ECON4150 - Introductory Econometrics Seminar 6

Stock and Watson EE10.1

April 28, 2015

Guns data set

- Some U.S. states have enacted "shall-issue" laws which allow citizens to carry concealed weapons
- We are going to investigate the effect of shall-issue laws on violent crime rates.
- Two opposite hipotesis: "more guns less crime" and "more guns more crime"
- In this exercise we use the data set Guns.dta.
- Balanced panel of data on 50 US states, plus the District of Columbia for the years 1977- 1999
- ullet There are a total of 51 states imes 23 years = 1173 observations

Variables Definition

Variable Definitions

Variable	Definition
vio	violent crime rate (incidents per 100,000 members of the population)
rob	robbery rate (incidents per 100,000)
mur	murder rate (incidents per 100,000)
shall	= 1 if the state has a shall-carry law in effect in that year
	= 0 otherwise
incarc_rate	incarceration rate in the state in the previous year (sentenced
	prisoners per 100,000 residents; value for the previous year)
density	population per square mile of land area, divided by 1000
avginc	real per capita personal income in the state, in thousands of dollars
pop	state population, in millions of people
pm1029	percent of state population that is male, ages 10 to 29
pw1064	percent of state population that is white, ages 10 to 64
pb1064	percent of state population that is black, ages 10 to 64
stateid	ID number of states (Alabama = 1, Alaska = 2, etc.)
year	Year (1977-1999)

Panel data aspect

So far cross-section data

	stateid	vio	mur	rob	incarc_rate	pb1064	pw1064	pm1029
1	1	414.4	14.2	96.8	83	8.384873	55.12291	18.17441
2	2	443.2	10.8	96.8	63	7.835672	62.1413	22.35269
3	4	494.2	9.5	138.2	125	3.436422	66.61958	18.72813
4	5	322.9	8.8	83.2	115	5.325915	59.49733	17.29585
5	6	706	11.5	287	85	5.011663	65.51936	18.78835
6	8	511.9	6.3	170.7	87	2.026637	72.37897	20.01816

• Panel data, long form

	stateid	year	vio	mur	rob	incarc_rate	pb1064	pw1064
1	1	77	414.4	14.2	96.8	83	8.384873	55.12291
2	1	78	419.1	13.3	99.1	94	8.352101	55.14367
3	1	79	413.3	13.2	109.5	144	8.329575	55.13586
4	1	80	448.5	13.2	132.1	141	8.408386	54.91259
5	2	77	443.2	10.8	96.8	63	7.835672	62.1413
6	2	78	441.9	12.9	91.3	75	7.945618	62.65009
7	2	79	491.1	13.3	109.6	127	8.057296	63.09599
8	2	80	436	8.9	81.8	133	8.123558	62.7206
9	4	77	494.2	9.5	138.2	125	3.436422	66.61958
10	4	78	552.1	9.4	162.9	129	3.457416	66.48782
11	4	79	593	8.9	175.7	146	3.473442	66.49958
12	4	80	650.9	10.3	193.6	139	3.592809	66.15736

```
gen ln_vio=ln(vio)
regress ln_vio shall, robust

Linear regression

Number of obs = 1173
F( 1, 1171) = 86.86
Prob > F = 0.0000
R-squared = 0.0866
Root MSE = .61735

| Robust
ln_vio | Coef. Std. Err. t P>|t| [95% Conf. Interval]

shall | -.4429646  .0475283   -9.32  0.000   -.5362148  -.3497144
_cons | 6.134919  .0193039  317.81  0.000  6.097045  6.172793
```

additional information for tables

```
estadd local State "-" , replace
added macro:
              e(State) : "-"
 /*
        adds a variable called state with the "value" "-" to display that we did not
        control for state fixed effects
*/
estadd local Time "-" , replace
added macro:
               e(Time) : "-"
 /*
         adds a variable called time with the "value" "-" to display that we did not
         control for time fixed effects
*/
estadd local Control "-" . replace
added macro:
            e(Control) : "-"
/*
          adds a variable called time with the "value" "-" to display that we did not
          control for other control variables
```

