CREDIT CARD FRAUD DETECTION

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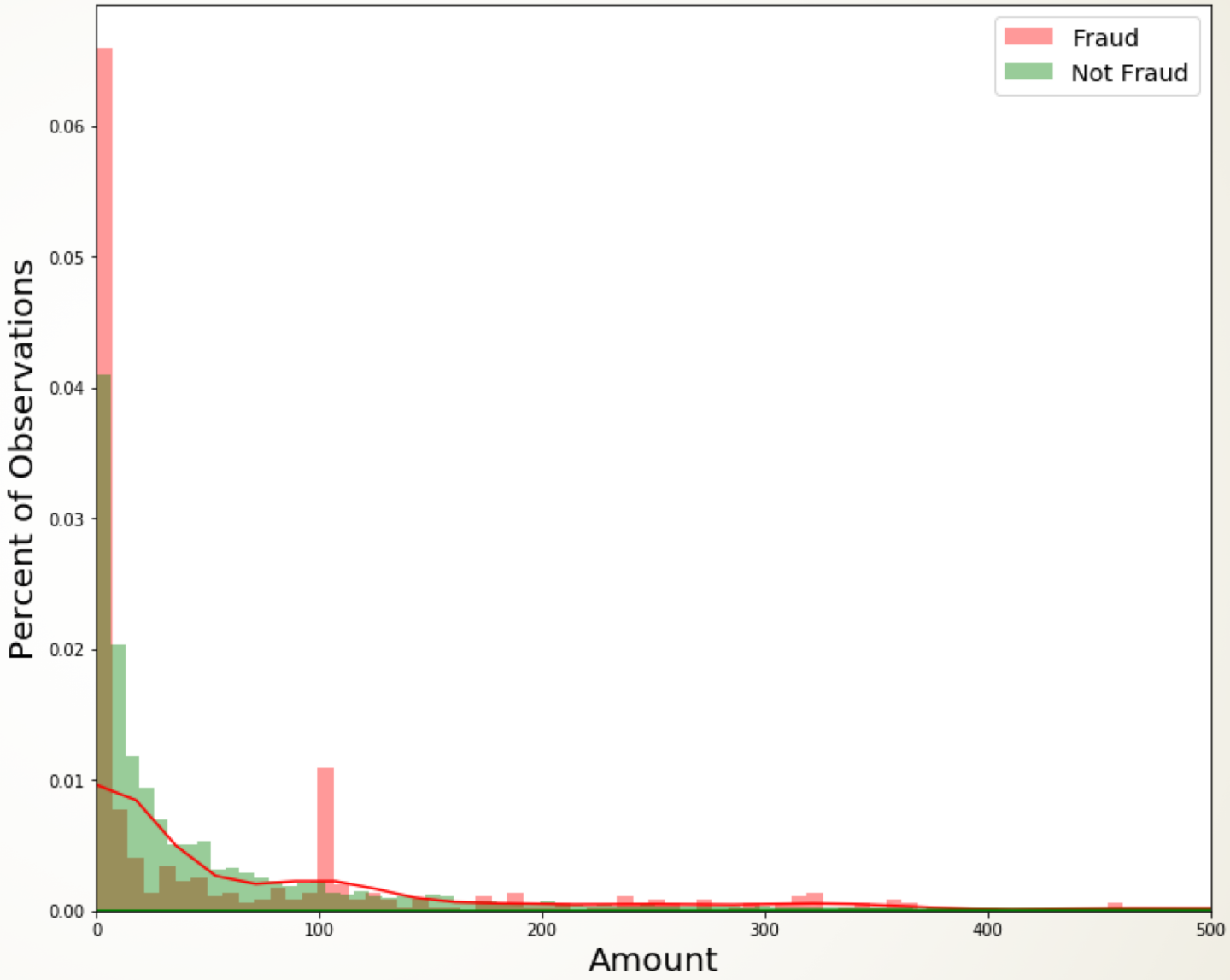
Problem Statement

- Can I build a model that predicts whether a credit card charge is fraudulent or not?
- This has a clear use cases because financial institutions are liable when fraudulent charges occur.

