

Title: European Bank's Customer Churn Analysis & Prediction

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Introduction & Executive Summary

What is Churn?

Churn is a term that refers to the number of customers that a company(bank) loses over a specific time period either by closing their account.

Cost of Customer Acquisition

5-25 times more expensive than retaining existing customers

Profit from Customer Retention (5%)

25-95 % profit increase

Problem Statement:

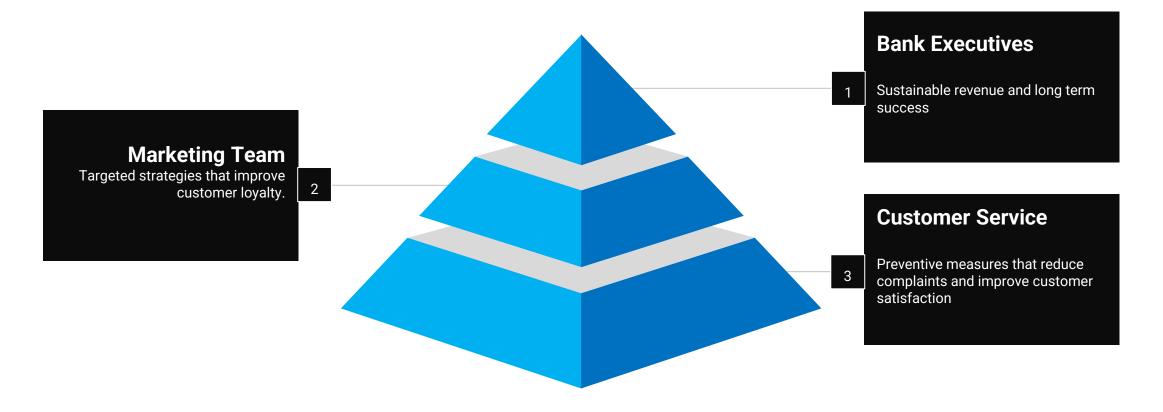
Customer churn is a critical issue for businesses in a competitive markets. Understanding and predicting churn can help bank to strategize proactive measures on customer retention, improve customer satisfaction and service, and therefore increasing revenue.

Oliver Wyman. (2021, March). Banking-as-a-Service enables up to 95 percent cost saving in acquiring customers. Retrieved from https://www.oliverwyman.com/media-center/2021/mar/banking-as-a-service-enables-up-to-95-percent-cost-saving-in-acquiring-customers.htm

Rajendran, S., Devarajan, R., & Elangovan, G. (2023, January). Customer churn prediction using machine learning approaches. 2023 International Conference on Engineering, Technology, and Applied Science (ICECONF). https://doi.org/10.1109/ICECONF57129.2023.1008381



Stakeholders & Goals



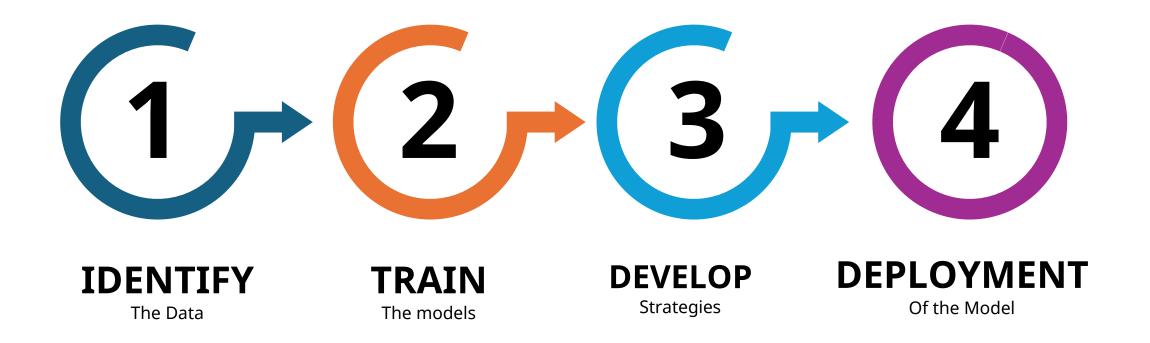
Goals:

To improve customer retention

Develop ML models to predict churn

Provide actionable insights for bank management







Identify Data

Dataset: 10,000 rows x 14 columns

Missing Data: Steps to handle missing data.

(Geography & gender)

Encoding Categorical Features:

- Label Encoding for binary categorical features (Gender).
- One-Hot Encoding for multi-class categorical features (Geography).

