

CS 777: Big Data Analytics

LAPD CRIME ANALYSIS

Term Project

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The Dataset

The LAPD police report dataset provides a comprehensive overview of crime incidents in Los Angeles from 2020 onward.

Originating from original crime reports transcribed from paper documents, potential data inaccuracies may exist due to the manual transcription process.

Date Rptd	DATE OCC	TIME OCC	AREA	AREA NAME	Rpt Dist No	Part 1-2	Crm Cd	Crm Cd Desc	Mocodes	Vict Age	Vict Sex	Vict Descent	Premis Cd
01/08/2020	01/08/2020	2230	03	Southwest	0377	2	624	BATTERY - SIMPLE ...	0444 0913	36	F	B	501
01/02/2020	01/01/2020	0330	01	Central	0163	2	624	BATTERY - SIMPLE ...	0416 1822 1414	25	M	H	102
04/14/2020	02/13/2020	1200	01	Central	0155	2	845	SEX OFFENDER REGI...	1501	0	X	X	726
01/01/2020	01/01/2020	1730	15	N Hollywood	1543	2	745	VANDALISM - MISDE...	0329 1402	76	F	W	502
01/01/2020	01/01/2020	0415	19	Mission	1998	2	740	VANDALISM - FELON...	0329	31	X	X	409

only showing top 5 rows



Data Cleaning and Processing

1.

Column Handling:

- Dropped irrelevant columns.
- Renamed for clarity.
- Trimmed off extra white spaces.

2.

Formatting:

- Date columns standardized.
- 'LOCATION' and 'Cross Street' merged.
- Mapped 'VictDescent' for interpretability.

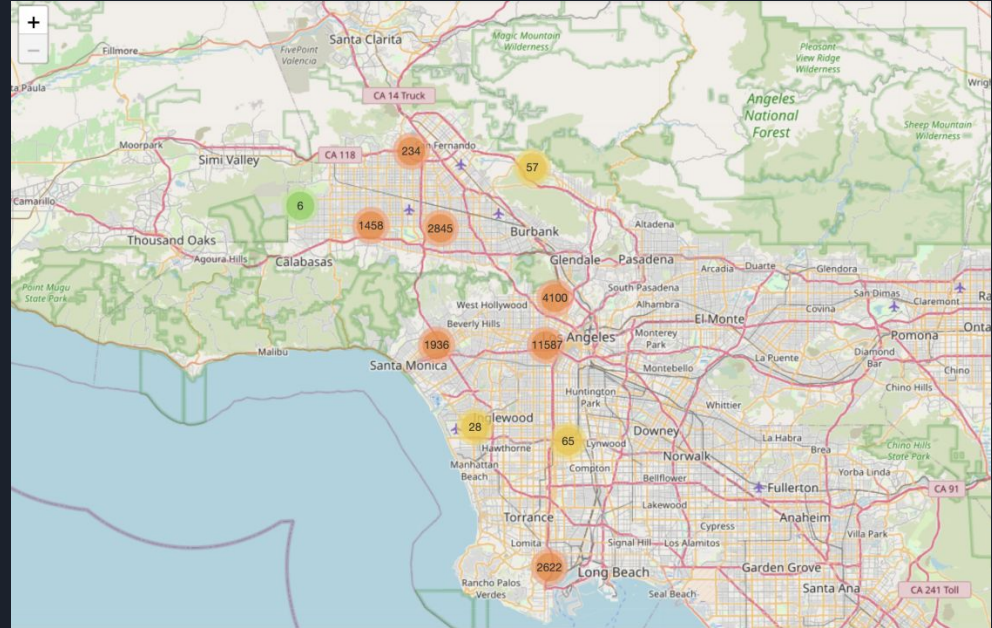
3.

Handling Nulls:

- 'PremisCd' or 'PremisDesc' null rows dropped.

