Policing Lab Research

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Big Picture Goal:

Merge Census data with the pre-existing officer-involved shooting dataset (OIS_master)

More Specifically...

Find the smallest geographical unit (Census block) where each shooting incident occurred and gather relevant demographic features of the units from Census Bureau data

OIS_master dataset

4086 officer-involved shootings in 37 cities in the US

- Incident ID
- City
- Date
- Time
- *Location (Address)
- *Coordinates (lat/long)

- Offender ID
- *Offender Race
- Offender Fatality (Fatal / Struck / Not-Struck)
- *Officer Race
- *Officer Fatality (Fatal / Struck / Not-Struck)

Basic Process

1

Block-level GEOIDs

Use the Census and FCC APIs to get GEOIDs for every shooting location based off of address and coordinates respectively

2

Census Demographic Information

Use the get_decennial() function in the *tidycensus* package to get block-level demographic info for every block (identified by a 15-digit GEOID) in every county/state combination that had a shooting incident

3

Join the two datasets

Join the OIS_master data with the block-level Census data so that each shooting incident has relevant demographic features

Methods

Stage 1

Gathering Census Information

- Used the get_decennial() function from the package tidycensus (API call to the decennial Census) on each county-state combination individually for a total of the 37 cities and obtained block-level data from the Decennial Census: Summary File 1 (2010), sf1.
- Used variables beginning with "P" which give us population info down to the block level
- Each observation in the result is a block that has a 15-digit GEOID as the identifier and variables we want

| - | GEOID | NAME | birac [‡] | blackhisp | blackmale | female = | hispmale [‡] | male |
|-----|-----------------|--|--------------------|-----------|-----------|----------|-----------------------|------|
| / | 060750101001000 | Block 1000, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 0 | 0 | 0 |
| 2 | 060750101001001 | ock 1001, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 13 | 1 | 31 |
| 3 | 060750101001002 | Block 1002, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 0 | 0 | 0 |
| 4 | 060750101001003 | Block 1003, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 0 | 0 | 0 |
| 5 | 060750101001004 | Block 1004, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 1 | 0 | 0 |
| 6 | 060750101001005 | Block 1005, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 0 | 0 | 0 |
| 7 | 060750101001006 | Block 1006, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 1 | 0 | 1 |
| 8 | 060750101001007 | Blo k 1007, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 0 | 0 | 0 |
| 9 | 060750101001008 | Block 1008, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 0 | 0 | 0 |
| 10 | 060750101001009 | Bock 1009, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 0 | 0 | 1 |
| Įı. | 060750101001010 | lock 1010, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 1 | 0 | 0 |
| 12 | 060750101001011 | Block 1011, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 1 | 0 | 0 |
| 13 | 60750101001012 | Block 1012, Block Group 1, Census Tract 101, San Fra | 0 | 0 | 0 | 0 | 0 | 0 |

Ex: San Francisco

Stage 1 (continued)

Gathering Census Information

- Repeating this process individually for each of the 37 cities was tedious
- Created an efficient and replicable version of the code where we looped over vectors containing the counties and respective states and assigned the result to the city names
- Saved all the cities to a list to later join with the OIS_master

```
# retrieve data from sf1
for(i in 1:length(county)){
 assign(cityname[i], get_decennial(geography = "block", county = county[i], state = state[i],
                                    variables = c(totpop = "P001001".
                                                   male = "P012002",
                                                   female = "P012026".
                                                   blackmale = "P012B003"
                                                   hispmale = "P012H002"
                                                   totblack = "P003003",
                                                   tothisp = "P004003",
                                                   totwhite = "P008003"
                                                   totasian = "P003005"
                                                   totnative = "P008005"
                                                   birac = "P008011",
                                                   blackhisp = "P005012".
                                                   toturban = "P002002".
                                                   singleparent_f = "P018006",
                                                   singleparent_m = "P018005"),
                                    year = 2010, geometry = TRUE, summary_var = "P001001")
         %>% spread(key = "variable", value = "value"))
```

