# UC San Diego

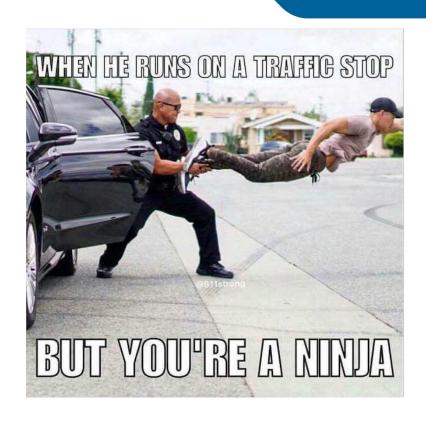
# Police Vehicle Stops Data Analysis ECE 143 - Group 5

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# Agenda

- Overview and Motivation
- Dataset, Data processing and EDA
- Temporal Analysis
- Geospatial Analysis
- Demographic Analysis
- Summary and Conclusion



#### **Overview And Motivation**

- San Diego is densely populated urban area with huge vehicle traffic.
- Police department is continuously monitoring for better public safety.
- Can we find few insights which helps them for better resource allocation?
- Many report says that Police department has negative bias towards few demographic groups.
- Does the dataset follow this trend?



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# **Dataset and Data Processing**



#### **Dataset**

- The Dataset gives records of car stops/pullovers by the police from Jan 2014 to June 2018 in San Diego.
- Police Vehicle Stops City of San Diego Open Data Portal.
- The data also describes attributes related to the record.
- The Data is in CSV format and can be represented in Tabular form.

Columns are features/Attributes related to the pullover.

A row is a pullover record

stop_id	stop_cause	service_area	subject_race	subject_sex	subject_age	date_time	date_stop	time_stop	sd_resident	arrested	searched	search_details_id	search_details_type	search_details_description
<b>0</b> 1044975	Moving Violation	110	W	0.0	24	2014-01-01 01:25:00	2014-01- 01	1:25	Y	N	N	1208956.0	ActionTaken	Citation
<b>1</b> 1044976	Moving Violation	320	W	0.0	42	2014-01-01 05:47:00	2014-01- 01	5:47	Y	N	Ň	1208957.0	ActionTaken	Verbal Warning
<b>2</b> 1044977	Moving Violation	320	L	0.0	29	2014-01-01 07:46:00	2014-01- 01	7:46	γ	N	N	1208958.0	ActionTaken	Verbal Warning
<b>3</b> 1044978	Moving Violation	610	W	0.0	23	2014-01-01 08:10:00	2014-01- 01	8:10	γ	N	N	1208959.0	ActionTaken	Citation
<b>4</b> 1044980	Equipment Violation		Н	0.0	35	2014-01-01 08:35:00	2014-01- 01	8:35	N	N	N	1208961.0	ActionTaken	Citation

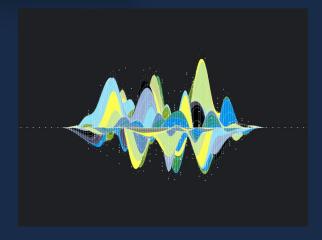
### **Data processing**

- Combining multiple CSV Files on both rows and Columns.
- Removing Columns that fails to provide any insights.
  - For eg: the 'stop ID' fails to provide meaningful insights.
- Removing NaN values.
- Encoded data replaced with original non-decoded entries.
  - For eg: For all races encoded as letters in the CSV data.

We replaced that with the original race tag.

'W' in CSV is replaced with 'White'.

# **Exploratory Data Analysis(EDA)**



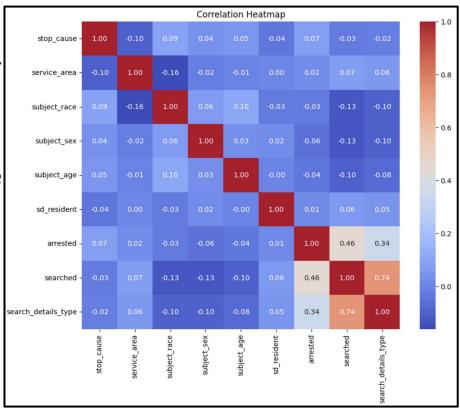
# **Exploratory Data analysis(EDA)**

- The main purpose of EDA is to help look at data before making any assumptions.
- EDA also helps to identify important features/Attributes.
- Important features help us to form meaningful inferences.
- List of features:

stop_id	action	arrested		
stop_cause	subject_age	searched		
service_area	date_stop	search_details_id		
subject_race	time_stop	search_details_type		
subject_sex	sd_resident	search_details_descri ption		

