

## Project Report

# The Impact of 'Shall-Carry' law and guns on Crime Rate in the U.S.



## Applied Econometrics and Time Series Analysis

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## Introduction:

The impact of incarceration rate and guns on crime has always engendered many public debates in America. Many states have adopted the right-to-carry laws, also known as the shall-issue laws, as an effort toward regulation of guns. A shall-carry law requires that government issue concealed carry handgun permits to any applicant who meets the necessary criteria. These criteria are that the applicant must be an adult, have no history of mental illness and no significant criminal record and should successfully complete a course in firearms safety training (if required by law). If these criteria are met, then the granting authority has no discretion in awarding the license. Many believe this would lead to lower crime rates as restrictions would be placed to own a gun. Whereas, many others who are against this law suggest that crime rate would increase as criminals would be more willing to commit a crime since there would be fewer guns issues, which would have been a threat to them.

The objective of this report is to evaluate the effect of higher incarceration rate and the effect of the shall-issue law on the reported crime rate, specifically robbery rate, murder rate and violent crime rate (all measured as incidents per 100,000 members of the population). A balanced panel of data of the 50 states of the United States, plus the District of Columbia, from 1977 through 1999 is used for the analysis.

## Data understanding:

We started out by performing a univariate analysis to gain an understanding about the variables in the dataset.

Missing data: There were no missing data to be dealt with.

## Univariate Analysis:

. summarize vio mur rob incarc_rate pb1064 pw1064 pm1029 pop						
> avginc density shall						
Variable	Obs	Mean	Std. Dev.	Min	Max	
vio	1,173	503.0747	334.2772	47	2921.8	
mur	1,173	7.665132	7.52271	.2	80.6	
rob	1,173	161.8202	170.51	6.4	1635.1	
incarc_rate	1,173	226.5797	178.8881	19	1913	
pb1064	1,173	5.336217	4.885688	.2482066	26.97957	
pw1064	1,173	62.94543	9.761527	21.78043	76.52575	
pm1029	1,173	16.08113	1.732143	12.21368	22.35269	
pop	1,173	4.816341	5.252115	.402753	33.14512	
avginc	1,173	13.7248	2.554543	8.554884	23.64671	
density	1,173	.3520382	1.355472	.0007071	11.10212	
shall	1,173	.2429668	.4290581	0	1	

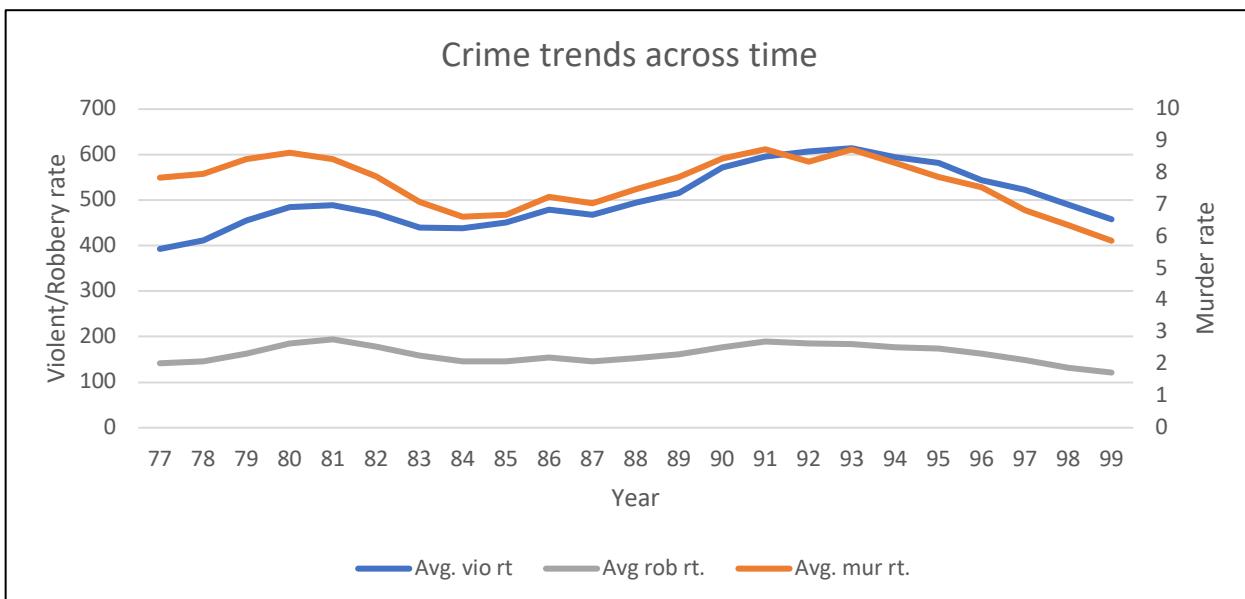
## Corellation matrix:

We created a corelation matrix to understand the relationship between variables. On a broad level all types of crimes are strongly corealted to each other. Percentage of white population is negatively corelated with crime and percentage of black population is positively corelated with crime. Population density has a strong corelation with crime whereas averageincome has a weak corelation.

. corr vio	mur	rob	incarc_rate	pb1064	pw1064	pm1029	pop	avginc	density	shall	
	vio	mur	rob	incarc~e	pb1064	pw1064	pm1029	pop	avginc	density	shall
vio	<b>1.0000</b>										
mur	<b>0.8265</b>	<b>1.0000</b>									
rob	<b>0.9071</b>	<b>0.7976</b>	<b>1.0000</b>								
incarc_rate	<b>0.7027</b>	<b>0.7096</b>	<b>0.5668</b>	<b>1.0000</b>							
pb1064	<b>0.5698</b>	<b>0.6018</b>	<b>0.5812</b>	<b>0.5308</b>	<b>1.0000</b>						
pw1064	<b>-0.5730</b>	<b>-0.6154</b>	<b>-0.5842</b>	<b>-0.5271</b>	<b>-0.9820</b>	<b>1.0000</b>					
pm1029	<b>-0.1696</b>	<b>0.0150</b>	<b>-0.0860</b>	<b>-0.4463</b>	<b>0.0162</b>	<b>-0.0126</b>	<b>1.0000</b>				
pop	<b>0.3190</b>	<b>0.0999</b>	<b>0.3172</b>	<b>0.0953</b>	<b>0.0581</b>	<b>-0.0654</b>	<b>-0.0975</b>	<b>1.0000</b>			
avginc	<b>0.4080</b>	<b>0.2206</b>	<b>0.4148</b>	<b>0.4615</b>	<b>0.2627</b>	<b>-0.1912</b>	<b>-0.5279</b>	<b>0.2152</b>	<b>1.0000</b>		
density	<b>0.6647</b>	<b>0.7486</b>	<b>0.7818</b>	<b>0.5593</b>	<b>0.5432</b>	<b>-0.5551</b>	<b>-0.0637</b>	<b>-0.0780</b>	<b>0.3433</b>	<b>1.0000</b>	
shall	<b>-0.2069</b>	<b>-0.1794</b>	<b>-0.2125</b>	<b>0.0424</b>	<b>-0.1839</b>	<b>0.2123</b>	<b>-0.2772</b>	<b>-0.1244</b>	<b>-0.0000</b>	<b>-0.1126</b>	<b>1.0000</b>

## Exploratory Data Analysis:

### Crimes across time:



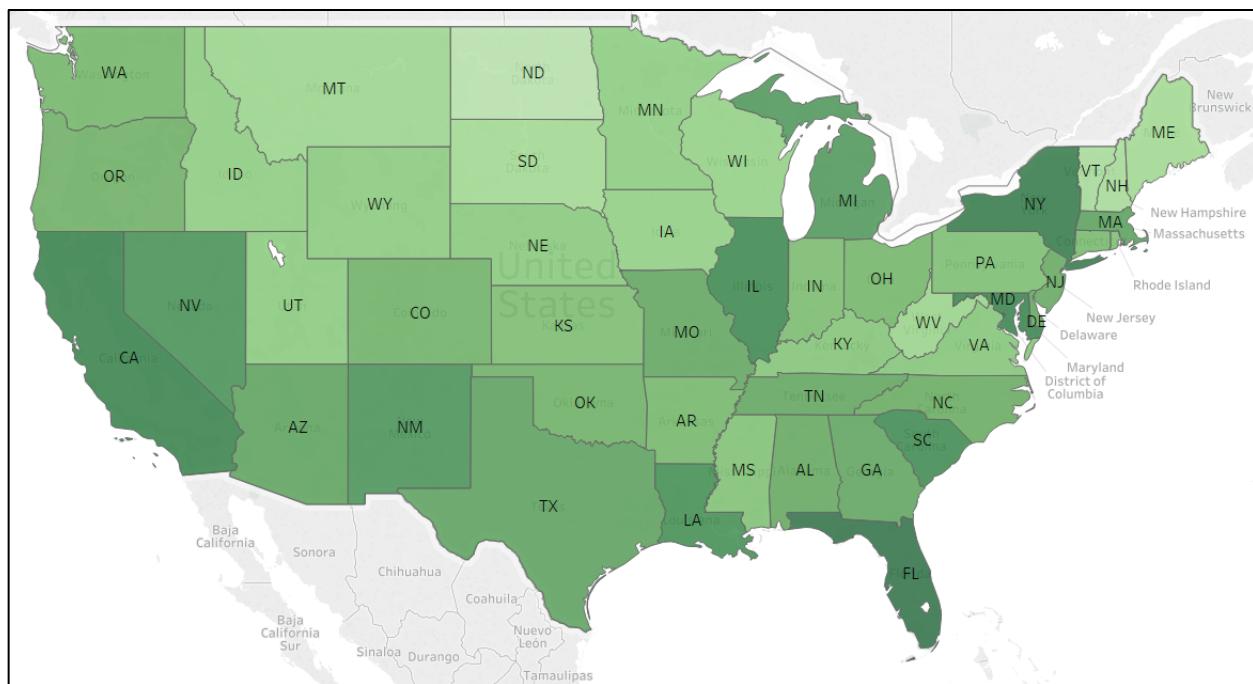
First step was to understand the distribution of data across time. Here we can see that the three nationwide averages for violent crime rate, Robbery rate and murder rate all follow a similar trend

across time. The distribution looks like bimodal, with a peak in crime in the years 1980 and 1991 with a trough at 1984 and continuous decrease post 1993.

Since the three distributions across time are similar we will be summing up the values across year to create a crime score per year for the remaining exploratory data analysis.

### Crimes across states:

#### Violent Crimes:



Average violent crime rate was calculated across time and the states were plotted.