regress ln_vio shall incarc_rate density avginc pop pb1064 pw1064 pm1029, robust

Linear regression

Number of obs = 1173F(8, 1164) = 95.67Prob > F = 0.0000R-squared = 0.5643Root MSE = .42769

 ln_vio	Coef.	Robust Std. Err.	t	P> t	[95% Conf.	Interval]
shall	3683869	.0347879	-10.59	0.000	436641	3001329
incarc_rate	.0016126	.0001807	8.92	0.000	.0012581	.0019672
density	.0266885	.0143494	1.86	0.063	0014651	.054842
avginc	.0012051	.0072778	0.17	0.869	013074	.0154842
pop	.0427098	.0031466	13.57	0.000	.0365361	.0488836
pb1064	.0808526	.0199924	4.04	0.000	.0416274	.1200778
pw1064	.0312005	.0097271	3.21	0.001	.012116	.0502851
pm1029	.0088709	.0120604	0.74	0.462	0147917	.0325334
_cons	2.981738	.6090198	4.90	0.000	1.786839	4.176638

```
est sto reg2
estadd local State "-" , replace
estadd local Time "-" , replace
estadd local Control "Yes" , replace
```

a., i ii.

Table: dependent variable is In(vio)

	1	2
shall	-0.443*** (0.0475)	-0.368*** (0.0348)
Control variables	_	Yes

Standard errors in parentheses

* p < 0.05, ** p < 0.01, *** p < 0.001

- The coefficient in regression (2) equals -0.368, which suggests that shall-issue law reduce the violent crime rate by 36%.
- The coefficient in (1) is -0.443; in (2) it is -0.369. Both are highly statistically significant.
- Adding the control variables results in a small drop in the estimated coefficient.
- The "real world" significance of the coefficient remains stable

2	ш.
a.,	

 Possible omitted variables that vary between states but not over time Attitudes towards guns and crime, quality of police and other crime-prevention programs.

```
xtreg ln_vio shall incarc_rate density avginc pop pb1064 pw1064 pm1029, fe robust
Fixed-effects (within) regression
                                       Number of obs =
                                                          1173
Group variable: stateid
                                                            51
                                       Number of groups =
R-sq: within = 0.2178
                                        Obs per group: min =
     between = 0.0033
                                                           23.0
                                                     avg =
     overall = 0.0001
                                                     max =
                                                              23
                                       F(8.50)
                                                       = 34.10
                                        Prob > F
                                                            0.0000
corr(u_i, Xb) = -0.3687
                         (Std. Err. adjusted for 51 clusters in stateid)
                       Robust
              Coef. Std. Err. t P>|t| [95% Conf. Interval]
    ln_vio |
     shall | -.0461415 .0417616 -1.10 0.275 -.1300223
                                                           0377392
incarc_rate | -.000071 .0002504 -0.28 0.778 -.0005739 .0004318
    density | -.1722901 .1376129 -1.25 0.216 -.4486936 .1041135
            -.0092037 .0129649 -0.71 0.481
                                             -.0352445 .016837
     avginc |
            .0115247 .014224 0.81 0.422 -.0170452 .0400945
       pop
            .1042804 .0326849 3.19 0.002
                                             .0386308 .1699301
    pb1064
            .0408611 .0134585 3.04 0.004 .0138289 .0678932
    pw1064 |
     pm1029 | -.0502725 .0206949 -2.43 0.019 -.0918394 -.0087057
                       .7701057 5.02 0.000 2.319214
            3.866017
                                                           5.412819
     _cons
       rho | .94712779 (fraction of variance due to u i)
est sto reg3
estadd local State "Yes" , replace
estadd local Time "-", replace
estadd local Control "Yes", replace
```

4□ > 4□ > 4□ > 4 = > 4 = > = 90

\\ i.state create a dummy for each state in the regression
reg ln_vio shall incarc_rate density avginc pop pb1064 pw1064 pm1029 i.state, robust