Drop: Row number, Customer ID, Surname

Responding Variable: Exited

Variables

Numerical

- 1) Credit Score
- 2) Age
- 3) Balance
- 4) Estimated Salary

Categorical

- 1)Geography
- 2)Gender
- 3)Tenure
- 4)No of products
- 5) Has Credit card
- 6)Active member



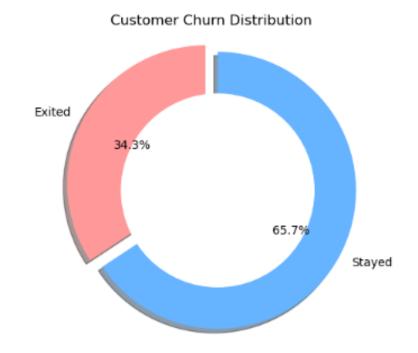
Exploratory Data Analysis Findings

Customer Churn Rate

Total Customer: 18, 869

Customer Churn Rate: 34.3%

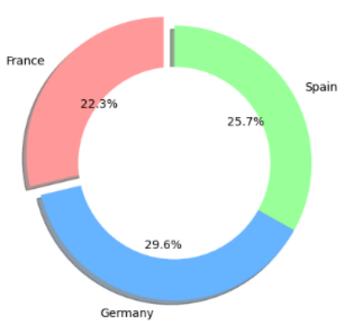
	Count	Percentage
Stayed	12, 427	65.7%
Exited	6,442	34.3%





Customer Churn Rate by Country

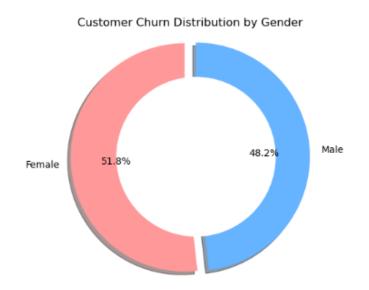
Customer Churn Distribution by Country

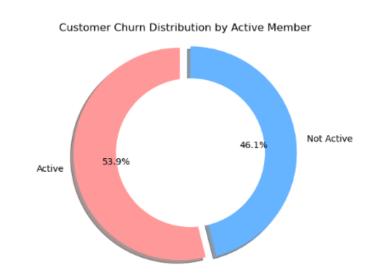


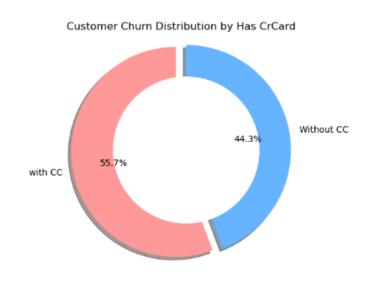
	Count	Churn Count	Percentage
France	10.352	2.304	22.3%
Germany	7,635	2,257	29.6%
Spain	7,324	1,881	25.7%
Total	18,869	6,442	100%

Geography: Higher exits in Germany, lowest in France.









	Count	Percentage
Female	3,336	51.8%
Male	3,105	48.2%

Gender: Higher exit proportion for females.

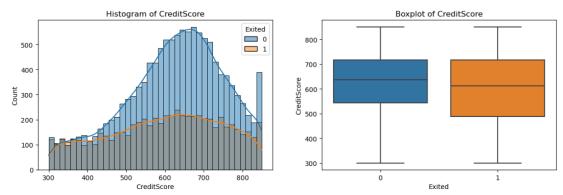
	Count	Percentage
Active	3,475	53.9%
Not Active	2,967	46.1%

Is Active Member: Higher exit proportion for not active members.

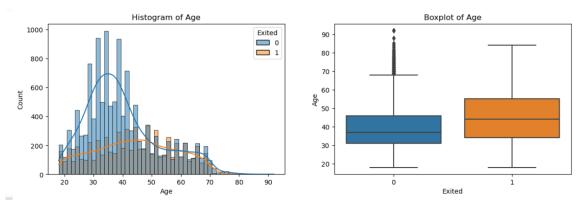
	Count	Percentage
With CC	3,589	55.7%
Without CC	2,853	44.3%

CrCard: Higher exit proportion with CC

Continuous Numerical Variables

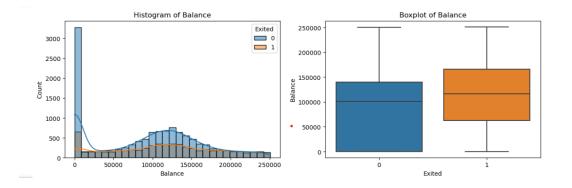


Credit Score: Slightly lower median for exited customers. **Median credit score for customers who exited: 611.0** compared to 637.0 for those who did not exit.

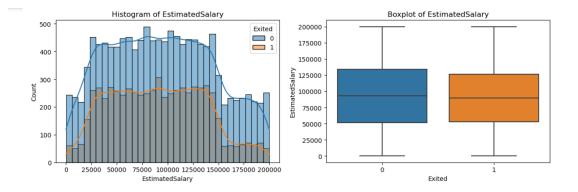


Age: Older customers are more likely to exit. **Mean age of customers who exited is 44** compared to **39 for those who did not exit**.





Balance: Higher median balance for exited customers. Range of balance for customers who exited: 0 to 250,898.09 with a **median balance of** \$116,760.60



Estimated Salary: Similar distributions, suggesting it might not be a significant factor for churn.