Ex of more efficient get_decennial() call

Stage 2

Finding the GEOID

First tried the US Census
 Geocoder which takes an
 address batch as an input
 and returns the GEOID of
 each address. However,
 many addresses were
 formatted incorrectly so we
 had around 1600 NA values
 for the GEOID column (40%)

Correctly Formatted Addresses

| 3653 | 13581 | 180014 | 12/29/10 | 15:40 | 626 Bacon St |
|------|-------|--------|----------|-------|---------------------|
| 3654 | 13583 | 180015 | 1/4/11 | 10:33 | 1380 Howard St |
| 3655 | 13585 | 180016 | 6/7/11 | 17:23 | 65 Buena Vista East |
| 3656 | 13586 | 180016 | 6/7/11 | 17:23 | 65 Buena Vista East |
| 3657 | 13587 | 180017 | 6/29/11 | 7:55 | 1155 Ellis St |
| 3658 | 13588 | 180017 | 6/29/11 | 7:55 | 1155 Ellis St |

Incorrectly Formatted Addresses

| 13653 180047 1/4/15 17:21 Mission Station 13654 180048 2/26/15 21:47 2800 block of Folsom St 13655 180048 2/26/15 21:47 2800 block of Folsom St | |
|---|------|
| 2505 10 2720,23 22117 2503 2503 2503 | |
| 13655 180048 2/26/15 21:47 2800 block of Folsom St | |
| | |
| 13656 180049 3/17/15 19:07 1500 block of Van Ness Ave at Pine St | |
| 13657 180049 3/17/15 19:07 1500 block of Van Ness Ave at Pine St | |
| 13658 180050 7/26/15 7:00 Collector Ramp, San Bruno Ave/US-10 | 1 NB |
| 13659 180051 9/18/15 8:00 San Mateo, California | |

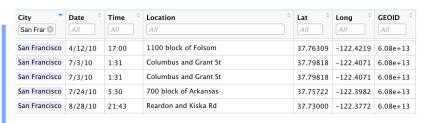
Ex: San Francisco

Stage 2 (continued)

Finding the GEOID

- Then used the Federal Communications Commission (FCC) geocoding API which takes coordinates as an input and
- This cut the number of NAs in the GEOID column down to 559

Incorrect address with coordinates



Incorrect address with no coordinates (559)

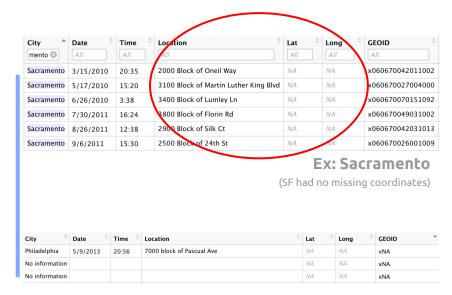
| City | Date [‡] | Time [‡] | Location | Lat [‡] | Long [‡] | GEOID [‡] |
|---------------|-------------------|-------------------|--|------------------|-------------------|--------------------|
| San Frar 💿 | All | All | All | All | All | All |
| San Francisco | 7/26/15 | 7:00 | Collector Ramp, San Bruno Ave/US-101 NB | NA | NA | NA |
| San Francisco | 9/18/15 | 8:00 | San Mateo, California | NA | NA | NA |
| San Francisco | 10/24/15 | 8:00 | Main Gate, Ave of the Palms, Treasure Island | NA | NA | NA |
| San Francisco | 10/24/15 | 8:00 | Main Gate, Ave of the Palms, Treasure Island | NA | NA | NA |

Ex: San Francisco

Stage 2 (continued)

Finding the GEOID

- Finally used the ggmaps package with the Google Maps API to find coordinates of remaining data points that had incorrectly formatted addresses and no coordinates
- Then used FCC API to find GEOID; this resulted in only 3 missing GEOIDs
 - one in Philadelphia and two cities that were unspecified



Final Result - only 3 NAs

Dataset Functionality

Variables (flexible)

- total population
- total male
- total female
- total black male
- total hispanic male
- total black population
- total hispanic population
- total white population
- total asian population
- total native population
- total bi-racial (black/white) population
- total black hispanic
- total urban population
- single parent household (mother present)
- single parent household (father present)

