## Police Overtime - D

#### **Problem Statement:**

The problem at hand is to conduct an in-depth analysis of how the Boston Police Department (BPD) allocates and spends its operating budget, with a specific focus on the registration and utilization of overtime. The BPD operates with a budget exceeding \$400 million, and understanding how this budget is distributed is of paramount importance to ensure accountability, transparency, and to address potential issues of financial excess and wasteful spending. Additionally, there is room for creative data science work, including exploring sociological, environmental, and political dimensions related to policing and its impact.

#### **Data Collection:**

Primarily, the stakeholders have provided us access to the following 4 datasets on which we perform our exploratory data analysis-

- 1. <u>Employee earnings data (search police)</u> This dataset contains information on the earnings of all employees of the City of Boston, including job title, name, department name, regular pay, gross pay, injury pay, "other pay" and overtime pay.
- **2.** Campaign contribution data This dataset has information on the contributions made to the house representatives, senator and CC contributions, with a focus on contributions made by the police force.
- 3. BPD field activity data Spanning from 2011 to 2022, this dataset details interactions between the Boston Police Department (BPD) and the public via three record management systems: OLD RMS, NEW RMS, and MARK43. It sheds light on the frequency and duration of stops, highlighting specific officers and supervisors involved. The data aids in recognizing patterns, potential discrepancies, and mapping active officers and supervisors using key variables like 'stop\_duration', 'contact\_officer', and 'supervisor'. It offers a comprehensive view of BPD's field activities and trends over the years.
- 4. Overtime data from 2012-2022 This dataset provides valuable information to discover the discrepancies between the amount of authorized overtime hours and the amount of hours actually worked. Also, it provides insights on the overtime details for the duty performed by the police in the Court.

# **Analysis / Tasks Performed:**

## **Earnings data:**

Performed analysis on the earning-data.csv dataset which includes general earnings data for city of Boston employees, and answered the following questions:

How much BPD officer pay came from injury pay? What percentage of officers took injury pay in a given year?

Percentage of BPD Officers Who Took Injury Pay: 15.07%

Total Injury Pay for BPD Officers: \$23,628,467.04 Percentage of Total Gross Pay from Injury Pay: 5.82%

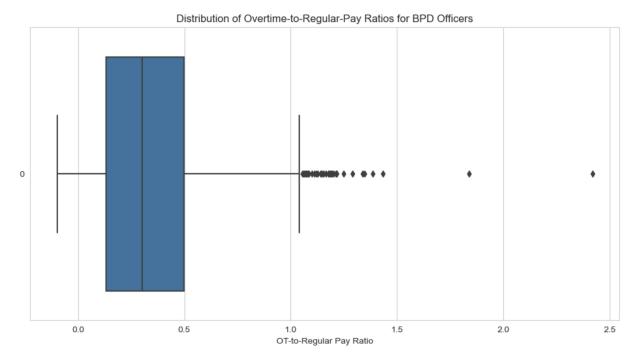
Identifying instances of financial excess in BPD spending, Characterizing wasteful BPD overtime practices

Performed analysis on the ratio of overtime to regular pay, and identified the top officers who took the most overtime pay, as well as created a boxplot to show the distribution of how officers take overtime pay:

Percentage of BPD Officers Who Took Overtime Pay: 79.77%

Top 10 officers who took the most overtime pay:

NAME REGULAR OVERTIME	OT_to_Regular_Ratio
20438 Stewart, Greta E. 1977.92 4785.93	2.419678
11518 Woodley,Lorenzo I 11775.11 21638.24	1.837625
1846 Webster, Geneese 65294.61 93654.82	1.434342
111 Acosta, Jose L 109502.02 151608.89	1.384531
3 Demesmin, Stanley 145775.26 196515.25	1.348070
Jones, Craig D 108038.52 144457.87	1.337096
203 Christie, Albert C 109502.02 141222.99	1.289684
14 Barrett, Thomas E. 130930.12 163494.70	1.248717
387 Cornelius, Devon M 92637.86 112645.22	1.215974
260 Shikoluk, Michael 99922.46 121258.11	1.213522



It appears that there is a heavy skew in the amount of overtime pay taken by officers, with a large amount of outliers well above the average. In addition, the average ratio of overtime to regular pay is quite high, hovering around 30% of the regular pay being overtime.

# **BPD** field activity data:

Upon investigating BPD field activity data. As the record system of BPD has changed twice from the past 10 years (old RMS, new RMS, Mark43), started with conducting a thorough analysis on the combined dataset, which encompasses detailed stop-and-frisk records for a specific region over the span of 2011 to 2022. I addressed the following research inquiries:

Understanding and Cleaning stop duration Data:

Examined the 'stop\_duration' variable, which initially contained a mix of string descriptors and numerical values.