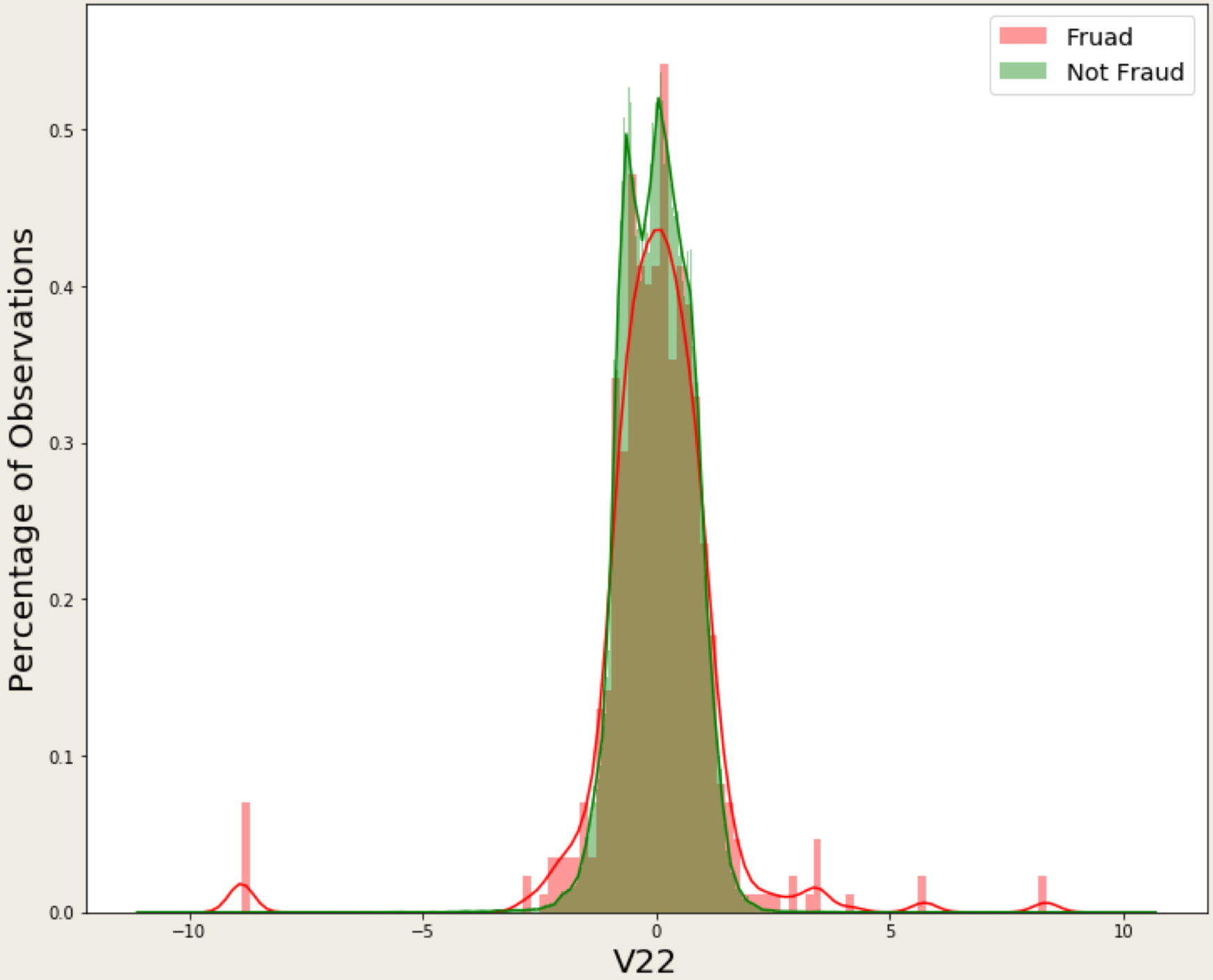
The Data

- 284,807 observations from European cardholders in 2013
- 492 Fraud | 284,315 non- fraud
- Amount and Time column
- V1 – V28 PCA’d columns.