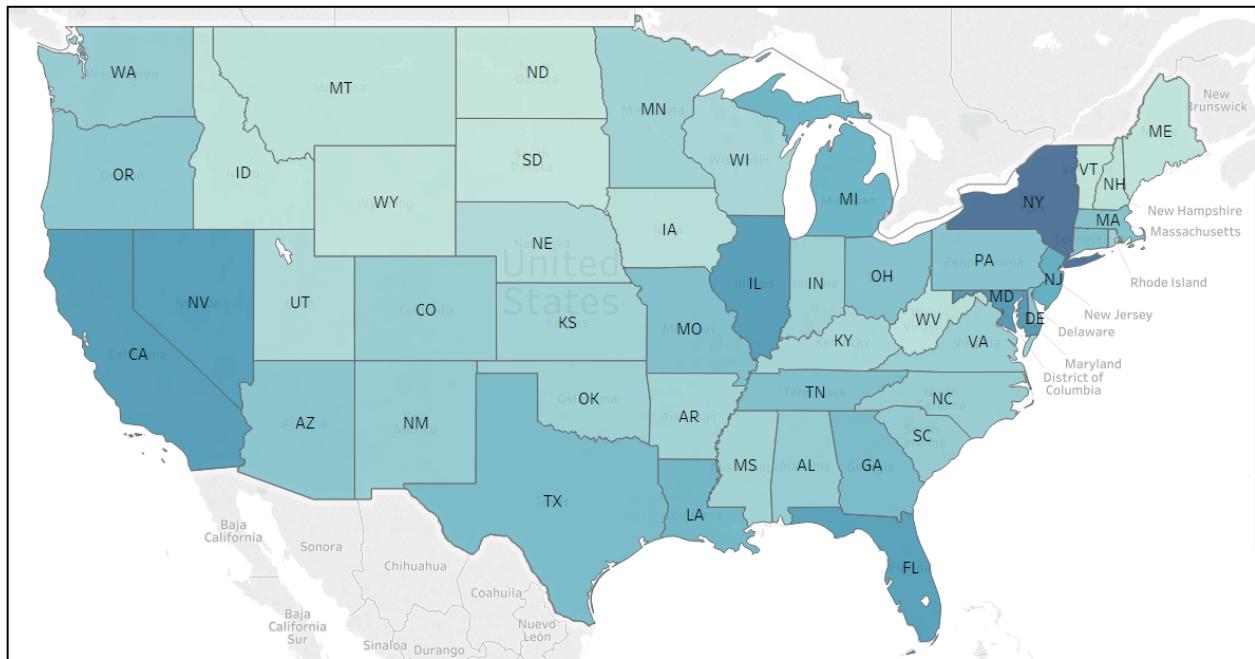
#### Top states:

1. DC
2. Florida
3. New York

#### Bottom state:

1. North Dakota
2. New Hampshire
3. Vermont

Robbery:



Average robbery rate was calculated across time and the states were plotted.

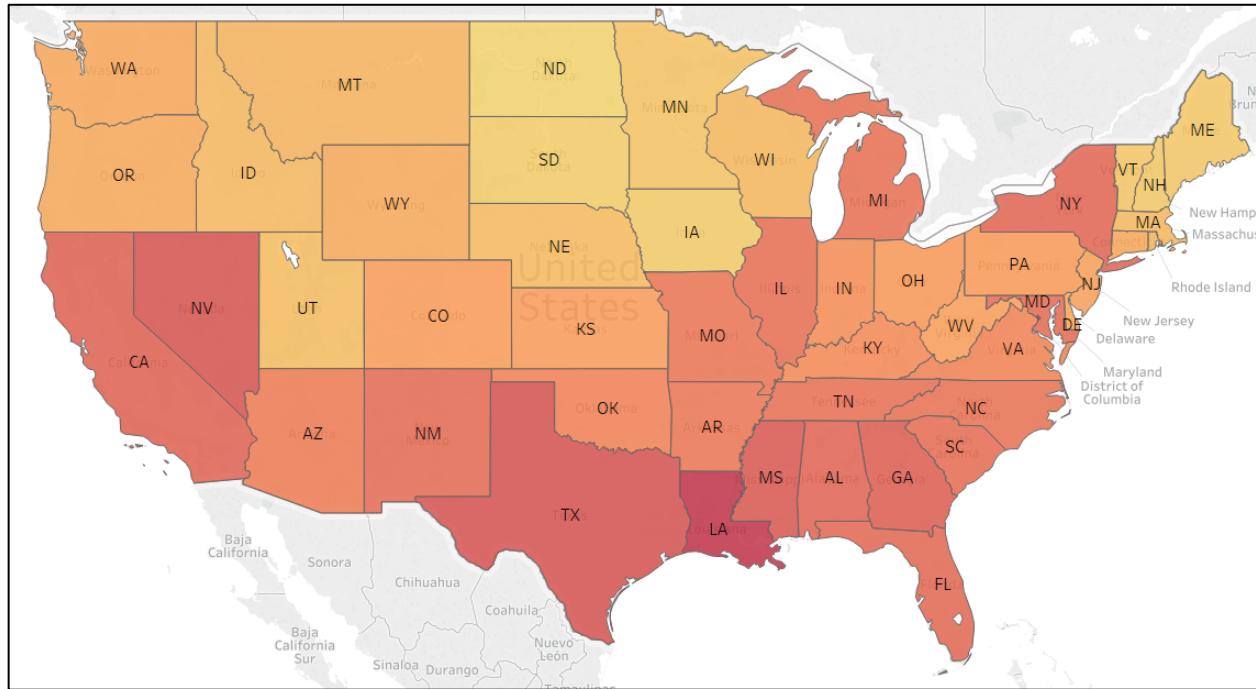
Top states:

1. DC
2. Maryland
3. New York

Bottom state:

1. North Dakota
2. South Dakota
3. Vermont

Murder:



Average murder rate was calculated across time and the states were plotted.

Top states:

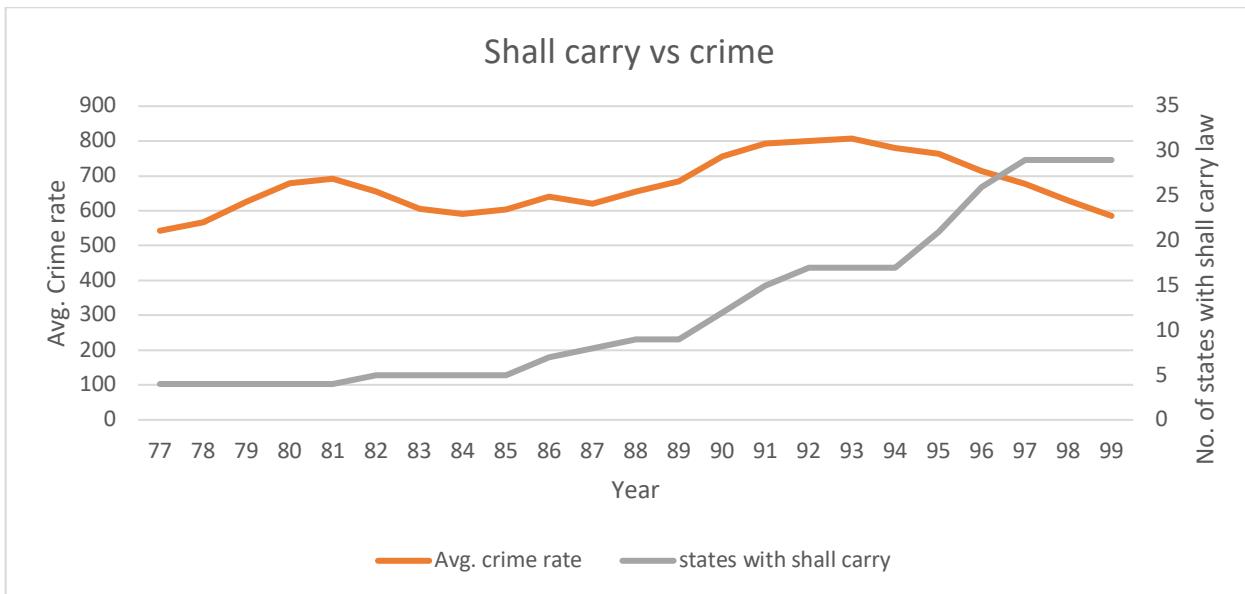
1. DC
2. Louisiana
3. Texas

Bottom state:

1. North Dakota
2. South Dakota
3. Iowa

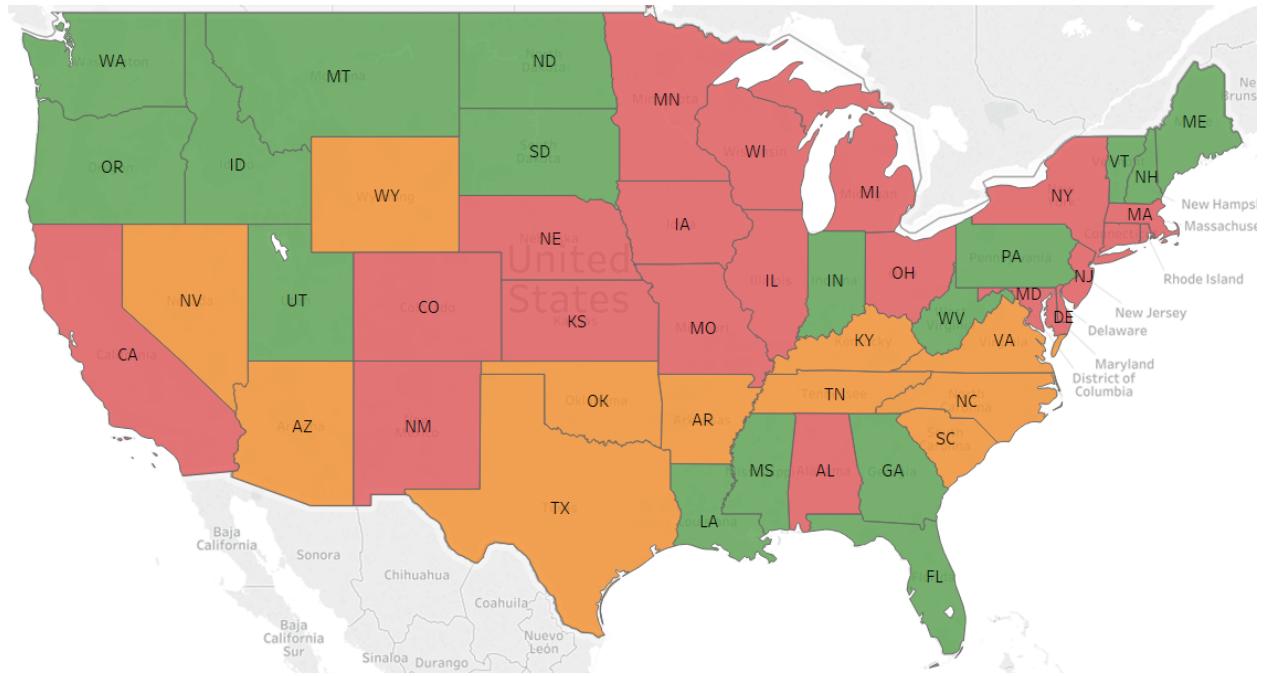
Overall DC is among the top states for all three types of crimes and North Dakota is among the bottom three for all types of crimes. Most of the sun belt states have higher crime rate compared to the rest of USA.

