Analysis of Covid-Era Policing Policies on Philadelphia Arrests Ayina Anyachebelu, Anthony Steimle, Benjamin Yao,

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Introduction

Analysis Overview

- Looking at how Philadelphia arrest rates change due to pandemic-era policing policies.
 - What happens when pandemic-era policies are implemented?
 - What happens when pandemic-era policies are removed?

Relevance

- Analysis focuses on city of Philadelphia
- 2020 Stay-At-Home order unprecedented policy with unique implications
- Analyzing crime & arrest trends increasingly pertinent as police forces fall under increasing scrutiny



Event Research

Pandemic Court Clog

Philadelphia courts procedures were largely put on hold, leading to overcrowded jails and pressure from civil rights organizations like the ACLU for new policing policies.

Philly police to halt narcotics arrests, other charges during COVID outbreak

With courts closed by pandemic, Philly police stop low-level arrests to manage jail crowding

"If an officer believes that

MAY 02, 2020

$^{safety,\,the\,officer\,will\,notif}$ Philadelphia police to resume making circumstances and utilize arrests for nonviolent crimes following determining the appropria pause due to COVID-19

A stoppage on arrests for certain crimes due to coronavirus has come to an end, the Police Commissioner announced Friday

Memo Release

Philadelphia Police Commissioner, Danielle M. Outlaw released an internal memo (which was leaked to the public). These policies were repealed weeks later.

pandemic, Beginning today (March 17, 2020), the City is suspending all he public, as the City tries to limit the potential for exposure to Covid addition, beginning Wednesday, March 18, 2020, only essential employee

In tandem with the City's enhanced response, the Police Department will also be making modifications to several of our existing protocols. These

Effective Today (March 17, 2020), during the 4pm x 12am tour.

s Theft from auto § Burglary § Vandalism § All Bench warrants

§ Stolen auto & Economic crimes (had checks, fraud

If an officer believes that releasing the offender would pose a threat to public safety, the officer will notify a supervisor, who will review the totality of the circumstances and utilize discretion, in the interest of public

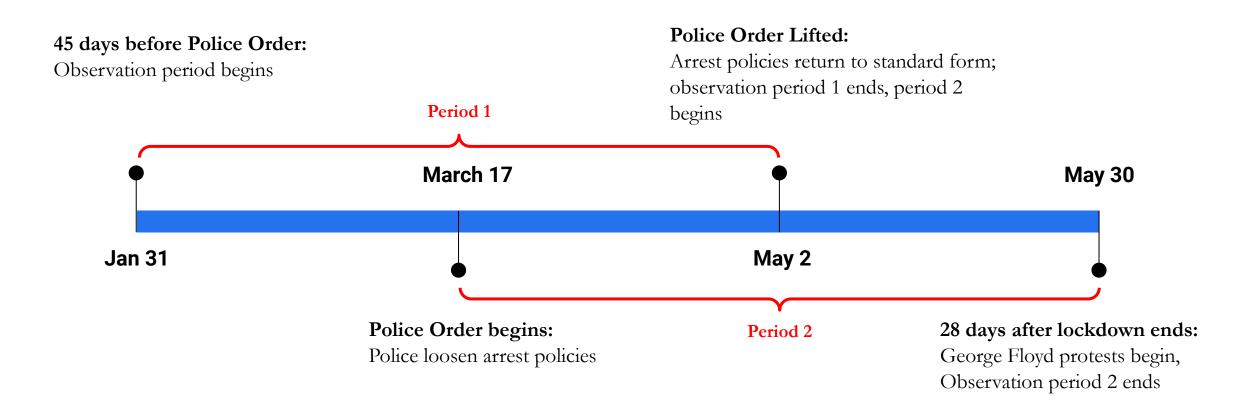
- § All Narcotics Offenses
- § Theft from persons
- § Retail Theft
- s Theft from auto
- § Burglary
- § Vandalism
- § All Bench warrants
- § Stolen auto
- § Economic crimes (bad checks, fraud)
- § Prostitution







Context Timeline



Period 1: "Lockdown Period"

Period 2: "Post-Lockdown Period"



Hypothesis

We hypothesize that the internal police memo caused a significant decrease in petty arrests after the effective date of the internal police memo, and the policy lift caused higher than typical arrests after the conclusion of policy due to compensate for arrest delays.



