

# Credit Card Fraud Detection

Valentina Aparicio · Hayley Cromer · Chloe Griffin · Abbigale Outlaw

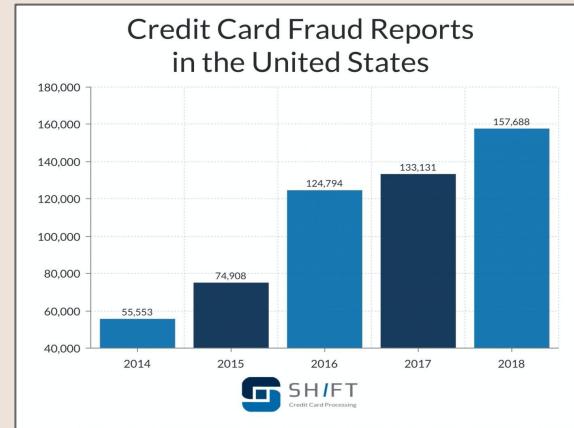
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# Credit Card Fraud Detection

## Problem:

- Number of fraudulent transactions keeps rising
- Lots of money lost daily due to credit card fraud



## Project Goal:

- Find new creative ways to detect fraud data
- Compare various detection methods

# Results

- Preferred Method:  
Random Forest
- SMOTE to handle  
imbalanced data
- Importance of specific  
variables
- Orange 3 for data  
visualization

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# Approach

1. Researched our anonymous industry partner
2. Explored several methods and techniques we could use.
3. Implemented those into several Python and Orange 3 programs.
4. Improved and updated the programs in order to get better results.
5. Compared the final outputs, scores and overall performance of the programs.
6. Made final conclusions on preferred methods and recommendations.



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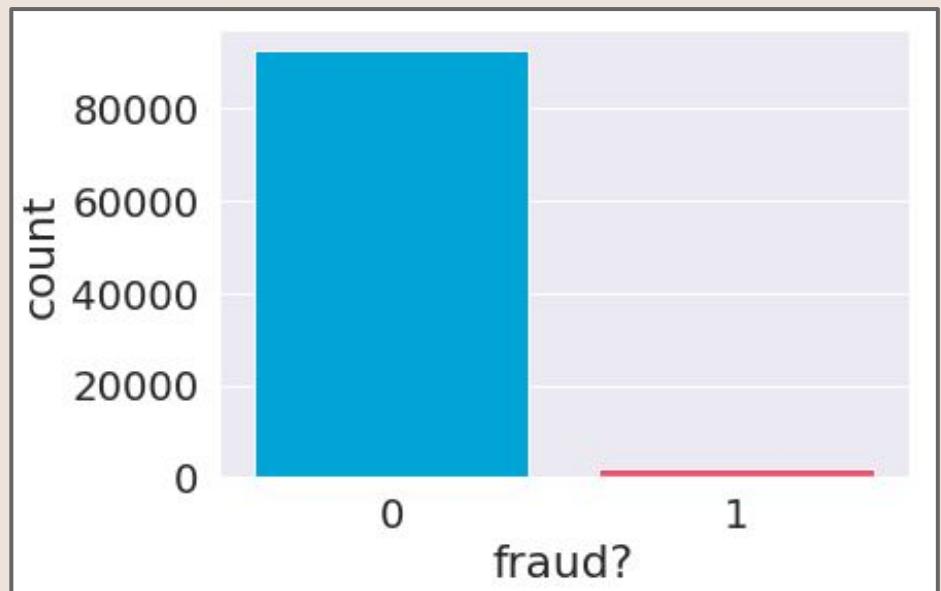
# Data Sets

## "CCDataMining1"

- 20 variables (columns)
- 65,534 samples (rows)

## "CCDataMining2"

- 20 variables (columns)
- 100,000 samples (rows)

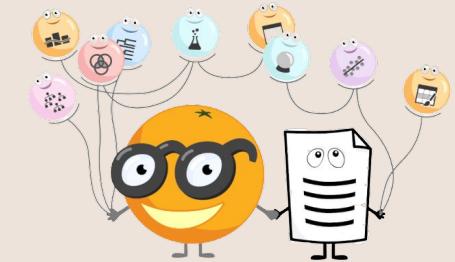


"CCDataMining1:" fraud vs not fraud

# Program Development

## Orange 3 Software

- Component based data mining framework.
- Data visualization and data analysis.



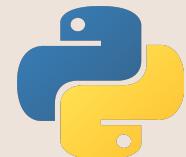
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## Jupyter Notebooks

- Uses code cells to run an IPython kernel.
- Already used within partner financial institution.