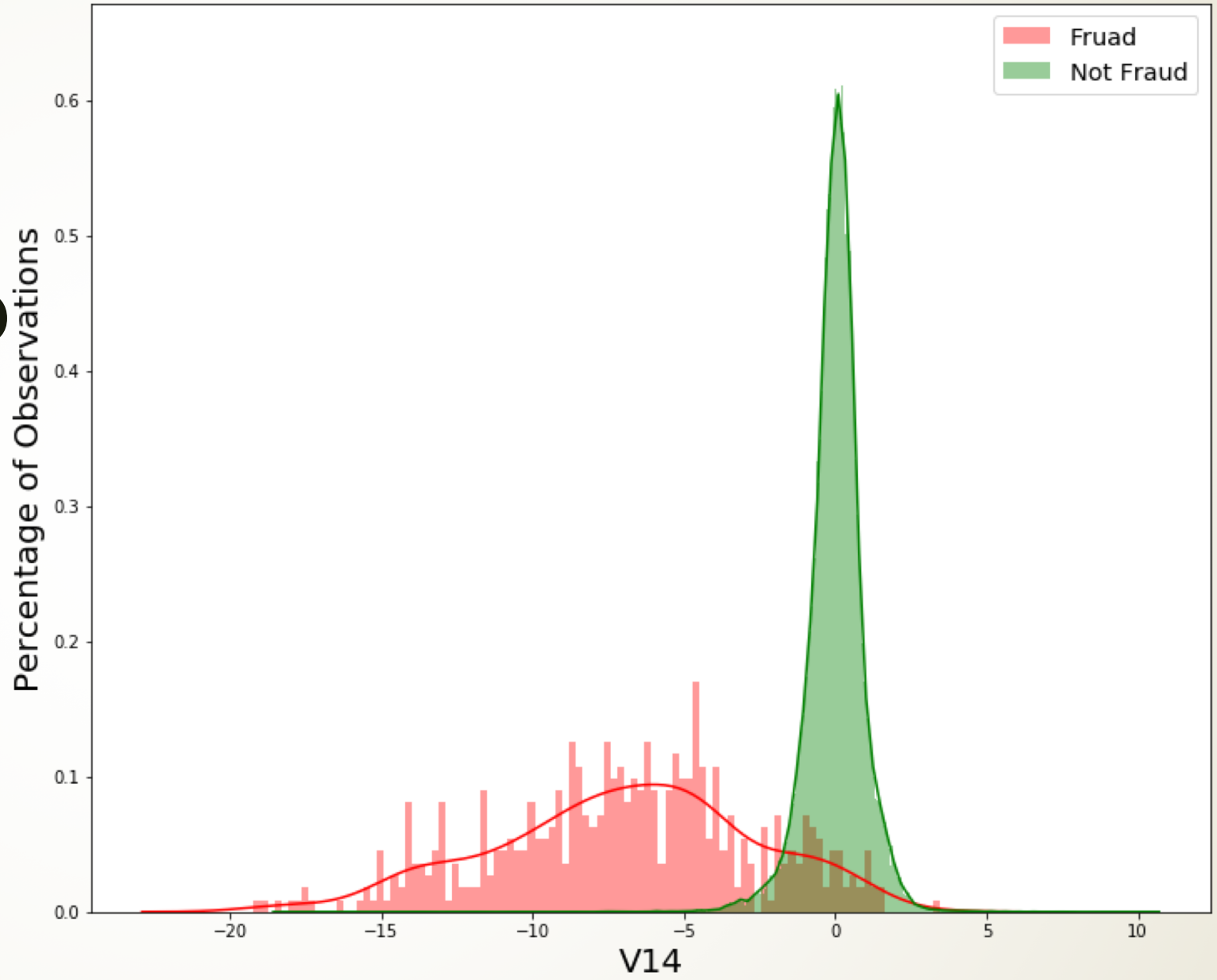
Most Charges are low dollar amount.