Data & Methods



Methodology: Data Sourcing

Data Search

- Looking for dataset that contained overall arrest data, petty crime example data (e.g. narcotics/drug arrests), and violent crime data (e.g. firearm arrests)
- First searched through cleaner data databases like Kaggle
 - Turned to Philadelphia District Attorney's Office
 - Provided an overwhelming amount of data, from incidents, arrests, case length, charges across different categories of time and divided by district, zipcode, etc.
 - Decided to use citywide arrest data across various petty/violent crimes
 - Day-by-day reports allowed observation of local effects

Data Cleaning Process

- Narrowed scope of data to specific section of each year
- Chose drug and firearm arrests as a popular possession-related petty crime with enough daily volume to observe significant difference

		Daily by Police	Daily by ZIP Code	Weekly by Census	Quarterly by Census
Dataset	Daily Citywide	District	(ZCTA)	Tract	Tract
Incident Data	OpenDataPhilly				
Arrests Data	CSV	CSV	CSV	CSV	CSV
Charges Data	CSV	CSV	CSV	CSV	CSV
Case Outcome Data	CSV	CSV	CSV	CSV	CSV
Bail Data	CSV	CSV	CSV	CSV	CSV
Future Years of	CSV	CSV	CSV	CSV	CSV



Philadelphia District Attorney's Office

Summary Case Outcomes Data	CSV	CSV	CSV	CSV	CSV
Summary Case Length	CSV	CSV	CSV	CSV	CSV



Methodology: Technical Analysis

Research Method: Simple Difference-in-Differences

Treatment: Philadelphia in Year 2020

Control: Philadelphia in Years 2014 - 2019

Regression Design: Amount of arrests = Intercept + B1(Post) + B2(Treatment) + B3(Treatment x Post)

• Regressions were done twice, once where the Post-Period represents after the police order and another where the post-period represents after the order was lifted

Parallel Trends

- We saw parallel trends hold before our 1st event of interest (March 17th Police Order)
- However, parallel trends did not hold before our 2nd event of interest (May 2nd Order Lift)

Checking for Confounding variables and other Arrests

- We limited our windows to exclude other significant events e.g George Floyd Protests
- Checking if results were due to the police order (petty-crime arrest) versus a general police shift
 - We examined non-petty offenses e.g Firearm Possession Arrests

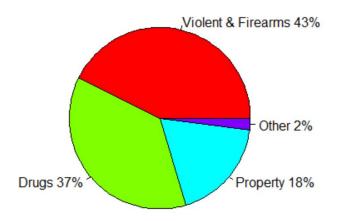




Summary Statistics

2020 Total Number of Arrests: 22,240

Categorized Offenses in 2020



	All Arrests	Drug Possession	Firearm Possession
Minimum	8.0	0	0
1st Quantile	46.0	1.0	3.0
Median	66.5	3.0	5.0
Mean	72.1	5.7	5.7
3rd Quantile	95.5	9.5	8.0
Maximum	241.0	39.0	22.0



Results



Lockdown: Total Arrests

Equation:
$$\hat{Y} = a + \beta 1 X_1 + \beta 2 X_2 + \beta 3 X_1 X_2 + \epsilon$$