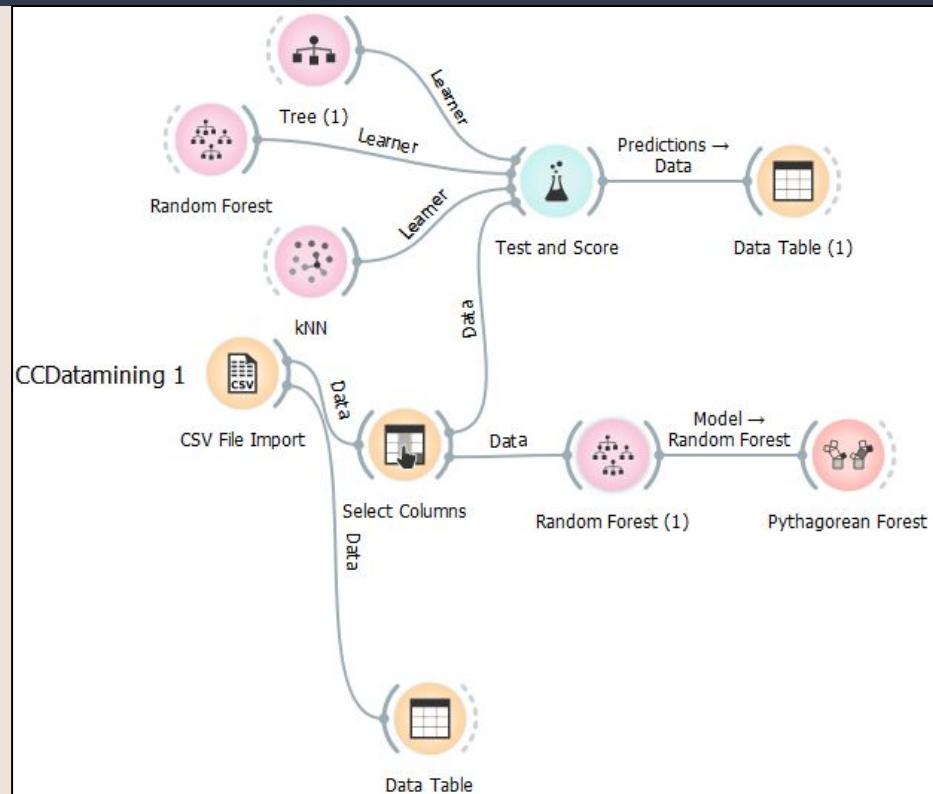
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# Method Testing - Orange 3

- 1) Inputs data from CCDatamining1
- 2) Selects target variable
- 3) Tests and Scores three methods
- 4) Gives results and illustrates predictions