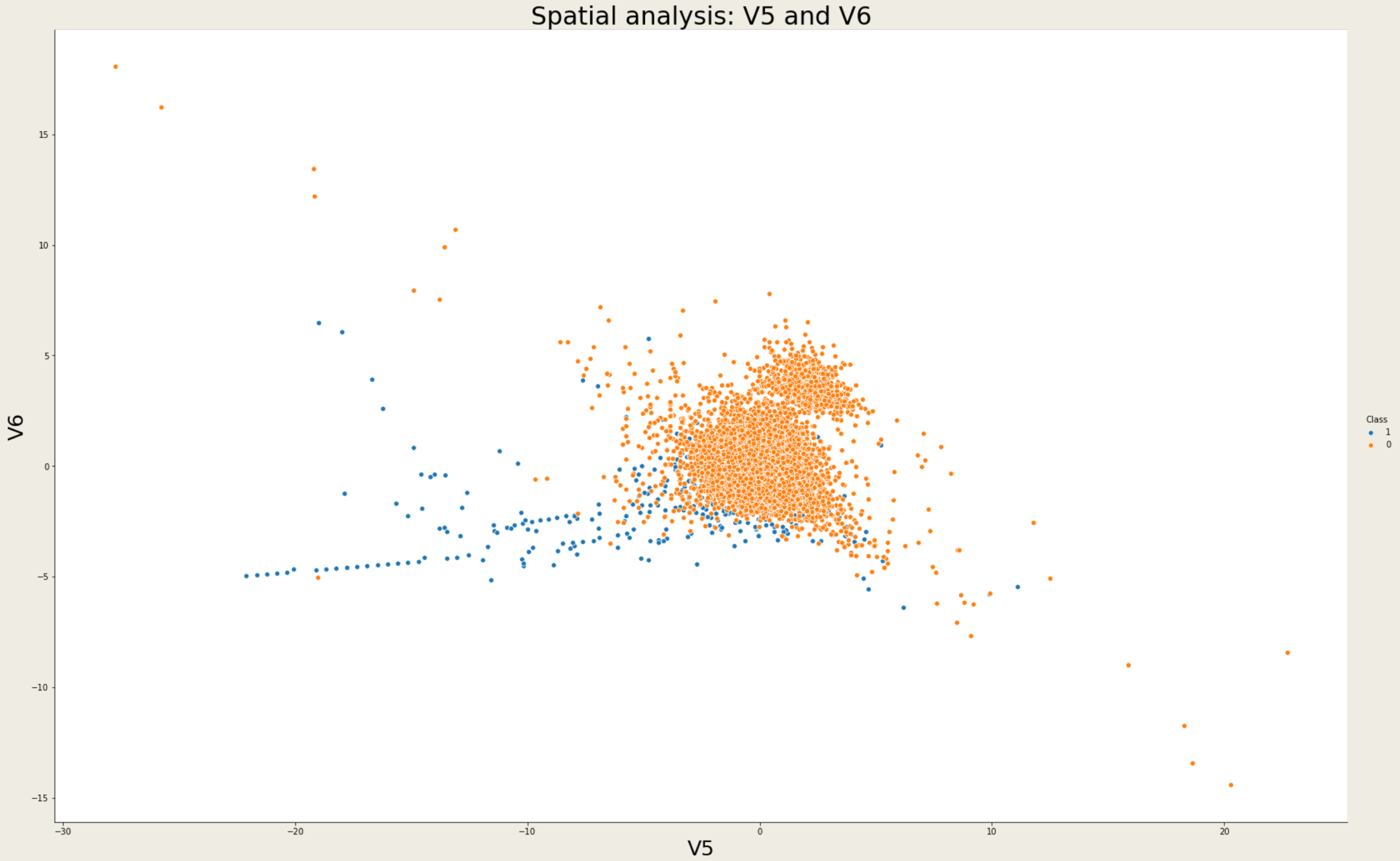


Variables with low correlation to class like V22 follow roughly the same shape.



