



Bank Customer Churn Prediction

BAN 5573

Visual Analytics and Business Intelligence

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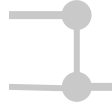
December 2022





Background

Motivation



Introduction



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Descriptive Analysis



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Dataset main information

Results



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Conclusion



Introduction

Background

- Churn customers are those who stopped being customers during certain periods
- Credit score, geography, gender, and age play a certain role in churn.

Motivation

- What are the values of churn of customers associated with different factors?
- Predict whether a customer will exit the bank or not, based on his/her characteristics.



Dataset main information

kaggle.com/datasets/mathchi/churn-for-bank-customers

kaggle

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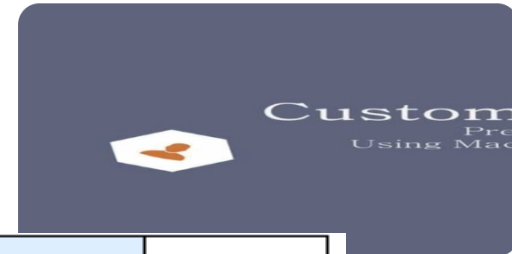
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New Notebook

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Churn for Bank Customers

Predict customer churn in a bank



Data Set Characteristics:	Multivariate	Number of Instances:	10000
Attribute Characteristics:	Categorical, Numerical, Boolean, String	Number of Attributes:	14
Associated Tasks:	Classification	Missing Values?	No



