



# Credit Card Fraud Detection

Real-world application of advanced  
Machine Learning Classification  
Techniques and Quantum Simulation

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# Background

Classical Classification ML Techniques:

- Feature Selection with Cross Validation [AND \_\_]
- Basic: Trees/RF/Boosting, QDA, Logistic Regression, SVM, Naive Bayes
- Advanced: Resampling Imbalanced Data, Neural Nets

Quantum-Computing Enhanced ML:

- ??
- ??



# Project Objective



To use an ensemble of machine learning techniques and methods to solve a real-world problem, credit card fraud detection. A key component of this project was the exploration of QC-enhanced ML.



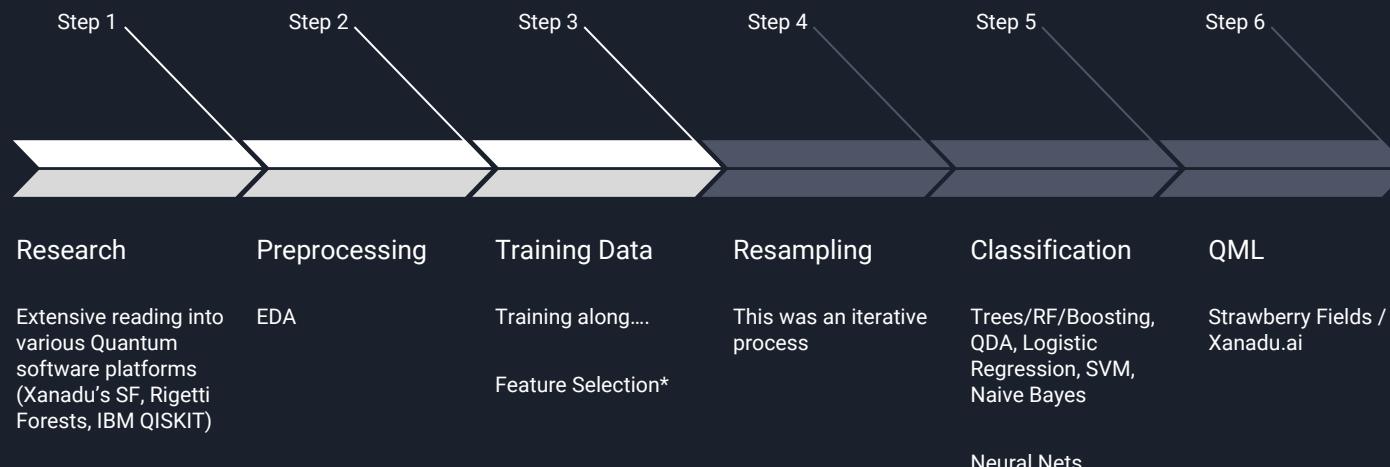
# Target Audience

Prospective employers, fellow classmates, &  
NYC Data Science Academy educators





# Project Timeline



# Research

Credit Card Data & Quantum Computing

- Kaggle competition
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# State of Quantum Technology

01

Hardware - Adiabatic (annealing), Gate Model (circuit model)

02

Software - quantum cloud platforms; optimization, machine learning & sampling problems



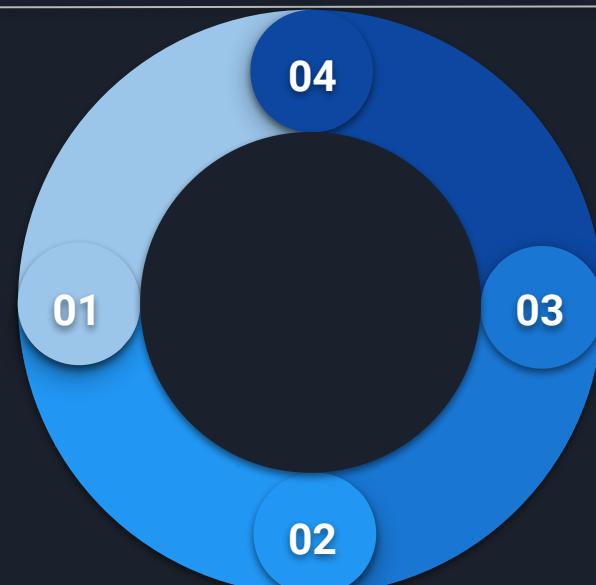
# Preprocessing & Exploratory Data Analysis

Data tidying

Data  
Manipulation

Analysis

Data  
Visualization





# Preprocessing & Exploratory Data Analysis

- 01 Data Tidying - PCA Transform already done;
- 02 Data Manipulation -
- 03 Data Visualization - normalize confusion matrix
- 04 Analysis - ROC AUC; CV vs Public LB



# Exploratory Data Analysis (EDA)



# Training the Data



# Feature Selection



# Resampling Imbalanced Data



# Modelling and Predictive Analysis

Why we explored the following  
models:

d



# **Classification Prediction Model**

## Logistic Regression

Also known as a Generalized Linear  
Model (GLM)



