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## **Report on Interactive Visualization Findings**

Highlighting the high and low socio-economy status communities with different colors and plot TRRs on them. Set up a time slider to see how it changes over time.

Our theme is about the relationship between geo locations, eco-socio status, and over-policing. In our previous analysis, we have seen the positive correlation between low economic communities and high complaint reports (which indicates more misconduct behaviors). In this report, we would like to further look at key metrics(number of CRs and TRRs) in different communities with different economic social statuses. More importantly, we want to monitor the changes over time, and locate the trend in recent years.

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### Median Income 20000 30000 40000 50000 60000 70000 80000 90000 T Number of CR counts 7,000 6,500 6,000 Https://observablehq.com/@liux2/findings



#### Plot 1

So, we created this plot with multiple dimensions in years, beat areas, median incomes, the number of tactical response reports, and the number of complaint reports.

As you can see, low-income areas(yellow bubbles in the plot) are generally having more CR and TRR reports. High-income areas, despite 2 outliers typically have fewer. From the time perspective, we have seen a reduced increasing speed for both records over the years for most of the beat area. Especially for the "rich" area, most of them stayed around the origin point.

From this chart, we can find there are multiple shallow colored points( low income community) on the top right corner, it means for these areas, police are receiving significant high amount of complains and they would prefer to use technique weapons to agains the citizens. Therefore, we can conclude, there is over-policing in low socio-eco status neighborhoods.

Also from this chart, we can find there are multiple shallow colored

points(low-income community) on the top right corner, which means for these areas, police are receiving a significantly high amount of complaints and they would prefer to use technique weapons to against the citizens. Therefore, we can assume there is over-policing in low socio-eco status neighborhoods. attendance, in different neighborhoods. Set up a time sider to see how it changes over time.

In this section, we discuss the potential relations between officer attendant rate in each beat area each year and the tactical response. We assume that the attendance rate could affect the tactical response rate of the area. To observe the attendance rate each year, we created the animation of officer attendance rate each year, as the following plot shows.







Plot 2

In the graph, we observed that areas such as 123, 271, 101, and 105 have consistently low attendant rates. By comparing with the records in checkpoint 2, we can know

#### Visualizations



#### Plot 3

This observation also complemented the findings in the previous question, highincome communities tend to receive less tactical responses from police officers, and they tend to file more complaints to the officers.

#### Bar Chart Race of Trr over different Beats

This chart animates the number of Trr in different beats from 2004 to 2016. `

Replay

# 2004

#### Plot 2

With this bar chart race of trr over different beats, we can clearly find such two leading beats, 132 and 65. Associated with Plot 2 in Observable Notebook and the median income map in our Checkpoints 2, we can find that beat 65 and 132, especial beat 132 belongs to the low socio-eco status neighborhoods and it contains a high police attendance rate. The high police attendance rate is important because it implies that police officers pay attention and focus on the security in such areas. With the leading trrs in such low socio-eco status neighborhoods like the beat 132, we can assume that there is over-policing in some low-income areas.

#### Conclusion

With the above four plots, we can roughly have an idea of the patterns of police misconduct. Police officers tend to give less tactical responses to high-income areas. To avoid biases, police misconduct can be viewed from both public narratives and

nolice narratives. which means an area with both high complaint rates and high https://observablehq.com/@liux2/findings

tactical response can be viewed as the police misconduct area. From our observation, police misconduct areas tend to have less income.

## Appendix

The setups for the plots is as follows.

#### Plot 1

```
update = undefined
```

currentData = ▶ Array(41) [Object, Object, Object, Object, Object, Object, Object

```
x = f(n)
```

```
y = f(n)
```

radius = f(n)

color = f(t)

xAxis = f(g)

yAxis = f(g)

grid = f(g)

```
dataAt = f(date)
```

```
valueAt = f(values, date)
```

data =  $\blacktriangleright$  Array(41) [Object, Object, Object,

```
dates = \blacktriangleright Array(144) [2005-01-01, 2005-02-01, 2005-03-01, 2005-04-01, 2005-05-01,
```

```
parseSeries = f(series)
```

bisectDate = f(t, n, e, i)

```
margin = \triangleright Object {top: 20, right: 20, bottom: 35, left: 40}
```

height = 560

```
d3 = ▶ Object {format: f(t), formatPrefix: f(t, n), timeFormat: f(t), timeParse:
```

import {Scrubber} from "@mbostock/scrubber"

#### Plot 2

```
selectedDateWatcher = "2020"
```

d32 = ▶ Object {event: null, format: f(t), formatPrefix: f(t, n), timeFormat: f(t

import {legend} from "@d3/color-legend"

import {Legend, Swatches} from "@d3/color-legend"

topojson = ▶ Object {bbox: f(e), feature: f(e, t), merge: f(e), mergeArcs: f(e, t

districts = ▶ Object {type: "FeatureCollection", features: Array(277)}

trr\_data = ▶ Object {2014: Object, 2015: Object, 2016: Object, 2017: Object, 2018

#### Plot4

data\_1 = ▶ Array(11877) [Object, Object, Object, Object, Object, Object, Object,

 $duration_1 = 100$ 

// import {Scrubber} from "@mbostock/scrubber"

names\_1 = ▶ Set(161) {132, 65, 93, 77, 126, 194, 67, 133, 64, 145, 177, 134, 94, datevalues\_1 = ▶ Array(148) [Array(2), Array(2), Array(2), Array(2), Array(2), Ar

 $rank_1 = f(value)$ 

 $k_1 = 10$ 

```
keyframes = ► Array(1471) [Array(2), Array(2), Array(2), Array(2), Array(2), Array
nameframes_1 = > Array(161) [Array(2), Array(2), Array(2), Array(2), Array(2), Ar
data_1 = FileAttachment("beat_trr-1.csv").csv({typed: true})
next_1 = ▶ Map(236670) {Object => Object, Object => Object, Object => Object, Obj
bars = f(svg)
labels = f(svg)
textTween = f(a, b)
formatNumber = f(t)
tickFormat = undefined
axis = f(svg)
ticker = f(svg)
formatDate = f(e)
color_1 = f(d)
x_1 = f(n)
y_1 = f(i)
height 1 = 531
barSize = 48
```

margin\_1 = > Object {top: 16, right: 6, bottom: 6, left: 0}

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