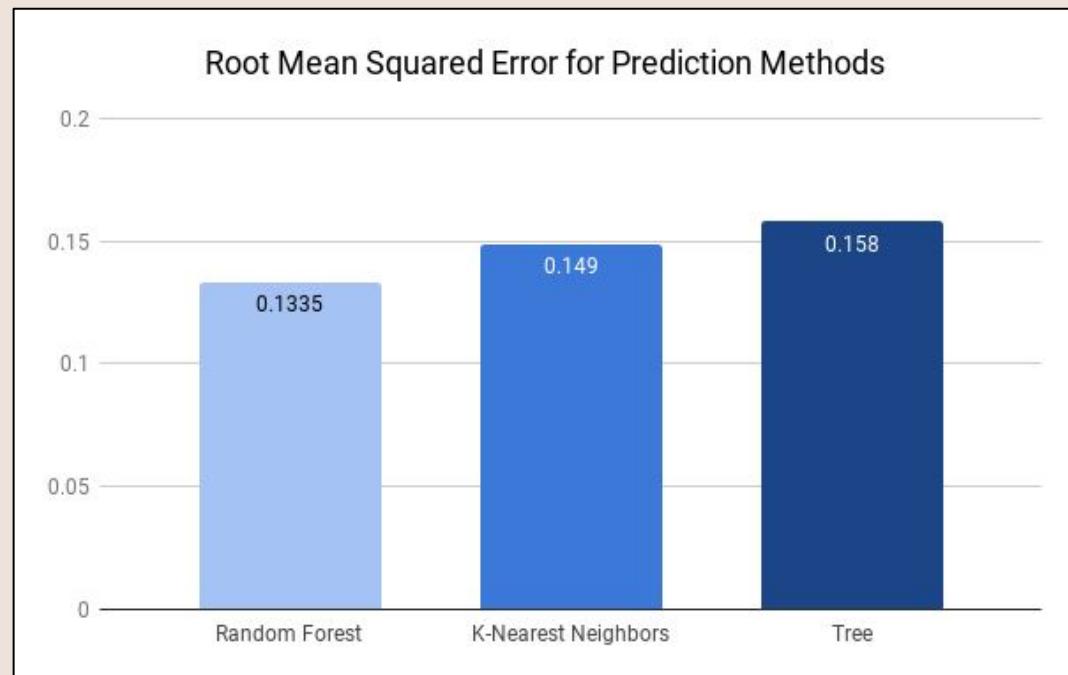


# Method Testing - Interpreting Results

Root Mean Squared Error:

$$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^n (X_{obs,i} - X_{model,i})^2}$$

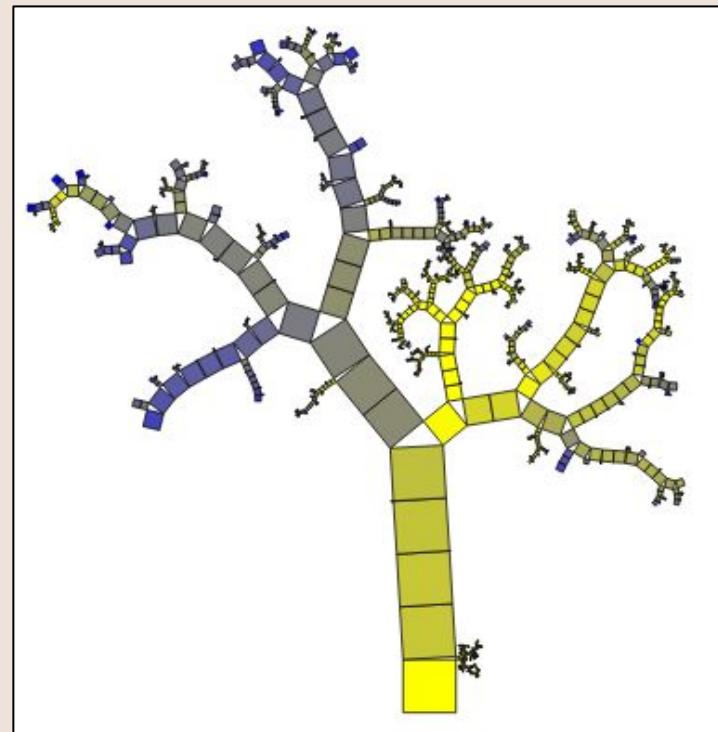
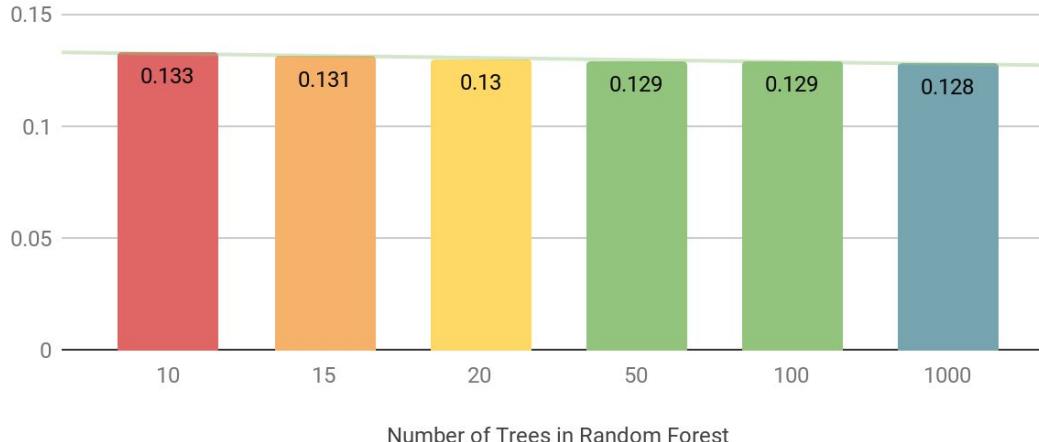
- Initial method comparison
- Least error: Random Forest



# Method Testing - Random Forest Trees

- More trees = Less error = Longer runtime
- Visual illustration of Random Forest tree

