

Untitled

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Abstract

This paper is a replication of Lott and Mustard 1997 paper on effects of right to carry laws on crime rates. We use state level data instead of county data used in the original paper. We perform twoway fixed effects regression. In addition we also use bacon decomposition to weight the matrices efficiently. We also use Callaway-Santanna estimator and Sun and Abraham event study in this context. Finally we see if these new additons change any conclusion made in the paper significantly.

Overview

The original paper used cross-sectional time series data for US counties over 1977-1992 to estimate crime and deaths. They conclude that right to carry laws reduce violent crimes and if the policy was implemented in 1992 then 1500 murders could have been avoided. They also report that this law would not cause increases in “accidental” deaths that is the someone killing a person who didn’t mean any harm in self-defense. They go on to report various other factors such as social benefit, annual gain, other crime rates etc but we focus on just the crime volume using 4 different techniques.

When the law was introduced there were compelling argument from both ends of the spectrum that were bolstered by either anecdotes or data. The arguments basically were: a)will this lead to those citizens committing more crime who otherwise did not or b) will this be a safeguard against criminals who intended to commit some sort of crime. When cases of killing in self-defense occur, it become even more crucial to test this theory.

Bacon decomposition is a modification on TWFE as in that it introduces a more sophisticated way to weight the 2*2 matrices than TWFE. TWFE naively assumed equal matrices without accounting for the fact that changing the length of panel will change the weights and therefore change the estimates.

Callaway and Santanna also develop on a weighting mechanism. They calculate average group treatment effect. Their major success is in the fact that this is non-parametric regression. This way is agnostic of time. Parallel trends conditional on time invariant co variate are expected to overlay in a way that can be computed by the propensity score. Sun and Abraham use differential timing of the event studies to measure changes in the treatment by computing lead and lags. This is sort of like Callaway and Santanna in the sense that they also focus on group specific ATT. These techniques primarily enhance the estimation because they take into account the heterogeneity treatment effect.

Data Collection

We collected this data from the replication kit provided by the authors. The same data set was used to publish their paper. The data is from FBI crime reports. It gives state level crime rate, crime count, arrest rate, length of the crime and many more variables. The categories of crime used for the purpose of this paper are violent crime, murder, burglary, larceny, theft, rape, robbery, assault and property crime. A key variable in the regression is “shall” which is an indicator variable identifying shall as 1 if the carry gun laws are in effect.

Below plot and tables provides a summary of the crime statistics across states over the years 1977-1992