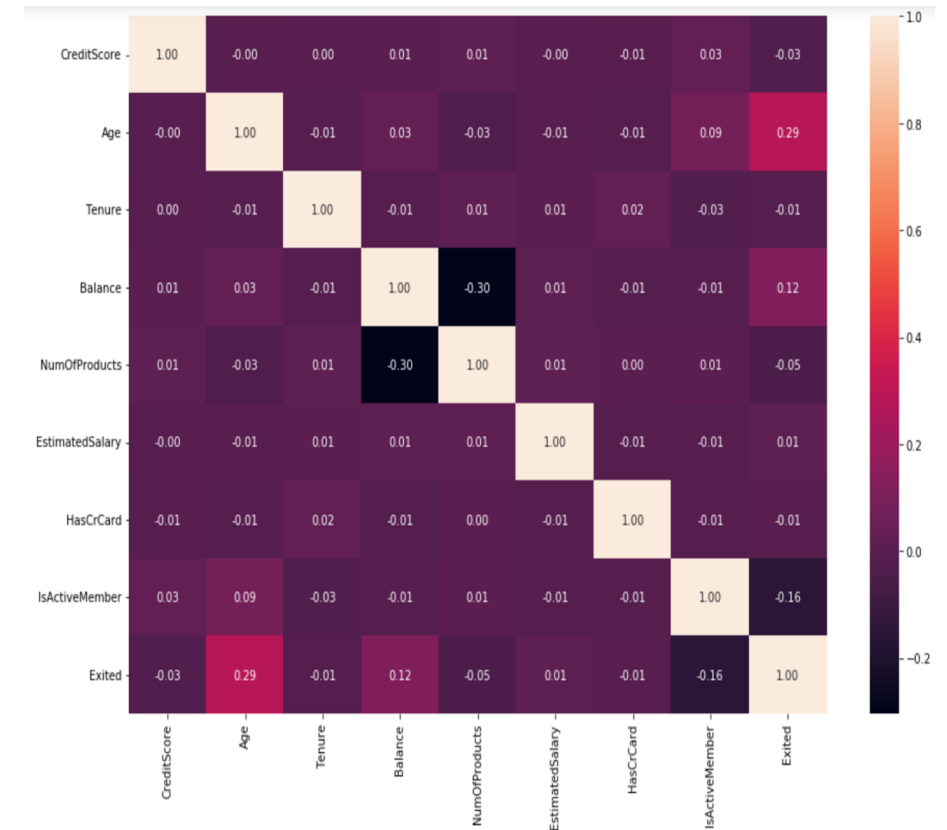
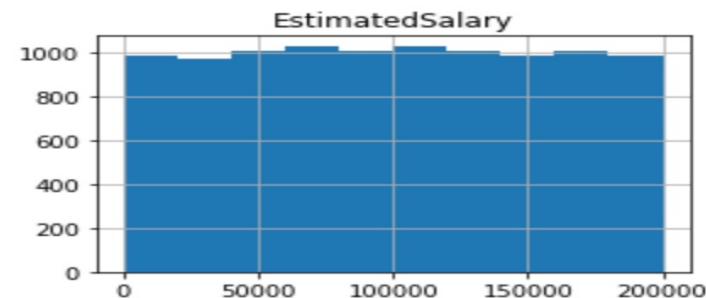
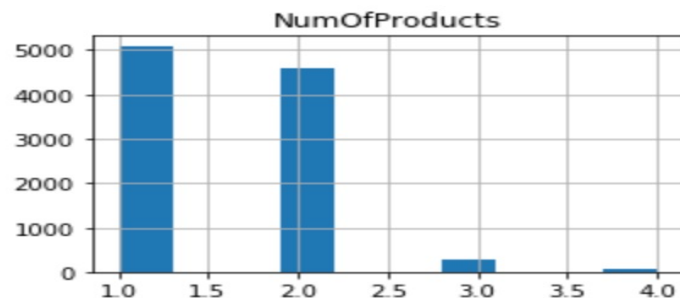
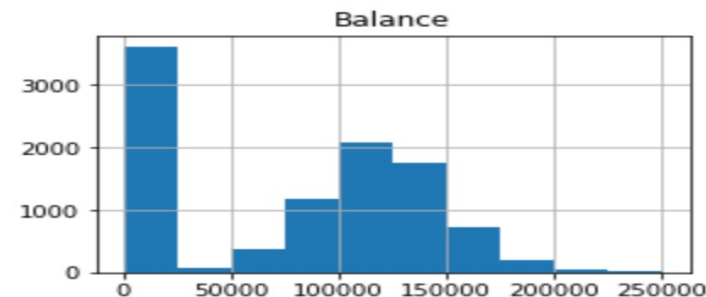
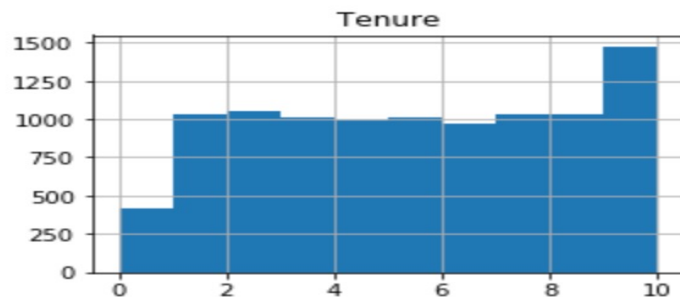
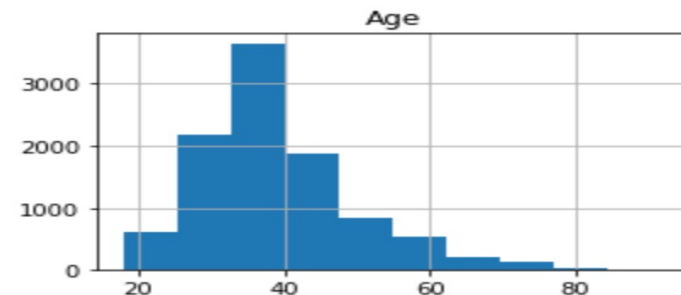
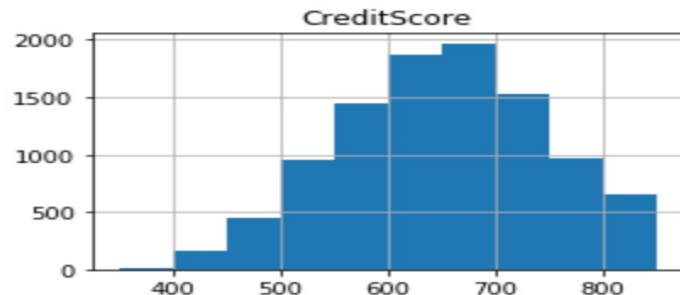
Dataset Features Description

- RowNumber
- CustomerId
- Surname
- CreditScore
- Geography
- Gender
- Age
- Tenure

- Balance
- NumOfProducts
- HasCrCard
- IsActiveMember
- EstimatedSalary
- Exited