# **Classification Prediction Model**

## Gaussian Naive Bayes



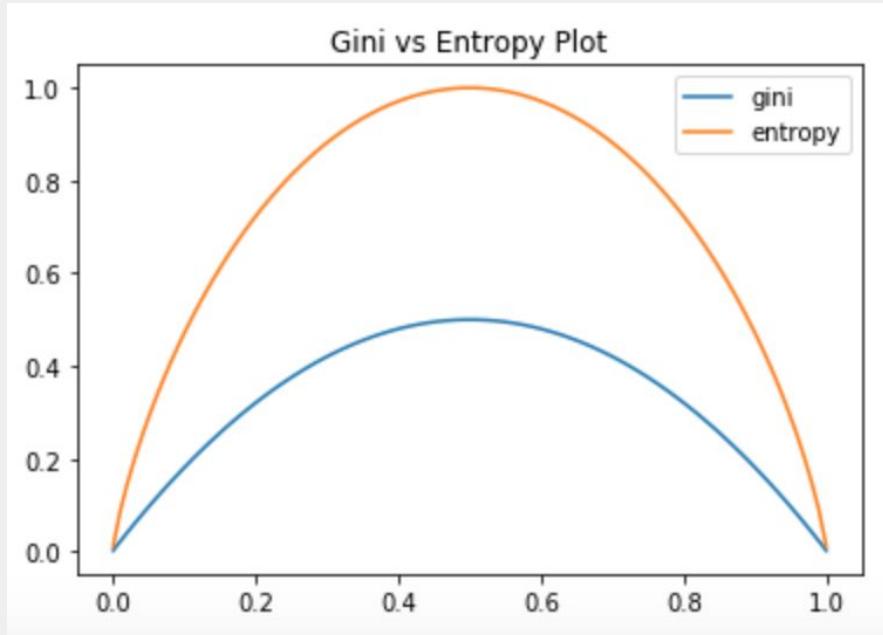
# **Classification Prediction Model**

## Support Vector Machines (SVM)

# Classification Prediction Model

## Decision Trees

Trees: Gini Coefficient and Information Gain





# Classification Prediction Model

## Decision Trees

### Trees: Bagging / RF

- While Decision Trees are not always an optimal approach, they are useful for their interpretability. By aggregating different approaches, we were able to improve our model's accuracy.
- RF approach averages over a collection of de-correlated Trees. With a big enough # of trees, we do not have to worry about overfitting. RMSE 0.1562
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# Classification Prediction Model

## Decision Trees

### Trees: Boosting

- Boosting with Cross Validation
- Boosting
  - Trees are grown sequentially; each tree grown based on information from previously grown trees
  - Tuning Parameters
  - XGBoost (high accuracy, scalability, faster)  
RMSE 0.1294



# **Classification Prediction Model**

## Bayesian Optimization



# **Classification Prediction Model**

## Neural Nets



# **Classification Prediction Model**

## Quantum Algorithms



# Ensemble Procedure

Voting vs. Average vs. Stacking

Procedures

Feature Engineering most essential

Resampling key for this kind of problem. Not enough time to play around with sequencing..



## Ensemble Methods

Stacking/averaging a diverse ensemble of ML algorithms to improve model's predictive accuracy.



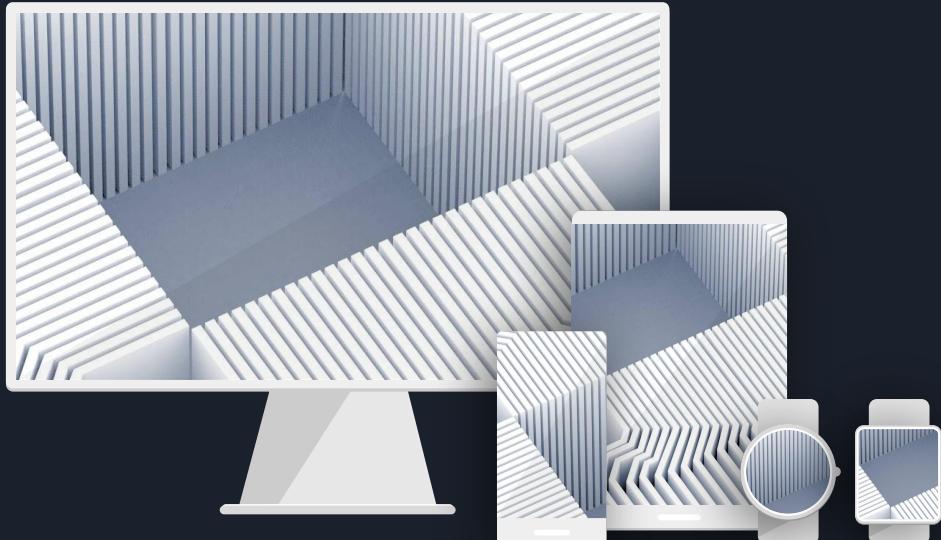
# Conclusions and Moving Forward

- 01 [Takeaways from the process, specifically: ]
- 02 [Takeaways from classical ML algorithms, generally:]
- 03 With more time, I would explore other software platforms in detail. Quantum algorithms are still in their nascent phase. Bridging the gap between quantum computing and ML will be a major challenge for this industry.
- 04 Future Applications: Quantum Machine Learning algorithms for financial services, big pharma, & logistical problem-solving are among the major applications of quantum technology so keep these developments in mind as you continue your data science journey.

Thank you!

“Anyone who is not shocked by  
quantum theory has not understood it.”

- Niels Bohr



# License