Variables with high correlation to class like V14 follow different shapes for fraud and non-fraud




Spatial analysis: V5 and V6
V6
V5
15
10
5
0
−5
−10
−15
−30
−20
−10
0
10
20
Class
1
0

What does my model want to optimize

- False negative: Fraud occurred and was not caught
- False positive: Fraud did not occur but we acted as it did.

- False negatives are worst case scenario so we want a model that optimizes our recall score.

Training and Testing my Data

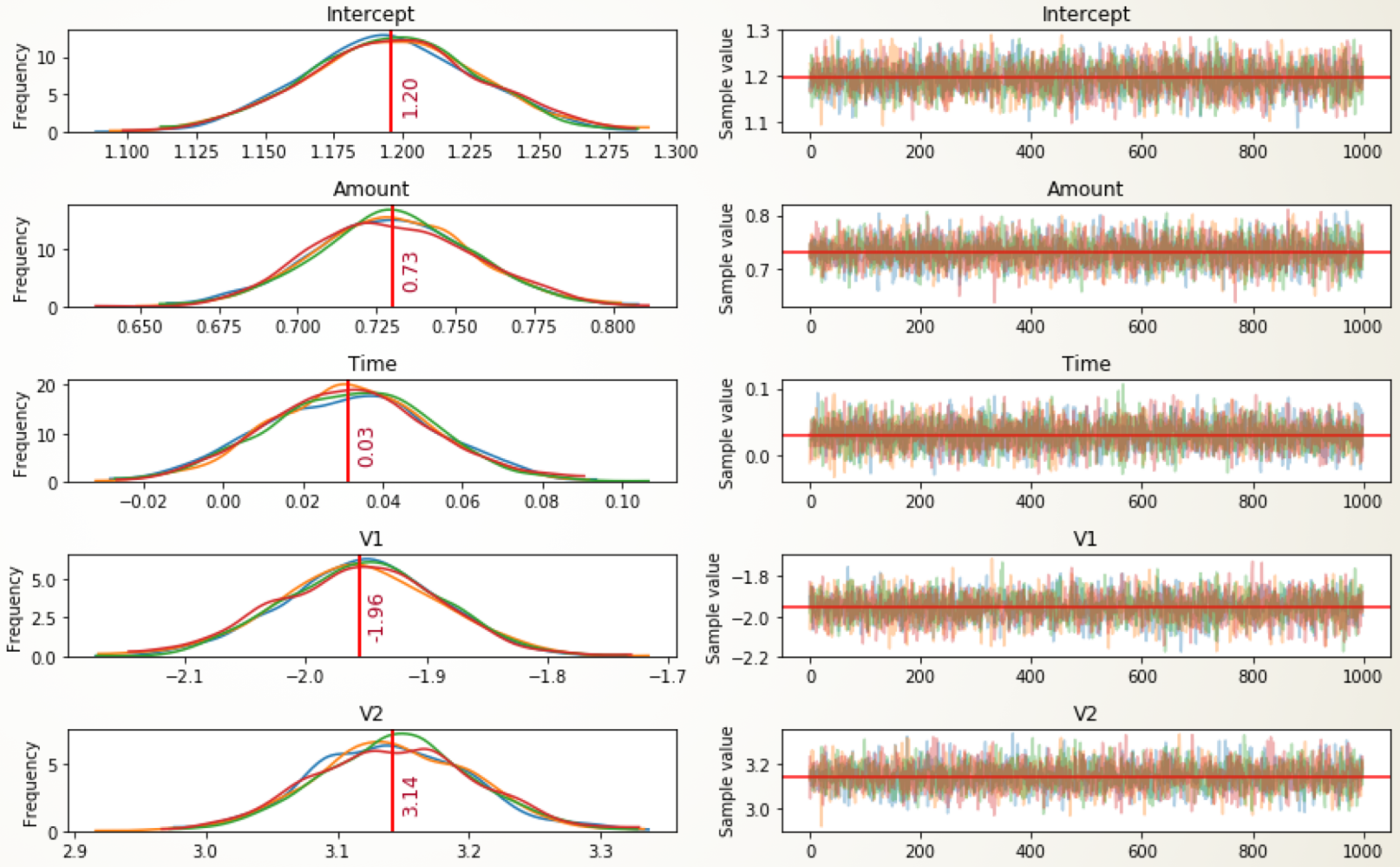
- I created a subset of my data by randomly selecting 50,000 rows and training my data and then testing my models on the entire dataset.
- I also used SMOTE to underrepresent the majority class and overrepresent the minority class.
- Scaled my data.

