



CHICAGO CAR CRASH PROJECT - ML



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PROBLEM STATEMENT



- The primary objective of this study is to identify and understand the main contributing factors to car accidents in Chicago, with a specific focus on driver behavior.
- By analyzing various modeling techniques and their results, the aim is to determine the primary causes of car accidents and provide insights that can contribute to the development of effective strategies for accident prevention and road safety improvement.



DATA

Data in this case is obtained from
<https://data.cityofchicago.org/Transportation/Traffic-Crashes-Crashes/85ca-t3if>

TECHNIQUES



Data Collection

Data Preprocessing:

Exploratory Data Analysis

Feature Selection

Modeling Techniques:

Regression Analysis:

Classification Analysis:

Clustering Analysis

Model Evaluation

Interpretation and Recommendations



RESULT

RESULTS

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MODELLING

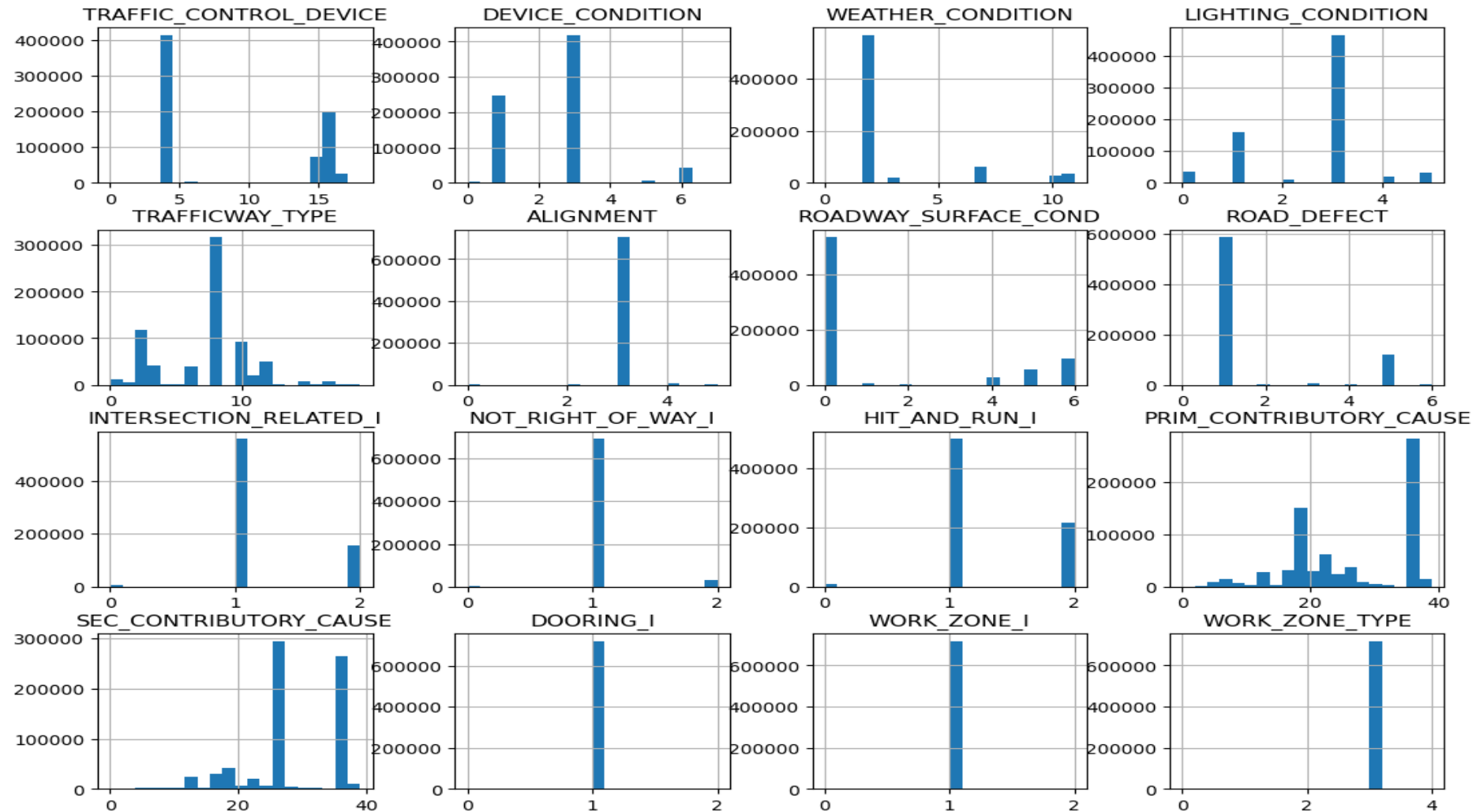
REMARKS



VISUALISATIONS

Some of the visualizations done

Distribution of Numeric Features



MODELLING: RANDOMCLASSIFIER

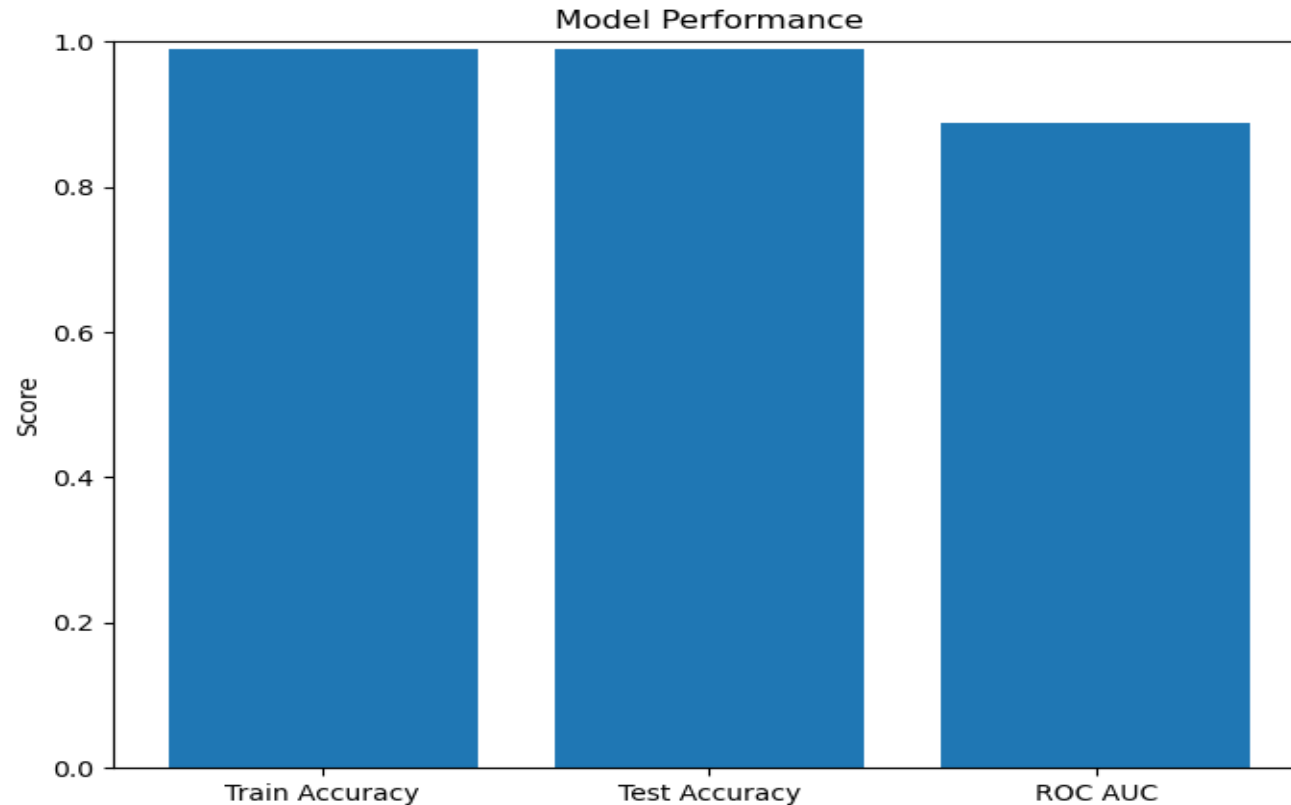
RandomClassifier has been used to train and test