Equation:
$$\hat{Y} = 109.9 - 6.08X_1 + 2.94X_2 - 77.30X_1X_2 + \varepsilon$$

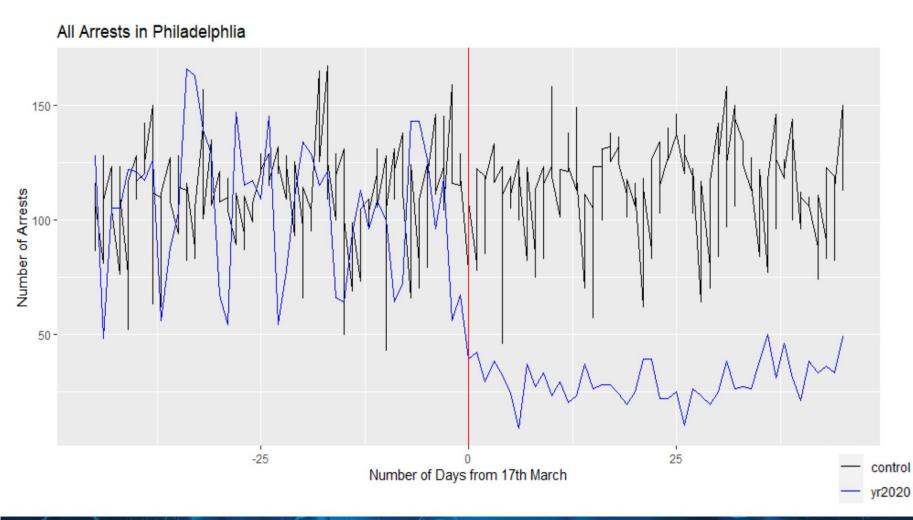
- interaction term shows statistical significance
- sample size = 180 (90*2)

```
Call:
lm(formula = number_of_arrests ~ Treatment + Post + TreatmentxPost
    data = all_the_arrests)
Residuals:
   Min
            1Q Median
-66.907 -9.574 2.402 13.108 62.174
Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)
               109.907
                            1.909 57.586
                            3.817 -1.593
Treatment
                -6.081
                                            0.112
                 2 941
                            2.714 1.084
                                            0.279
TreatmentxPost -77.301
                            5.428 -14.241
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 22.42 on 360 degrees of freedom
Multiple R-squared: 0.5899, Adjusted R-squared: 0.5865
F-statistic: 172.6 on 3 and 360 DF, p-value: < 2.2e-16
```

- X_1 = Dummy variable for Treatment Group (1 if year 2020, 0 otherwise)
- X_2 = Dummy variable for being in the lockdown (1 for after March 17, 0 otherwise)
- X_1X_2 = interaction term