# **EDA (Finding Important Features)**

- Find inter-dependency between features.
- If 2 features are highly correlated use 1.
- 'Searched' and 'Search\_ details\_type' highly correlated so we can take only one



# **EDA(Important Features)**

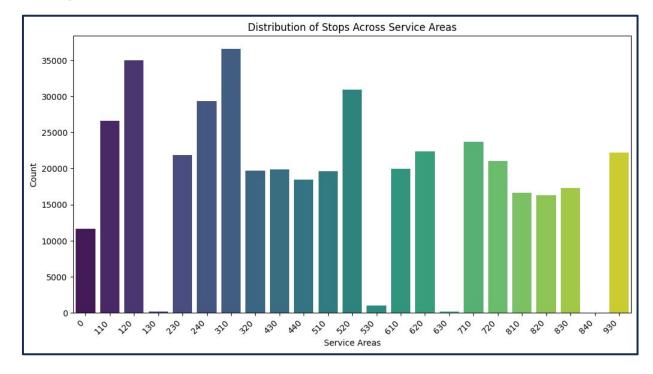
- We can't include all features in the correlation Heatmap.
- Features related to 'Time' are important for Temporal Analysis.
- Binary Features are less likely to have meaningful insights.
- So some of the Important features:

Stop_cause	Subject_race	Date_stop			
Service_area	Subject_age	Time_stop			

 Later we decompose Date and time in sub columns of months, year, hours for plotting purposes.

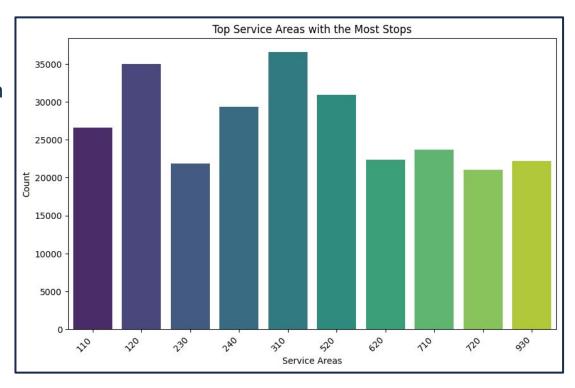
# **EDA(Service Area)**

- Stops per Service Area
   Codes tell us which area
   has more pull over rate.
- This information can be used for Geospatial analysis of the Data.



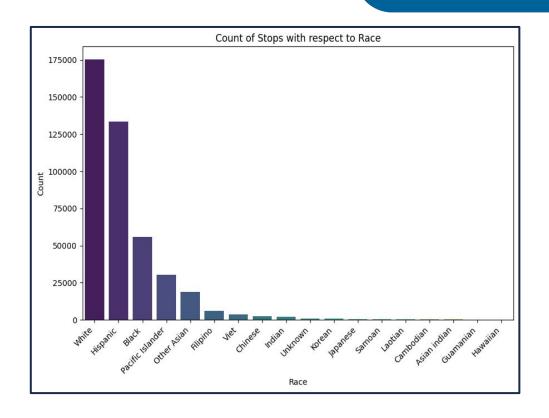
# **EDA(Service Area)**

- Top ten service areas which has the most number of stops.
- We have referred to official San diego Service codes.