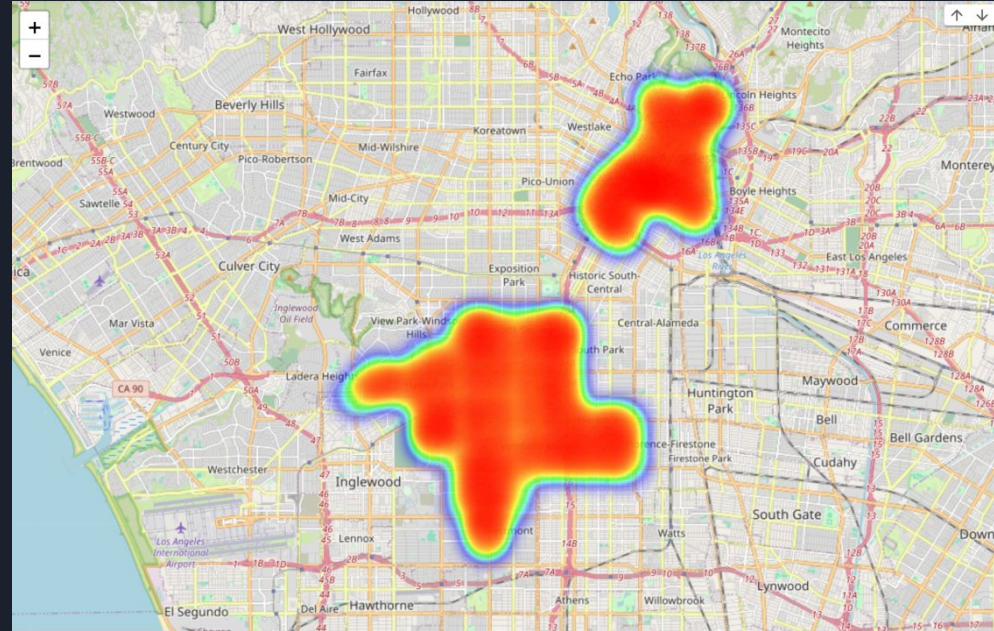
Geospatial Analysis

1. Extracted geographical coordinates (latitude and longitude).
2. Downloaded the geojson file for LA.
3. Plotted crime locations on the map, with each point representing an incident.
4. With the help of this geospatial map, we observed that Central LA has the highest concentration of crimes.
5. Generated heatmaps to visually represent the intensity of crime in different areas.



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Integration with PostgreSQL and Tableau

- Leveraged the Python library psycopg2 to seamlessly transfer our data into PostgreSQL database.
- The primary motivation was to establish a connection between Tableau and our dataset where PostgreSQL was a required intermediate. This integration facilitates diverse visualizations in alignment with our three key business questions.



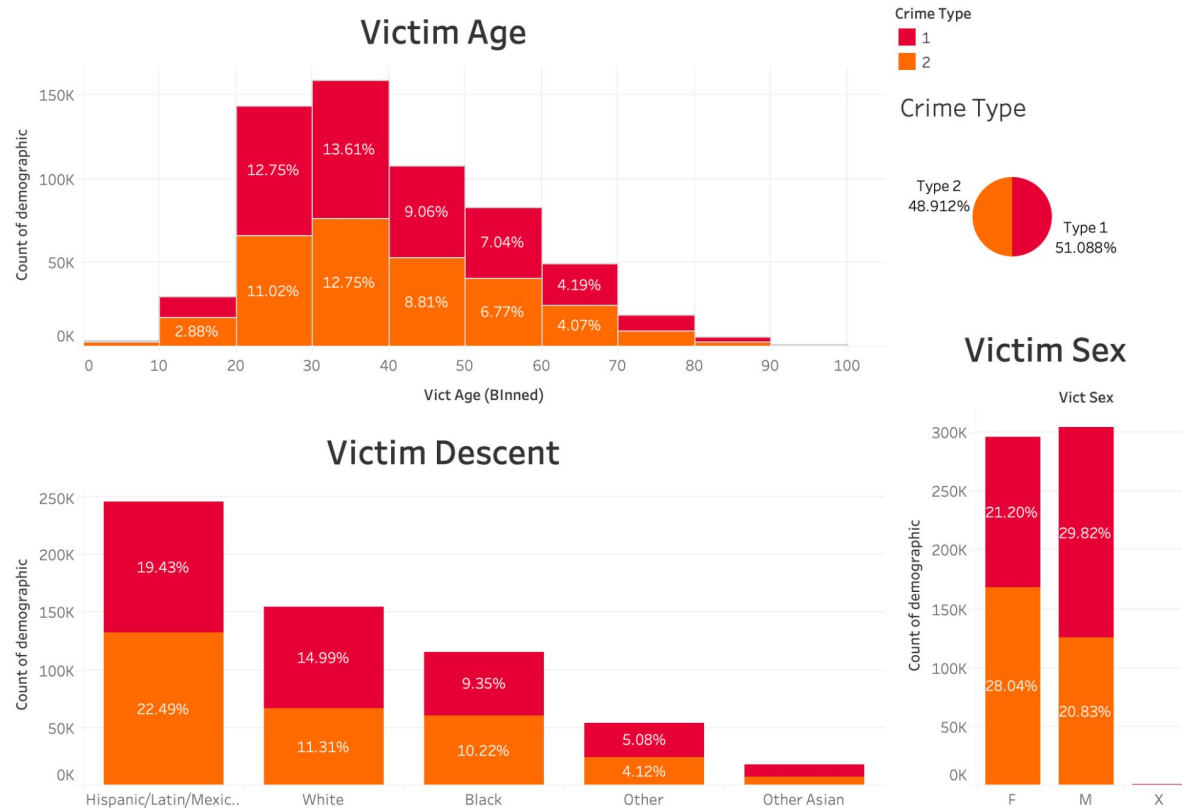
Business Questions & Tableau Dashboards

Investigated whether an ***individual's demographic profile***, encompassing age, ethnicity, and gender, correlates with and potentially indicates a ***higher susceptibility to specific crime types***.

