# **Credit Card Fraud**

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#### **Topic Overview**

- According to Data Breach Index, more than 5 million records are stolen on a daily basis
- Fraud-> common for both Card-Present and Card-not present type of payments

- Credit Card Fraud Detection
  - Identify & reject fraudulent purchase attempts
  - Significance prevent fraudulent purchase attempt charges

Matplotlib & ML models used to determine trends in fraudulent purchases



#### **Data Overview**

Kaggle Credit Card Fraud Detection dataset:
 <a href="https://www.kaggle.com/datasets/dhanushnarayananr/credit-card-fraud">https://www.kaggle.com/datasets/dhanushnarayananr/credit-card-fraud</a>

- 1,000,000 entries
- 8 features

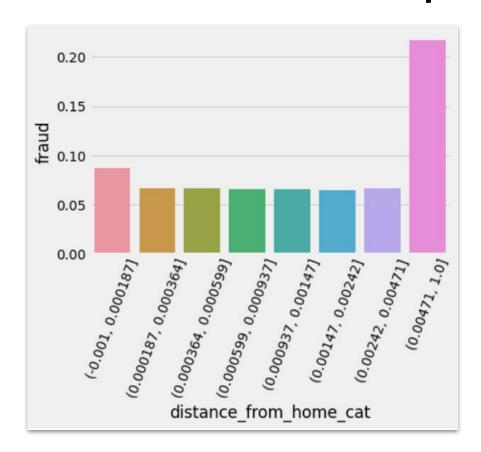
distance\_from\_home distance\_from\_last\_transaction ratio\_to\_median\_purchase\_price repeat\_retailer used\_chip used\_pin\_number online\_order fraud

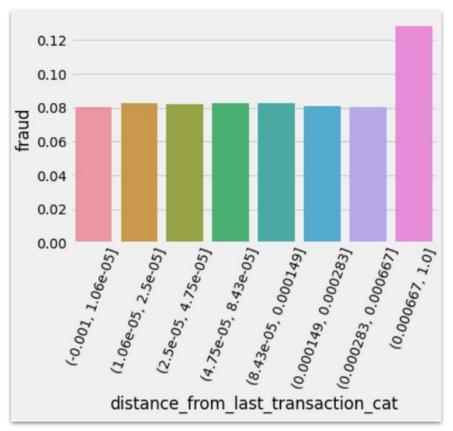
Target variable: fraud

### **Pandas**

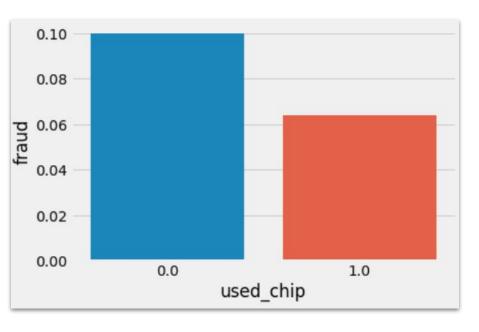
	distance_from_home	$distance\_from\_last\_transaction$	ratio_to_median_purchase_price	repeat_retailer	used_chip	used_pin_number	online_order	fraud
0	0.005443	0.000026	0.007250	1.0	1.0	0.0	0.0	0.0
1	0.001018	0.000015	0.004816	1.0	0.0	0.0	0.0	0.0
2	0.000478	0.000068	0.001581	1.0	0.0	0.0	1.0	0.0
3	0.000211	0.000473	0.001338	1.0	1.0	0.0	1.0	0.0
4	0.004156	0.000048	0.008284	1.0	1.0	0.0	1.0	0.0

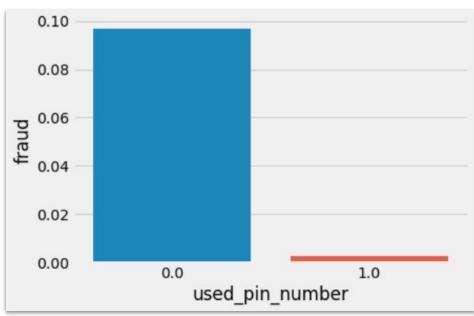
### **Dataset Overview: Matplotlib**





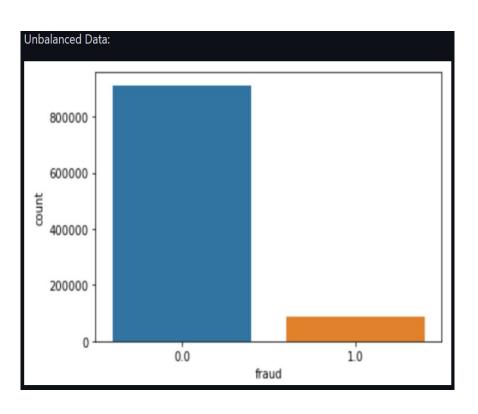
### Continued...

