

**Team 3**

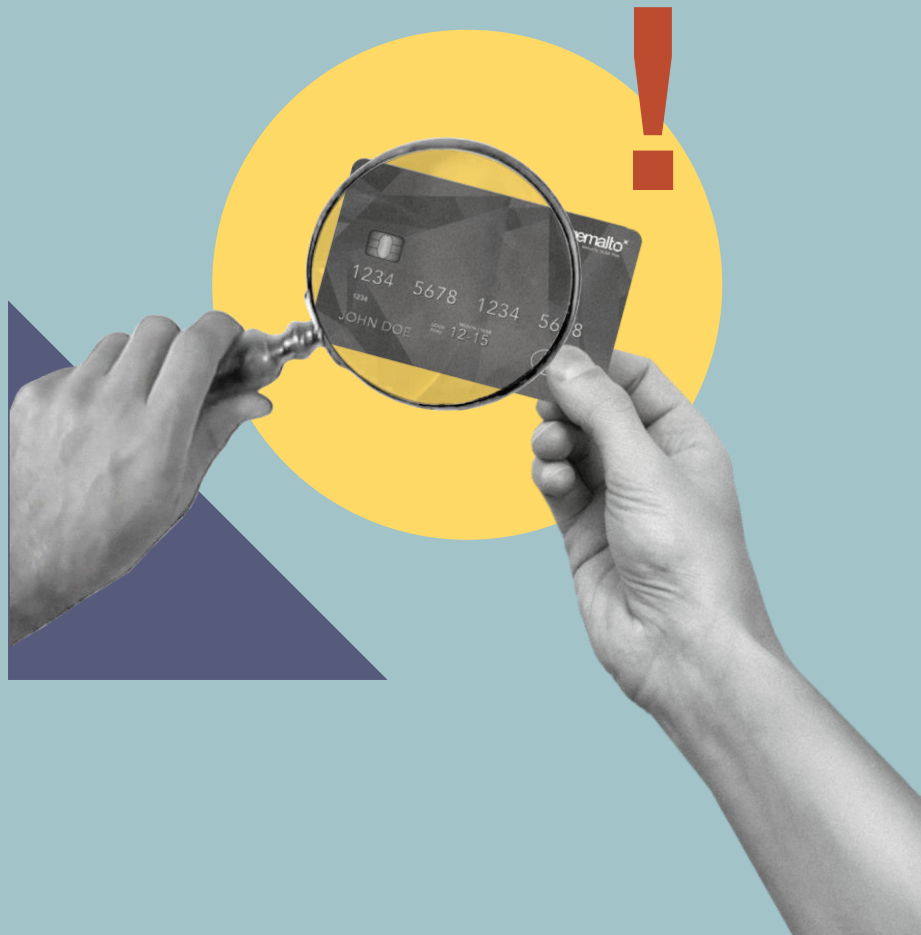
# Detecting Credit Card Fraud Using Machine Learning

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## What is credit card fraud?

According to the FBI, credit card fraud *"... is the unauthorized use of a credit or debit card, or similar payment tool to fraudulently obtain money or property."*

## What are the effects?

Aside from the loss of the card holder, fraud can cause companies to suffer loss of brand value which impacts their growth plans, leading to erosion in shareholder value.