Root Mean Squared Error by Number of Trees in Random Forest

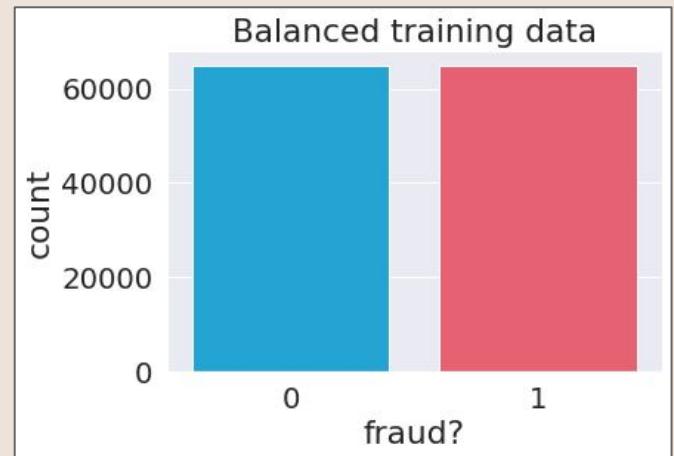


# Random Forest in Python

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- Built using pandas and Scikit-Learn RandomForestClassifier library and imbalanced-learn SMOTE on Jupyter Notebooks.
- What does the program do?
  - Provides features comparison,
  - Splits into training and testing sets,
  - Uses SMOTE to deal with the imbalanced data,
  - Trains the model, and
  - Tests predictions.



# Random Forest Results

	precision	recall	f1-score	support
0	0.98	1.00	0.99	27776
1	0.73	0.32	0.45	629
accuracy			0.98	28405
macro avg	0.86	0.66	0.72	28405
weighted avg	0.98	0.98	0.98	28405

Tests provides:

- Recall
- Precision, and
- F1 scores.

CCDatamining1 Results

	precision	recall	f1-score	support
0	0.98	1.00	0.99	29204
1	0.73	0.43	0.54	796
accuracy			0.98	30000
macro avg	0.86	0.71	0.77	30000
weighted avg	0.98	0.98	0.98	30000