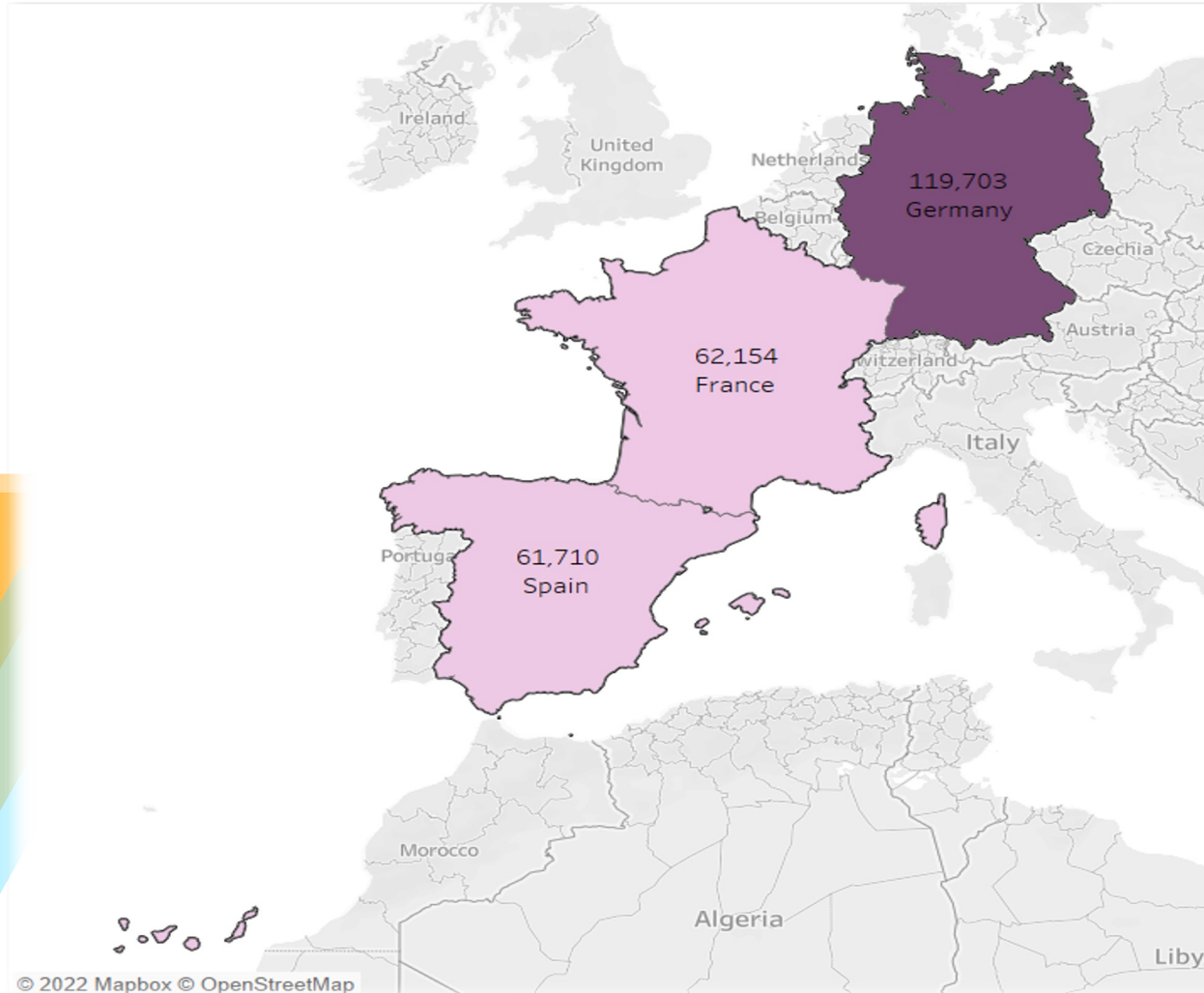
Descriptive Analysis





Visualization of Customers' Information in Different Countries

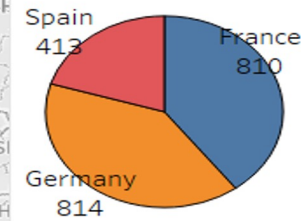
Balance median in each country



Median Balance

61,710 119,703

Exited Customers



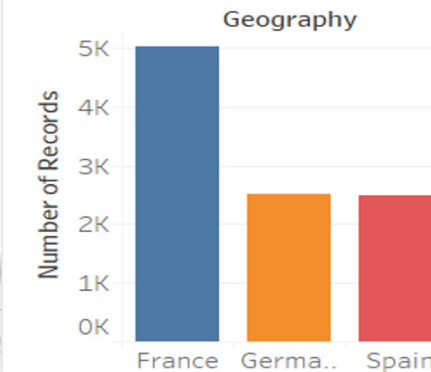