Linear regression

Number of obs = 1173F(58, 1114) = 364.90 Prob > F = 0.0000 R-squared = 0.9411 Root MSE = .16072

ln_vio	Coef.	Robust Std. Err.	t	P> t	[95% Conf.	. Interval]
shall	0461415	.0199433	-2.31	0.021	0852721	007011
incarc_rate	000071	.0000973	-0.73	0.466	0002619	.0001199
density	1722901	.1048789	-1.64	0.101	3780725	.0334923
avginc	0092037	.0067335	-1.37	0.172	0224155	.004008
pop I	.0115247	.0097044	1.19	0.235	0075162	.0305655
pb1064	.1042804	.0165552	6.30	0.000	.0717976	.1367633
pw1064	.0408611	.0053859	7.59	0.000	.0302935	.0514287
pm1029	0502725	.0077908	-6.45	0.000	0655588	0349863
· i						
stateid						
2	.0559649	.0788371	0.71	0.478	098721	.2106508
56 I	4804004	.1293103	-3.72	0.000	7341196	2266813
i						
_cons	4.036775	.3845839	10.50	0.000	3.282185	4.791366

\\ same result on shall and estimate for each dummy, but we are not interested on those estimates

Table: dependent variable is ln(vio)

	1	2	3
shall	-0.443*** (0.0475)	-0.368*** (0.0348)	-0.0461 (0.0418)
Control variables State fixed effects	_ _	Yes —	Yes Yes

Standard errors in parentheses

- * p < 0.05, ** p < 0.01, *** p < 0.001
- The results change when we add state fixed effects.
- The absolute value of the coefficient on shall falls to 0.046, a large reduction in the coefficient from 0.369 without fixed effects.
- Evidently there was important omitted variable bias in the specification without fixed effects.
- The estimate of the effect of shall issue laws on the violent crime rate is no longer statistically significantly different from zero
- The regression model with fixed effects is more credible because this controls for unobserved characteristics that vary between states but that are constant over time

```
Fixed-effects (within) regression
                                               Number of obs
                                                                         1173
Group variable: stateid
                                               Number of groups =
                                                                           51
R-sq: within = 0.4180
                                               Obs per group: min =
                                                                           23
       between = 0.0419
                                                              avg =
                                                                         23.0
       overall = 0.0009
                                                                           23
                                                              max =
                                               F(30,50)
                                                                        56.86
corr(u_i, Xb) = -0.2929
                                               Prob > F
                                                                       0.0000
```

(Std. Err. adjusted for 51 clusters in stateid)

1		Robust				
ln_vio	Coef.	Std. Err.			[95% Conf.	Interval]
shall	0279935	.0407168	-0.69	0.495	1097757	.0537886
_Iyear_78	.0585261	.0161556	3.62	0.001	.0260767	.0909755
_Iyear_99	.4328776	.2862197	1.51	0.137	1420117	1.007767
incarc_rate	.000076	.0002079	0.37	0.716	0003416	.0004935
density	091555	.1238622	-0.74	0.463	3403396	. 1572296
avginc	.0009587	.0164931	0.06	0.954	0321688	.0340861
pop	0047544	.0152294	-0.31	0.756	0353436	.0258347
pb1064	.0291862	.0495407	0.59	0.558	0703192	.1286916
pw1064	.0092501	.0237564	0.39	0.699	0384659	.0569662
pm1029	.0733254	.0524733	1.40	0.168	0320704	.1787211
_cons	3.765525	1.152108	3.27	0.002	1.451448	6.079603
sigma_u	.6663043					
sigma_e	.1400264					
rho	.95770338	(fraction	of waria	nce due t	(i n o	
1110	. 551 10556	(114661011	or varian	ree due t	,o u_1/	

est sto reg4
estadd local State "Yes" , replace
estadd local Time "Yes" , replace
estadd local Control "Yes" , replace

```
/*
 Testing the joint significance of the year dummies: _Iyear*,
 with * you consider all the variables starting with _Iyear
*/
  testparm _Iyear*
 (1) _{Iyear_78} = 0
 (2) _{Iyear_79} = 0
 (3) _{Iyear_80} = 0
 (4) _{Iyear_81} = 0
 (18) _Iyear_95 = 0
 (19) _{Iyear_96} = 0
 (20) _{Iyear_97} = 0
 (21) _Iyear_98 = 0
 (22) _Iyear_99 = 0
       F(22, 50) = 21.62
           Prob > F = 0.0000
estadd scalar F_year = r(F)
added scalar:
             e(F_{year}) = 21.621583
```