CCDatamining2 Results

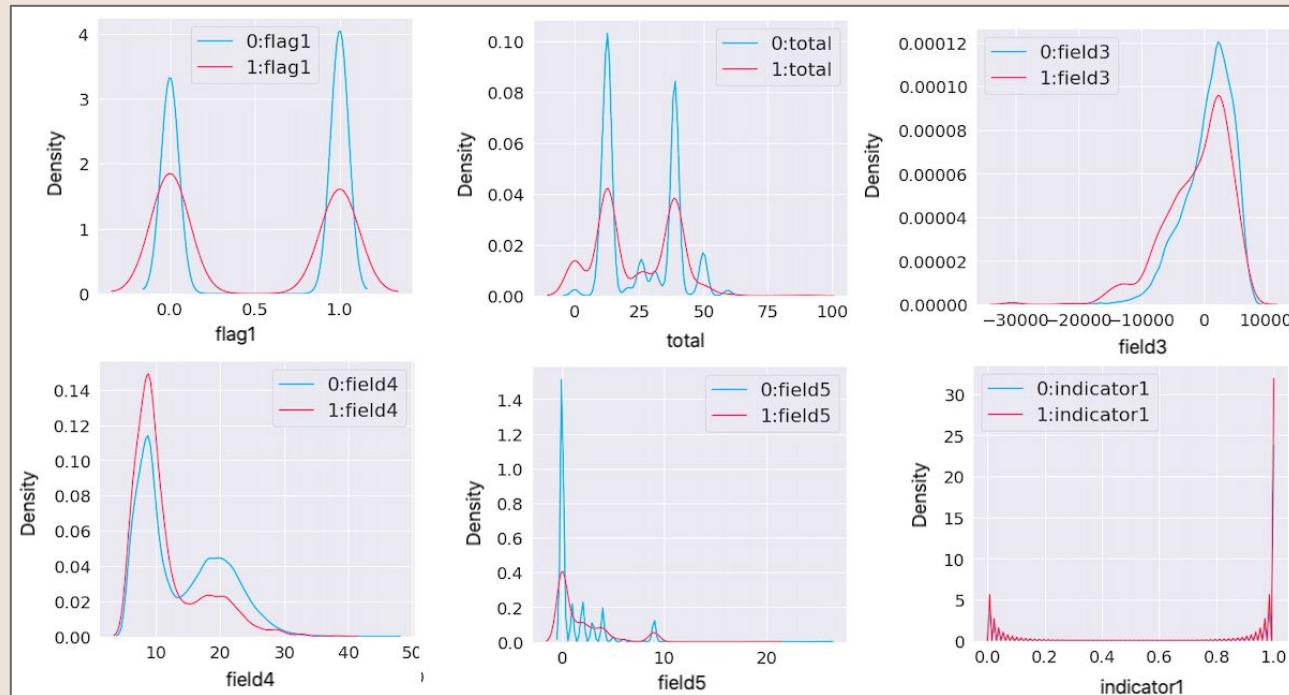
# CCDataMining1 Features Comparison

## Field 3 & Field 4

- Similar for normal and fraudulent
- Not good indicators

## Total & Field 5

- Differences in values
- May be better indicators



# K-Nearest Neighbors

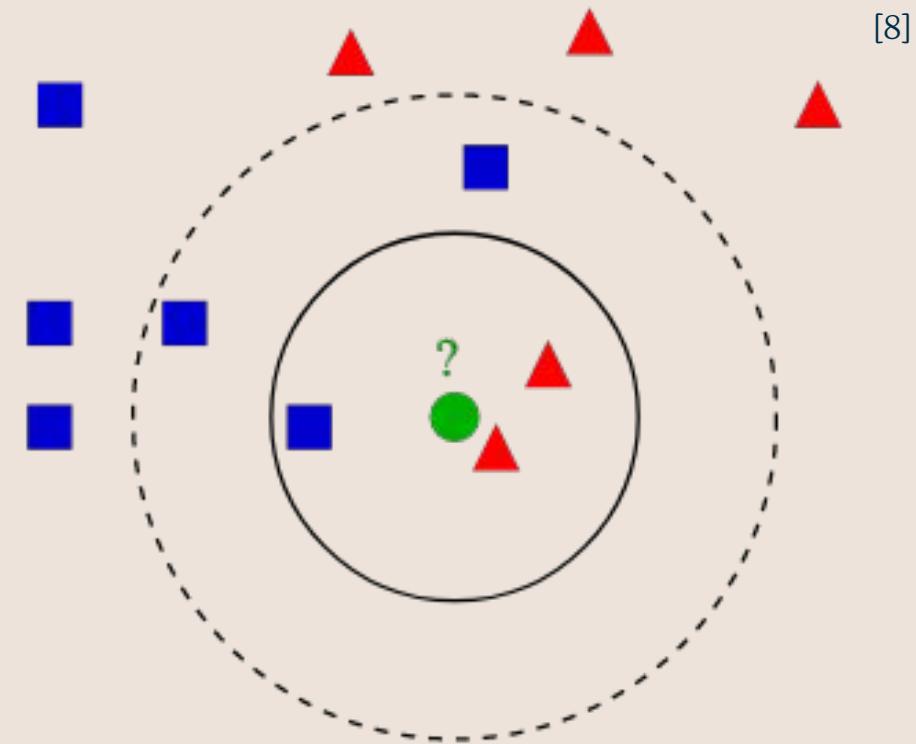
- A non-parametric method for classification and regression