Table 1: State-wise implementation dates of the right to carry guns law

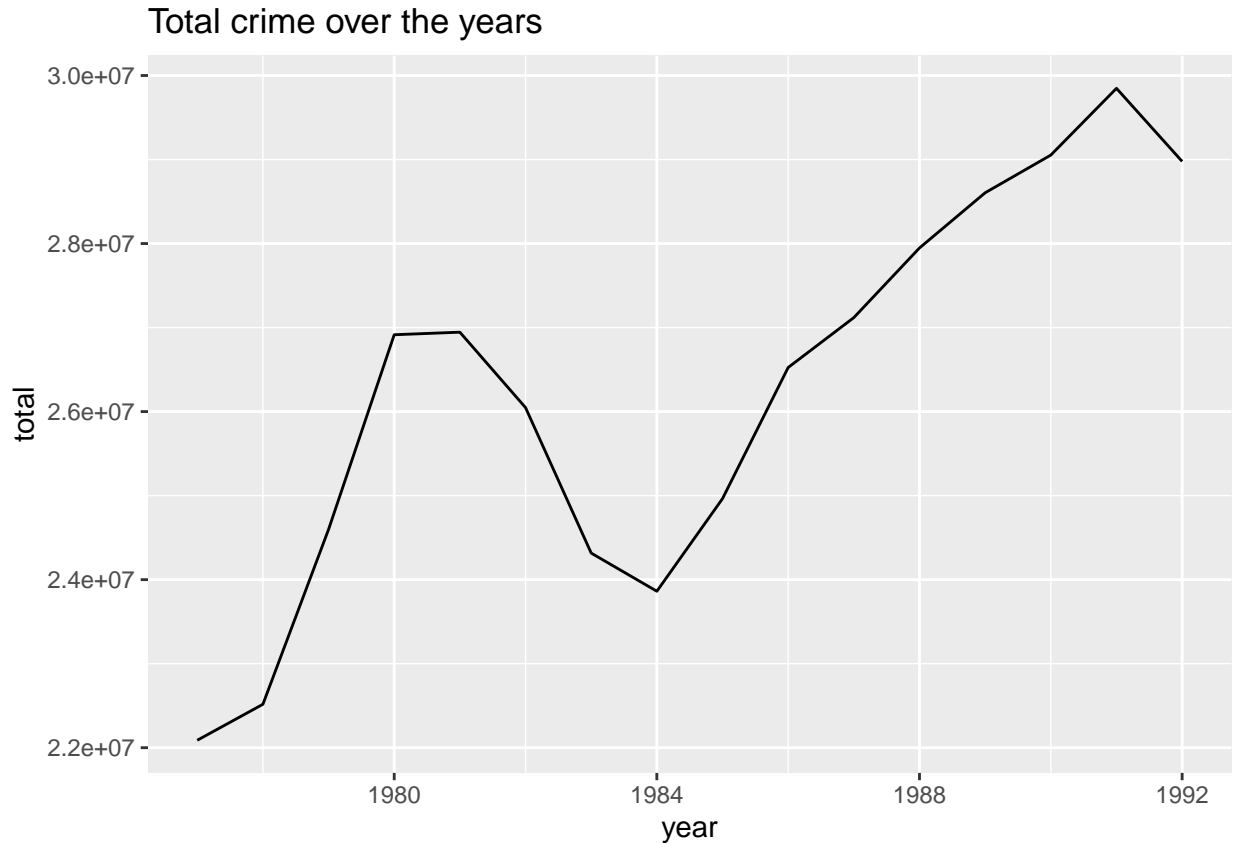
State	Year
Alabama	before 1977
Connecticut	before 1977
New Hampshire	before 1977
North Dakota	before 1977
South Dakota	before 1977
Vermont	before 1977
Washington	before 1977
Indiana	1981
Maine	1986
Florida	1988
Virginia	1989
Georgia	1990
Pennsylvania	1990
West Virginia	1990
Idaho	1991
Mississippi	1991
Oregon	1991
Montana	1992

Table 2: Summary of crime rate statistics categorized by the type of crime

Crime_Type	Mean	SD	Max	Min
Violent Crime	483.93	318.94	2856.13	47.56
Property Crime	4618.34	1210.46	8644.89	2036.21
Murder	7.77	6.88	81.25	0.56
Rape	33.98	15.07	100.60	7.39
Aggravated Assault	278.76	159.65	1466.66	31.48
Robbery	163.42	176.25	1632.78	6.50
Auto Theft	410.30	231.15	1560.81	97.41
Burglary	1239.34	417.76	2871.15	366.53
Larceny	2968.71	751.02	5272.23	1193.93

Table 3: Summary of no of arrests statistics categorized by the type of crime

Crime_Type	Mean	SD	Max	Min
Violent Crime	41.22	22.17	558.81	9.03
Property Crime	16.93	4.56	58.56	3.16
Murder	91.15	56.05	1363.16	14.29
Rape	41.08	17.37	310.63	8.88
Aggravated Assault	44.95	16.82	190.93	7.08
Robbery	31.66	13.52	195.20	5.88
Auto Theft	22.31	37.74	394.28	1.07
Burglary	13.84	4.41	28.71	0.62
Larceny	18.54	5.09	77.27	4.64



Differenet model estimations: TWFE, BACON, CALLAWAY AND SANTANNA

We begin with TWFE estimation which is in fact the regression methodology used in the original paper. Given the staggered natur of the policy implementation, TWFE would be a go-to option here. log of crime rate is the dependent variable to enhance interpretation and scale. We do this regression for all 9 categories of crime. We then move to Bacon decomposition and instantly notice how it led to better results to due much more effective weighting. We have two groups that form into sepearate 2*2 matrix - Earlier vs Later Treated and Later vs Earlier Treated. For purpose of this paper we perform bacon decomposition without controls. We notice here that 2nd matrice is weighted much more highly than the 1st ma trice. Now this is where the TWFE would have given us equal weightage for more and therefore can be misleading. Next we move to estimating group effects useing callaway and santanna. A group here would be a state. Treatment for all the 4 approaches is defined as 1 when of the law is implemented if not 0. Callaway and Santanna method gives higher standard errors but the direction of ATT are more or less the same as TWFE and bacone except for a few. Given the non-parametric nature of this estimator the interpretation of coefficients is tricky.