Lockdown: Total Arrests

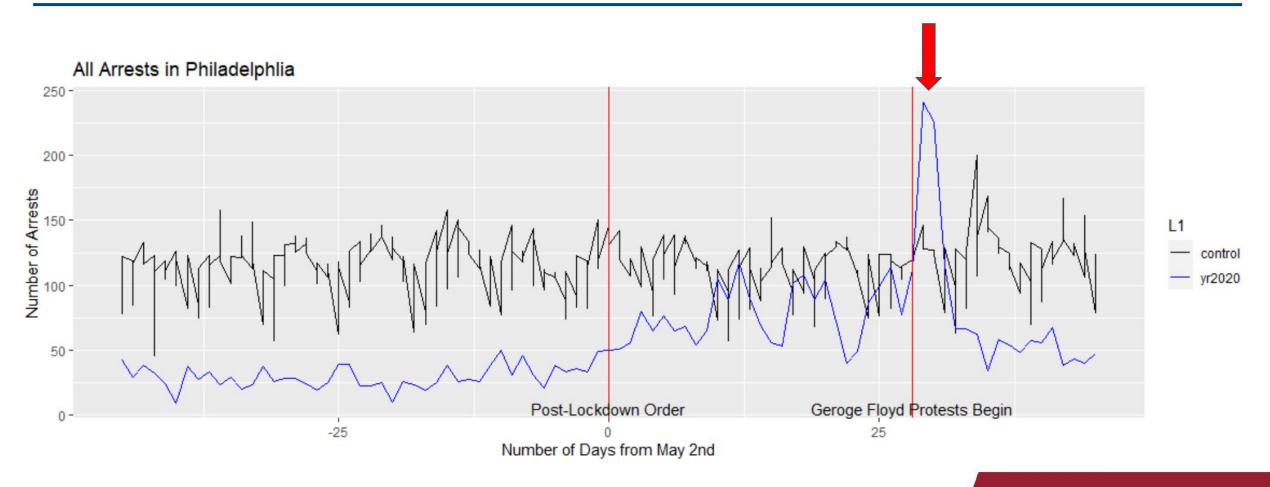


• significant drop when the lockdown happened

Parallel trendObserved



Post Lockdown: Total Arrests





Post Lockdown: Total Arrests

Equation:
$$\hat{Y} = 113.3 - 83.39X_1 - 1.39X_2 + 50.41X_1X_2 + \varepsilon$$

X₁: Dummy variable for Treatment Group (1 if year 2020, 0 otherwise)

X₂: Dummy variable for being in postlockdown period (1 for after May 2, 0 otherwise)

 X_1X_2 : Interaction term

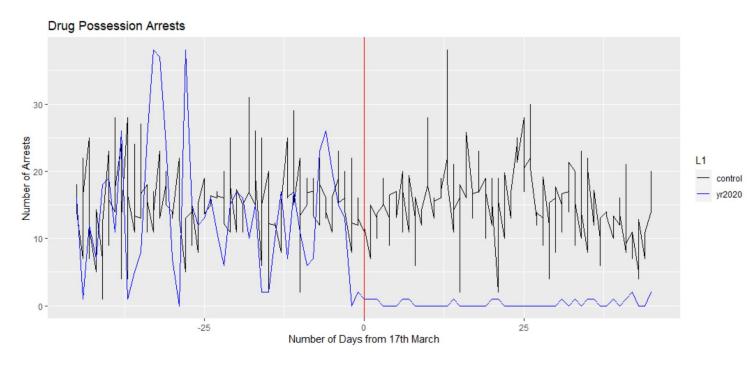
P- Value of interaction: <2e-16

```
lm(formula = number_of_arrests ~ Treatment + Post_Post + TreatmentxPost_Post,
    data = all_post_arrests_w)
Residuals:
             1Q Median
-67.303 -9.011 2.059 12.088 44.697
Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
                     113.303
                                 1.671 67.797
(Intercept)
                     -83.390
                                                 <2e-16 ***
Treatment
                     -1.394
                                 2.717 -0.513
                                                  0.608
Post_Post
TreatmentxPost_Post 50.410
                                        9.277 <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```





Lockdown: Drug Arrests



Equation:
$$\hat{Y} = 15.30 - 1.86X_1 - 0.30X_2 - 12.75X_1X_2 + \varepsilon$$

```
Call:
lm(formula = number_of_arrests ~ Treatment + Post + TreatmentxPost
    data = all_drugs_poss)
Residuals:
     Min
               10 Median
-14.2995 -2.9963 -0.3778
                           2.1779 24.5652
Coefficients:
               Estimate Std. Error t value Pr(>|t|)
(Intercept)
                15.2995
                           0.5010 30.540
                                            <2e-16 ***
                -1.8647
                                            0.0635 .
Treatment
                -0.3032
                           0.7124 -0.426
                                            0.6706
Post
TreatmentxPost -12.7538
                           1.4248 -8.951
                                            <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 5.885 on 360 degrees of freedom
Multiple R-squared: 0.4037, Adjusted R-squared: 0.3987
F-statistic: 81.25 on 3 and 360 DF, p-value: < 2.2e-16
```

X₁: If Treated = 1, 0 otherwise
X₂: If After Policing Order = 1, 0 otherwise

X₁**X**₂: Interaction term between

Treatment and Post



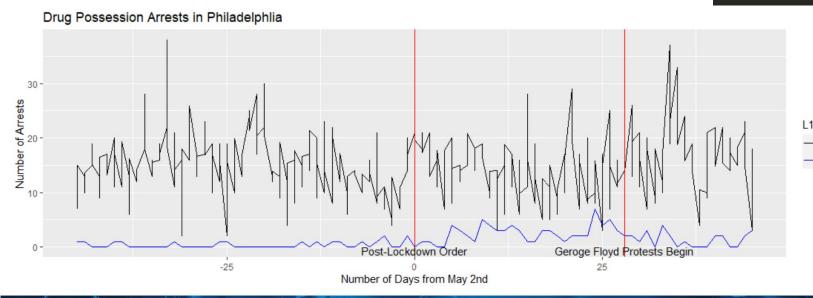
Post Lockdown: Drug Arrests