## Shall carry law vs crime:



Nation's average crime rate and the number of states with shall carry law doesn't show a clear trend. In the years from 1989 to 1992 there is an increase in the number of states which passed the shall carry law however there is an increase in the nationwide average for crime. Where as from 1994 to 1997 the increase in number of states with the shall carry law is corelated to the decrease in crime. 1997 to 1999 where no new states implemented the law, there is still a decrease in the number of crimes, suggesting other variables are driving crime rate.

Grouping states based on adoption of Shall carry law:

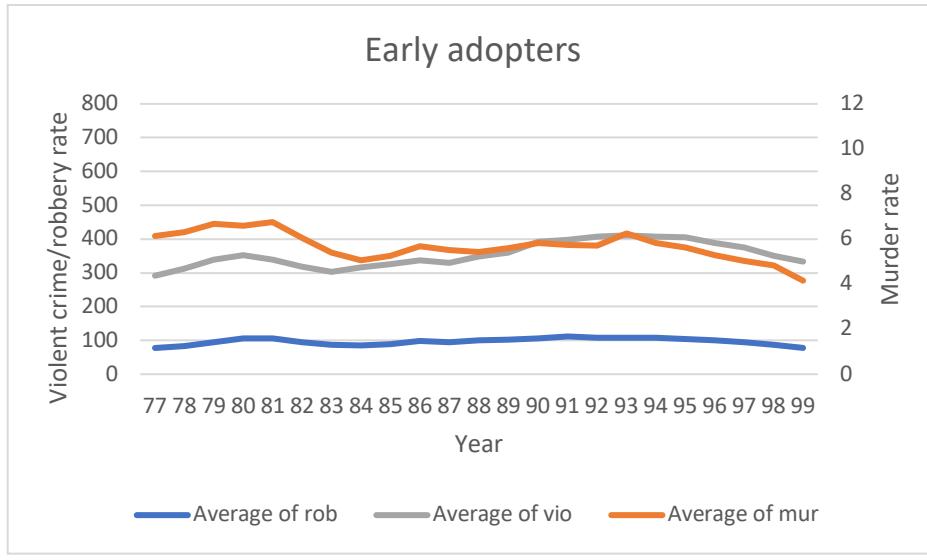
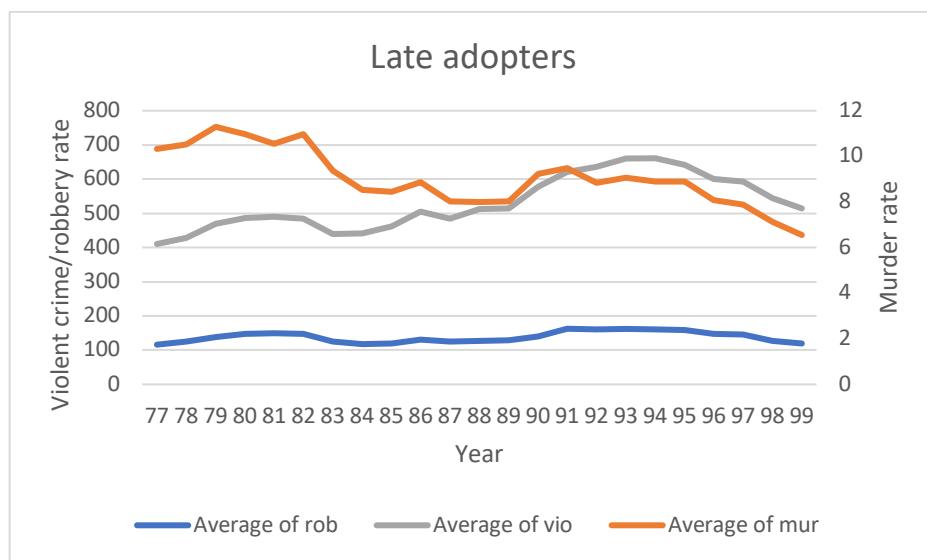
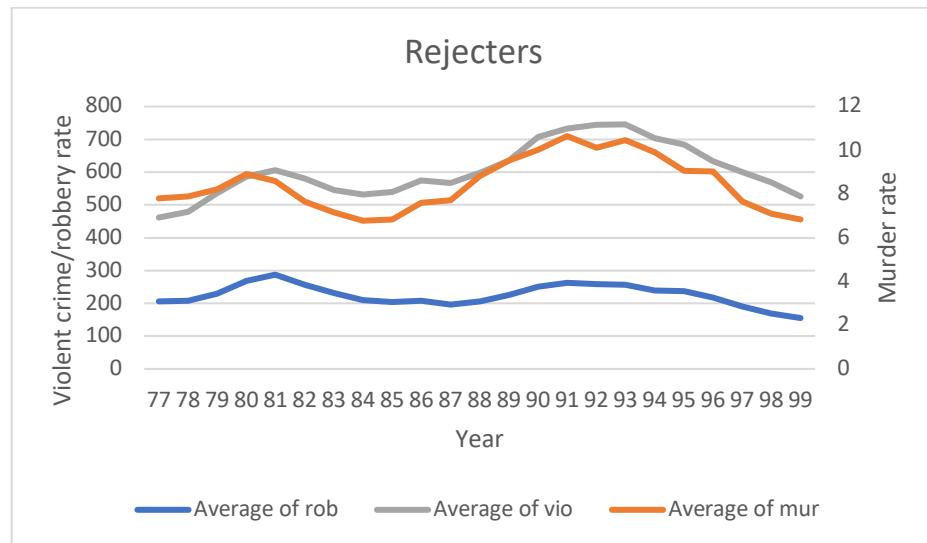


State	Average of shall	Percentile Rank	Group
AL	0	0	1
CA	0	0	1
CO	0	0	1
CT	0	0	1
DC	0	0	1
DE	0	0	1
HI	0	0	1
IA	0	0	1
IL	0	0	1
KS	0	0	1
MA	0	0	1
MD	0	0	1
MI	0	0	1
MN	0	0	1
MO	0	0	1
NE	0	0	1
NJ	0	0	1
NM	0	0	1
NY	0	0	1
OH	0	0	1
RI	0	0	1
WI	0	0	1

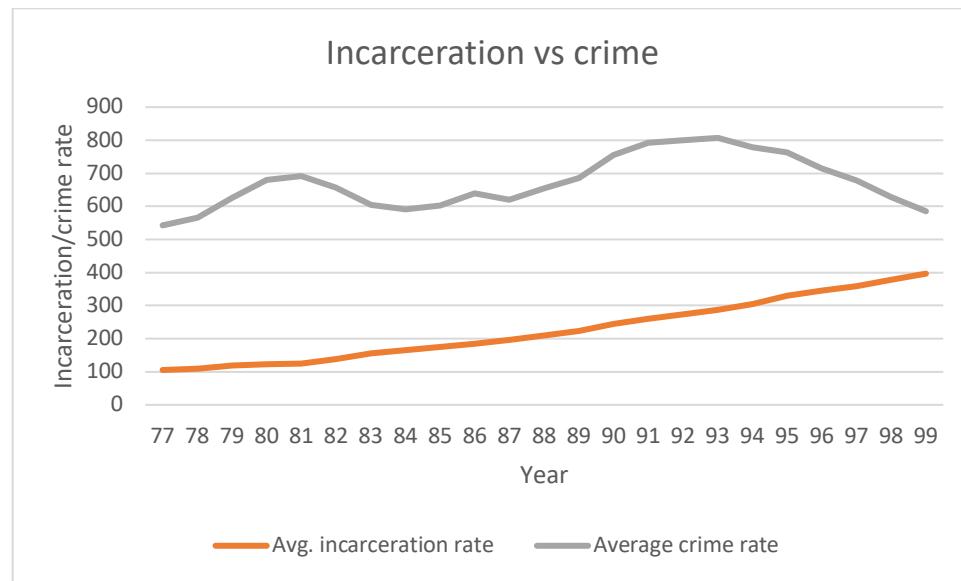
State	Average of shall	Percentile Rank	Group
KY	0.130435	0.44	2
SC	0.130435	0.44	2
TX	0.130435	0.44	2
AR	0.173913	0.5	2
NC	0.173913	0.5	2
NV	0.173913	0.5	2
OK	0.173913	0.5	2
VA	0.173913	0.5	2
AK	0.217391	0.6	2
AZ	0.217391	0.6	2
TN	0.217391	0.6	2
WY	0.217391	0.6	2

State	Average of shall	Percentile Rank	Group
LA	0.347826	0.68	3
MT	0.347826	0.68	3
ID	0.391304	0.72	3
MS	0.391304	0.72	3
OR	0.391304	0.72	3
GA	0.434783	0.78	3
PA	0.434783	0.78	3
WV	0.434783	0.78	3
FL	0.521739	0.84	3
UT	0.565217	0.86	3
ND	0.608696	0.88	3
SD	0.608696	0.88	3
ME	0.782609	0.92	3
IN	1	0.94	3
NH	1	0.94	3
VT	1	0.94	3
WA	1	0.94	3

Once the states were split based on their adoption of shall carry law, we observed some difference in trends. The group of states that rejected the shall carry law on an average has higher crime than the states that adopted shall carry. Shall carry law seems to affect murders the most as the increase in murders from 1985 – 1991 in the rejecter states is not present in the adopter states. Also the late adopter states show a sharper drop in the number of murders form 1983 than the other two groups.



Incarceration vs crime rate:



A bivariate between the previous year's incarceration rate and the current years crime does not show any conclusive trends. Incarceration has been on a steady rise however crime rate has been varying.

Conclusion from EDA:

1. Shall law seems to have a different impact on different types of crimes
2. Incarceration over time does not seem to have a clear impact on crime rate

Approach:

We will create three different models with the dependent variable being each type of crime:

## Model Building

Violent Crime Rate:

Model 1:

We first run a Pooled Effect model with all the independent variables. The results are shown below:

. reg vio incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall						
Source	SS	df	MS	Number of obs	=	1,173
Model	94776406.5	8	11847050.8	F(8, 1164)	=	381.10
Residual	36184330.1	1,164	31086.1942	Prob > F	=	0.0000
				R-squared	=	0.7237
				Adj R-squared	=	0.7218
Total	130960737	1,172	111741.243	Root MSE	=	176.31
voie	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.8155948	.044174	18.46	0.000	.7289253	.9022642
pb1064	11.29196	6.864417	1.64	0.100	-2.176053	24.75998
pw1064	2.702383	3.453603	0.78	0.434	-4.0736	9.478366
pm1029	9.316543	4.441374	2.10	0.036	.602549	18.03054
pop	18.50567	1.054859	17.54	0.000	16.43603	20.57531
avginc	1.234759	3.207344	0.38	0.700	-5.058062	7.527581
density	94.66882	5.428439	17.44	0.000	84.0182	105.3194
i.shall	-92.72125	13.4257	-6.91	0.000	-119.0625	-66.37997
_cons	-178.7774	224.0104	-0.80	0.425	-618.2867	260.7319

Our next step is to run the same model using the robust standard errors to account for the incorrect standard errors.