Logistical Information

- Incident ID
- City
- Date
- Time
- *I ocation
- *Coordinates (lat/long)
- GEOID (15 digit block-level)
- Offender ID
- *Offender Race
- Offender Fatality
- *Officer Race
- *Officer Fatality

Findings

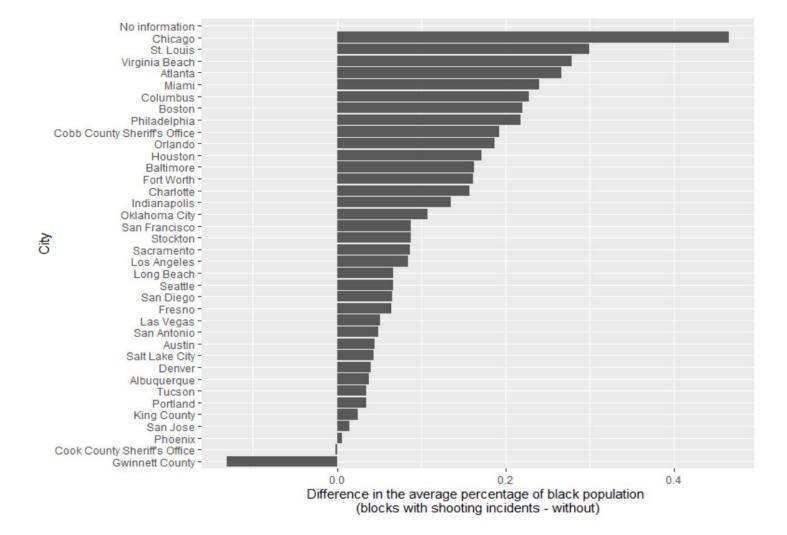
Top 5 Cities with the Largest Differences in Black Population Proportion

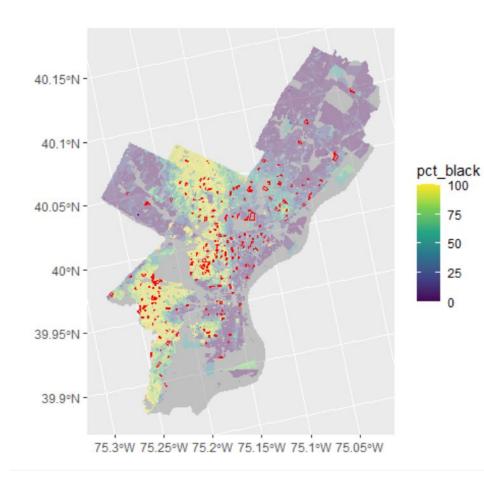
| | Black Population Proportion | | | | | |
|---------------------------|---------------------------------------|-------------------------|------------|--|--|--|
| City | Blocks with shooting | Blocks without shooting | Difference | | | |
| Chicago | 0.7120 | 0.2463 | 0.4657 | | | |
| St. Louis | 0.7817 | 0.4816 | 0.3001 | | | |
| Virginia Beach | 0.4192 | 0.1400 | 0.2792 | | | |
| Atlanta | 0.7420 | 0.4759 | 0.2661 | | | |
| Miami | 0.4502 | 0.2103 | 0.2398 | | | |
| Note: Difference is calcu | lated by blocks with shooting inciden | ts - blocks without | | | | |

Black population percentage in blocks with/without shooting incidents

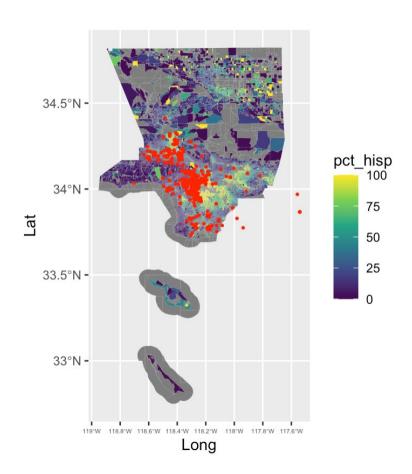
In top 5 cities with the largest differences







Map of Philadelphia officer-involved shootings colored according to percent black population per block



Map of Los Angeles officer-involved shootings colored according to percent Hispanic population per block