Sun and Abraham

Here we look at all the crime classes and plot their 95% confidence interval The study the heterogeneous effect by studying the lead and lags. Post treatment effects of all the crimes are more or less following the same trend which is a good learning and therefore we can say that using gthe leads and the lags the parallel trend scan be overlayed with relief.

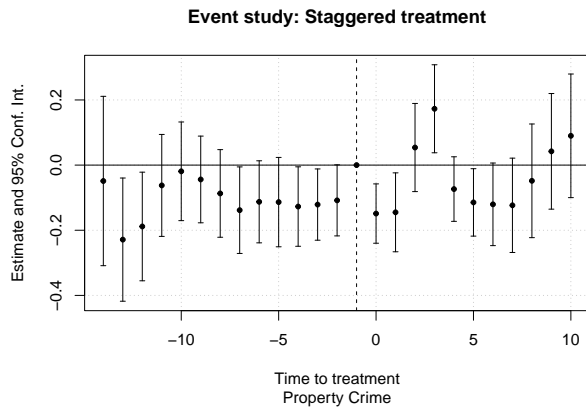
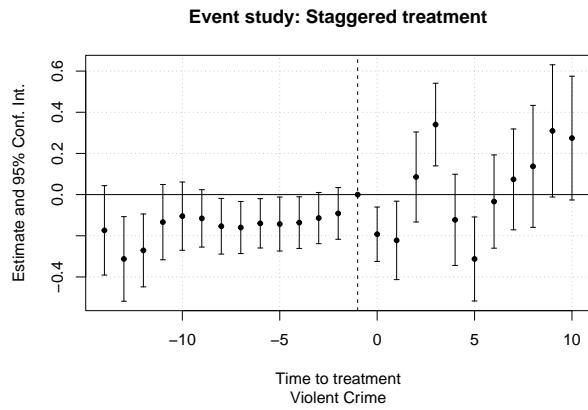
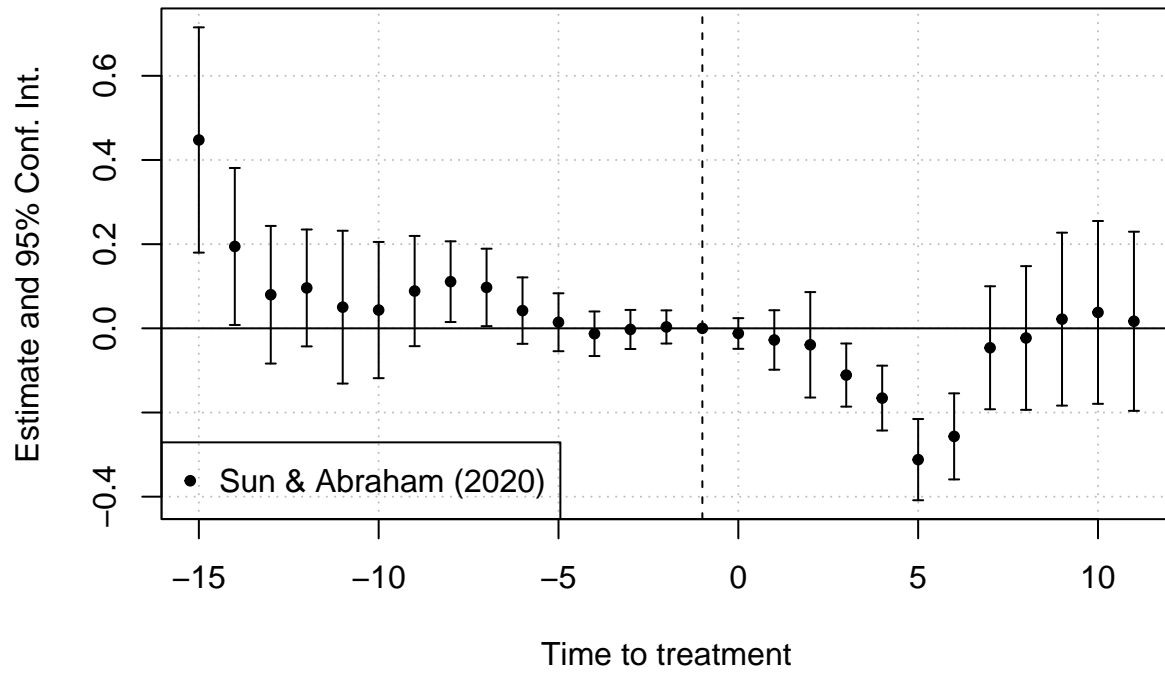
Table 4: Effects

