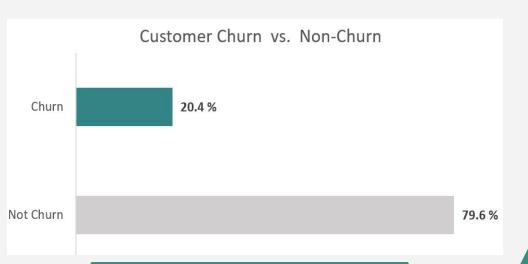
# PREDICTING BANK CUSTOMER CHURN

CHETANA VYAS



#### Why does it matter?



#### **Involuntary Churn**

- Closing the business
- Outdated Equipments



#### **Avoidable Churn**

- Poor Customer Service
- Rigid Pricing
  - Security Threats
- Complicated Interface

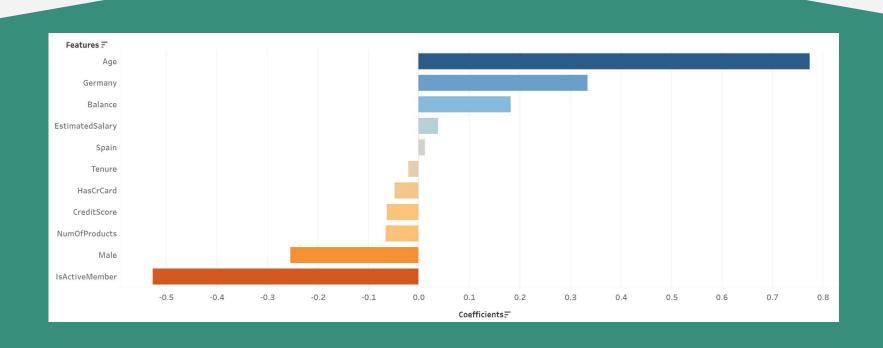
# Baseline Model -Logistic Regression

F\_Beta Score=0.80

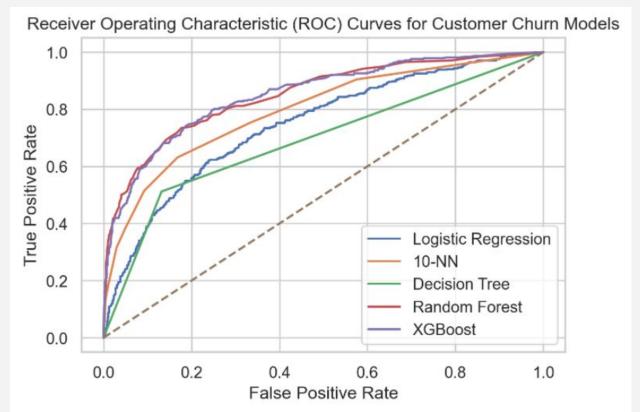
Beta = 2.5
( prefer recall )

**Recall = 0.21** 

#### **FEATURE IMPORTANCE**



# Classification Model Comparisons (ROC AUC)



- Random Forest
- XGBoost (Extreme Gradient Boosting)

#### Classification Model Comparisons (F Beta Score)









# **Extreme Gradient Boosting**

F\_Beta Score=0.89 (Beta = 2.5)