. reg vio incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, robust						
Linear regression				Number of obs	=	1,173
				F(8, 1164)	=	135.12
				Prob > F	=	0.0000
				R-squared	=	0.7237
				Root MSE	=	176.31
voie	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.8155948	.0911446	8.95	0.000	.6367687	.9944208
pb1064	11.29196	8.354651	1.35	0.177	-5.099897	27.68382
pw1064	2.702383	4.126139	0.65	0.513	-5.393119	10.79788
pm1029	9.316543	4.778368	1.95	0.051	-.0586355	18.69172
pop	18.50567	1.40521	13.17	0.000	15.74865	21.2627
avginc	1.234759	3.291204	0.38	0.708	-5.222597	7.692116
density	94.66882	8.261852	11.46	0.000	78.45904	110.8786
i.shall	-92.72125	14.0256	-6.61	0.000	-120.2395	-65.20297
_cons	-178.7774	270.354	-0.66	0.509	-709.2131	351.6584

Due to the time effect included in panel data, a variable x in time t has an influence over xt+1. Therefore, the assumption that the error terms between different time periods are not serially correlated is violated. Additionally, the variance of the error term may also be different over time.

To account for the problem of serial correlation which causes the coefficient estimates to become inefficient and the standard errors to be misleading, the next step is to run the panel data as a fixed and random effects model.

### Model 2:

A random effects model is run next as it has more degrees of freedom than the fixed effect model. Below are the results from the random effects model:

. xtreg vio incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, re						
Random-effects GLS regression		Number of obs = 1,173				
Group variable: stateid		Number of groups = 51				
R-sq:		Obs per group:				
within = 0.1665		min = 23				
between = 0.7632		avg = 23.0				
overall = 0.6967		max = 23				
		Wald chi2(8) = 399.18				
corr(u_i, X) = 0 (assumed)		Prob > chi2 = 0.0000				
<hr/>						
vio	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
incarc_rate	.3931534	.0397895	9.88	0.000	.3151674	.4711393
pb1064	20.61077	7.196501	2.86	0.004	6.505887	34.71565
pw1064	8.026554	3.093081	2.60	0.009	1.964226	14.08888
pm1029	-9.640604	3.623266	-2.66	0.008	-16.74207	-2.539133
pop	13.15851	3.07476	4.28	0.000	7.132095	19.18493
avginc	-6.025157	3.514785	-1.71	0.086	-12.91401	.8636953
density	104.7166	15.97841	6.55	0.000	73.39954	136.0337
i.shall	-37.19344	11.6025	-3.21	0.001	-59.93392	-14.45296
_cons	-54.70189	231.1279	-0.24	0.813	-507.7042	398.3004
sigma_u	133.02093					
sigma_e	98.505026					
rho	.64583883	(fraction of variance due to u_i)				

Model 3:

The fixed effects model is run next. The results are as follows:

. xtreg vio incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe						
Fixed-effects (within) regression						Number of obs = 1,173
Group variable: stateid						Number of groups = 51
R-sq:						Obs per group:
within = 0.2038						min = 23
between = 0.3446						avg = 23.0
overall = 0.2658						max = 23
						F(8, 1114) = 35.64
corr(u_i, Xb) = -0.8388						Prob > F = 0.0000
vio	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.0967879	.0573672	1.69	0.092	-0.015772	.2093479
pb1064	12.80803	10.88266	1.18	0.239	-8.54479	34.16085
pw1064	10.32325	3.110094	3.32	0.001	4.220951	16.42556
pm1029	-23.86385	3.924761	-6.08	0.000	-31.56461	-16.16309
pop	12.24406	5.346751	2.29	0.022	1.753224	22.7349
avginc	-4.066645	3.621108	-1.12	0.262	-11.17161	3.038316
density	-155.9533	52.11761	-2.99	0.003	-258.213	-53.69351
i.shall	-18.57508	11.56324	-1.61	0.108	-41.26327	4.113097
_cons	203.0109	235.8217	0.86	0.389	-259.6938	665.7157
sigma_u	500.37955					
sigma_e	98.505026					
rho	.96269175	(fraction of variance due to u_i)				
F test that all u_i=0: F(50, 1114) = 52.30						Prob > F = 0.0000

### Hausman Test:

Hausman's test is run next to test if the dependent variables of the random effects model are correlated with its error. Below are the results:

	Coefficients			
	(b) fixed	(B) random	(b-B) Difference	sqrt(diag(V_b-V_B)) S.E.
incarc_rate	.0967879	.3931534	-.2963654	.0413254
pb1064	12.80803	20.61077	-7.802738	8.163496
pw1064	10.32325	8.026554	2.2967	.3248615
pm1029	-23.86385	-9.640604	-14.22325	1.508542
pop	12.24406	13.15851	-.9144534	4.374196
avginc	-4.066645	-6.025157	1.958512	.8710399
density	-155.9533	104.7166	-260.6699	49.60782
1.shall	-18.57508	-37.19344	18.61836	.

b = consistent under  $H_0$  and  $H_a$ ; obtained from xtreg  
 B = inconsistent under  $H_a$ , efficient under  $H_0$ ; obtained from xtreg

Test:  $H_0$ : difference in coefficients not systematic

```

chi2(8) = (b-B)'[(V_b-V_B)^(-1)](b-B)
          =       64.49
Prob>chi2 =      0.0000
(V_b-V_B is not positive definite)
    
```

The p-value is < 0.00. So, we reject the null hypothesis that the coefficients of the fixed effect and the random effects are the same. Hence, the fixed effects model is selected.

Model 4:

The fixed effect model is then run, with the Cluster Robust Standard Errors to get more efficient coefficients.

Robust						
vio	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.0967879	.1567361	0.62	0.540	-.2180259	.4116017
pb1064	12.80803	26.48154	0.48	0.631	-40.38171	65.99777
pw1064	10.32325	6.148925	1.68	0.099	-2.027226	22.67373
pm1029	-23.86385	9.45301	-2.52	0.015	-42.85078	-4.876919
pop	12.24406	9.561442	1.28	0.206	-6.96066	31.44878
avginc	-4.066645	6.276427	-0.65	0.520	-16.67322	8.539929
density	-155.9533	113.4665	-1.37	0.175	-383.8573	71.95084
i.shall	-18.57508	19.47905	-0.95	0.345	-57.69992	20.54975
_cons	203.0109	368.3606	0.55	0.584	-536.8631	942.8849
sigma_u	500.37955					
sigma_e	98.505026					
rho	.96269175	(fraction of variance due to u_i)				

None of the variables except for the percent of males in the age group 10 to 29 (pm1029) are significant at a 95% confidence level.

Model 5:

We then use log variables to account for the skewness observed in the histogram of the variables

Robust						
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
log_vio						
log_incrate	-.0633165	.0669849	-0.95	0.349	-.1978596	.0712267
pb1064	.1167926	.0320688	3.64	0.001	.0523805	.1812046
pw1064	.0413543	.0143821	2.88	0.006	.0124671	.0702415
pm1029	-.0643953	.0259147	-2.48	0.016	-.1164465	-.0123441
log_pop	-.1258555	.1625252	-0.77	0.442	-.452297	.200586
avginc	-.0065709	.0132301	-0.50	0.622	-.0331442	.0200025
density	-.1285301	.0788056	-1.63	0.109	-.2868158	.0297557
i.shall	-.037671	.0435654	-0.86	0.391	-.1251748	.0498327
_cons	4.446236	.8064566	5.51	0.000	2.82642	6.066051
sigma_u						
sigma_e						
rho	.9524155	(fraction of variance due to u_i)				

We see that both percent of black people as well as white people in the age group 10 to 64 (pb1064 and pw1064) have become significant at the 95% confidence level.

### Model 6:

Dummy variables for each year is then added to check for time effects on the fixed effects model. The results are given below:

Robust						
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
log_vio						
log_incrate	-.1042194	.0691802	-1.51	0.138	-.243172	.0347332
pb1064	.0073456	.053356	0.14	0.891	-.099823	.1145142
pw1064	.0000793	.0249287	0.00	0.997	-.0499914	.05015
pm1029	.0804462	.0526424	1.53	0.133	-.0252892	.1861817
log_pop	-.1936537	.1932852	-1.00	0.321	-.5818785	.1945711
avginc	-.0005546	.0162047	-0.03	0.973	-.0331028	.0319935
density	-.1041343	.0796751	-1.31	0.197	-.2641666	.0558979
i.year	-.02888	.0393632	-0.73	0.467	-.1079432	.0501833
year						
78	.0674092	.0160657	4.20	0.000	.0351404	.099678
79	.1846194	.0275775	6.69	0.000	.1292283	.2400104
80	.2430197	.0399504	6.08	0.000	.162777	.3232624
81	.2509862	.0433771	5.79	0.000	.1638608	.3381116
82	.2438153	.0545544	4.47	0.000	.1342396	.353391
83	.2229004	.0673075	3.31	0.002	.0877094	.3580915
84	.2667187	.0822677	3.24	0.002	.1014791	.4319583
85	.326206	.0976441	3.34	0.002	.1300821	.52233
86	.4154293	.1142612	3.64	0.001	.1859289	.6449297
87	.4249587	.1312944	3.24	0.002	.1612461	.6886713
88	.4970081	.145989	3.40	0.001	.2037805	.7902356
89	.5623485	.1604123	3.51	0.001	.2401509	.8845462
90	.6952929	.2055371	3.38	0.001	.2824595	1.108126
91	.7585728	.2150783	3.53	0.001	.3265754	1.19057
92	.8009791	.2275746	3.52	0.001	.3438821	1.258076
93	.8321949	.2375269	3.50	0.001	.3551081	1.309282
94	.8278492	.2461974	3.36	0.001	.3333472	1.322351
95	.8328327	.2576959	3.23	0.002	.3152352	1.35043
96	.7876821	.2695923	2.92	0.005	.24619	1.329174
97	.7763838	.2774862	2.80	0.007	.2190363	1.333731
98	.7311231	.2891897	2.53	0.015	.1502684	1.311978
99	.6810194	.2995348	2.27	0.027	.0793861	1.282653
_cons	4.994805	1.318697	3.79	0.000	2.346123	7.643486
sigma_u	.8266663					
sigma_e	.13895235					
rho	.97252288	(fraction of variance due to u_i)				

F-Test:

An F-test is performed on the year dummy-variables, to assess if the fixed effect model is affected by the time component of the panel data.

```
. testparm i.year

( 1) 78.year = 0
( 2) 79.year = 0
( 3) 80.year = 0
( 4) 81.year = 0
( 5) 82.year = 0
( 6) 83.year = 0
( 7) 84.year = 0
( 8) 85.year = 0
( 9) 86.year = 0
(10) 87.year = 0
(11) 88.year = 0
(12) 89.year = 0
(13) 90.year = 0
(14) 91.year = 0
(15) 92.year = 0
(16) 93.year = 0
(17) 94.year = 0
(18) 95.year = 0
(19) 96.year = 0
(20) 97.year = 0
(21) 98.year = 0
(22) 99.year = 0

      F( 22,      50) =   21.71
      Prob > F =    0.0000
```

The results show that the time effects do indeed affect the fixed effect model.