Mapped string values to a median time, and converted everything to minutes for standardization.

Analyzed the distribution of stop durations and identified potential outliers or erroneous data entries.

**Key Findings:** 

Total number of cases: 211,522

Cases with non-empty stop\_duration: 35,971

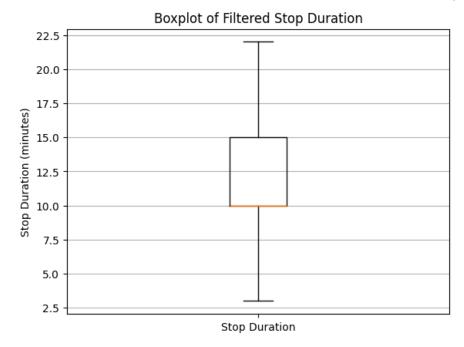
Average stop\_duration: 49.47 minutes

count	37810.00000	
mean	50.27818	
std	2187.32763	
min	1.00000	
25%	10.00000	
50%	10.00000	
75%	15.00000	
max	389700.00000	

Name: stop\_duration, dtype: float64

drop outliers by calculating the IQR (1.5)

count 33314.000000 mean 11.174971 std 4.489778 min 3.000000 25% 10.000000 50% 10.000000 75% 15.000000 max 22.000000 Name: stop\_duration, dtype: float64



Analysis of Officer and Supervisor Activity:

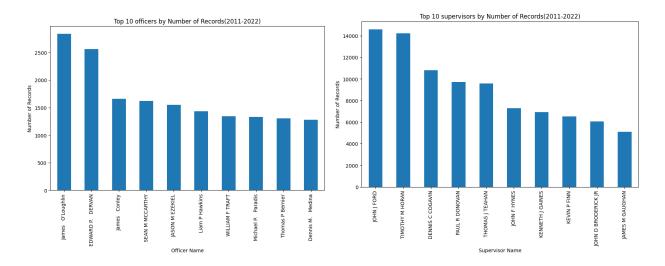
Explored the data to understand which officers and supervisors had the highest number of records.

Initially utilized the unique IDs of officers and supervisors for this analysis.

Later, enhanced the visual representation by mapping IDs to actual names to offer a clearer understanding of the most active personnel.

### Challenges Addressed:

Dealt with potential naming inconsistencies by ensuring that a unique mapping between IDs and names was maintained. This helped in addressing potential issues arising from multiple naming conventions for the same individual.



### **Campaign Contribution Data & Overtime Data:**

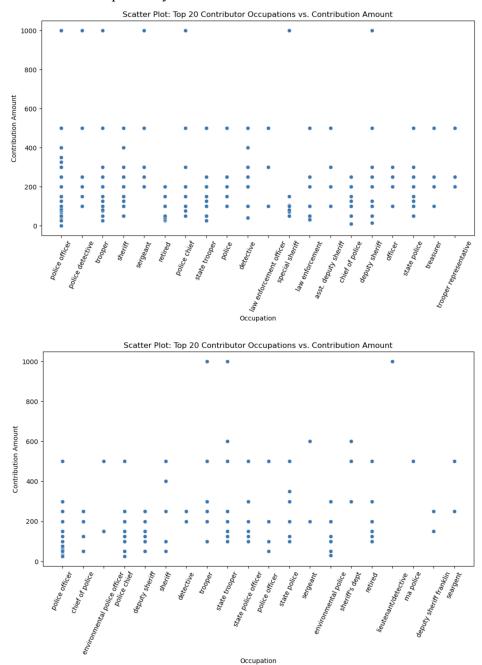
Exploratory data analysis was performed on the following two datasets as well, <u>Campaign</u> <u>contribution data</u> and <u>Overtime data from 2012-2022</u>, with which we were able to answer quite a few of the base questions.

In the Campaign contribution data we first explored the data which provides us with details of the contributions made to the House Reps and Senators. Notably this dataset had 3 pieces of vital information, the contributors name, the recipient's name and the amount received by the recipient/house rep/Senator. We find the name and amount received by the top 10 recipients.