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Summary of EDA

Customers who are more likely to churn are:

- 1) High balance customers
- 2) Older Customers
- 3) Customer with Fewer products
- 4) Non active customers
- 5) Customers in Germany



Approach

Data Collection & Preprocessing

Exploratory Data Analysis (EDA)

Develop ML Model

Data Preprocessing

- Handling Missing Values: Mode values for 'Geography', 'Gender'
- Encoding: Label Encoding, One-Hot Encoding
- •Class Imbalance: SMOTE
- Scaling: StandardScaler

Exploratory Data Analysis

• Histograms, Boxplots, donut chart

Model Training

- Data Splitting: Training, Testing sets
- •Algorithms: Logistic Regression, Random Forest, Gradient Boosting, AdaBoost, KNN

Model Tuning

 Hyperparameter Tuning: Random Forest, Gradient Boosting

Model Evaluation

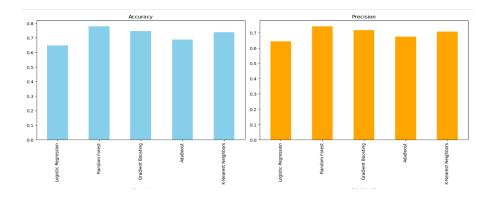
 Performance Metrics: Accuracy, Precision, Recall, AUC

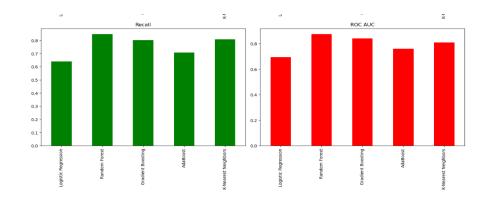


Model Performance

Model Performance:

	Accuracy	Precision	Kecall	AUC
Logistic Regression	0.644136	0.641945	0.633849	0.694120
Random Forest	0.775498	0.740157	0.841334	0.871649
Gradient Boosting	0.743311	0.713357	0.803906	0.836956
AdaBoost	0.696641	0.682412	0.722945	0.762206
K-Nearest Neighbors	0.728023	0.693899	0.805126	0.803041





Evaluation Results:

Random Forest: Best performance in Accuracy (77.5%),
Precision (74.0%),
Recall (84.1%),
AUC (0.872).

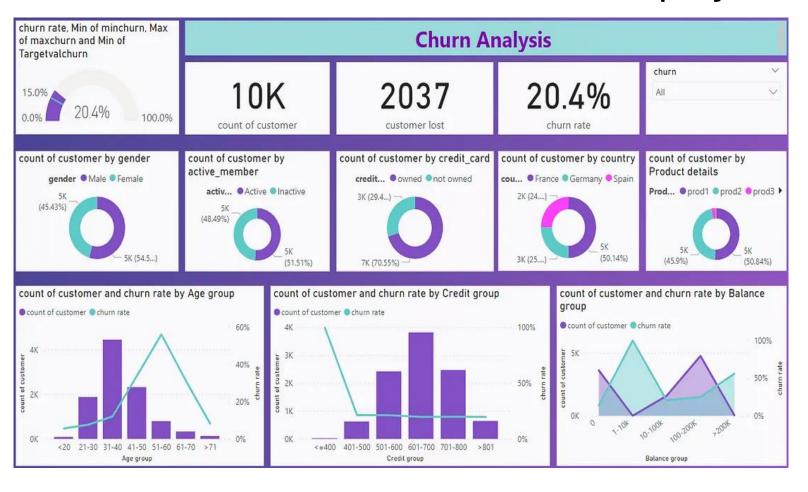
Gradient Boosting: Strong alternative.

Conclusion:

The analysis highlights the critical factors influencing customer churn and identify effective machine learning model in predicting churn.



Recommendation on Future Work: Deployment



Dashboard with the machine learning model developed

Objective: Real-time monitoring of customer behavior and churn rate.

Review the analysis in Biweekly management meeting

Make sure strategies aligns with the data pattern reflection on customer behaviour.

Features:

Live (Churn rate, customer count, customer lost)

Donut charts for:

 customers and churn rate by gender, activity status, credit card status, country, churn rate, no of products.

Line bar charts for:

 customers and churn rate by age group, credit score, balance.



Actionable Insights for Customer Retention Strategies:

Customers who are more likely to Churn	Retention Strategies
High balance customers	Personalized Services: Assign dedicated relationship managers. Exclusive Benefits: Offer premium banking services and rewards. (Low interest loan, insurance, credit card with special benefits. Reward Programs: Implement loyalty programs.
Older Customers	Financial Planning Services: Provide retirement planning and investment advice. Long-Term Relationship Building: Focus on trust and personalized interactions. Health and Wellness Programs: Offer health-related financial services.
Customers with Fewer Products	Cross-Selling Campaigns: Encourage adoption of additional products. Bundling Discounts: Offer incentives for product bundling. Product Education: Educate on benefits of additional services.
Non-Active Customers	Reactivation Campaigns: Launch targeted campaigns with incentives. Digital Engagement: Enhance digital platforms for convenience. Exclusive Benefits: Offer premium banking services and rewards. (Low interest loan, insurance, credit card with special benefits
Germany Customers	Localized Offerings: Customize marketing and products. Community Engagement: Participate in local initiatives.



Thank you



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