### Model 7:

To understand the effect of non-linear relationship of the independent variables on violent crime rates, interaction variables were added. The only significant interaction was shall\*percentage of black population. The results are as follows:

```
. xtreg log_vio pw1064 avginc pm1029 log_pop c.log_incrate density i.shall##c.p  
> b1064 i.year, fe vce(cluster stateid)

Fixed-effects (within) regression                               Number of obs     =    1,173
Group variable: stateid                                     Number of groups  =      51

R-sq:                                                               Obs per group:
    within  = 0.4727                                         min =          23
    between = 0.4730                                         avg =        23.0
    overall = 0.2895                                         max =          23

                                                F(31, 50)      =     74.84
corr(u_i, Xb)  = -0.7825                                     Prob > F       =   0.0000

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

log_vio	Robust					
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
pw1064	.0023161	.0247531	0.09	0.926	-.047402	.0520342
avginc	-.0040728	.0142901	-0.29	0.777	-.0327754	.0246298
pm1029	.0982148	.0517161	1.90	0.063	-.00566	.2020896
log_pop	-.2238996	.1834917	-1.22	0.228	-.5924535	.1446543
log_incrate	-.1014184	.0658296	-1.54	0.130	-.233641	.0308042
density	-.0651606	.0724121	-0.90	0.373	-.2106046	.0802835
1.shall	-.2185981	.0630807	-3.47	0.001	-.3452995	-.0918967
pb1064	-.0123516	.0545962	-0.23	0.822	-.1220113	.0973081
shall# c.pb1064						
1	.0357479	.008361	4.28	0.000	.0189544	.0525415
year						
78	.0733303	.0168294	4.36	0.000	.0395276	.1071331
79	.1945539	.0289443	6.72	0.000	.1364176	.2526903
80	.2567862	.0413274	6.21	0.000	.1737777	.3397946
81	.2682949	.0445629	6.02	0.000	.1787876	.3578022
82	.2700751	.055902	4.83	0.000	.1577927	.3823575
83	.2563013	.0689346	3.72	0.001	.1178421	.3947605
84	.3091804	.0848483	3.64	0.001	.1387576	.4796032
85	.376629	.1008441	3.73	0.000	.1740777	.5791804
86	.4784283	.1167863	4.10	0.000	.243856	.7130006
87	.4984314	.1336427	3.73	0.000	.230002	.7668607
88	.5783546	.1486463	3.89	0.000	.2797897	.8769196
89	.6510437	.1630797	3.99	0.000	.3234884	.978599
90	.7864282	.2106893	3.73	0.000	.3632462	1.20961
91	.8550086	.2203393	3.88	0.000	.4124441	1.297573
92	.9009866	.2328843	3.87	0.000	.4332248	1.368748
93	.9362999	.2428609	3.86	0.000	.4484994	1.4241
94	.9364301	.2520807	3.71	0.001	.4301112	1.442749
95	.9452227	.2658301	3.56	0.001	.4112871	1.479158
96	.8971862	.2780709	3.23	0.002	.3386643	1.455708
97	.885901	.2865608	3.09	0.003	.3103267	1.461475
98	.843688	.2990441	2.82	0.007	.2430402	1.444336
99	.7955219	.3104489	2.56	0.013	.1719669	1.419077
_cons	4.671467	1.296198	3.60	0.001	2.067976	7.274958
sigma_u	.8573313					
sigma_e	.13334069					
rho	.97638175	(fraction of variance due to u_i)				

The adjusted R-squared value was calculated at .4584

The linear variable percentage of males is significant at the 10% significance level, while the variable shall and the interaction effect (shall\*percentage of black population) along with the time periods from 1978 through 1999 were found to be significant at the 95% confidence level.

The interpretations of the coefficients are given below:

- As the years increase the violent crime rate increases.
- When a state has the shall law issued, there is an associated substantial decrease in violent crime rate.
- The percentage of males in the population has an increasing effect on the violent crime rate.
- When the shall law is implemented in a state, the percentage of black population on violent crime rate increases.

## Murder Rate

Model 1:

We first run a Pooled Effect model with all the independent variables. The results are shown below:

. reg mur incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall						
Source	SS	df	MS	Number of obs	=	1,173
Model	51892.2351	8	6486.52939	F(8, 1164)	=	523.14
Residual	14432.6089	1,164	12.3991486	Prob > F	=	0.0000
Total	66324.844	1,172	56.5911638	R-squared	=	0.7824
				Adj R-squared	=	0.7809
				Root MSE	=	3.5212
mur	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.0241046	.0008822	27.32	0.000	.0223736	.0258355
pb1064	.030323	.1370932	0.22	0.825	-.2386544	.2993004
pw1064	-.0326635	.0689739	-0.47	0.636	-.1679905	.1026635
pm1029	1.011478	.0887012	11.40	0.000	.8374463	1.185511
pop	.1661031	.0210672	7.88	0.000	.1247691	.207437
avginc	-.3346584	.0640557	-5.22	0.000	-.4603359	-.2089809
density	2.496043	.1084144	23.02	0.000	2.283334	2.708753
i.shall	-1.077353	.2681323	-4.02	0.000	-1.60343	-.5512763
_cons	-8.99181	4.473839	-2.01	0.045	-17.7695	-.21412

Our next step is to run the same model using the robust standard errors in order to account for the incorrect standard errors.

<pre>. reg mur incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, robust</pre>						
Linear regression						
					Number of obs	= 1,173
					F(8, 1164)	= 138.76
					Prob > F	= 0.0000
					R-squared	= 0.7824
					Root MSE	= 3.5212
<hr/>						
mur	Robust					
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.0241046	.0020404	11.81	0.000	.0201012	.0281079
pb1064	.030323	.1491101	0.20	0.839	-.2622317	.3228777
pw1064	-.0326635	.0695919	-0.47	0.639	-.169203	.1038761
pm1029	1.011478	.1151811	8.78	0.000	.7854927	1.237464
pop	.1661031	.0263041	6.31	0.000	.1144942	.2177119
avginc	-.3346584	.070592	-4.74	0.000	-.4731603	-.1961565
density	2.496043	.3271928	7.63	0.000	1.85409	3.137997
1.shall	-1.077353	.2202453	-4.89	0.000	-1.509475	-.6452307
_cons	-8.99181	4.232699	-2.12	0.034	-17.29638	-.687236

Due to the time effect included in a panel data, a variable x in time t has an influence over xt+1. Therefore, the assumption that the error terms between different time periods are not correlated is violated. Additionally, the variance of the error term may also be different over time.

To account for the problem of serial correlation which causes the coefficient estimates to become inefficient and the standard errors to be misleading, the next step is to run the panel data as a fixed and random effects model.

Model 2:

A random effects model is run next as it has more degrees of freedom than the fixed effect model.  
Below are the results from the random effects model:

. xtreg mur incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, re						
Random-effects GLS regression		Number of obs = 1,173				
Group variable: stateid		Number of groups = 51				
R-sq:		Obs per group:				
within = 0.1641		min = 23				
between = 0.8928		avg = 23.0				
overall = 0.7575		max = 23				
		Wald chi2(8) = 753.31				
corr(u_i, X) = 0 (assumed)		Prob > chi2 = 0.0000				
<hr/>						
mur		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
incarc_rate		.0212448	.0010772	19.72	0.000	.0191335 .023356
pb1064		-.382155	.1770604	-2.16	0.031	-.729187 -.035123
pw1064		-.1868612	.0859997	-2.17	0.030	-.3554174 -.0183049
pm1029		1.237827	.1009954	12.26	0.000	1.039879 1.435774
pop		-.0084263	.0508474	-0.17	0.868	-.1080853 .0912327
avginc		.1282011	.0921206	1.39	0.164	-.052352 .3087542
density		2.264481	.2402161	9.43	0.000	1.793666 2.735296
i.shall		-1.3641	.3260685	-4.18	0.000	-2.003183 -.7250177
_cons		-5.437536	6.152176	-0.88	0.377	-17.49558 6.620508
<hr/>						
sigma_u		1.6370471				
sigma_e		2.7074164				
rho		.26772388 (fraction of variance due to u_i)				

Model 3:

The fixed effects model is run next. The results are as follows:

. xtreg mur incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe						
Fixed-effects (within) regression						Number of obs = 1,173
Group variable: stateid						Number of groups = 51
R-sq:						Obs per group:
within = 0.3204						min = 23
between = 0.7745						avg = 23.0
overall = 0.5752						max = 23
						F(8, 1114) = 65.66
corr(u_i, Xb) = -0.9787						Prob > F = 0.0000
<hr/>						
mur		Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
incarc_rate		.0079266	.0015767	5.03	0.000	.0048329 .0110203
pb1064		-1.783213	.2991105	-5.96	0.000	-2.370096 -1.196329
pw1064		-.0710444	.0854811	-0.83	0.406	-.2387666 .0966777
pm1029		.5870231	.1078723	5.44	0.000	.3753673 .7986789
pop		-.5212885	.1469558	-3.55	0.000	-.8096298 -.2329472
avginc		.5839008	.0995264	5.87	0.000	.3886205 .779181
density		-6.825431	1.432456	-4.76	0.000	-9.636045 -4.014816
i.shall		-.6550583	.3178163	-2.06	0.040	-1.278644 -.0314723
_cons		7.47542	6.481573	1.15	0.249	-5.242047 20.19289
<hr/>						
sigma_u		20.321614				
sigma_e		2.7074164				
rho		.98255975	(fraction of variance due to u_i)			
F test that all u_i=0: F(50, 1114) = 17.10						Prob > F = 0.0000

### Hausman Test:

Hausman's test is run next to test if the dependent variables of the random effects model are correlated with its error. Below are the results:

```
. hausman fixed2 random2
```

	Coefficients		(b-B)	sqrt(diag(V_b-V_B))
	(b) fixed2	(B) random2	Difference	S.E.
incarc_rate	.0079266	.0212448	-.0133182	.0011514
pb1064	-1.783213	-.382155	-1.401058	.2410741
pw1064	-.0710444	-.1868612	.1158167	.
pml029	.5870231	1.237827	-.6508035	.0378992
pop	-.5212885	-.0084263	-.5128622	.1378787
avginc	.5839008	.1282011	.4556996	.0376735
density	-6.825431	2.264481	-9.089911	1.41217
1.shall	-.6550583	-1.3641	.7090419	.

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)
          =      313.25
Prob>chi2 =    0.0000
(V_b-V_B is not positive definite)
```

The p-value is < 0.00. So, we reject the null hypothesis that the coefficients of the fixed effect and the random effects are the same. Hence, the fixed effects model is selected.