## Advantages

- Strong consistency rate
- Simple to implement

## Disadvantage

- Slows as the size of the data set increases



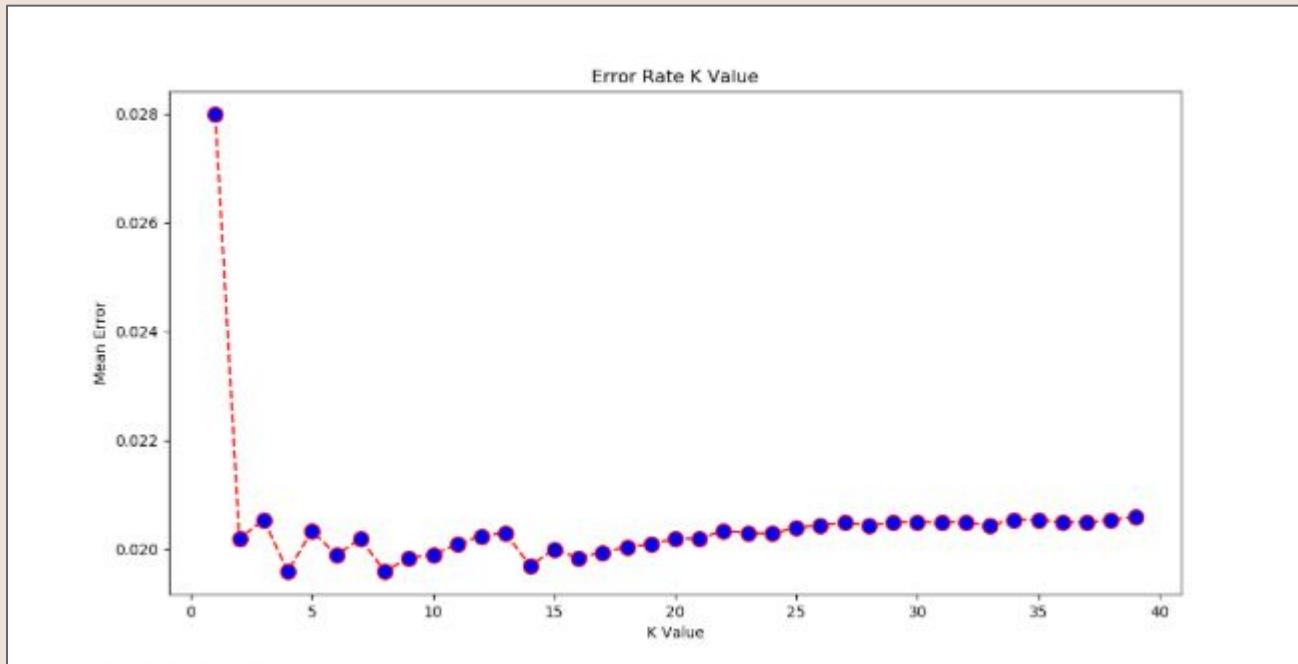
# K-Nearest Neighbors Program

- Built using Scikit-Learn KNeighborsClassifier library
- 2 versions of the program
  - Splits the data
  - Trains the model
  - Determines the “score” for non-fraud and fraud predictions
  - Prints classification report

	precision	recall	f1-score	support
0	0.98	1.00	0.99	19413
1	0.76	0.34	0.47	587
accuracy			0.98	20000
macro avg	0.87	0.67	0.73	20000
weighted avg	0.97	0.98	0.97	20000

Classification Report, CCDataMining2 (k=4)

# K-Nearest Neighbors Program



*k* error comparison with CCDataMining2

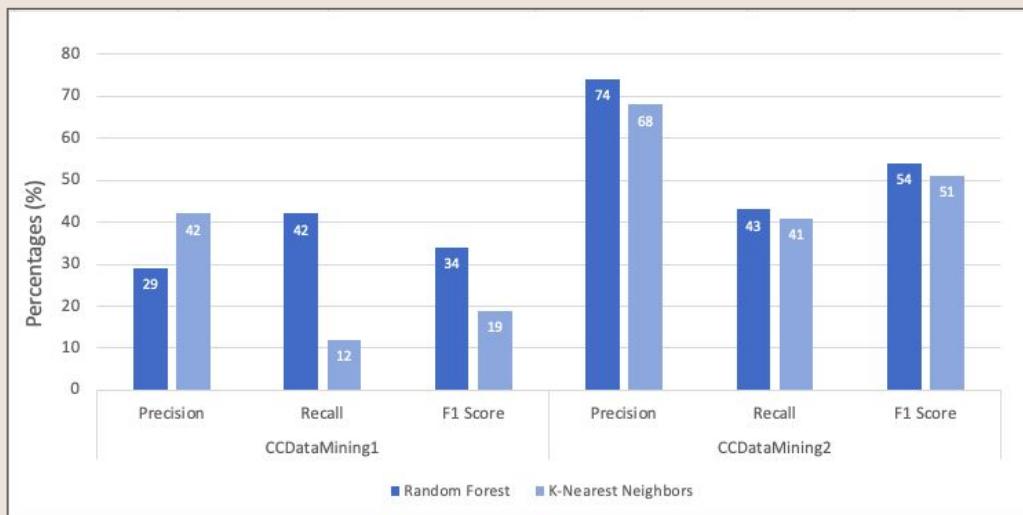
# Project Recap

## Problem

- Credit Card Fraud causes companies to lose a great deal of money everyday.
- Our goal was to compare methods and provide suggestions to improve the detection process.

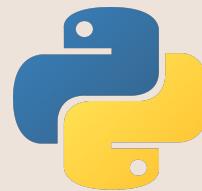
## Results

- The Random Forest method outperformed K-Nearest Neighbors on Python and Orange 3 softwares.



# Suggested Future Work

- Try other data sets within existing programs,
- Determine time and recall balance for Random Forest Programs, and
- Consider implementing Orange 3 and SMOTE programs



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# Acknowledgements

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Faculty Advisor: Jessica Sorrells

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Thank you!