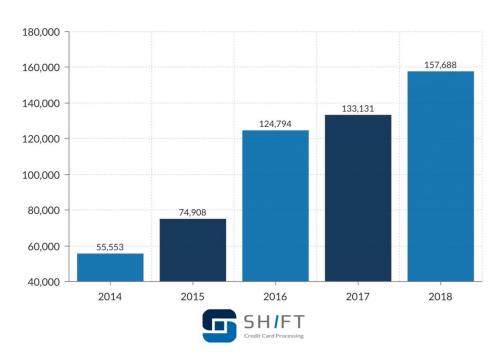
# Fraud Detection With Supervised Learning

Project 3
Benjamin Lu

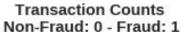
#### **Credit Card Fraud Statistics**

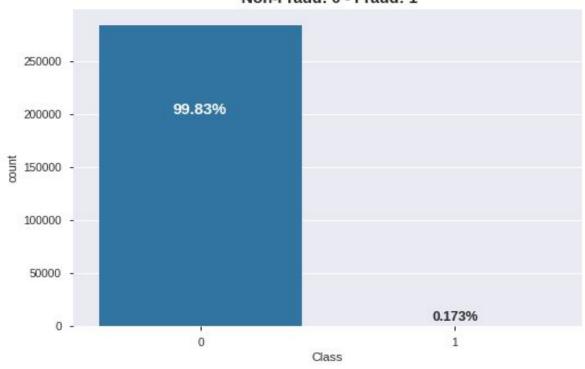
- Worldwide losses of \$24.26 Billion in 2018
- Credit card fraud increased by 18.4 percent over 2017
- USA is the global leader as the most credit card fraud prone country with 38.6 percent of reported fraud losses in 2018



Pertinent and Growing Problem that Banks want to Address

#### **Fraud Detection Dataset**





#### High Class Imbalance Problem

#### Methodology Adopted for Imbalanced Data

# Data Preparation

- stratified test splits
- Standard scaling

# Data Resampling

- RUS Random Under Sampling for 50:50 dataframe
- SMOTE Synthetic minority oversampling
- ADASYN Oversampling near decision boundaries

## Modelling & Testing

- Logistic regression, Random forests, SVC, KNN, XGBoost
- Hyper parameter tuning and cross validation

## **RUS - Training Results**

Dataset size: ~700

	Precision	Recall	F1-Score	AUC ROC
Logistic Reg	0.914	0.97	0.941	0.944
Random Forest	0.888	0.972	0.928	0.935
SVC	0.911	0.968	0.939	0.942
KNN	0.886	0.969	0.926	0.932
XGBoost	0.997	1	0.999	0.999