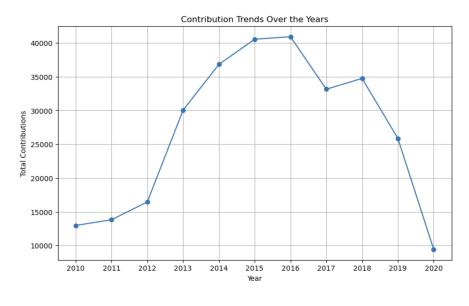
30750.00
25275.00
14500.00
11132.73
10885.00
10310.00
9700.00
8500.00
8350.00
6950.00

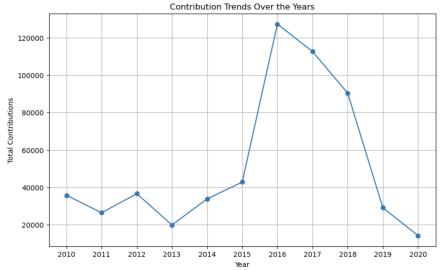
Top 10 names who Matching Names	have received	the	most	contributions
Michael Moore	163730.0			
Stanley Rosenberg	100800.0			
Karen Spilka	72800.0			
Michael Rodrigues	46830.0			
Walter Timilty	29925.0			
Michael Rush ´	24920.0			
Bruce Tarr	22750.0			
Patrick O'Connor	17850.0			
John Velis	8750.0			
Marc Pacheco	8400.0			

We create a scatter plot to observe contributor occupation vs. contribution amount for the house rep and Senator datasets respectively.



We also analyze the amount of contributions received each year by the House Reps and the Senators, which can provide insights into how much impact the incoming elections have on the amount of contributions.





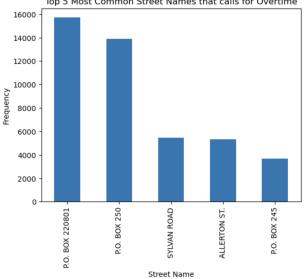
In the overtime dataset we have two kinds of data. One, based on the yearly details ranging from 2012-2022 on how overtime has been used by the police officers and another based on the police officer's overtime based on his services in the Court. The main fields found upon exploring these datasets are Employee (Officer name), hours worked and hours paid.

For the yearly dataset, we find the total amount of overpaid overtime hours and find the average rank of the officer who has applied for overtime in that year. On multiplying this we get answers to some of the base questions in the project. For instance, from the analysis of financial excess of BPD spending we can conclude about the amount of money the state would save in the year of 2022 if the police officers were paid according to their actual worked hours amounts to **6,781,003.98** \$

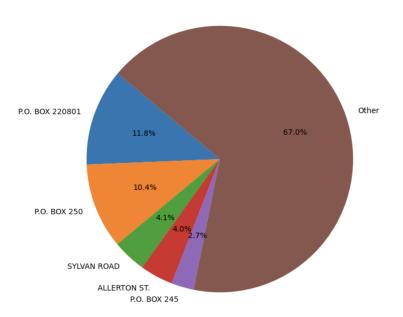
We also have extracted other crucial insights as shown by the images below for the year of 2012-

Top 5 street names with a count of the times an offcer has done overtime:
P.O. BOX 220801 15704
P.O. BOX 250 13870
SYLVAN ROAD 5462
ALLERTON ST. 5313
P.O. BOX 245 3662
Name: Street Name, dtype: int64

Top 5 Most Common Street Names that calls for Overtime



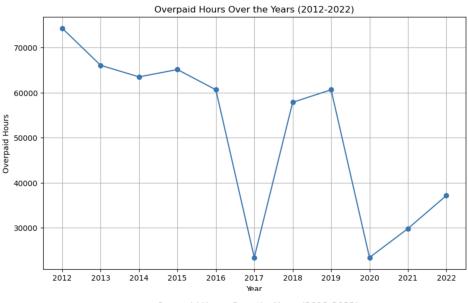
Top 5 Most Common Street Names vs. Others

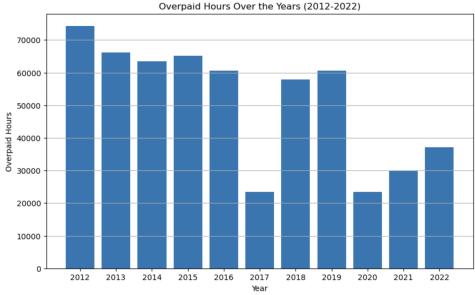


The overpaid hours of the top 5 most overpaid officers: Employee BURCH, CHARLES K. 826.77

641.50 FITZPATRICK, DENIS J RUSSELL, PATRICK 632.37 CHEN, WILLIAM L 601.77 549.25 DORCH, KENNETH C

For the Court overtime dataset, we calculate the total overpaid hours due to discrepancies in the hours worked vs. hours paid of the Officers working overtime in the Court. Further, we answer a few more of the base questions of the project by visually comparing yearly overpaid hours as shown below and we also find the distribution of ratios of overtime worked vs. overtime paid to find potential outliers in the court OT database, which has been shown below for the year 2022.