Modeling

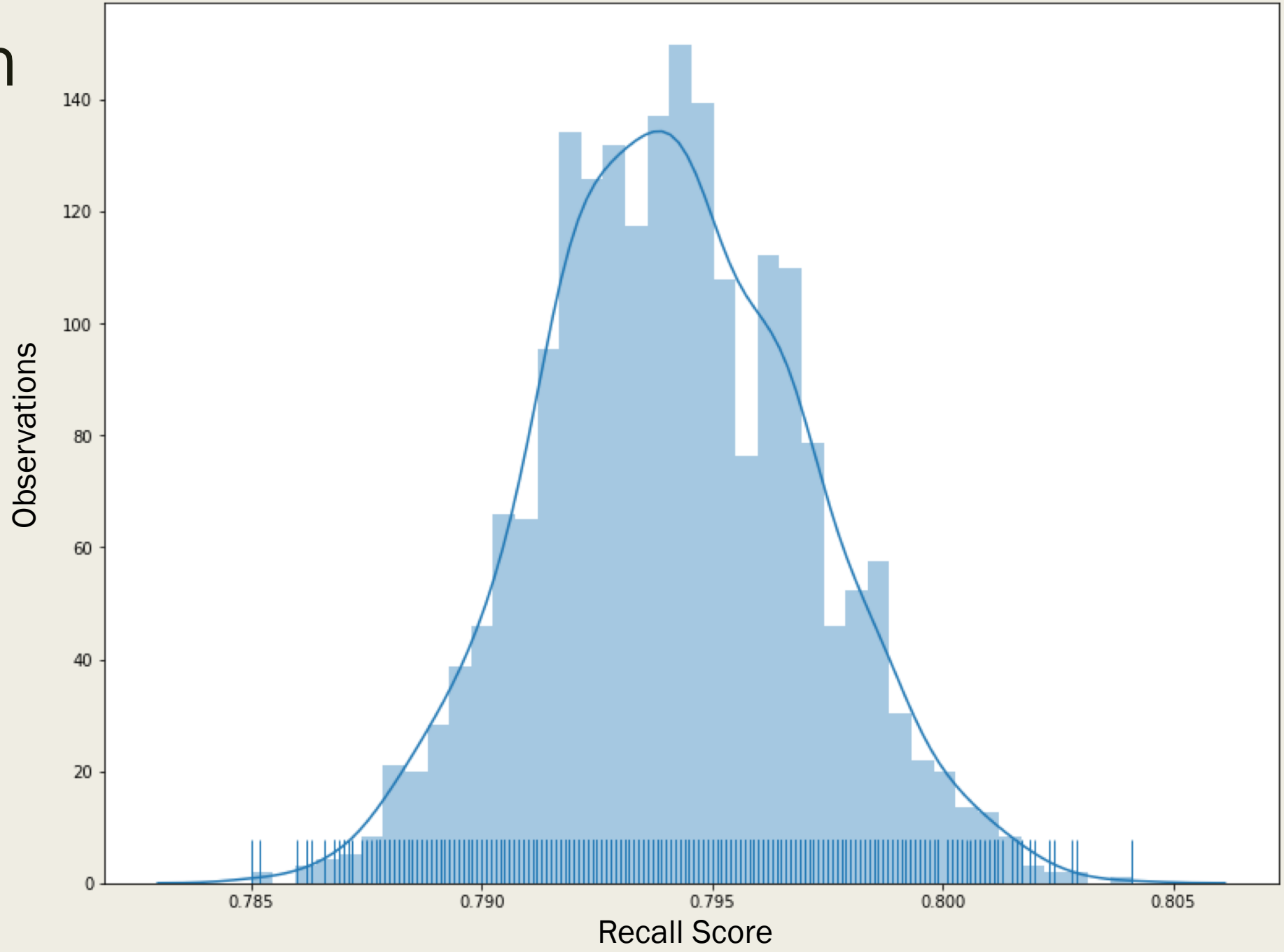
- Logistic Regression
- Random Forest
- Gradient Boosting
- XG Boost
- K Nearest Neighbors
- Support Vector Machine
- K-Means