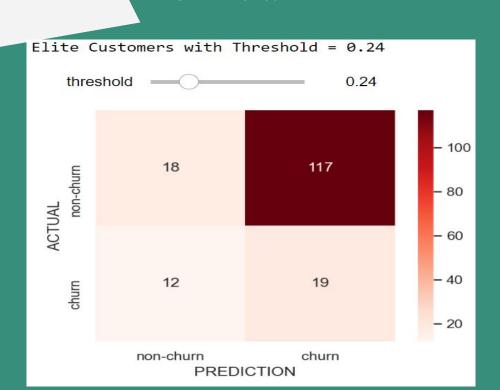
Recall = 0.54



#### ELITE CUSTOMERS

PRECISION = 14%

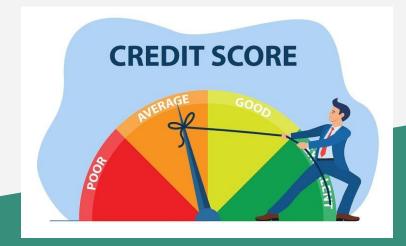
**RECALL** = **62**%





# Scoring the Random Forest Model on customers having

- Bank balance > 100K
- > Credit Score > 750



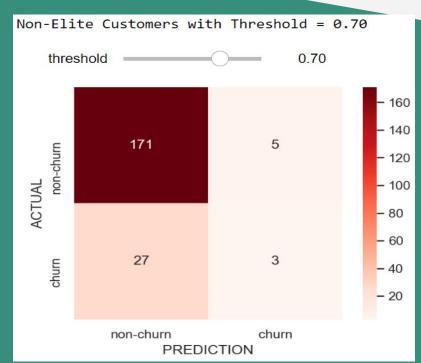
# PRECISION = 38% RECALL = 10%

**NON - ELITE** 

**CUSTOMERS** 

## Scoring the Random Forest Model on customers having

- Bank balance < 10K</p>
- Credit Score < 600</p>



# SO WHAT DO WE DO WITH OUR BINARY CLASSIFICATION MODEL?

#### RECOMMENDATIONS

#### DAILY CHURN DETECTION

Build powerful
Machine Learning
Models to analyze
customer behaviour



### CHURN PREVENTION

Loyalty & Retention Programs

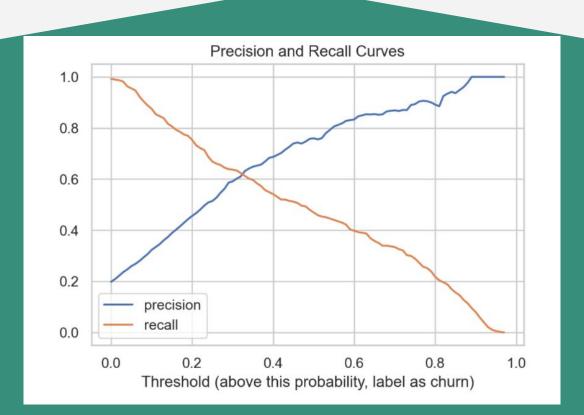
### CONTINUOUS OPTIMIZATION

On-demand access to predicted customers at risk of churning



#### **APPENDIX - Precision & Recall Curve**

Random Forest Classifier



#### **APPENDIX - Row data**

| Exited | CreditScore | Age | Tenure | Balance   | NumOfProducts | HasCrCard | IsActiveMember | EstimatedSalary | Germany | Spain | Male |
|--------|-------------|-----|--------|-----------|---------------|-----------|----------------|-----------------|---------|-------|------|
| 1      | 619         | 42  | 2      | 0.00      | 1             | 1         | 1              | 101348.88       | 0       | 0     | 0    |
| 0      | 608         | 41  | 1      | 83807.86  | 1             | 0         | 1              | 112542.58       | 0       | 1     | 0    |
| 1      | 502         | 42  | 8      | 159660.80 | 3             | 1         | 0              | 113931.57       | 0       | 0     | 0    |
| 0      | 699         | 39  | 1      | 0.00      | 2             | 0         | 0              | 93826.63        | 0       | 0     | 0    |
| 0      | 850         | 43  | 2      | 125510.82 | 1             | 1         | 1              | 79084.10        | 0       | 1     | 0    |
|        |             |     |        | •••       |               |           |                |                 |         |       |      |
| 0      | 771         | 39  | 5      | 0.00      | 2             | 1         | 0              | 96270.64        | 0       | 0     | 1    |
| 0      | 516         | 35  | 10     | 57369.61  | 1             | 1         | 1              | 101699.77       | 0       | 0     | 1    |
| 1      | 709         | 36  | 7      | 0.00      | 1             | 0         | 1              | 42085.58        | 0       | 0     | 0    |
| 1      | 772         | 42  | 3      | 75075.31  | 2             | 1         | 0              | 92888.52        | 1       | 0     | 1    |
| 0      | 792         | 28  | 4      | 130142.79 | 1             | 1         | 0              | 38190.78        | 0       | 0     | 0    |
|        |             |     |        |           |               |           |                |                 |         |       |      |

#### **TOOLS**





Seaborn

#### **CLASSIFICATION ALGORITHMS**