```
Equation: \hat{Y} = 15.12 - 14.75X_1 - 1.17X_2 + 3.3719X_1X_2 + \varepsilon
```

```
Call:
lm(formula = number_of_arrests ~ Treatment + Post_Post + TreatmentxPost_Post,
    data = all_post_drugs_w)
Residuals:
             10 Median
-13.124 -2.124 -0.247
                         2.043 22.876
Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
                    15.1244
                                 0.4055 37.296
(Intercept)
                    -14.7548
                                 0.8110 -18.192
                                                  <2e-16 ***
Treatment
Post_Post
                     -1 1700
                                 0.6593 -1.775
                                                 0.0770 .
                     3.3719
                                 1.3185
                                         2.557
                                                 0.0111 *
TreatmentxPost_Post
Signif. codes: 0 '*** 0.001 '** 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

ror: 4.764 on 292 degrees of freedom

n 3 and 292 DF, p-value: < 2.2e-16



X₁: If Treated = 1, 0 otherwise
X₂: If After Policing Lift Order
= 1, 0 otherwise

Adjusted R-squared: 0.6031

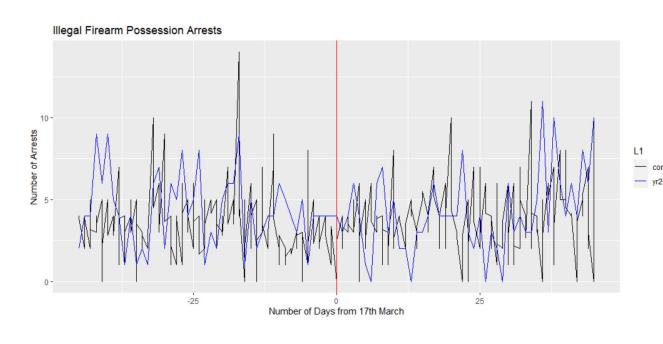
X₁X₂: Interaction term between Treatment and Post



Balance Check: Firearm Arrests



Lockdown: Firearm Arrests



Regression:

$$\hat{Y} = 3.457 + 0.848X_1 + 0.439X_2 - 0.543X_1X_2 + \varepsilon$$

 \mathbf{X}_1 : If Treated = 1, 0 otherwise

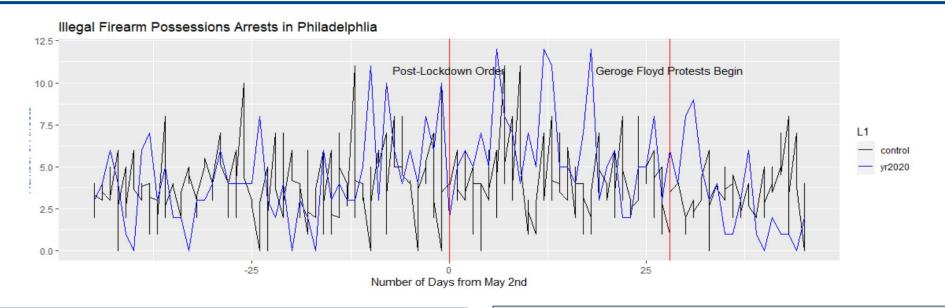
 \mathbf{X}_2 : If After instituting order = 1, 0 otherwise

X₁X₂: Interaction term between Treatment and **post**

P-Value of Treatment X Post: 0.292



Post Lockdown: Firearm Arrests



Regression:

$$\hat{Y} = 3.912 + 0.242X_1 + 0.105X_2 + 1.885X_1X_2 + \varepsilon$$

 \mathbf{X}_1 : If Treated = 1, 0 otherwise

 \mathbf{X}_2 : If After Lifting order = 1, 0 otherwise

X₁X₂: Interaction term between Treatment and Post

P-Value of Treatment X Post: 0.00233



Results Summary

	Lockdown Period (relative to control)	Post Lockdown (relative to control)	Post Lockdown (relative to pre period)
Total Arrests			
Drug Arrests			
Firearm Arrests			