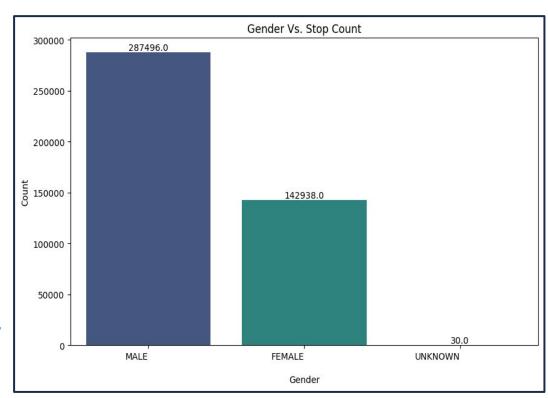
# **EDA(Subject Race)**

- Higher chance of white individuals being pulled over, without considering population demographics.
- San Diego Demographics: Whites 45.9%, Hispanics 33.5%, Blacks 4.7%.
- Pullover rate for blacks almost 3 times in proportion, despite their population being approximately 1/10th of whites.
- <u>City Demographic Profiles San</u>
   <u>Diego County</u>

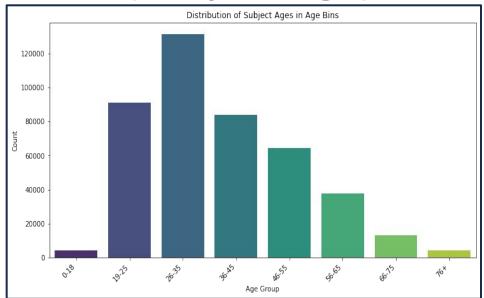


### **Stops Based on Genders**

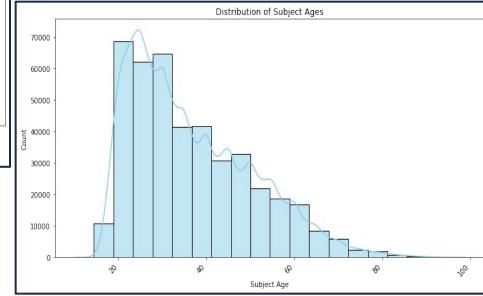
- Observe that we can find more number of Male Drivers pulled over as compared to Female Drivers.
- But the population of Males and Females is roughly equal in San Diego.
- So we might interpret that Male drivers are twice as likely to drive recklessly or being pulled over.
- <u>City Demographic Profiles San</u>
   <u>Diego County</u>



# **EDA(Subject Age)**



 Most stops observed in the age group of 26-35.  This age group aligns with higher numbers in San Diego demographics.



#### **EDA**

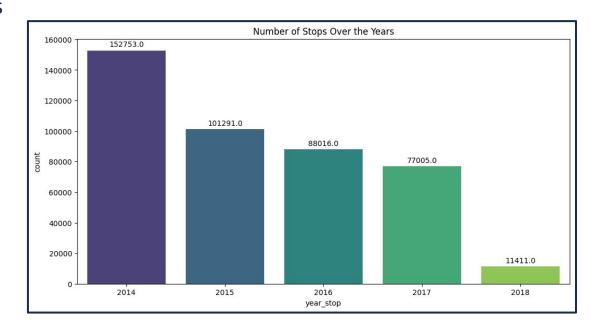
- We now have figured out our important features and done basic EDA.
- Further we can move on to work on detailed insights based on
  - Temporal Analysis
  - Geospatial Analysis
  - Demographic Analysis



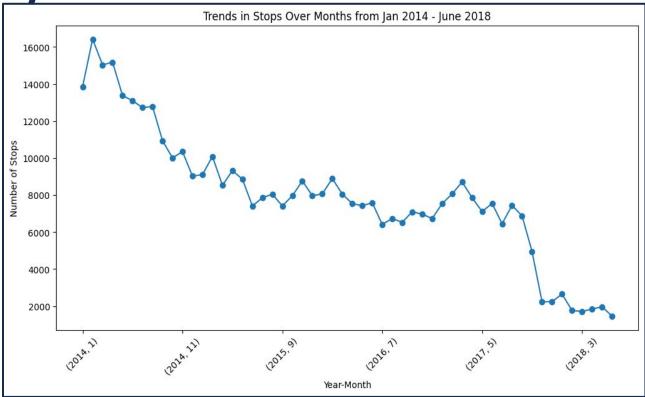
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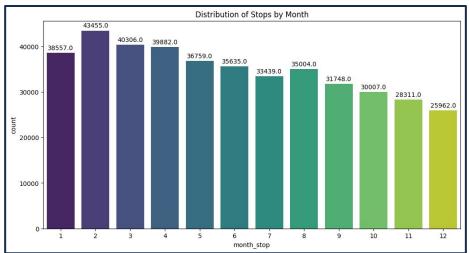
- The number of stops per year is following a negative trend.
- There are two possible reasons for this:
  - More adherence to the law
  - Fewer Cops on roads
- 2018 cannot be considered as the data is collected till June.
- Census of State and Local Law
   Enforcement Agencies, 2018 –
   Statistical Tables