Geography

France
Germany
Spain

Sum of Exited

2,037

Sheet 12

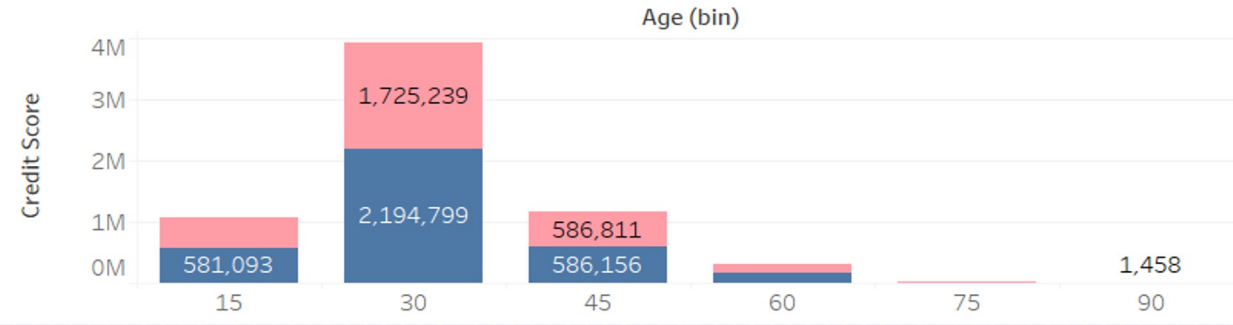




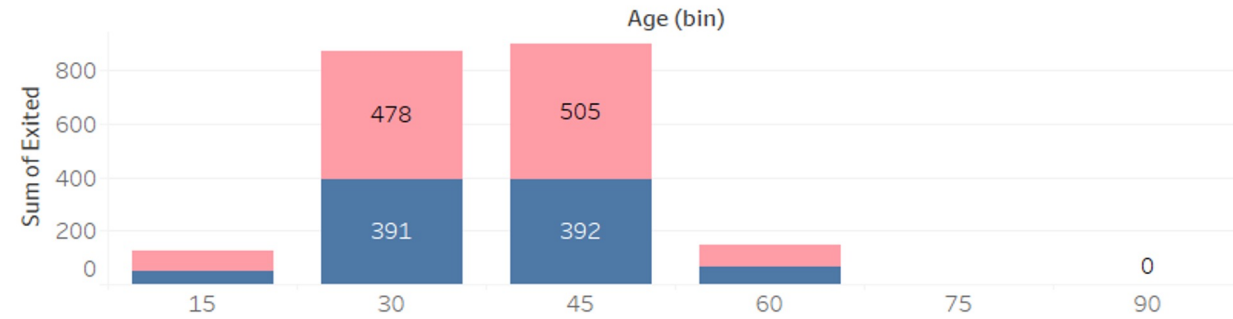
Customer Behavioral Pattern in Different Age Groups

Gender
Female
Male

Credit Score on the basis of gender.