Table: dependent variable is In(vio)

	1	2	3	4
shall	-0.443***	-0.368***	-0.0461	-0.0280
	(0.0475)	(0.0348)	(0.0418)	(0.0407)
Control variables	_	Yes	Yes	Yes
State fixed effects	_	—	Yes	Yes
Time fixed effects F-time dummies	_	-	_	<i>Yes</i> 21.62

Standard errors in parentheses

- The absolute value of the coefficient on shall falls further to 0.028, the coefficient is not significantly different from zero.
- The time effects are jointly statistically significant, so this regression seems better specified than the regression in part (b).

^{*} p < 0.05, ** p < 0.01, *** p < 0.001

```
// replicate with ln_rob and ln_mur
gen ln_rob=ln(rob)
regress ln_rob shall, robust
est sto reg1
estadd local State "-" , replace
estadd local Time "-", replace
estadd local Control "-" , replace
regress ln_rob shall incarc_rate density avginc pop pb1064 pw1064 pm1029, robust
est sto reg2
estadd local State "-" , replace
estadd local Time "-", replace
estadd local Control "Yes" , replace
xtreg ln_rob shall incarc_rate density avginc pop pb1064 pw1064 pm1029, fe robust
est sto reg3
estadd local State "Yes" , replace
estadd local Time "-" , replace
estadd local Control "Yes" , replace
xi: xtreg ln_rob shall i.year incarc_rate density avginc pop pb1064 pw1064 pm1029, fe robust
est sto reg4
estadd local State "Yes" , replace
estadd local Time "Yes", replace
estadd local Control "Yes", replace
testparm Ivear*
estadd scalar F vear = r(F)
esttab reg* . label se beta(2) keep(shall) ///
  s(Control State Time F_year , label("Control variables" "State fixed effects" "Time fixed effects" "F-time dummies")) ///
 mtitles("1" "2" "3" "4") nonumbers title(Dependent variable is ln(rob))
log close
```

dependent variable is In(rob)

	1	2	3	4			
shall	-0.773*** (0.0693)	-0.529*** (0.0510)	-0.00782 (0.0552)	0.0268 (0.0522)			
Control variables State fixed effects	_ _	Yes —	Yes Yes	Yes Yes			
Time fixed effects F-time dummies	_	_	_	<i>Yes</i> 25.86			

dependent variable is In(dmur)

	1	2	3	4
shall	-0.473***	-0.313***	-0.0608	-0.0150
	(0.0485)	(0.0357)	(0.0370)	(0.0382)
Control variables	_	Yes	Yes	Yes
State fixed effects		—	Yes	Yes
Time fixed effects F-time dummies	-	_	_	Yes 19.61

Standard errors in parentheses

The results are similar to the results using violent crimes:

- There is a large estimated effect of concealed weapons laws in specifications (1) and (2).
- This effect is however due to omitted variable bias because the effect disappears when state and time effects are added.

^{*} p < 0.05, ** p < 0.01, *** p < 0.001

e: Remaining threats to internal validity

Definition A statistical analysis is said to have internal validity if the statistical inferences about causal effects are valid for the population and setting being studied. The estimator should be unbiased and consistent and the standard errors are computed in a way that makes confidence intervals have the desired confidence levels. Threats to internal validity

- Functional form misspecification
- Omitted variable bias
- Measurement error
- Sample selection
- Simultaneous causality
- Simultaneous causanty
- $\bullet \ \ \text{Heteroskedasticity and/or correlated error terms} \rightarrow \text{violation i.d.d.}$



e

- Simultaneous causality: violent crimes this may induce policy makers to change concealed weapons laws
- Omitted variables: There might be important variables that vary between states and over time that are omitted from the regression model. For example other policy measures that are related to the implementation of shall issue laws and that affect crime rates

- The most credible results include both state fixed effects and time fixed effects.
- These results indicate that there is no significant effect of concealed weapon laws on crime rates