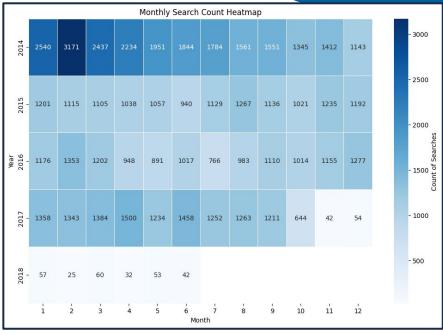


 In Support of previous Bar graph, we can see a decreasing trend over years.

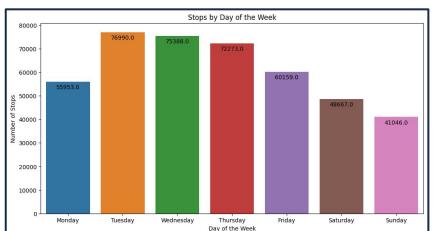


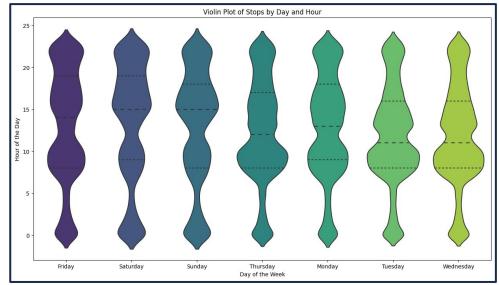
- The number of stops per month is following a negative trend.
- NBC News Article



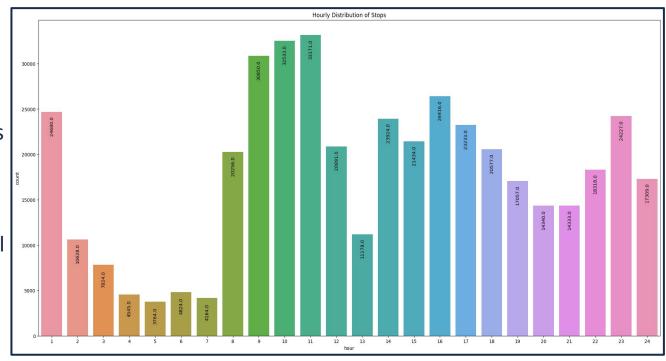


- Here we find that the Stop rate decreases from Tuesday to Sunday, Monday being an exception.
- No special reason but maybe due to public being more attentive to start.



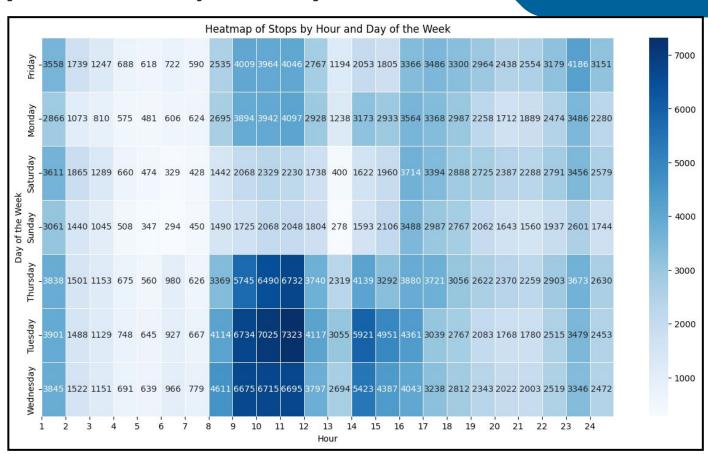


- Bar chart: Hours of the day vs Count.
- Higher number of stops during rush/office hours (8 to 11).
- Mornings are relatively quiet.
- Late nights show a small rise, possibly people returning from work.