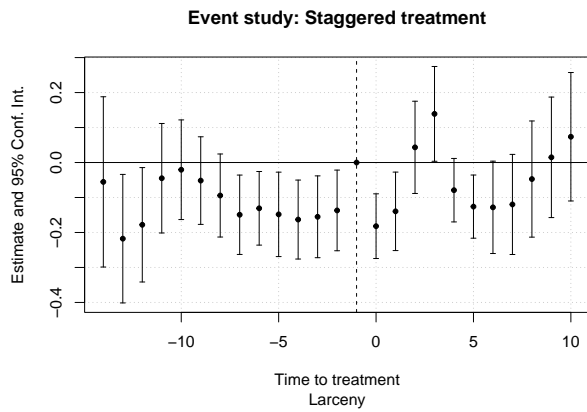
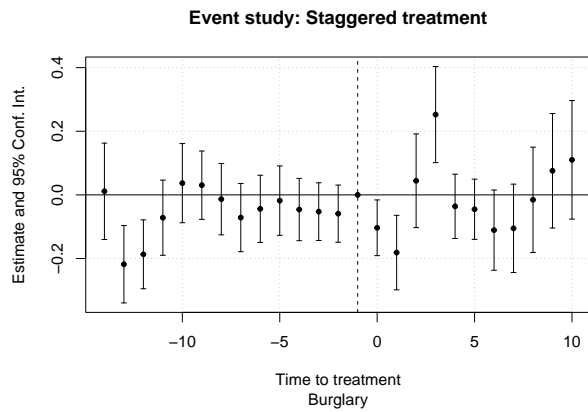
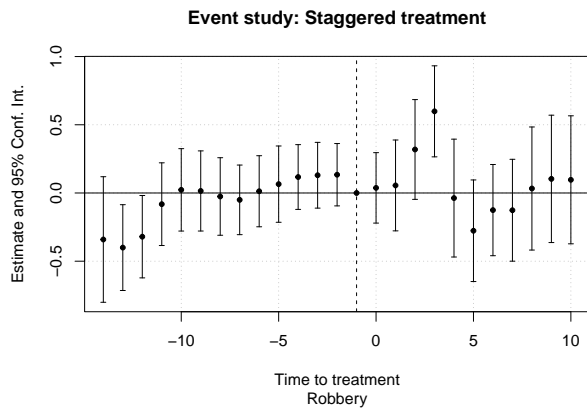
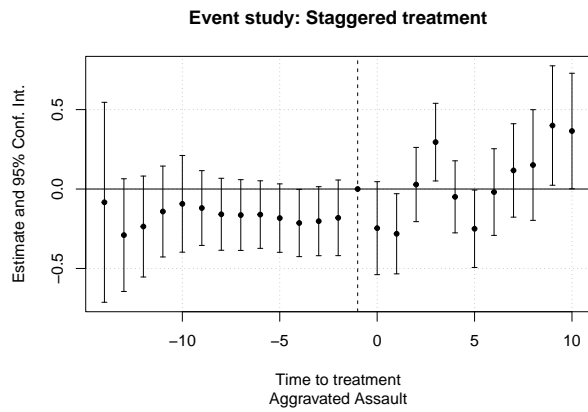
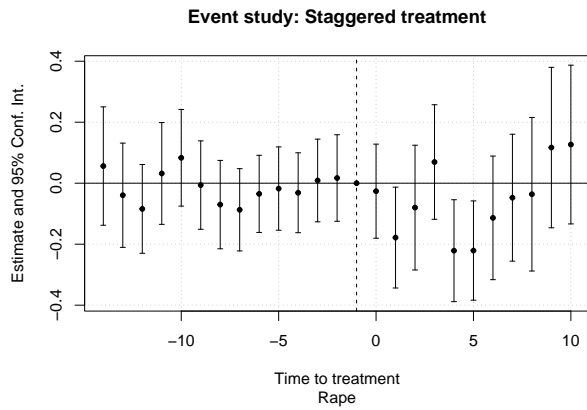
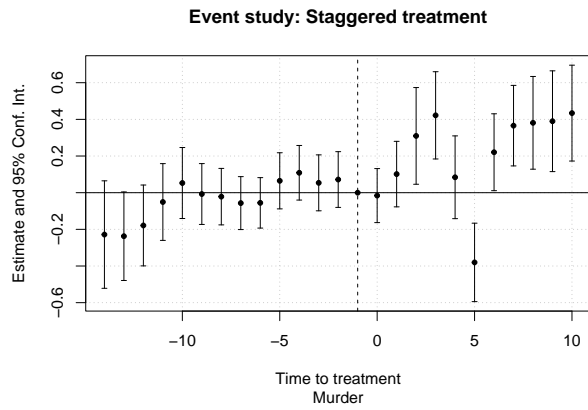
outcome	TWFE	CallowaySantAnna
lvio	-0.057	-0.01
	(0.023)	(0.025)
lmur	0.009	0.013
	(0.014)	(0.012)
laga	-0.05	-0.049
	(0.04)	(0.025)
lbur	-0.049	0.02
	(0.028)	(0.029)
laut	-0.051	0.005
	(0.03)	(0.046)
lpro	-0.018	0.039
	(0.031)	(0.035)
lrap	-0.024	-0.016
	(0.019)	(0.016)
lrob	0.014	0.03
	(0.013)	(0.016)
llar	0.034	0.011
	(0.03)	(0.043)