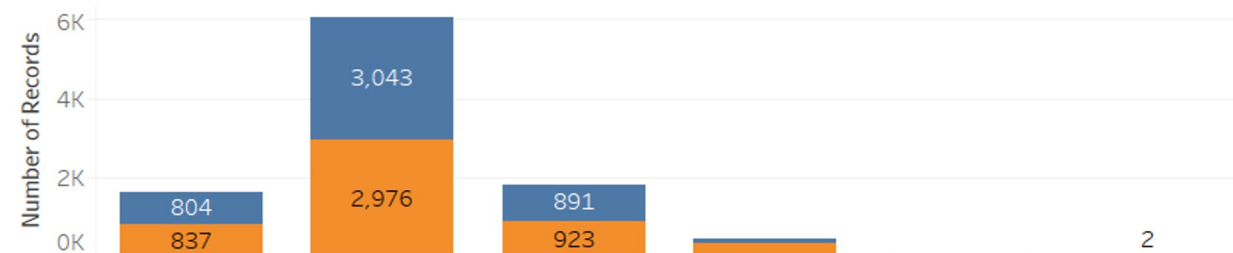


Sum of Exited in each age group



Active vs. InActive members in age group

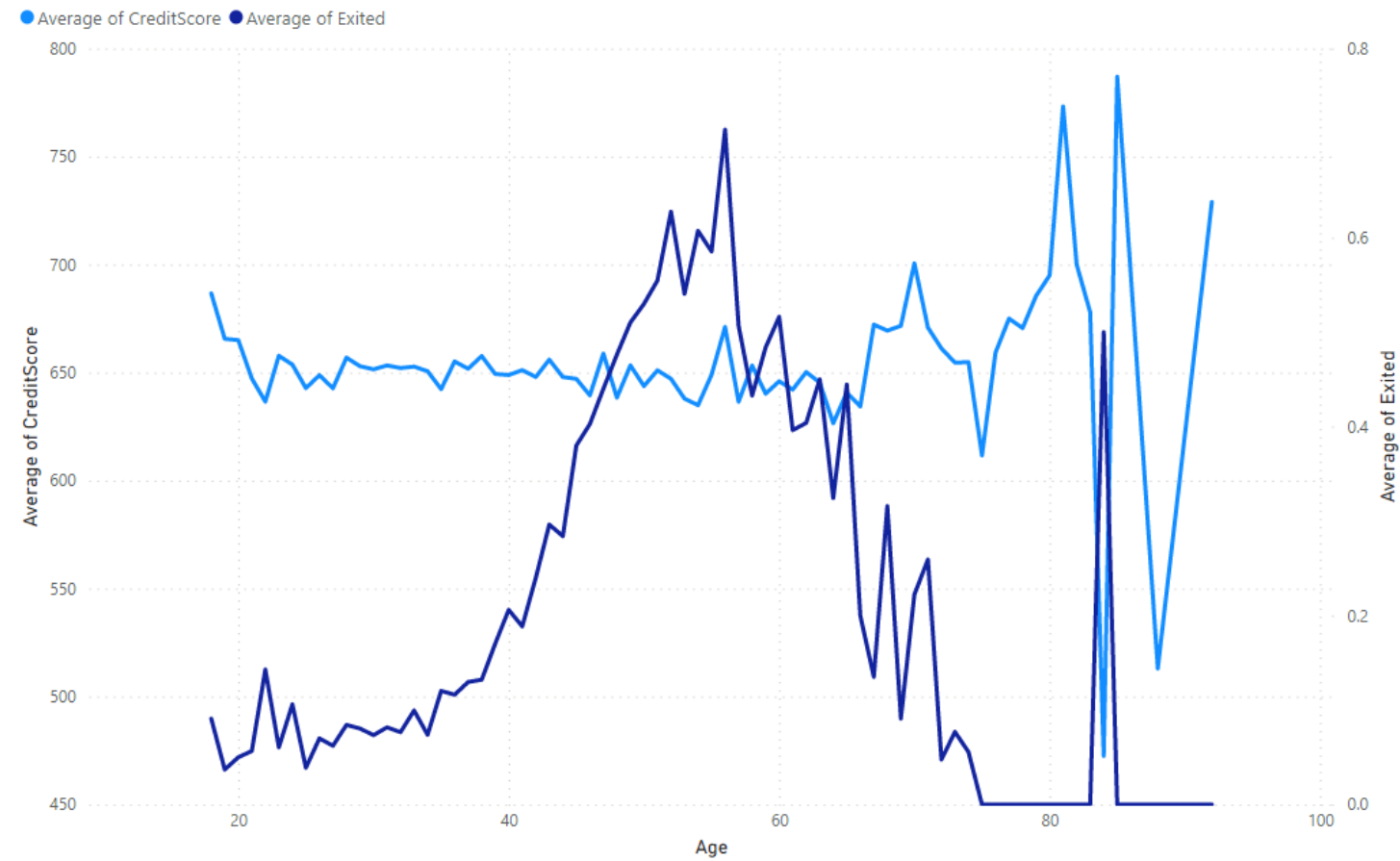
Is Active Member
0
1





Average credit score and average exited age

Average of CreditScore and Average of Exited by Age





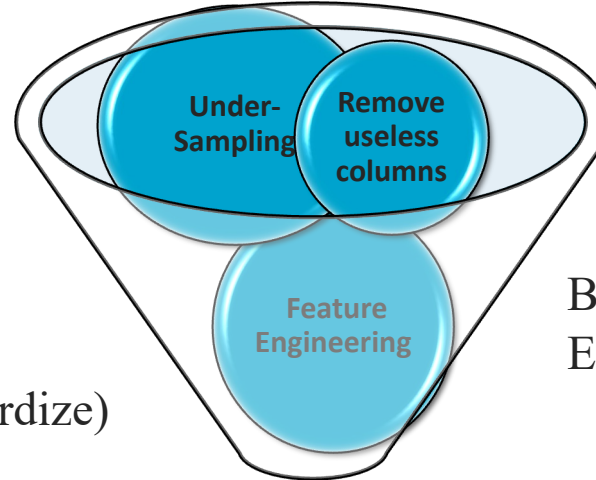
Data Preprocessing

0 7963
1 2037



0 2037
1 2037

Scale (Standardize)



Columns:

1- RowNumber, 2- CustomerId,
3- Surname

Balance: Poor, Good, Excelent, Top
EstimatedSalary: Poor, Good, Excelent, Top

Supervised Learning

Decision Tree

Random Forest

Logistic Regression

K Nearest Neighbor

Voting Classifier



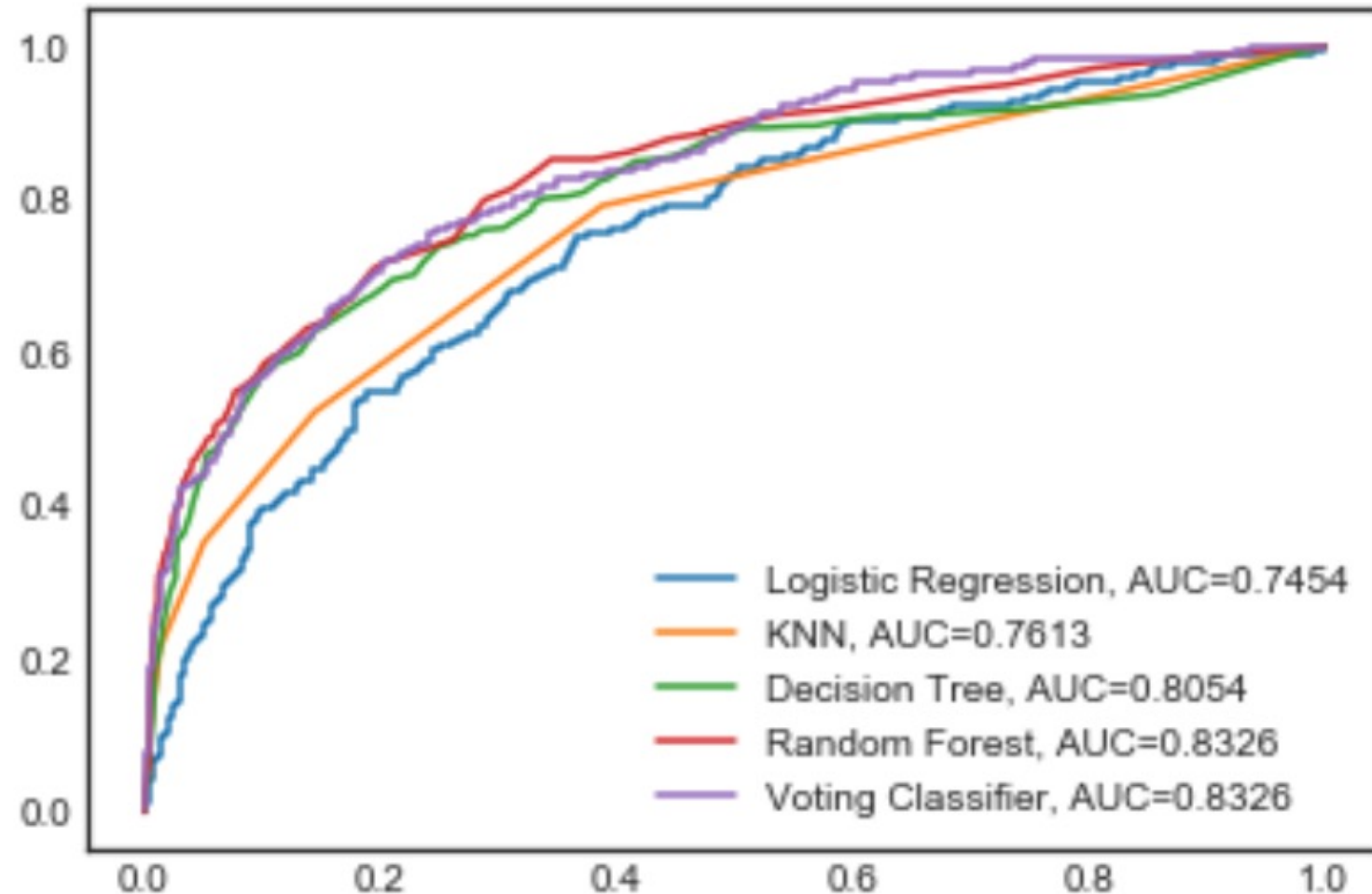
Predictive Analysis

Model	Accuracy	Precision	Recall	F1-score
Decision Tree	0.85	0.79	0.69	0.72
Random Forest	0.86	0.81	0.70	0.73
Logistic Regression	0.80	0.69	0.56	0.56
KNN	0.83	0.75	0.65	0.68
Voting Classifier	0.85	0.82	0.67	0.71



Predictive Analysis

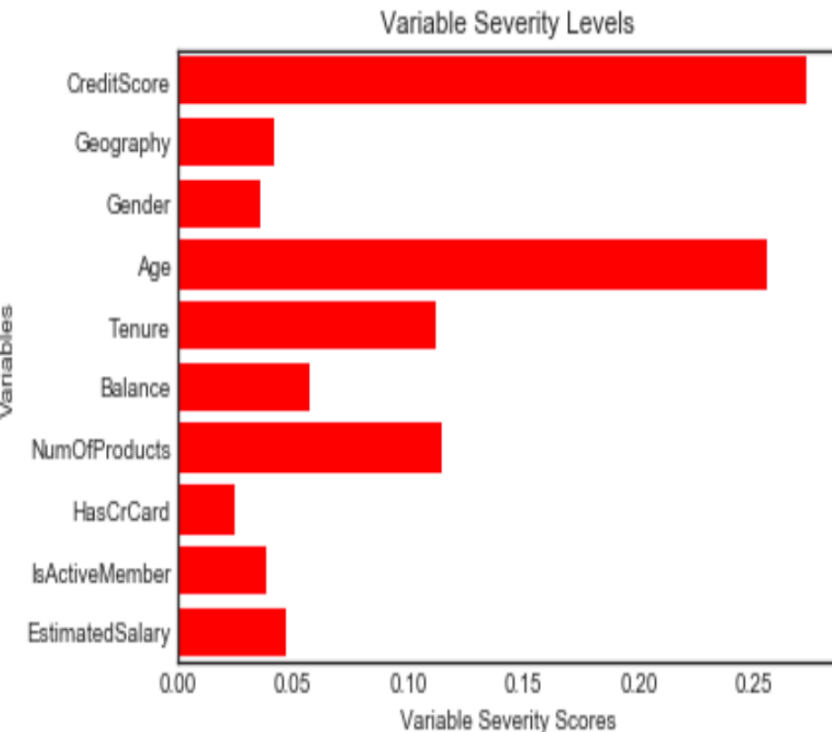
**Compare Models Using
the area under the
ROC curve**