#### **Temporal Analysis - Heatmap for Days vs Hours**

- Detailed view by days and hours using heatmap.
- Weekdays, especially office hours, show higher numbers of pullovers.
- Weekends deviate from this pattern with lower pullover rates.



### **Temporal Analysis - Inferences**

#### In Summary we can say:

- Decrease in stops and pullovers from 2014 to 2018.
- General trend of more stops on weekdays, particularly during office hours.
- Feb 2014 shows a spike, but excluding it, a downward monthly trend is observed.
- Recommendation: Increase police workforce during office hours, optimize resource allocation from March to December, especially during festive days.



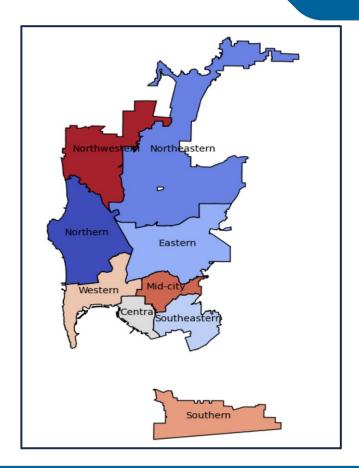
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# **Geospatial Analysis**



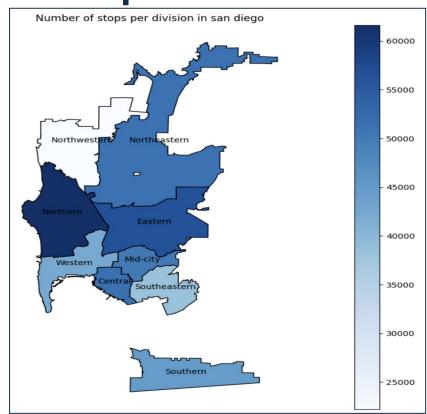
### **Geospatial Analysis**

- Which parts of San Diego has more stops?
- Which part of San Diego has highest probability of arrest?
- Note:- The police departments has divided San Diego into 9 divisions



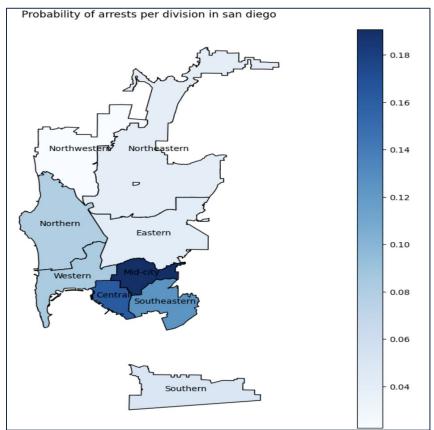
### **Geospatial Analysis - Most Stops**

- Northern division has highest stops.
- Many tourist attractions.
- More population compared to other regions.



**Geospatial Analysis - Most Arrests** 

- Mid-city division has highest probability of Arrests.
- Mid-city has highest probability of arrests, this could be because the area experiences high crime rate, rules violations.



### **Geospatial Analysis - Inference**

 Northern division has more stops, due huge population and many tourist attractions.

Helps to identify the requirement of resource allocation.

 Mid-city has highest probability of arrests, this could be because the area experiences high crime rate, rules violations.

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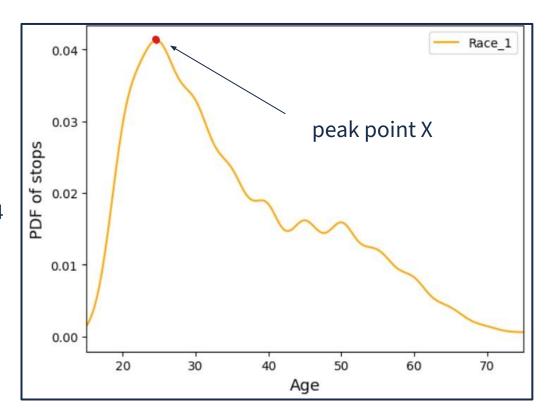
# Demographic Analysis

Main Question :- Is there Bias and Discrimination against a Race or age group?