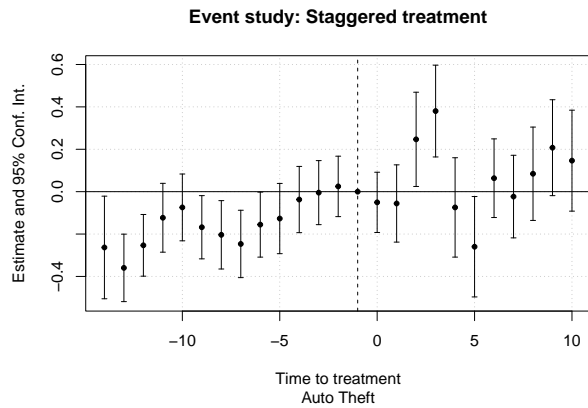
Table 5: Bacon Decomposition Estimates

Crime Variable	Type	Average Estimate	Weight
lvio	Earlier vs Later Treated	0.07561323	0.06838103
lvio	Later vs Earlier Treated	-0.07645011	0.02339212
lpro	Earlier vs Later Treated	-0.0105168	0.06838103
lpro	Later vs Earlier Treated	0.006443744	0.02339212
lmur	Earlier vs Later Treated	0.07973674	0.06838103
lmur	Later vs Earlier Treated	0.001787755	0.02339212
lrap	Earlier vs Later Treated	-0.03864389	0.06838103
lrap	Later vs Earlier Treated	-0.08243419	0.02339212
laga	Earlier vs Later Treated	0.1164473	0.06838103
laga	Later vs Earlier Treated	-0.1471737	0.02339212
lrob	Earlier vs Later Treated	0.1077515	0.06838103
lrob	Later vs Earlier Treated	0.08954899	0.02339212
lbur	Earlier vs Later Treated	-0.03396534	0.06838103
lbur	Later vs Earlier Treated	-0.0556313	0.02339212
llar	Earlier vs Later Treated	-0.006083384	0.06838103
llar	Later vs Earlier Treated	0.02077019	0.02339212
laut	Earlier vs Later Treated	0.08312628	0.06838103
laut	Later vs Earlier Treated	0.0868026	0.02339212

Event study: Staggered treatment







Conclusion

Overall the techniques suggest the same conclusion that derived by Lott and Mustard in their original paper however the magnitudes are different and rightfully so. These improvements are very crucial in policymaking. This also goes to say that the field of economics is constantly updating and upgrading. Therefore, it might be a good practice to revisit the old conclusions drawn in papers written before especially when the topic is still relevant. For example Callaway and Santana show opposite impact on property crime rate than TWFE and Bacon. So this needs hard thinking as to which one to be followed and more importantly why the difference has risen.