May Result in Significant Information Loss

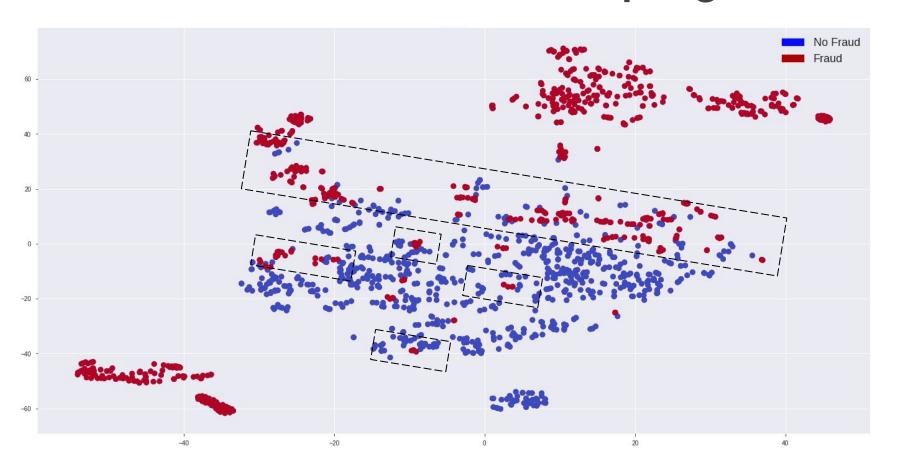
### **ADASYN + RUS - Training Results**

• Dataset size: ~1500

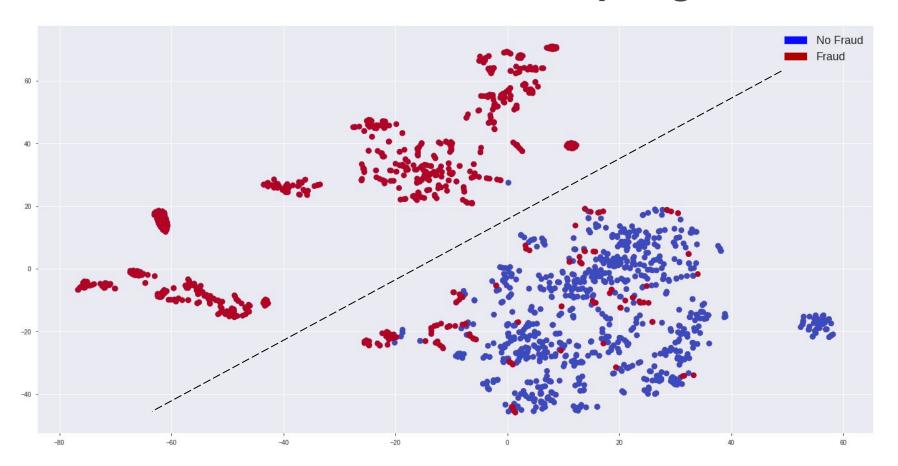
	Precision	Recall	F1-Score	AUC ROC
Logistic Reg	0.861	0.936	0.897	0.904
Random Forest	0.884	0.972	0.926	0.933
SVC	0.918	0.937	0.928	0.929
KNN	0.927	0.913	0.92	0.92
XGBoost	0.999	0.994	0.996	0.996

#### Results Significantly Poorer

## **ADASYN + Random Under Sampling**



## **SMOTE + Random Under Sampling**



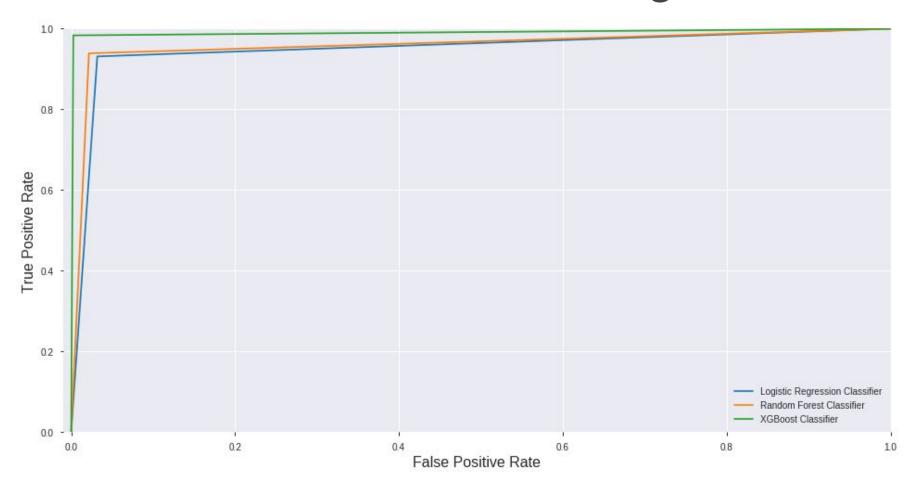
## **Training Results SMOTE + RUS**

Dataset size: ~1500

	Precision	Recall	F1-Score	AUC ROC
Logistic Reg	0.926	0.972	0.948	0.951
Random Forest	0.93	0.979	0.954	0.956
SVC	0.934	0.964	0.949	0.95
KNN	0.938	0.951	0.944	0.945
XGBoost	0.991	1	0.996	0.996

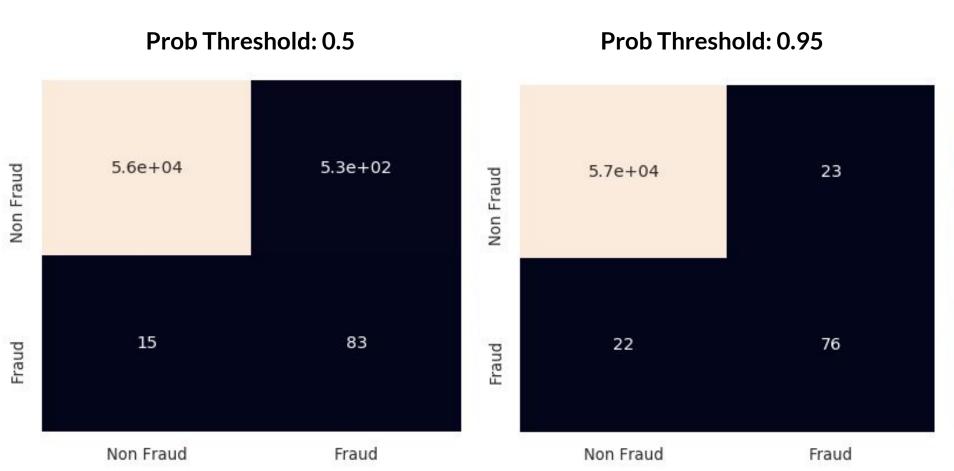
**Best Training Performance Achieved** 

## **ROC - Area Under Curve (Training)**



# Validating

#### **Confusion Matrix**



#### Recommendations

 Able to achieve balance between precision and recall by adjusting threshold

 Necessary to balance fraud detection as well as customer satisfaction

## **Further Study**

#### More Feature Engineering

- currently model is blind to value weight decisions by transaction value
- Outlier removal

#### **Comparison Over Time and Variables**

Looking at Customer Profiles and determine outliers

#### **Testing of Different Models**

- Ensembling
- Artificial Neural Networks

### **Questions?**