# Features Used



## Personal

Gender  
Age



## Location

City  
City population  
Merchant distance



## Transaction

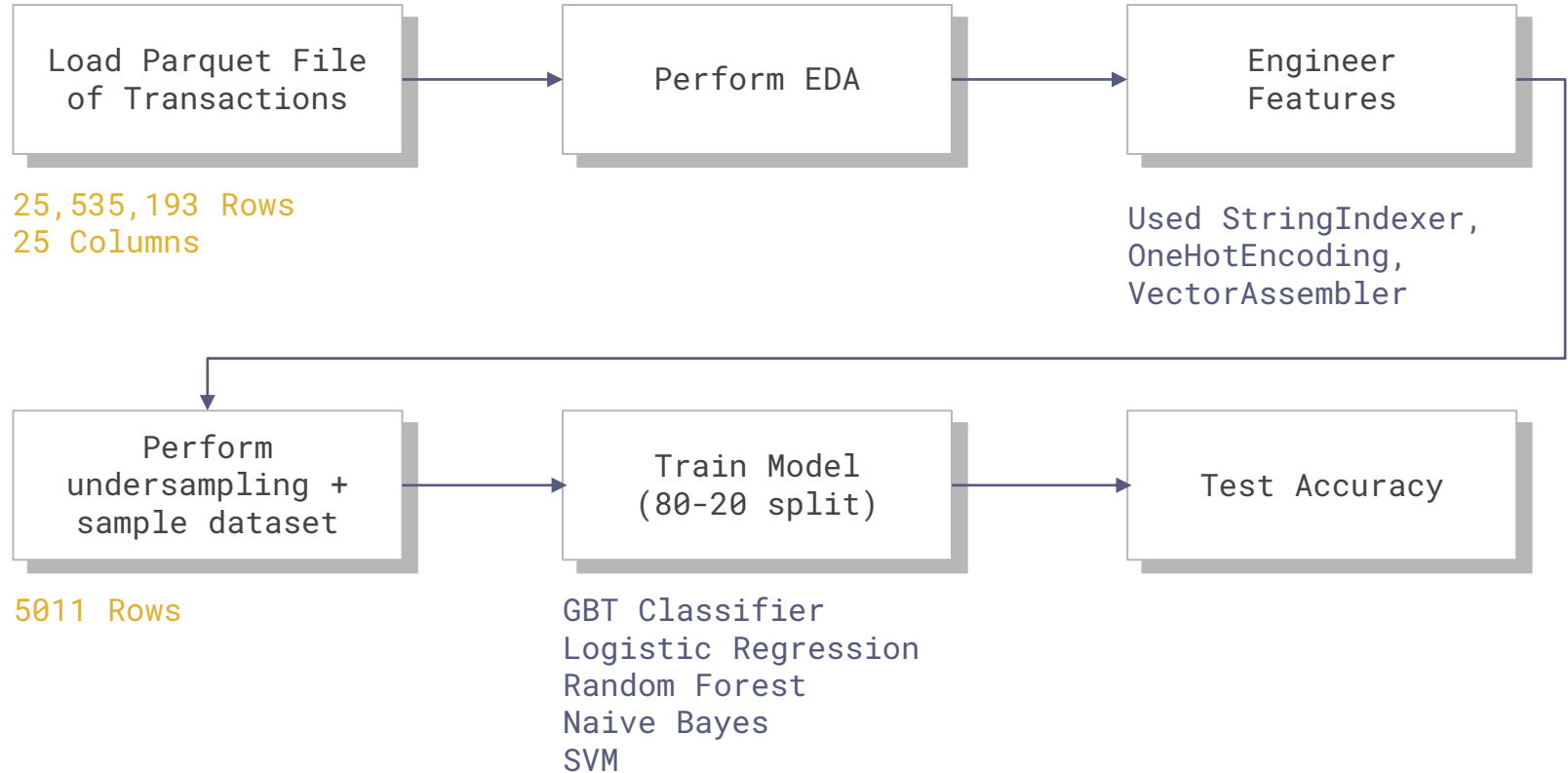
Category  
Type  
Amount



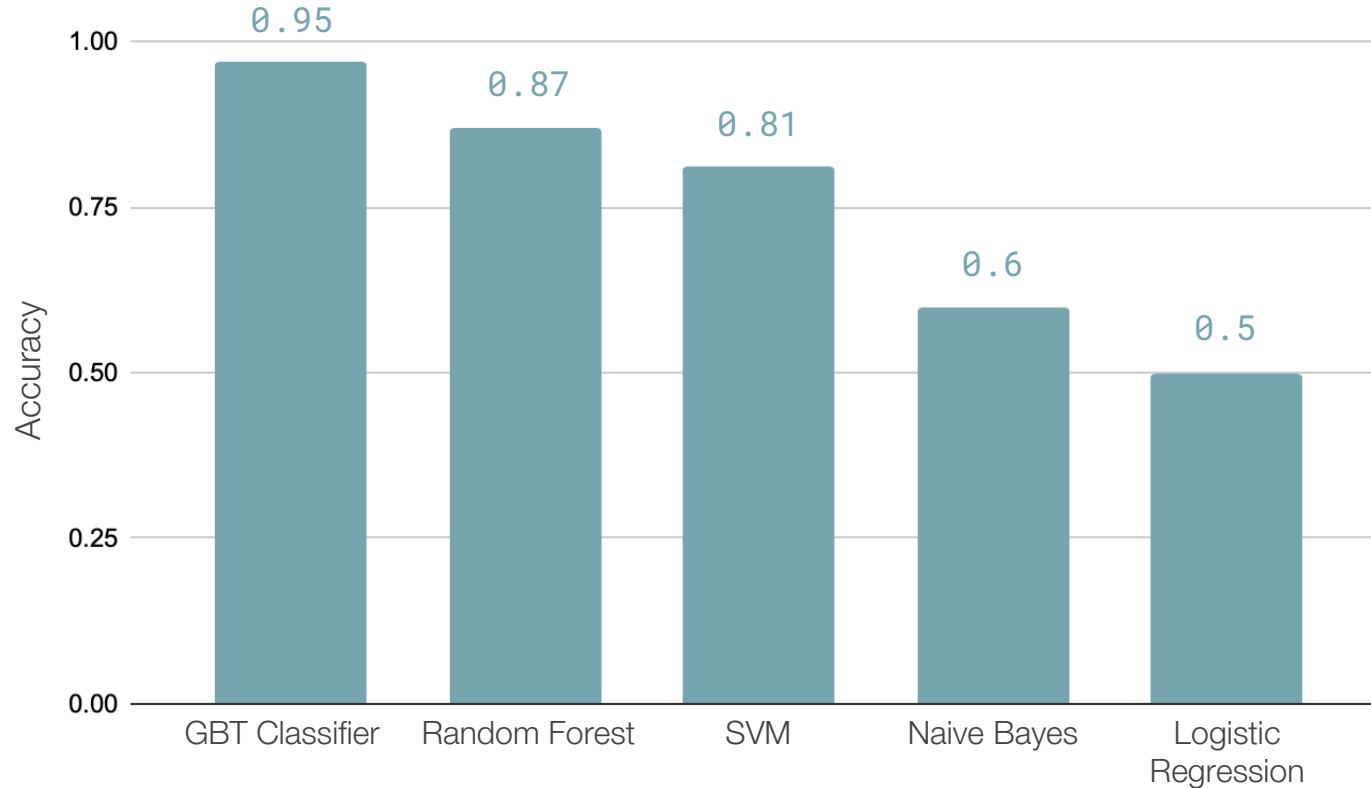
## Time

Year  
Month  
Day of week  
Hour

# Methodology



## How did the models perform?



Accuracies for 5 runs of the **GBT Classifier** model using different samples:

1st run 0.97360

2nd run 0.957248

3rd run 0.94820

4th run 0.961662

5th run 0.941618

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avg **0.956466**

# Taking a closer look at the GBT Classifier model

## Confusion Matrix

| Label     | Prediction | Count |
|-----------|------------|-------|
| Fraud     | Fraud      | 518   |
| Fraud     | Not fraud  | 9     |
| Not fraud | Fraud      | 18    |
| Not fraud | Not Fraud  | 486   |

## Classification Report

|               | Precision   | Recall      | F1-score | Support |
|---------------|-------------|-------------|----------|---------|
| 0 (not fraud) | <b>0.98</b> | 0.96        | 0.97     | 504     |
| 1 (fraud)     | 0.97        | <b>0.98</b> | 0.97     | 527     |
| accuracy      |             |             | 0.97     | 1031    |
| macro avg     | 0.97        | 0.97        | 0.97     | 1031    |
| weighted avg  | 0.97        | 0.97        | 0.97     | 1031    |



2% of genuine transactions are labeled as fraud.

(False positive)