Explored how crime data can be leveraged to understand the ***most prevalent crimes and crime rates*** based on ***geographic areas***.

Examined the relationship between ***the time of occurrence and the time of reporting*** for different types of crimes, seeking to uncover any dependencies.

BUSINESS QUESTION 1: Does an individual's demographic profile, including age, ethnicity, and gender, correlate with and potentially indicate a higher susceptibility to specific crime type?

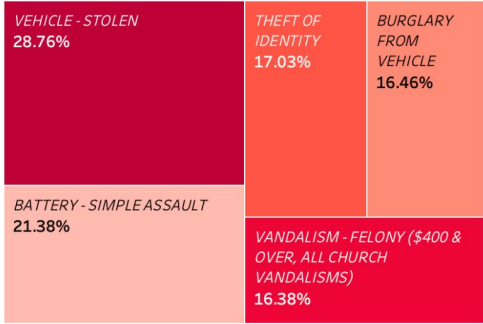


INFERENCE:

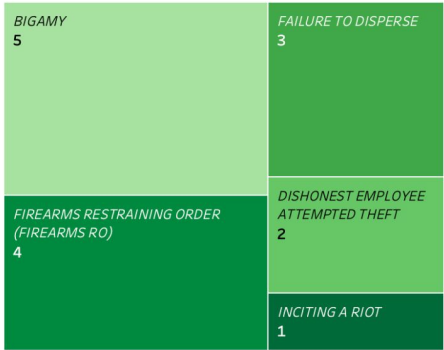
The data indicates that 30-40 year old Hispanic/Latin/Mexican males are most likely to be victims of Type 1 crimes.

BUSINESS QUESTION 2: How to leverage crime data to understand most prevalent crimes and crime rates based on the areas?

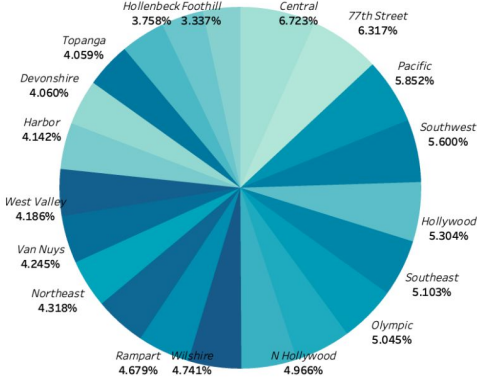
Top 5 Crimes Committed



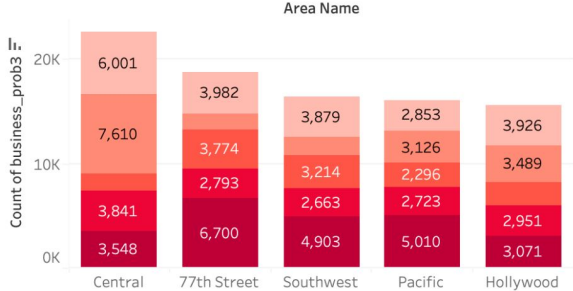
Least 5 Crimes Committed



Area-Wise Incidence Rates



Crime Hotspots: Top 5 Offenses Across Key Areas



INFERENCE:

Stolen vehicles dominate crime in Los Angeles, notably in Central LA, where property crimes like burglary from vehicles are on the rise. Despite lower overall crime rates in Foothill, the challenge of stolen vehicles persists citywide. The rarity of inciting a riot suggests a generally stable societal environment.

BUSINESS PROBLEM 3: Is there a relation between the difference in the time of occurrence and report of a crime dependent on the crime?