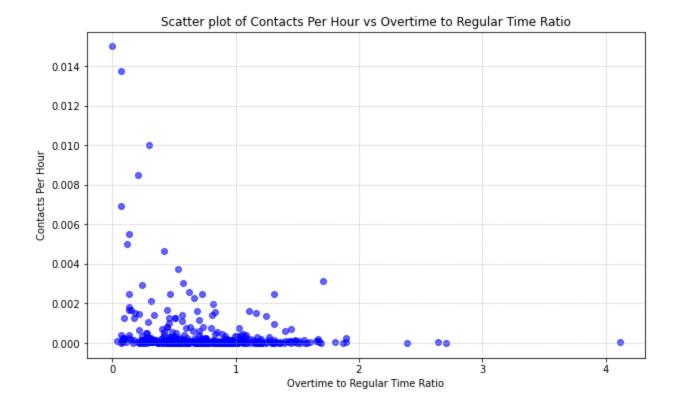


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Mean Overtime Ratio: 0.38
Median Overtime Ratio: 0.25
Standard Deviation of Overtime Ratio: 0.29
Potential Outliers:
                              NAME
                                    Average_Overtime_Ratio
14431 Agudelo-Echevarria, Marily
                                                    0.\overline{0}62500
8063
                Becker, Matthew F.
                                                    0.093750
14740
            Charles-Sampson, Azadi
                                                    0.093750
13188
                  Crabbe, David D.
                                                    0.093750
10355
                 Fantaroni, Evan M
                                                    0.062500
                    Foley, James M
                                                    0.093750
581
5035
            Hyppolite, Jonathan A
                                                    0.062500
13884
                                                    0.093750
                  Lawless, Timothy
                                                    0.062500
6697
            Legacy, Mary Katherine
14836
                                                   0.093750
            Lonergan, Ryan Charles
                                                    0.062500
6592
                     Mendes, Kevin
                Milton, Christa A.
14832
                                                   0.062500
                                                   0.062500
             Murphy, Daniel Brian
8293
1745
                    Powell, Nadine
                                                   0.093750
                   Ross,Allison D
                                                    0.062500
4622
6696
        Samuel-Lenehan, Angelique
                                                    0.062500
9943
        Van Orman, Haley Margaret
                                                    0.062500
13950
                     West, Shawn L
                                                    0.062500
7061
                  Yanovitch, Scott
                                                    0.083333
                      Distribution of Overtime Ratios (2022)
               0.2
                              0.4
                                             0.6
                                                            0.8
                                                                          1.0
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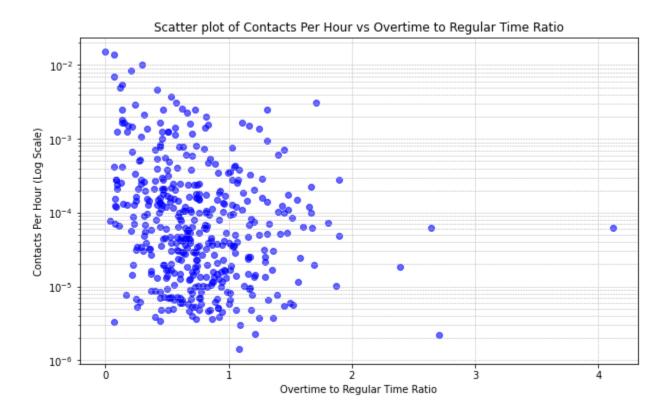
### **Overtime Hours and Field Contact:**

In the interest of seeing the use of overtime hours, we decided to quickly see if a relationship exists between field contacts per hour, and the overtime/regular-time ratio. Since there was only a hunch there would be a correlation, we exclusively looked at the year 2022. Plotting the relationship we produced this plot:

Overtime Ratio (WRKDHRS / OTHOURS)



Seen as the plot above did not expose too much detail due to extreme outliers, we produced another plot, this time with the contacts per hour graphed on a logarithmic scale:



As we can see, there is a clear negative correlation, potentially showing the decrease of officer productivity during overtime hours. Granted, the sample size is not large enough to draw such a concrete conclusion, but there is now more than enough reason to conduct more research into the potential relation.

### **Individual Contributions:**

- 1. Aaron Zheng: Analyzed and performed analysis on the earning-data.csv dataset which includes general earnings data for city of Boston employees, and answered some of the base questions.
- 2. Chen Yang: Performed an analysis on the BPD field activity data and addressed a few of the challenges.
- 3. Rithik Bhandary: Performed exploratory data analysis on the campaign contribution dataset and the overtime dataset which answer all of the base questions for the overtime dataset. Plotted trends and found valuable insights with figures.
- 4. Jakob Rundlett: Found correlation and produced graphs alluding to decrease in officer productivity during overtime hours.