Model 4:

The fixed effect model is then run, with the Cluster Robust Standard Errors to get more efficient coefficients.

xtreg mur incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe vce > (cluster stateid)						
Fixed-effects (within) regression Group variable: stateid			Number of obs = 1,173 Number of groups = 51			
R-sq:			Obs per group:			
within	= 0.3204			min =	23	
between	= 0.7745			avg =	23.0	
overall	= 0.5752			max =	23	
			F(8, 50)	=	2942.43	
corr(u_i, Xb)	= -0.9787		Prob > F	=	0.0000	
(Std. Err. adjusted for 51 clusters in stateid)						
mur	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.0079266	.0112422	0.71	0.484	-.0146541	.0305072
pb1064	-1.783213	2.063557	-0.86	0.392	-5.927989	2.361563
pw1064	-.0710444	.2061573	-0.34	0.732	-.4851234	.3430346
pm1029	.5870231	.4259051	1.38	0.174	-.2684324	1.442479
pop	-.5212885	.3999463	-1.30	0.198	-1.324604	.2820274
avginc	.5839008	.3060605	1.91	0.062	-.0308398	1.198641
density	-6.825431	10.92914	-0.62	0.535	-28.77726	15.1264
i.shall	-.6550583	.3940816	-1.66	0.103	-1.446595	.136478
_cons	7.47542	11.74442	0.64	0.527	-16.11394	31.06478
sigma_u	20.321614					
sigma_e	2.7074164					
rho	.98255975	(fraction of variance due to u_i)				

None of the variables are significant at a 95% confidence level.

Model 5:

We then use the log form of density to account for the skewness observed in the histogram.

Robust						
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.0087949	.0043522	2.02	0.049	.0000532	.0175367
pb1064	-2.524701	1.136673	-2.22	0.031	-4.807776	-.2416258
pw1064	-.0104977	.1470838	-0.07	0.943	-.3059243	.2849289
pm1029	.4505844	.30753	1.47	0.149	-.1671078	1.068277
log_pop	-190.3064	46.89719	-4.06	0.000	-284.5022	-96.11065
avginc	.5805541	.2367686	2.45	0.018	.1049903	1.056118
log_density	184.6991	47.20772	3.91	0.000	89.87961	279.5186
i.shall	-.6888283	.3563621	-1.93	0.059	-1.404603	.0269461
_cons	680.3142	174.8007	3.89	0.000	329.2166	1031.412
sigma_u	267.28259					
sigma_e	2.6513293					
rho	.99990161	(fraction of variance due to u_i)				

The variables incarceration rate, percentage of black population, log of population, average income and log of density are significant at the 95% confidence level.

Model 6:

Dummy variable for each year is then added to check for time effects on the fixed effect model. The results are given below:

```
. xtreg mur incarc_rate pb1064 pw1064 pm1029 log_pop avginc log_density i.shall  
> i.year, fe vce(cluster stateid)  
  
Fixed-effects (within) regression  
Group variable: stateid  
  
R-sq:  
    within = 0.4374  
    between = 0.1515  
    overall = 0.1263  
  
corr(u_i, Xb) = -0.9994  
  
Number of obs = 1,173  
Number of groups = 51  
  
Obs per group:  
    min = 23  
    avg = 23.0  
    max = 23  
  
F(30, 50) = 54.30  
Prob > F = 0.0000  
  
(Std. Err. adjusted for 51 clusters in stateid)
```

mur	Robust					
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.0105974	.0043489	2.44	0.018	.0018623	.0193325
pb1064	-2.917727	1.876761	-1.55	0.126	-6.687311	.8518581
pw1064	-.2606319	.4428649	-0.59	0.559	-1.150152	.6288884
pm1029	1.18437	1.076745	1.10	0.277	-.9783349	3.347075
log_pop	-136.8291	44.31992	-3.09	0.003	-225.8483	-47.8099
avginc	.9581055	.2575099	3.72	0.001	.4408816	1.475329
log_density	131.2615	44.31893	2.96	0.005	42.24434	220.2787
1.shall	-.1771192	.4208353	-0.42	0.676	-1.022392	.6681535
year						
78	.0048033	.2254295	0.02	0.983	-.4479851	.4575917
79	.8002806	.4146133	1.93	0.059	-.0324947	1.633056
80	1.701343	.7723262	2.20	0.032	.15008	3.252606
81	1.716899	1.053163	1.63	0.109	-.3984403	3.832239
82	1.533093	1.200636	1.28	0.208	-.878456	3.944642
83	.7634724	1.335948	0.57	0.570	-1.919857	3.446802
84	.0544027	1.488729	0.04	0.971	-2.935797	3.044602
85	.171753	1.605452	0.11	0.915	-3.052893	3.396399
86	.7795916	1.906351	0.41	0.684	-3.049426	4.60861
87	.6941002	2.224872	0.31	0.756	-3.774686	5.162886
88	1.209253	2.834292	0.43	0.671	-4.48359	6.902096
89	1.614317	3.210232	0.50	0.617	-4.833623	8.062258
90	2.623762	4.173563	0.63	0.532	-5.759085	11.00661
91	3.386994	4.507692	0.75	0.456	-5.666972	12.44096
92	2.990373	4.567859	0.65	0.516	-6.184443	12.16519
93	3.565938	4.787037	0.74	0.460	-6.049109	13.18099
94	3.090143	4.759568	0.65	0.519	-6.469732	12.65002
95	2.591256	4.762234	0.54	0.589	-6.973972	12.15648
96	2.201133	5.058536	0.44	0.665	-7.959235	12.3615
97	1.282473	4.88219	0.26	0.794	-8.523695	11.08864
98	.3123943	4.86412	0.06	0.949	-9.457477	10.08227
99	-.3849733	4.887195	-0.08	0.938	-10.20119	9.431247
_cons	485.4728	168.0552	2.89	0.006	147.9239	823.0217
sigma_u	188.98368					
sigma_e	2.4882256					
rho	.99982668	(fraction of variance due to u_i)				

The percentage of black population became insignificant after including the time effect.

### F-Test:

An F-test is performed on the year dummy-variables, to assess if the fixed effect model is affected by the time component of the panel data.

```
. testparm i.year

( 1) 78.year = 0
( 2) 79.year = 0
( 3) 80.year = 0
( 4) 81.year = 0
( 5) 82.year = 0
( 6) 83.year = 0
( 7) 84.year = 0
( 8) 85.year = 0
( 9) 86.year = 0
(10) 87.year = 0
(11) 88.year = 0
(12) 89.year = 0
(13) 90.year = 0
(14) 91.year = 0
(15) 92.year = 0
(16) 93.year = 0
(17) 94.year = 0
(18) 95.year = 0
(19) 96.year = 0
(20) 97.year = 0
(21) 98.year = 0
(22) 99.year = 0

F( 22,      50) =     8.49
                  Prob > F =  0.0000
```

The results show that the time effects do indeed affect the fixed effect model.

Model 7:

To understand the effect of non-linear relationship of the independent variables on murder rate, many combinations of interactions variables were included. The most meaningful interactions were found to be log of population\*incarceration rate and log of density\*percentage of black population. The results are as follows:

```
. xtreg mur pw1064 pm1029 c.log_pop##c.incarc_rate avginc c.log_density##c.pb10
> 64 i.shall i.year, fe vce(cluster stateid)

Fixed-effects (within) regression                               Number of obs     =      1,173
Group variable: stateid                                     Number of groups  =       51

R-sq:
    within  = 0.5224                                         Obs per group:
    between = 0.1432                                         min  =        23
    overall = 0.1200                                         avg  =     23.0
                                                       max  =        23

                                                F(32, 50)      =     374.15
corr(u_i, Xb)  = -0.9993                                     Prob > F        =   0.0000

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

mur	Robust					
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
pw1064	-.7906742	.3973302	-1.99	0.052	-1.588735	.007387
pm1029	1.580729	1.071888	1.47	0.147	-.5722208	3.733678
log_pop	-131.4554	37.50356	-3.51	0.001	-206.7836	-56.12732
incarc_rate	.0038021	.0021354	1.78	0.081	-.0004869	.0080912
c.log_pop# c. incarc_rate	-.003262	.0012722	-2.56	0.013	-.0058172	-.0007067
avginc	1.045786	.1746025	5.99	0.000	.6950868	1.396486
log_density	128.6868	37.93322	3.39	0.001	52.49571	204.8779
pb1064	-3.976124	1.339472	-2.97	0.005	-6.666532	-1.285715
c. log_density# c.pb1064	-.8507762	.1370476	-6.21	0.000	-1.126044	-.5755079
1.shall	-.2059363	.281548	-0.73	0.468	-.7714422	.3595696

year						
78	.0009581	.1846999	0.01	0.996	-.3700225	.3719388
79	.8534713	.3612955	2.36	0.022	.127788	1.579155
80	1.682538	.6939248	2.42	0.019	.2887491	3.076327
81	1.770138	.9067815	1.95	0.057	-.0511859	3.591463
82	1.694621	1.0194	1.66	0.103	-.3529034	3.742146
83	1.021242	1.206436	0.85	0.401	-1.401955	3.444439
84	.3052377	1.368936	0.22	0.824	-2.44435	3.054826
85	.4306723	1.503104	0.29	0.776	-2.588401	3.449745
86	1.048517	1.851297	0.57	0.574	-2.669923	4.766957
87	.9959423	2.180739	0.46	0.650	-3.3842	5.376085
88	1.537691	2.850245	0.54	0.592	-4.187194	7.262575
89	1.970244	3.260389	0.60	0.548	-4.57844	8.518928
90	4.13956	4.154494	1.00	0.324	-4.204987	12.48411
91	5.093141	4.462665	1.14	0.259	-3.870385	14.05667
92	4.811734	4.525828	1.06	0.293	-4.27866	13.90213
93	5.481632	4.727054	1.16	0.252	-4.012935	14.9762
94	5.149108	4.714888	1.09	0.280	-4.321023	14.61924
95	4.906196	4.726352	1.04	0.304	-4.586961	14.39935
96	4.650774	4.991216	0.93	0.356	-5.374378	14.67593
97	3.830479	4.798286	0.80	0.428	-5.807161	13.46812
98	2.996614	4.769789	0.63	0.533	-6.583789	12.57702
99	2.429816	4.800178	0.51	0.615	-7.211624	12.07126
_cons	497.0006	141.8224	3.50	0.001	212.1419	781.8593
sigma_u	177.07854					
sigma_e	2.2945713					
rho	.99983212	(fraction of variance due to u_i)				

The adjusted R-squared value was found to be .509.

It was found that the independent variables incarceration rate and percentage of white population was significant at the 10% significance level, while log of population, average income, log of density and percentage of black population were significant at the 95% confidence level. Additionally, the interaction effects log of population\*incarceration rate and log of density\*percentage of black population were also significant. The time dummy variables were only significant for the year 1979 and 1980 at the 5% significant level and 1981 at the 10% significance level.

The interpretations of the coefficients are below:

- Increase in percentage of white population was inversely related with murder rate.
- Increase in incarceration rate had a very negligible negative effect on murder rate.
- When population increased by 1%, there was an associated 1.31 unit decrease in murder rate.
- Increase in average income leads to increase in murder rate by 1.04 units.

- More densely populated areas lead to an increase in murder rate. In specific, 1% increase in log of density is associated with a 1.28 unit increase in murder rate.
- Increase in percentage of black population by 1 unit leads to decrease in murder rate by 3.97 units.

## Robbery

### Model 1

We first run a Pooled Effect model with all the independent variables. The results are shown below:

. reg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall						
Source	SS	df	MS	Number of obs	=	1,173
Model	26757434.5	8	3344679.31	F(8, 1164)	=	532.09
Residual	7316879.52	1,164	6285.97897	Prob > F	=	0.0000
Total	34074314	1,172	29073.6468	R-squared	=	0.7853
				Adj R-squared	=	0.7838
				Root MSE	=	79.284
rob	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.0862563	.0198641	4.34	0.000	.0472829	.1252298
pb1064	7.243095	3.086785	2.35	0.019	1.18681	13.29938
pw1064	1.502237	1.553013	0.97	0.334	-1.544781	4.549255
pm1029	3.501497	1.997193	1.75	0.080	-.4170035	7.419998
pop	10.98124	.4743482	23.15	0.000	10.05056	11.91191
avginc	3.414262	1.442275	2.37	0.018	.5845116	6.244012
density	84.22337	2.441055	34.50	0.000	79.434	89.01273
i.shall	-27.4472	6.037257	-4.55	0.000	-39.29232	-15.60207
_cons	-169.972	100.7328	-1.69	0.092	-367.6101	27.66618

Our next step is to run the same model using the robust standard errors in order to account for the incorrect standard errors.

Linear regression						
	Robust					
rob	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	.0862563	.0434361	1.99	0.047	.0010344	.1714782
pb1064	7.243095	3.584797	2.02	0.044	.2097078	14.27648
pw1064	1.502237	1.742477	0.86	0.389	-1.916509	4.920983
pm1029	3.501497	2.024115	1.73	0.084	-.4698241	7.472819
pop	10.98124	.7943054	13.82	0.000	9.422807	12.53967
avginc	3.414262	1.486466	2.30	0.022	.4978086	6.330715
density	84.22337	6.556728	12.85	0.000	71.35904	97.08769
1.shall	-27.4472	4.909143	-5.59	0.000	-37.07896	-17.81544
_cons	-169.972	114.7728	-1.48	0.139	-395.1567	55.2128

Due to the time effect included in a panel data, a variable x in time t has an influence over xt+1. Therefore, the assumption that the error terms between different time periods are not correlated is violated. Additionally, the variance of the error term may also be different over time.

To account for the problem of serial correlation which causes the coefficient estimates to become inefficient and the standard errors to be misleading, the next step is to run the panel data as a fixed and random effects model.

Model 2:

A random effects model is run next as it has more degrees of freedom than the fixed effect model.  
 Below are the results from the random effects model:

. xtreg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, re						
Random-effects GLS regression		Number of obs = 1,173				
Group variable: stateid		Number of groups = 51				
R-sq:		Obs per group:				
within = 0.0377		min = 23				
between = 0.8239		avg = 23.0				
overall = 0.7575		max = 23				
		Wald chi2(8) = 400.64				
corr(u_i, X) = 0 (assumed)		Prob > chi2 = 0.0000				
rob	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
incarc_rate	-.0137862	.0190538	-0.72	0.469	-.0511311	.0235586
pb1064	15.96248	3.326353	4.80	0.000	9.442944	22.48201
pw1064	5.346849	1.502161	3.56	0.000	2.402668	8.29103
pm1029	-1.905452	1.754347	-1.09	0.277	-5.343908	1.533005
pop	9.033511	1.288485	7.01	0.000	6.508126	11.5589
avginc	-5.937499	1.686247	-3.52	0.000	-9.242482	-2.632515
density	87.40912	6.398136	13.66	0.000	74.869	99.94924
i.shall	-3.250151	5.646313	-0.58	0.565	-14.31672	7.81642
_cons	-218.1524	111.0545	-1.96	0.049	-435.8153	-.489552
sigma_u	51.305528					
sigma_e	48.244039					
rho	.53072435	(fraction of variance due to u_i)				

Model 3:

The fixed effects model is run next. The results are as follows:

. xtreg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe						
Fixed-effects (within) regression						Number of obs = 1,173
Group variable: stateid						Number of groups = 51
R-sq:						Obs per group:
within = 0.0739						min = 23
between = 0.3313						avg = 23.0
overall = 0.2849						max = 23
						F(8, 1114) = 11.12
corr(u_i, Xb) = -0.7981						Prob > F = 0.0000
rob	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	-.1415018	.0280963	-5.04	0.000	-.1966294	-.0863742
pb1064	23.48345	5.329915	4.41	0.000	13.02564	33.94125
pw1064	5.610663	1.523207	3.68	0.000	2.621986	8.59934
pm1029	-7.975287	1.9222	-4.15	0.000	-11.74683	-4.203747
pop	3.223055	2.618636	1.23	0.219	-1.914961	8.36107
avginc	-7.471394	1.773482	-4.21	0.000	-10.95113	-3.991652
density	-62.3304	25.52524	-2.44	0.015	-112.4134	-12.24745
i.shall	7.899847	5.663237	1.39	0.163	-3.211967	19.01166
_cons	-49.3018	115.4966	-0.43	0.670	-275.9171	177.3135
sigma_u	228.38104					
sigma_e	48.244039					
rho	.95728236	(fraction of variance due to u_i)				
F test that all u_i=0: F(50, 1114) = 40.59						Prob > F = 0.0000

### Hausman Test:

Hausman's test is run next to test if the dependent variables of the random effects model are correlated with its error. Below are the results:

	Coefficients		(b-B)	sqrt(diag(V_b-V_B))
	(b) fixed3	(B) random3	Difference	S.E.
incarc_rate	-.1415018	-.0137862	-.1277156	.0206483
pb1064	23.48345	15.96248	7.52097	4.164537
pw1064	5.610663	5.346849	.2638138	.2523314
pm1029	-7.975287	-1.905452	-6.069836	.7855694
pop	3.223055	9.033511	-5.810456	2.279707
avginc	-7.471394	-5.937499	-1.533895	.5493716
density	-62.3304	87.40912	-149.7395	24.71035
1.shall	7.899847	-3.250151	11.15	.4374973

b = consistent under  $H_0$  and  $H_a$ ; obtained from xtreg  
 B = inconsistent under  $H_a$ , efficient under  $H_0$ ; obtained from xtreg

Test:  $H_0$ : difference in coefficients not systematic

```

chi2(8) = (b-B)' [(V_b-V_B)^(-1)] (b-B)
          =       65.95
Prob>chi2 =      0.0000
(V_b-V_B is not positive definite)
    
```

The p-value is < 0.00. So, we reject the null hypothesis that the coefficients of the fixed effect and the random effects are the same. Hence, the fixed effects model is selected.

Model 4:

The fixed effect model is then run, with the Cluster Robust Standard Errors to get more efficient coefficients.

```
. xtreg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe vce
> (cluster stateid)
```

```
Fixed-effects (within) regression                               Number of obs     =      1,173
Group variable: stateid                                    Number of groups  =       51

R-sq:                                                 Obs per group:
    within  = 0.0739                                         min =        23
    between = 0.3313                                         avg =      23.0
    overall = 0.2849                                         max =        23

                                                F(8, 50)          =     94.12
corr(u_i, Xb)  = -0.7981                                     Prob > F        = 0.0000

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

rob	Robust					[95% Conf. Interval]
	Coef.	Std. Err.	t	P> t		
incarc_rate	-.1415018	.0478384	-2.96	0.005	-.237588	-.0454156
pb1064	23.48345	11.62043	2.02	0.049	.1431261	46.82377
pw1064	5.610663	2.346839	2.39	0.021	.8968973	10.32443
pm1029	-7.975287	2.951379	-2.70	0.009	-13.90331	-2.047268
pop	3.223055	4.901349	0.66	0.514	-6.621595	13.0677
avginc	-7.471394	3.036872	-2.46	0.017	-13.57113	-1.371656
density	-62.3304	47.5462	-1.31	0.196	-157.8298	33.16895
i.shall	7.899847	8.25673	0.96	0.343	-8.684282	24.48398
_cons	-49.3018	147.181	-0.33	0.739	-344.9236	246.32
sigma_u	228.38104					
sigma_e	48.244039					
rho	.95728236	(fraction of variance due to u_i)				

The variables incarceration rate, percentage of black, white and male populations and average income are significant at the 95% confidence level.

Model 5:

Dummy variable for each year is then added to check for time effects on the fixed effect model. The results are given below:

```
. xtreg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall i.year,  
> fe vce(cluster stateid)  
  
Fixed-effects (within) regression  
Group variable: stateid  
  
R-sq:  
    within = 0.2017  
    between = 0.3673  
    overall = 0.2358  
  
corr(u_i, Xb) = -0.6995  
  
Number of obs = 1,173  
Number of groups = 51  
  
Obs per group:  
    min = 23  
    avg = 23.0  
    max = 23  
  
F(30,50) = 187.06  
Prob > F = 0.0000  
  
(Std. Err. adjusted for 51 clusters in stateid)
```

rob	Robust					
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
incarc_rate	-.1155623	.0709902	-1.63	0.110	-.2581504	.0270257
pb1064	4.915188	10.18847	0.48	0.632	-15.54897	25.37934
pw1064	-1.7858	7.179589	-0.25	0.805	-16.20643	12.63483
pm1029	11.91268	11.56907	1.03	0.308	-11.32449	35.14985
pop	-.0267939	5.157911	-0.01	0.996	-10.38676	10.33318
avginc	-2.278894	3.263392	-0.70	0.488	-8.83361	4.275823
density	-34.25171	64.47687	-0.53	0.598	-163.7573	95.25391
1.shall	14.47587	8.592869	1.68	0.098	-2.783415	31.73515
year						
78	7.657708	3.037794	2.52	0.015	1.556119	13.7593
79	26.01776	6.300022	4.13	0.000	13.36379	38.67173
80	49.30214	12.6902	3.89	0.000	23.81313	74.79115
81	59.80562	18.64722	3.21	0.002	22.35158	97.25966
82	46.94134	17.04638	2.75	0.008	12.70267	81.18001
83	32.83726	16.70774	1.97	0.055	-.7212213	66.39574
84	26.47968	16.73897	1.58	0.120	-7.141527	60.10089
85	31.14552	18.58551	1.68	0.100	-6.184572	68.47561
86	43.73714	21.28271	2.06	0.045	.9895713	86.48471
87	39.66934	24.01913	1.65	0.105	-8.574506	87.91318
88	51.9991	30.09511	1.73	0.090	-8.448699	112.4469
89	65.97257	36.3295	1.82	0.075	-6.997368	138.9425
90	87.54741	52.17075	1.68	0.100	-17.24063	192.3355
91	103.083	54.09403	1.91	0.062	-5.568071	211.734
92	103.0294	57.92358	1.78	0.081	-13.31355	219.3723
93	105.0671	59.52927	1.76	0.084	-14.50095	224.6352
94	101.728	60.86035	1.67	0.101	-20.51366	223.9696
95	102.0039	66.22315	1.54	0.130	-31.00922	235.017
96	91.80394	67.41183	1.36	0.179	-43.5967	227.2046
97	79.40119	64.08699	1.24	0.221	-49.32132	208.1237
98	66.65684	65.16497	1.02	0.311	-64.23085	197.5445
99	59.42903	68.02213	0.87	0.386	-77.19744	196.0555
_cons	62.50443	359.6663	0.17	0.863	-659.9066	784.9154
sigma_u	201.34205					
sigma_e	45.242494					
rho	.95193474	(fraction of variance due to u_i)				