Temporal Discrepancy Heatmap: Occurrence vs. Reporting Time

CRM AGNST CHLD (13 OR UNDER) (14-15 & SUSP 10 YRS OLDER) 2,977 hours	DISHONEST EMPLOYEE ATTEMPTED THEFT 2,088 hours	DOCUMENT FORGERY / STOLEN FELONY		CREDIT CARDS, FRAUD USE (\$950.01 & OVER)	GRAND THEFT /	HUMAN		HUMAN
		CONTRIBUTING						
SEX OFFENDER REGISTRANT OUT OF COMPLIANCE 2,704 hours	SEXUAL PENETRATION W/FOREIGN OBJECT 1,524 hours	THEFT OF IDENTITY			CREDIT CARDS, FRAUD USE (\$950 & UNDER)	CHILD		
	EMBEZZLEMENT, PETTY THEFT (\$950 & UNDER) 1,492 hours							
SEX UNLAWFUL(INC MUTUAL CONSENT, PENETRATION W/ FRGN OBJ 2,666 hours	EMBEZZLEMENT, GRAND THEFT (\$950.01 & OVER) 1,429 hours	SODOMY/SEXUAL CONTACT B/W PENIS OF ONE PERS	STALKING 711 hours					
				BRIBERY				
LEWD/LASCIVIOUS ACTS WITH CHILD 2,454 hours	ORAL COPULATION 1,399 hours	CHILD ANNOYING (17YRS & UNDER)						
	RAPE, FORCIBLE 1,343 hours	UNAUTHORIZED COMPUTER ACCESS						
BIGAMY 2,232 hours	DOCUMENT WORTHLESS (\$200.01 & OVER)	MANSLAUGHTER, NEGLIGENCE 1,238 hours	PEEPING TOM					

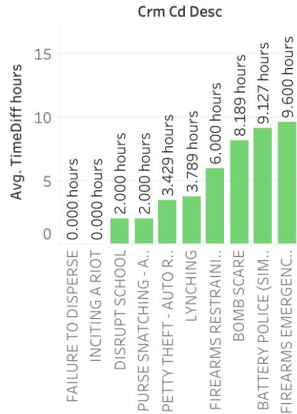
Average Time Difference in Hours



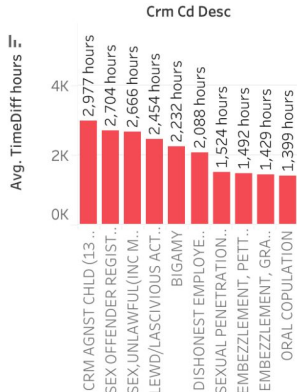
INFERENCE:

Crimes against children and sex offenses, requiring more time for reporting, underscore potential barriers in victim disclosure or societal reluctance. Conversely, swift reporting of school disruptions and purse snatching suggests heightened awareness and prompt community responsiveness to immediate threats

Analyzing Minimum Crime Reporting Times



Analyzing Maximum Crime Reporting Times



LA Crime Statistics Tableau Dashboard

<https://public.tableau.com/LACrimeAnalytics>

Machine Learning for Demographic Predictions using GCP

Accuracy: 0.5118049380458901

Classification Report:

		precision	recall	f1-score	support
American Indian/Alaskan Native		1.00	0.01	0.01	157
Asian Indian		0.00	0.00	0.00	80
Black		0.48	0.33	0.39	22992
Cambodian		0.00	0.00	0.00	7
Chinese		0.35	0.01	0.02	640
Filipino		0.00	0.00	0.00	696
Guamanian		0.00	0.00	0.00	11
Hawaiian		0.00	0.00	0.00	26
Hispanic/Latin/Mexican		0.54	0.74	0.63	49217
Japanese		0.25	0.00	0.01	227
Korean		0.19	0.02	0.03	849
Laotian		0.00	0.00	0.00	12
Other		0.24	0.03	0.05	10766
Other Asian		0.20	0.01	0.03	3528
Pacific Islander		0.00	0.00	0.00	47
Samoa		0.00	0.00	0.00	4
Vietnamese		0.00	0.00	0.00	163
White		0.48	0.56	0.52	30909
accuracy				0.51	120331
macro avg		0.21	0.09	0.09	120331
weighted avg		0.47	0.51	0.47	120331

Accuracy: 0.5711329582568083

Classification Report:

		precision	recall	f1-score	support
F		0.57	0.56	0.56	59256
M		0.58	0.58	0.58	60928
X		0.00	0.00	0.00	147
accuracy				0.57	120331
macro avg		0.38	0.38	0.38	120331
weighted avg		0.57	0.57	0.57	120331

Linear Regression with PCA MSE: 236.2902424485539

Decision Tree Regression with PCA MSE: 562.0412958005537

Decision Tree Regression with PCA MSE: 243.05583133485314

Random Forest Regression R-squared: 0.005272755796456696

Decision Tree Regression R-squared: -1.300203151801954

Linear Regression R-squared: 0.03296151994304031



Conclusion

- Successfully conducted comprehensive analysis of LAPD crime dataset to uncover insights into crime trends, demographics, and temporal patterns
- Developed interactive Tableau dashboard to enable data-driven decision making for law enforcement
- Achieved 50-60% accuracy in predicting victim demographics using machine learning models
- Identified opportunities to optimize resource allocation and enhance public safety strategies based on analysis



Future Scope

- Refining Machine Learning Models
- Exploration of Deep-Learning Integration
- Continuous Data Collection for Trend Detection