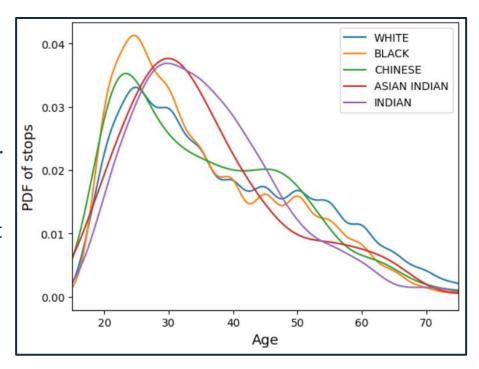
# **Demographic Analysis**

- Point X has coordinates.
  - o (27,0.04)
- P(stopping | Race, Age) =Y\_coordinate.
- Probability for current example
  - P(stopping | Race 1, 27) = 0.04
- More probability means more bias towards one age group or Race.



# **Demographic Analysis**

- From the combined graph,
   P(stopping | Black, 27) = 0.04 which is the highest.
- This means that a young black person will get stopped by the cops more often.
- Ideally the PDF for Races should be same. Only then we can say that Bias against anyone group is absent. But that is not the case here.



# **Demographic Analysis**

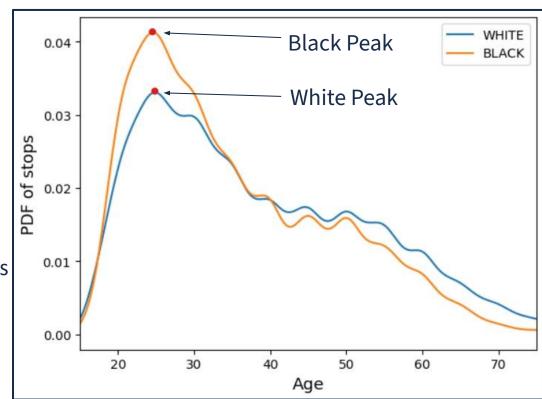
- P(stopping | Black, 27) = 0.04
- P(stopping | White, 27) ~ 0.032

i.e

P(stopping | Black, 27)>

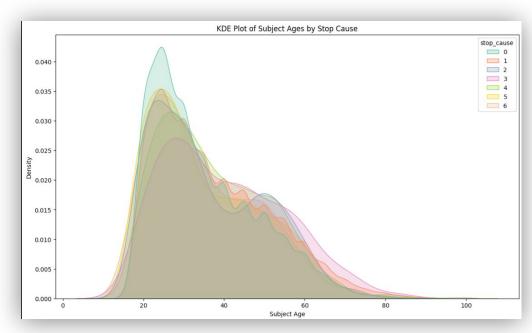
P(stopping|White, 27)

 This means a young black person is more likely to get stopped than a young white person.



### Age Vs Stop Cause

- Same trends as observed in the age vs Race graph.
- Youngsters are more likely to be stopped by the police than compared elderly folks.
- People usually stop driving cars/vehicles after 65. Hence a sudden drop in %.



## **Demographic Analysis-Inferences**

- The probability tells that a young Black person is more likely to get stopped by the cops.
- Does this mean that there is a Race/age bias?
- The answer is no, we still need more data to analyse.



# **Summary/Conclusion**

#### Temporal Analysis

By the trend of stops in a day and a year, we suggest that increasing police workforce on commute hours and holidays from March to December help to optimize the resource of police.

#### Geospatial Analysis

The division with higher stops require more resource allocations.

#### Demographic Analysis

The probability distribution indicates that there is a bias, but more evidence is needed to back the statement.

Outliers - Always follow the rules!!

#### Reference

- <u>City Demographic Profiles San Diego County</u>
- Racist Comments, Excessive Force and Offensive Behavior Revealed in San Diego
   Police Department Internal Affairs Cases
- https://voiceofsandiego.org/wp-content/uploads/2014/01/Vehicle-Stop-form.pdf
- RIPA police stop data race of persons stopped City of San Diego Open Data Portal