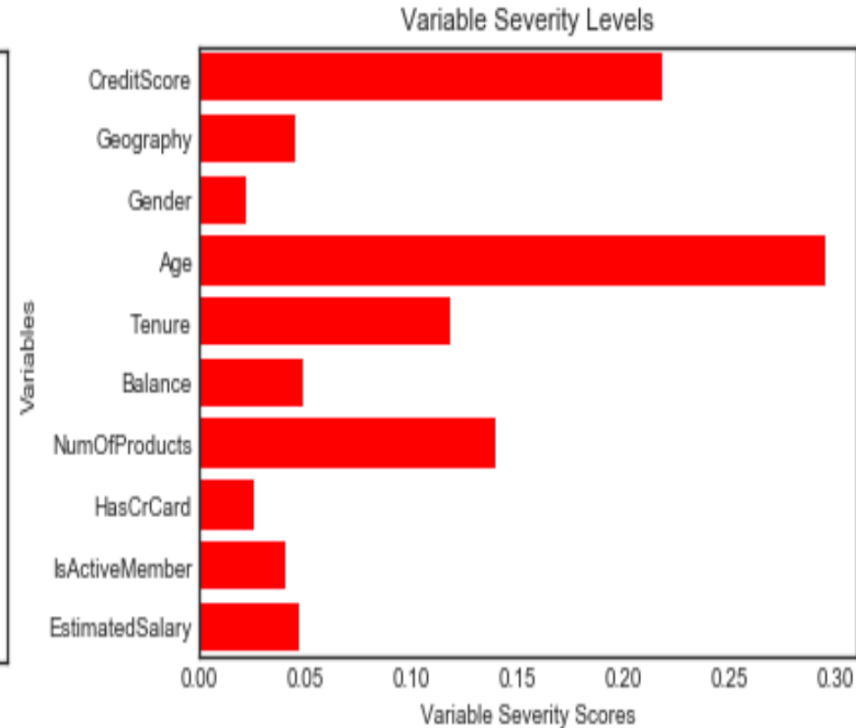


Feature Importance

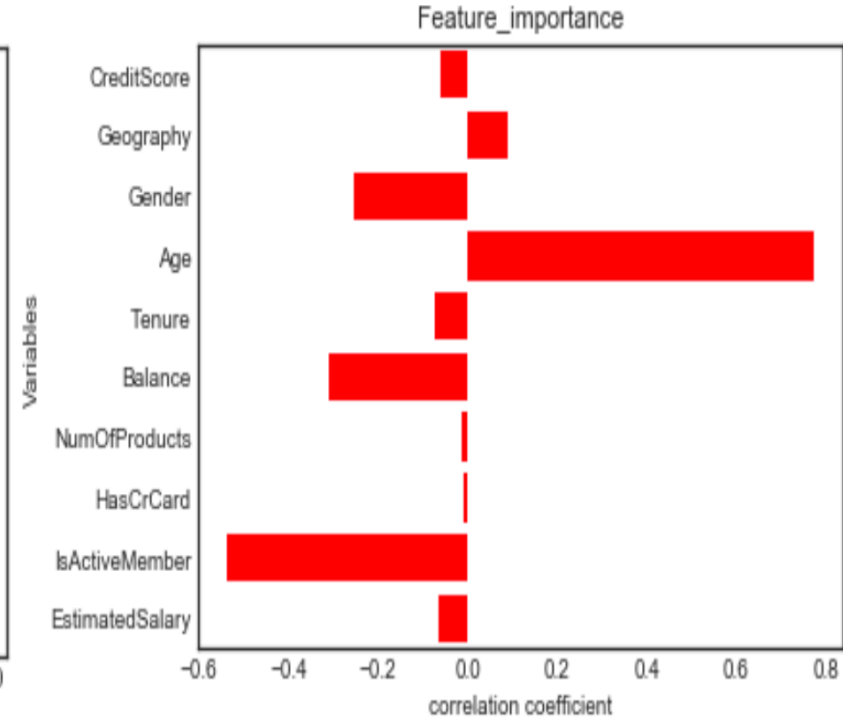
Decision Tree



Random Forest



Logistic Regression



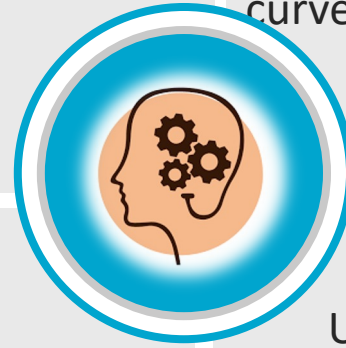




Result/Conclusion

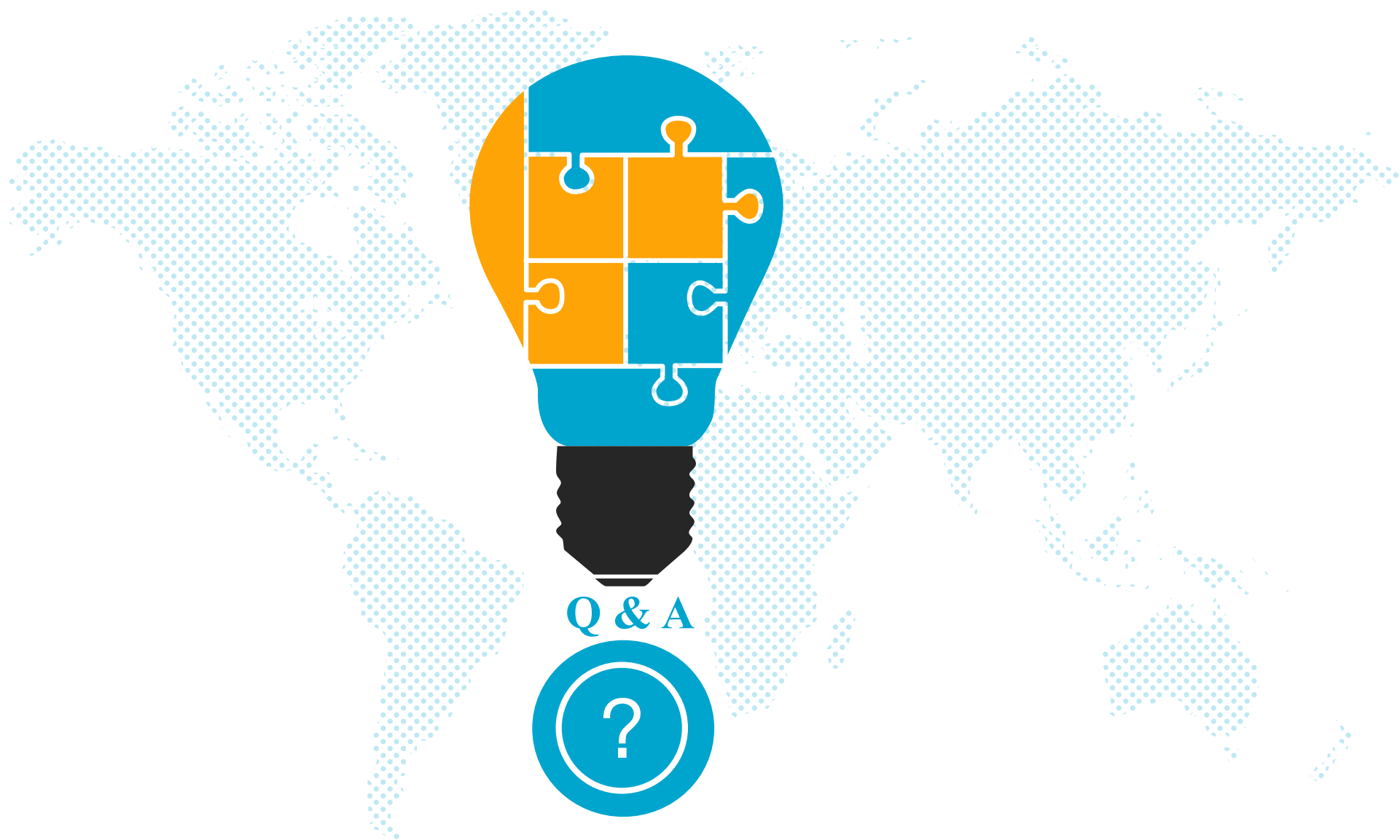
Customers' age, credit scores, and the number of products they use are the most important features for customer churn prediction.

Among models, Random Forest and voting model have the best performance based on accuracy, precision, recall, f-1 score, and area under the curve measurements.



Based on visualization, Almost half of the customers from Germany leave the bank. Therefore, we should investigate these two groups and find the reason so we can design proper strategies to prevent them from exiting

Using machine learning algorithms, we were able to predict 70% of exited customers. Offering promotions to them makes it possible to encourage them to stay.





THANK YOU