Train Accuracy: 0.9907520128378939

Test Accuracy : 0.9907520128378939

ROC Curve: 0.8871933925078359

RandomClassifier Model Bar Graph



MODELLING:KNN XGBOOST

Produces the poorest model results

```
from sklearn.metrics import roc_auc_score
from sklearn.preprocessing import LabelBinarizer

# Convert the target arrays to one-hot encoded format
lb = LabelBinarizer()
y_test_one_hot = lb.fit_transform(y_test)
y_test_pred_one_hot = lb.transform(y_test_pred)

# Calculate ROC AUC score
roc_auc = roc_auc_score(y_test_one_hot, y_test_pred_one_hot, multi_class='ovr')
print(f'ROC AUC Test: {roc_auc}')
```

✓ 0.8s

Python

```
ROC AUC Test: 0.49985988434849027
```

MODELLING: KNN:GRIDSEARCH

- The accuracy score represents the proportion of correctly classified instances out of the total number of instances in the test set. In this case, the best KNN model achieved an accuracy score of approximately 0.9805, which means that it correctly predicted the class labels for approximately 98.05% of the instances in the test data.

```
# Evaluate the model
accuracy = accuracy_score(y_test, y_test_pred)
print(f"Accuracy Score (Best Model): {accuracy}")
```



Python

```
Accuracy Score (Best Model): 0.9805079821818886
```

CLASSIFICATION REPORT

```
from sklearn.metrics import classification_report

# Calculate and print the classification report
classification_rep = classification_report(y_test, y_test_pred)
print("Classification Report (KNN):\n", classification_rep)
```



Python

Classification Report (KNN):

	precision	recall	f1-score	support
0	0.97	0.92	0.94	303
1	1.00	0.21	0.35	43
2	0.85	0.86	0.86	480
3	0.87	0.83	0.85	767
4	0.89	0.77	0.82	451
5	0.96	0.98	0.97	4061
6	0.96	0.99	0.97	7033
7	0.91	0.23	0.36	127
8	0.93	0.95	0.94	2528
9	0.94	0.88	0.91	1512
10	0.97	0.54	0.70	162
11	0.96	0.81	0.88	1935
12	0.94	1.00	0.97	11726
13	0.96	0.89	0.92	2269
14	0.95	0.77	0.85	690
15	0.95	0.77	0.85	1002
16	0.90	0.57	0.70	893
17	0.96	0.96	0.96	15259
18	0.88	0.88	0.88	88585

23	0.97	0.98	0.97	17381		
24	0.98	0.97	0.98	12011		
25	0.00	0.00	0.00	9		
26	0.99	1.00	0.99	18816		
27	1.00	0.15	0.26	34		
28	0.98	0.97	0.97	4589		
29	1.00	0.21	0.35	38		
30	0.94	0.97	0.95	2228		
31	0.99	0.43	0.60	171		
32	0.94	0.80	0.87	832		
33	0.89	0.88	0.89	927		
34	0.92	0.47	0.63	160		
35	0.91	0.53	0.67	241		
36	0.99	1.00	1.00	139472		
37	0.97	0.77	0.86	1793		
38	0.96	0.86	0.91	2099		
39	0.97	0.98	0.98	5685		
accuracy			0.98	361430		
macro avg			0.93	0.76	0.81	361430
weighted avg			0.98	0.98	0.98	361430

CONCLUSION



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- Based on the different models conducted and the analysis of their results, one of the primary causes for car accidents in Chicago could be identified as the "Driver's Behavior."- Several factors related to driver behavior, such as speeding, reckless driving, distracted driving, and impaired driving, consistently appeared as significant features in the models and had a strong impact on crash outcomes.
- These findings suggest that driver-related factors play a crucial role in contributing to car accidents in Chicago..



RECOMMENDATION

- - By analyzing the models' feature importance and coefficients, it can be inferred that addressing driver behavior through targeted interventions, awareness campaigns, and stricter enforcement of traffic regulations could potentially help mitigate the occurrence of car accidents in Chicago.
- - However, it's important to note that the primary cause of accidents can vary based on various factors such as the dataset used, the modeling techniques employed, and the specific context of the analysis.
- - It is recommended to consider a comprehensive approach that takes into account multiple factors, including road infrastructure, weather conditions, and other external influences, to gain a holistic understanding of the primary causes of car accidents in Chicago.

