysis

THANKS FOR YOUR ATTENTION!