

Project

Data Science

Agenda

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TEAM MEMBERS

CUSTOMER CHURN, OR THE ATTRITION OF CUSTOMERS, IS A CRITICAL CONCERN FOR BUSINESSES ACROSS VARIOUS INDUSTRIES. IDENTIFYING FACTORS THAT CONTRIBUTE TO CUSTOMER CHURN AND BUILDING PREDICTIVE MODELS TO ANTICIPATE AND MITIGATE THIS PHENOMENON ARE ESSENTIAL FOR MAINTAINING A STABLE AND THRIVING CUSTOMER BASE

DATASET DESCRIPTION

Our dataset consisting of :

- **ROWNUMBER**: AN IDENTIFIER FOR EACH ROW IN THE DATA SET.
- **CUSTOMERID**: A UNIQUE IDENTIFIER FOR EACH CUSTOMER.
- **SURNAME**: THE SURNAME OR LAST NAME OF THE CUSTOMER.
- **CREDITSCORE**: A NUMERICAL REPRESENTATION OF THE CREDITWORTHINESS OF THE CUSTOMER.
- **CREDITSCORE**: A NUMERICAL REPRESENTATION OF THE CREDITWORTHINESS OF THE CUSTOMER.
- **GEOGRAPHY**: THE GEOGRAPHICAL LOCATION OR COUNTRY WHERE THE CUSTOMER RESIDES.
- **GENDER**: THE GENDER OF THE CUSTOMER (MALE/FEMALE).
- **AGE**: THE AGE OF THE CUSTOMER.

DATASET DESCRIPTION

Our dataset consisting of :

- **TENURE:** THE NUMBER OF YEARS THE CUSTOMER HAS BEEN WITH THE BANK OR SERVICE.
- **BALANCE:** THE ACCOUNT BALANCE OF THE CUSTOMER.
- **NUMOFPRODUCTS:** THE NUMBER OF BANK PRODUCTS THE CUSTOMER IS USING.
- **HASCRCARD:** BINARY INDICATOR (0 OR 1) WHETHER THE CUSTOMER HAS A CREDIT CARD.
- **ISACTIVEMEMBER:** BINARY INDICATOR (0 OR 1) WHETHER THE CUSTOMER IS AN ACTIVE MEMBER.
- **ESTIMATEDSALARY:** THE ESTIMATED SALARY OF THE CUSTOMER.
- **EXITED:** THE TARGET VARIABLE INDICATING WHETHER THE CUSTOMER HAS CHURNED (1) OR NOT (0).

PREPROCESSING

- we use functions like (.info) &(.head) to describe our dataset
- Check Nulls & duplicate and remove it
- we dropped three unnecessary columns
- Making one-hot encode for Geography column
- Encode Gender column
- Calculate z-score for specific columns then detect the outliers and delete it

Each algorithm has strengths depending on the dataset's complexity and feature interactions.

RANDOMFOREST

- Handles mixed data types
- Works well with non-linear relationships

SVM

- Effective in high-dimensional spaces
- Works well for smaller datasets

KNN

- Simple and effective for smaller feature spaces

REASON : Because our data is a numerical data

RESULT

RANDOMFOREST:

```
RF Accuracy: 90.62%  
f1 score = 90.61775699785439  
recall score = 90.62306653539282  
precision score = 90.61370204480262  
confusion matrix:  
[[1059 110]  
 [ 104 1009]]
```

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SVM:

```
SVM Accuracy: 85.19%  
f1 score = 85.18131712745225  
recall score = 85.2935638157647  
precision score = 85.43791967414649  
confusion matrix:  
[[947 222]  
 [116 997]]
```

1

KNN:

```
KNN Accuracy: 82.87%  
f1 score = 82.78441984419844  
recall score = 83.08477384852935  
precision score = 83.95755630865156  
confusion matrix:  
[[ 867 302]  
 [ 89 1024]]
```

4

DEEP LEARNING MODEL :

```
Enhanced Neural Network Model Evaluation:  
Accuracy: 0.8584574934268185  
F1 Score: 0.8626116546150574  
Recall Score: 0.9110512129380054  
Precision Score: 0.8190630048465266  
Confusion Matrix:  
[[ 945 224]  
 [ 99 1014]]
```

2

RESULT

DATASET LINK

HERE IS THE LINK :

<https://www.kaggle.com/datasets/filippoo/deep-learning-az-ann>

DATASET LINK