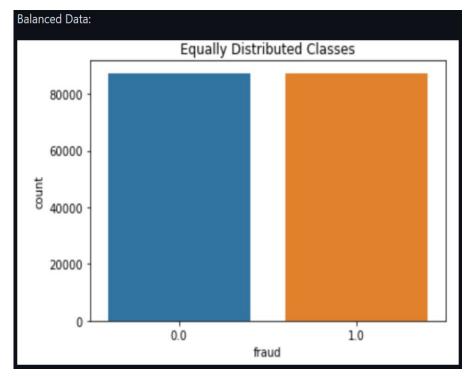




## **Balancing Data**

**SMOTE** used





### **Machine Learning Model**

#### Oversampling Accuracy & Precision

- Model Prediction
  - Logistic Regression
  - Random Forest Classifier
  - Decision Tree

SMOTE -> oversampling

High accuracy & precision results

```
Logistic Regression:
     print("Accuracy - " + str(accuracy score(y test, prediction)))
     print("Recall - " + str(recall score(y test, prediction)))
     print("precision - " + str(precision score(y test,prediction)))
 Accuracy - 0.9416435275992276
 Recall - 0.9498059519742087
 precision - 0.9346175065947139
Random Forest Classifier:
    print("Accuracy - " + str(accuracy score(y1 test,prediction1)))
    print("Recall - " + str(recall score(y1 test,prediction1)))
    print("precision - " + str(precision score(y1 test,prediction1)))
Accuracy - 0.9999934253636322
Recall - 0.9999868590501721
precision - 1.0
Decision Trees:
    print("Accuracy - " + str(accuracy_score(y2 test,prediction2)))
    print("Recall - " + str(recall score(y2 test,prediction2)))
    print("precision - " + str(precision score(y2 test,prediction2)))
Accuracy - 0.9999726056918067
Recall - 0.9999744273554232
precision - 0.9999707742272523
```

### **Machine Learning Model**

#### **Undersampling Accuracy & Precision**

- Model Prediction
  - Logistic Regression
  - Random Forest Classifier
  - Decision Tree

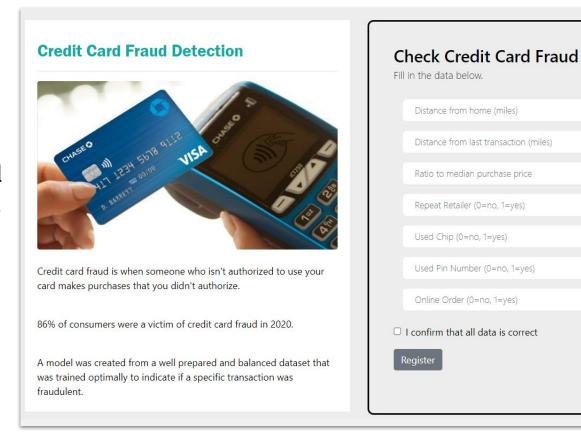
High accuracy & precision results

```
Logistic Regression:
     print("Accuracy - " + str(accuracy score(y test,prediction)))
     print("Recall - " + str(recall score(y test, prediction)))
     print("precision - " + str(precision score(y test,prediction)))
Accuracy - 0.9414214452427806
Recall - 0.9495286903999268
precision - 0.9343930115273775
Random Forest Classifier:
    print("Accuracy - " + str(accuracy score(y1 test,prediction1)))
    print("Recall - " + str(recall score(y1 test,prediction1)))
    print("precision - " + str(precision score(y1 test,prediction1)))
Accuracy - 1.0
Recall - 1.0
precision - 1.0
Decision Trees:
   print("Accuracy - " + str(accuracy_score(y2_test,prediction2)))
    print("Recall - " + str(recall_score(y2_test,prediction2)))
    print("precision - " + str(precision score(y2 test,prediction2)))
Accuracy - 0.9999427939437855
Recall - 1.0
precision - 0.9998856402241452
```

### Webpage Development

HTML & CSS

 Form uses ML model to reveal if purchase was fraudulent



#### Flask & Heroku Demo

 Undersampled Random Forest Classifier model chosen

 Flask app takes html form data and determines if the purchase was fraudulent



### **Questions?**