Conclusion

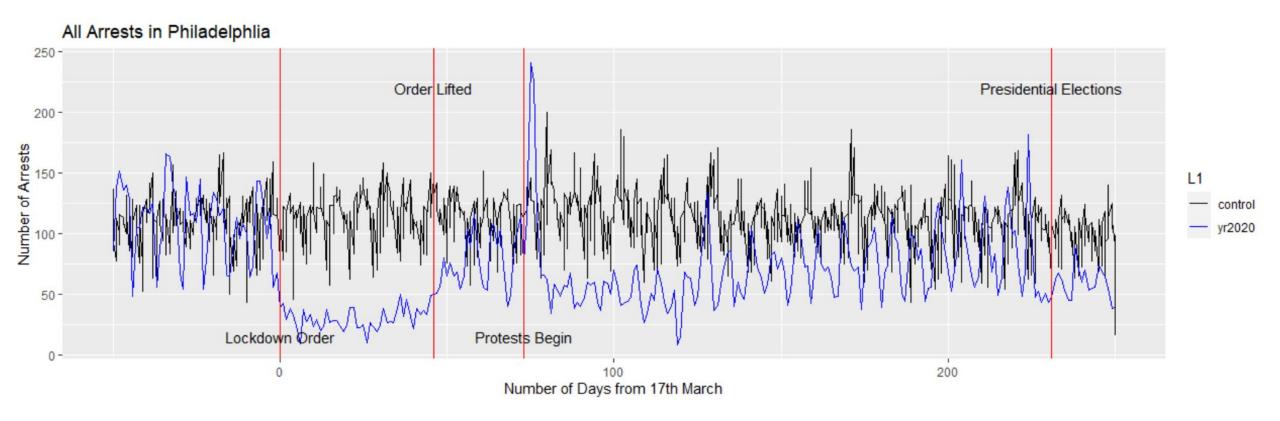
Hypothesis

We hypothesize that the internal police memo caused a significant decrease in petty arrests after the effective date of the internal police memo, and the policy lift caused higher than typical arrests after the conclusion of policy due to over-compensation.

Our analysis suggests that the memo was responsible for a significant decrease in petty arrests, but the lift of the policy did not necessarily lead to full compensation in the post-period.



Long Term Arrests





Limitations of Analysis

<u>Limitations of Future Projections</u>

- As a result of the George Floyd protests, increased scrutiny of police activities may have affected policing activities, changing arrest patterns after May 2
- Changes in implicit behavior caused by Covid-19
 (which continued to exist beyond the lockdown) may
 have had an impact on arrest rates beyond the change
 in explicit policing orders
- While we assume actual levels of petty crime remained the same (due to our balance checks), there is the possibility true petty crime did actually decrease

Limitations of Experimental Design

- Since Covid-19 affected the entire globe at once, it is difficult to find a location with a comparable parallel trend which did not experience covid effects
 - Control used to project 2020 may be imperfect



Implications & Further Research

- This analysis opens the door to broader causal questions:
 - Ones media attention increase the efficacy of policing policies?
 - O Did policies like the one investigated lead to tangible (public health) benefits/harm to society?
- Research the impact of pandemic-era arrest reduction policies on jail covid-positivity/mortality rates across different counties/states.
- Cross-check sources about the pandemic effects on true crime, and if the police were more effective at making arrests (arrest rates were higher/lower) in locations where arrest-reduction policies were in place.
- Find reliable reports of true changes in petty criminal behavior as a result of the pandemic.
- Check results against aggregated petty/violent crimes, or other specific crimes.
- Check for differences in responsiveness to internal policy changes in Philadelphia police department and other departments.
- Check for differences in responsiveness when changes are public/internal.



Thank You