### F-Test:

An F-test is performed on the year dummy-variables, to assess if the fixed effect model is affected by the time component of the panel data.

```
. testparm i.year

( 1) 78.year = 0
( 2) 79.year = 0
( 3) 80.year = 0
( 4) 81.year = 0
( 5) 82.year = 0
( 6) 83.year = 0
( 7) 84.year = 0
( 8) 85.year = 0
( 9) 86.year = 0
(10) 87.year = 0
(11) 88.year = 0
(12) 89.year = 0
(13) 90.year = 0
(14) 91.year = 0
(15) 92.year = 0
(16) 93.year = 0
(17) 94.year = 0
(18) 95.year = 0
(19) 96.year = 0
(20) 97.year = 0
(21) 98.year = 0
(22) 99.year = 0

      F( 22,      50) =     4.53
      Prob > F =     0.0000
```

The results show that the time effects do indeed affect the fixed effect model.

Model 6:

To understand the effect of non-linear relationship of the independent variables on violent crime rates, various interaction variables were added. While most of the interaction effects were found to be insignificant, the best model was found to include the interaction effects population\*percentage of black population, average income\*density and the density\*density. The output is as follows:

```
. xtreg rob pw1064 pm1029 c.pop##c.pb1064 c.avginc##c.density c.density#c.densi
> ty i.shall incarc_rate c.incarc_rate#c.pb1064 i.year, fe vce(cluster stateid)

Fixed-effects (within) regression                               Number of obs     =      1,173
Group variable: stateid                                     Number of groups  =        51

R-sq:                                                       Obs per group:
    within  = 0.3288                                         min =          23
    between = 0.7158                                         avg =       23.0
    overall = 0.6664                                         max =          23

                                                F(34, 50)           =   130109.72
corr(u_i, Xb)  = -0.9881                                     Prob > F        =      0.0000

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

rob	Robust					
	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
pw1064	.6210708	5.95493	0.10	0.917	-11.33976	12.5819
pm1029	18.31693	11.17369	1.64	0.107	-4.126098	40.75995
pop	27.85932	8.364489	3.33	0.002	11.05875	44.65989
pb1064	20.60148	14.79236	1.39	0.170	-9.109858	50.31282
c.pop# c.pb1064	-3.368779	.832841	-4.04	0.000	-5.04159	-1.695969
avginc density	-1.2223457 1134.294	2.983269 390.71	-0.41 2.90	0.683 0.005	-7.21553 349.5302	4.768615 1919.058
c.avginc# c.density	-7.888837	1.043358	-7.56	0.000	-9.984483	-5.793192
c.density# c.density	-54.54341	19.85128	-2.75	0.008	-94.41589	-14.67093
i.shall incarc_rate	3.055642 -.2364494	6.625512 .0729753	0.46 -3.24	0.647 0.002	-10.25209 -.3830247	16.36338 -.0898742
c. incarc_rate# c.pb1064	.0245751	.0048652	5.05	0.000	.0148031	.0343472

year						
78	6.78982	2.685558	2.53	0.015	1.395719	12.18392
79	22.82477	4.87604	4.68	0.000	13.03096	32.61859
80	43.66056	9.468083	4.61	0.000	24.64335	62.67776
81	54.42022	14.3207	3.80	0.000	25.65624	83.18419
82	43.35435	12.89936	3.36	0.001	17.44523	69.26347
83	30.60543	12.97822	2.36	0.022	4.537915	56.67295
84	26.96907	14.54208	1.85	0.070	-2.239549	56.1777
85	32.99647	16.96872	1.94	0.057	-1.086206	67.07914
86	47.90789	20.01846	2.39	0.020	7.699634	88.11614
87	45.95395	22.59815	2.03	0.047	.564242	91.34367
88	59.90605	27.24752	2.20	0.033	5.177796	114.6343
89	74.14719	32.23669	2.30	0.026	9.397897	138.8965
90	92.1776	44.8253	2.06	0.045	2.143344	182.2119
91	108.2644	47.07038	2.30	0.026	13.72079	202.8081
92	108.9246	50.3186	2.16	0.035	7.856742	209.9925
93	110.8022	51.98897	2.13	0.038	6.379286	215.2251
94	104.5985	51.56055	2.03	0.048	1.036096	208.1609
95	102.8208	53.74936	1.91	0.061	-5.138012	210.7795
96	94.08673	55.82316	1.69	0.098	-18.03739	206.2108
97	82.50341	54.21519	1.52	0.134	-26.39101	191.3978
98	68.25523	55.03791	1.24	0.221	-42.29166	178.8021
99	58.39949	56.45517	1.03	0.306	-54.99405	171.793
_cons	-602.5247	302.2424	-1.99	0.052	-1209.596	4.547086
sigma_u	590.70661					
sigma_e	41.560938					
rho	.99507414	(fraction of variance due to u_i)				

The adjusted R-squared value was found to be .3087.

We find that the independent variables density, population and incarceration rate are significant at the 95% confidence level along with the interaction effects population\*percentage of black population, average income\*density and density\*density. The time dummy variables for the years 1978 to 1996 are significant at the 90% confidence level.

The interpretations of the coefficients are below:

- Increase in density is associated with a substantial increase in robbery rate.
- Increase in population is also associated with an increase in robbery rate.
- When more people are incarcerated, there is a decline in the robbery rate.
- In a state with more population, if the percentage of black population increases, the rate at which robbery rate increases is diminishing. For a state with lower population, when the percentage of black people increases, increase in robberies is more compared to a state with a high population.
- When the average income of a densely populated area is high, robbery rate declines.

- As the population density of a state increases, the marginal rate at which it affects the increase in robbery rate is diminishing.

## Conclusion:

The pooled OLS model led to inflated estimators due to endogeneity, which was caused by the presence of individual state and time effects. These limitations were controlled for by using the state fixed and time fixed model. The state fixed aspect took into account each of their unique heterogeneity, whereas the time fixed aspect took into account effects that vary across time but not across states, like federal policies that affected all states.

There was also the limitation of limited data. With more data, even more variance could have been accounted for leading to more confident interpretations of the coefficients of the independent variables.

Since many of the features could not strongly explain the different types of crime rates, there exists a possibility of missing variables. Adding additional variables like unemployment rates, or a proxy for cultural attitude towards usage of guns may help engender a better model.

Based on the findings from the models created for the different crime rates, we find that the shall-issue law and the incarceration rate have no significant effect on murder rates and violent crime rates. We do find that shall-issue law significantly affects robbery rate at a 95% confidence level, by reducing it by 29.3 units when the law is adopted.

## Appendix

### Stata Code

\*\*Basic Analysis

```
summarize vio mur rob incarc_rate pb1064 pw1064 pm1029 pop avginc density shall  
xtset stateid year
```

```
xtsum vio mur rob incarc_rate pb1064 pw1064 pm1029 pop avginc density shall
```

```
corr vio mur rob incarc_rate pb1064 pw1064 pm1029 pop avginc density shall
```

\*\*Dummy variables for time

```
summarize i.year
```

```
testparm i.year
```

\*Log variables

```
histogram vio
```

```
gen log_vio = log(vio)
```

```
histogram log_vio
```

```
histogram mur
```

```
gen log_mur = log(mur)
```

```
histogram log_mur
```

```
histogram rob
```

```
gen log_rob = log(rob)
```

```
histogram log_rob
```

```
histogram pop
```

```
gen log_pop = log(pop)
```

```
histogram log_pop
```

```
histogram density
```

```
gen log_density = log(density)
```

```
histogram log_density
```

```

histogram incarc_rate
gen log_incrate = log(incarc_rate)
histogram log_incrate

**Violent Crime Rate

*Pooled OLS
reg vio incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall
reg vio incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, robust

*Random Effects Model
xtreg vio incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, re
estimates store random

*Fixed-Effects Model
xtreg vio incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe
estimates store fixed

*Hausman Test
hausman fixed random

*Robust Standard Errors
xtreg vio incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe vce(cluster stateid)
xtreg log_vio log_incrate pb1064 pw1064 pm1029 log_pop avginc density i.shall, fe vce(cluster stateid)

*Dummy variables for time
xtreg log_vio log_incrate pb1064 pw1064 pm1029 log_pop avginc density i.shall i.year, fe vce(cluster stateid)

*Interaction Variables
xtreg log_vio pw1064 avginc pm1029 log_pop c.log_incrate density i.shall##c.pb1064 i.year, fe
vce(cluster stateid)

ereturn list

**Murder Rate

*Pooled OLS

```

```

reg mur incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall
reg mur incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, robust
*Random Effects Model
xtreg mur incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, re
estimates store random2
*Fixed-Effects Model
xtreg mur incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe
estimates store fixed2
*Hausman Test
hausman fixed2 random2
*Robust Standard Errors
xtreg mur incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe vce(cluster stateid)
xtreg mur incarc_rate pb1064 pw1064 pm1029 log_pop avginc log_density i.shall, fe vce(cluster stateid)
*Dummy variables for time
xtreg mur incarc_rate pb1064 pw1064 pm1029 log_pop avginc log_density i.shall i.year, fe vce(cluster stateid)
*Interaction Variables
xtreg mur pw1064 pm1029 c.log_pop##c.incarc_rate avginc c.log_density##c.pb1064 i.shall i.year, fe
vce(cluster stateid)
ereturn list

**Robbery Rate
*Pooled OLS
reg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall
reg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, robust
*Random Effects Model
xtreg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, re
estimates store random3
*Fixed-Effects Model

```

```
xtreg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe  
estimates store fixed3  
  
*Hausman Test  
  
hausman fixed3 random3  
  
*Robust Standard Errors  
  
xtreg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall, fe vce(cluster stateid)  
  
*Dummy variables for time  
  
xtreg rob incarc_rate pb1064 pw1064 pm1029 pop avginc density i.shall i.year, fe vce(cluster stateid)  
  
*Interaction Variables  
  
xtreg rob pw1064 pm1029 c.pop##c.pb1064 c.avginc##c.density c.density#c.density i.shall incarc_rate  
c.incarc_rate#c.pb1064 i.year, fe vce(cluster stateid)  
  
ereturn list
```