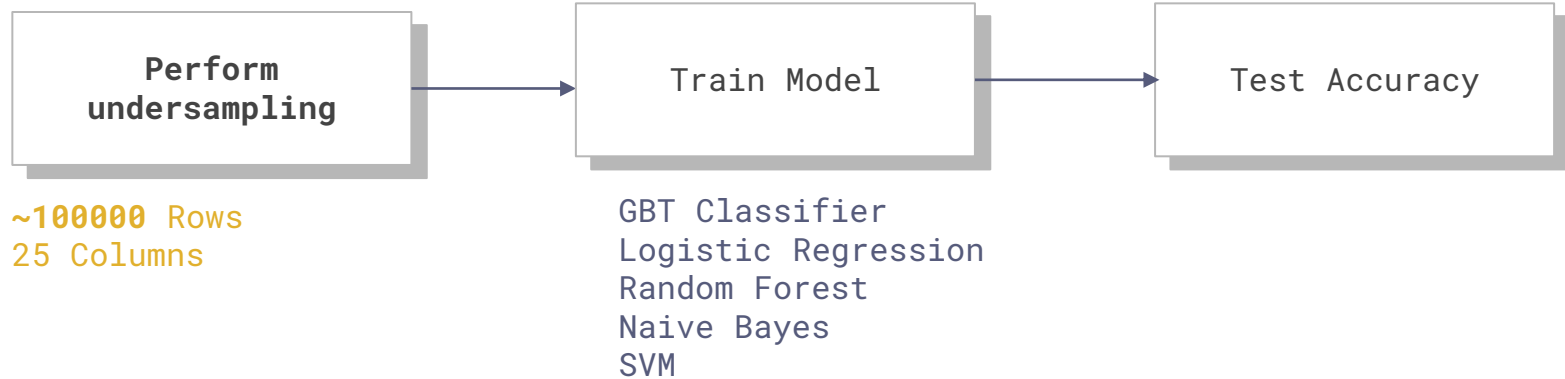


98% of fraudulent transactions are detected. 2% are not detected.

(False negative)

# Recommendations

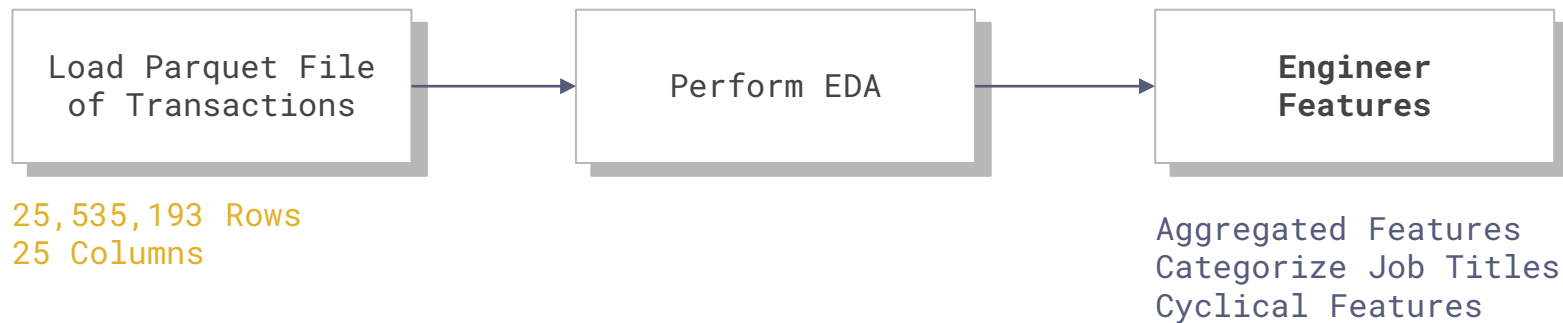
## 1 Use the entire undersampled balanced dataset with GBT Classifier



In the interest of **computational efficiency**, a smaller sample size of the dataset was used. It's suggested to use the entire dataset with undersampling in the next iteration of this project.

# Recommendations

## 2 Additional feature engineering



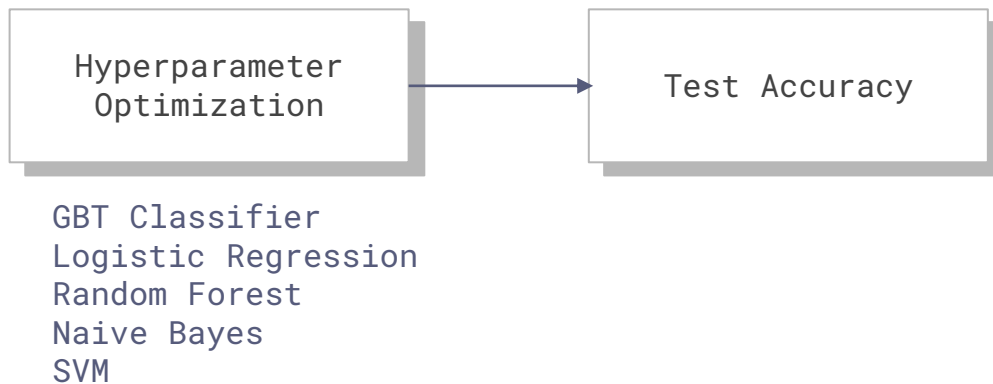
There are more features that can be extracted from the raw dataset such as:

- **aggregated features** like average amount spent per credit card number in a given timespan
- **categorizing job titles** using Natural Language Processing
- converting specific columns into **cyclical features** such as hours, day of week, month of year, etc.



# Recommendations

## 3 Model experimentation & hyperparameter optimization



- **Different models** could have been tested without the existing time constraints.
- **Hyperparameter optimization** could have also been applied to the models that displayed high accuracy.

# The End

## Thank You