Bayesian Logistic Regression

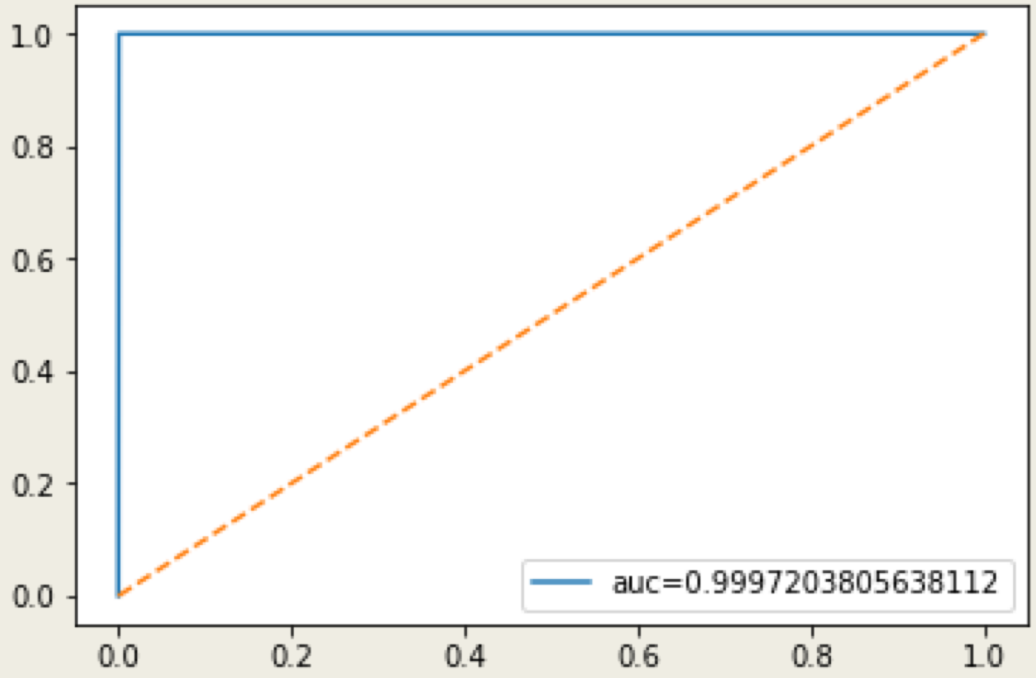
- For each Beta there is a distribution for possible values it could observe.



Distribution of Models



Conclusions



- Use Random Forest to model as the accuracy score is .9994 and I am able to get 0 false negatives.
- Bayesian models did not outperform regular logistic regression.
- In the future, I would love to get the anonymized data and explore those variables.