



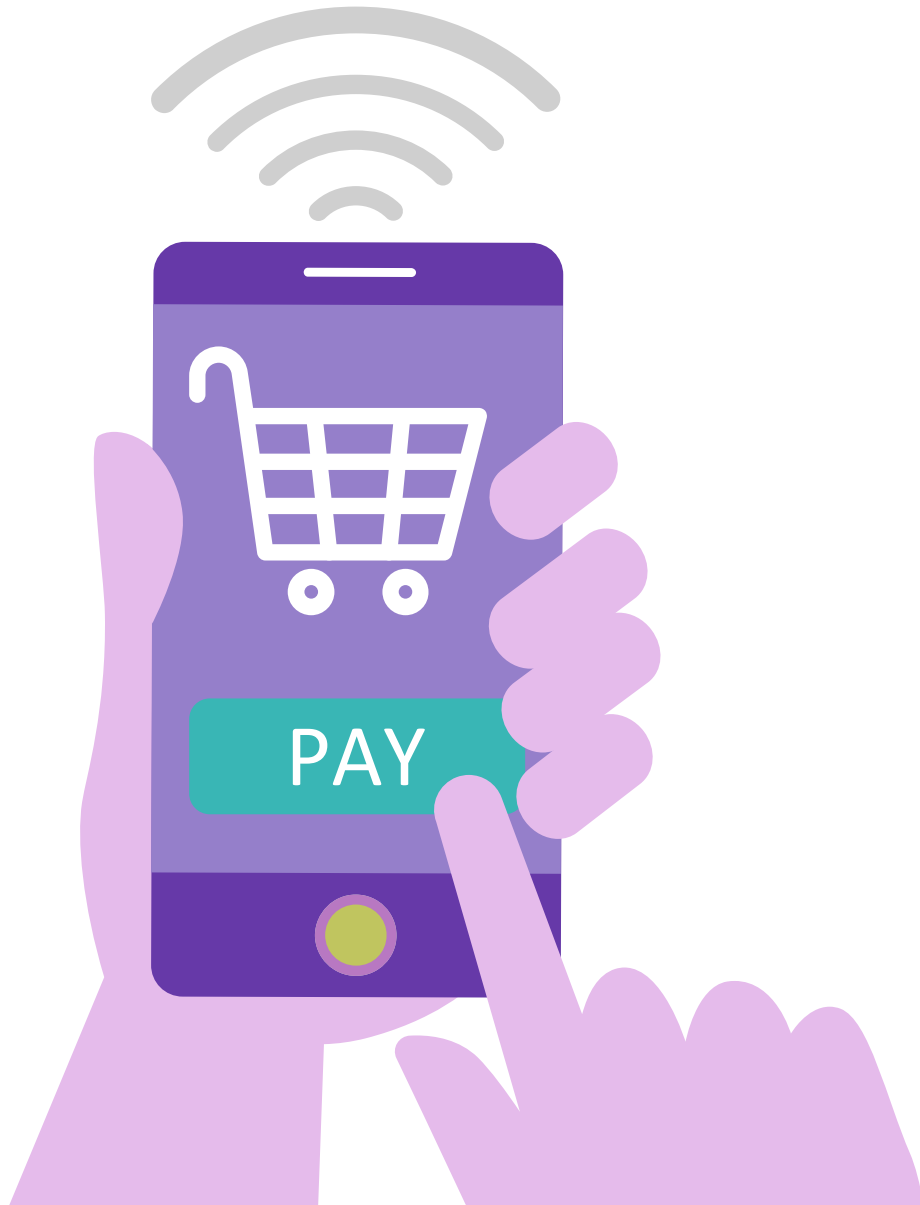
UNIVERSITÀ DI PISA  
Artificial Intelligence  
and Data Engineering

# CHURNPREDICT

Data Mining and Machine  
Learning project

Denny Meini





# PROJECT GOALS

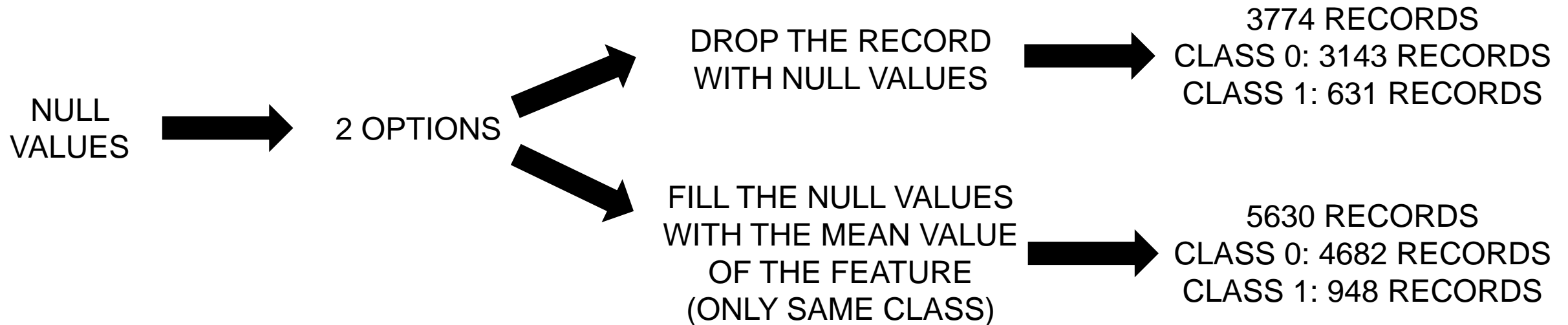
The goals of this project are to compare a list of classifiers in order to discover which of them gives us the best performance in terms of some parameters (f-score and average accuracy). Another goal is to build an application that uses the classifier which gave the best results. The dataset has some null values and an additional aim is to know if it is better to drop the incomplete records or to complete them.

# DATASET

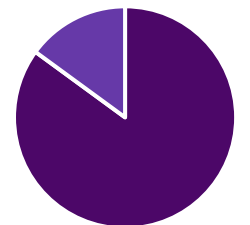
I took the dataset from kaggle

Link: <https://www.kaggle.com/datasets/ankitverma2010/ecommerce-customer-churn-analysis-and-prediction>

5630 records with 19 features and 1 binary target



IMBALANCED DATASET



Numerical feature  
Categorical feature

**CustomerID**

**Churn**

Output class [0: did not left, 1 left].

**Tenure**

The number of years the customer has been a customer.

**PreferredLoginDevice**

**CityTier**

The tier of the city (Chinese model) [1: big, 2: medium, 3: small].

**WarehouseToHome**

Distance between the warehouse and the customer's house.

**PreferredPaymentMode**

**Gender**

**HourSpendOnApp**

Number of hours the customer has spent on the app.



# DATASET

**NumberOfDeviceRegistered**

**PreferredOrderCat**

**SatisfactionScore**

Grade of satisfaction of the customer from 1 (not satisfied) to 5 (very satisfied).

**MaritalStatus**

**NumberOfAddress**

Number of address added by a customer.

**Complain**

If the customer complained or not during the last month [0: no, 1: yes].

**OrderAmountHikeFromlastYear**

Percentage increase regarding the orders from the previous year.

**CouponUsed**

Number of coupon the customer used last month.

**OrderCount**

Number of order the customer placed during last month.

**DaySinceLastOrder**

**CashbackAmount**

Average cashback of the customer last month.

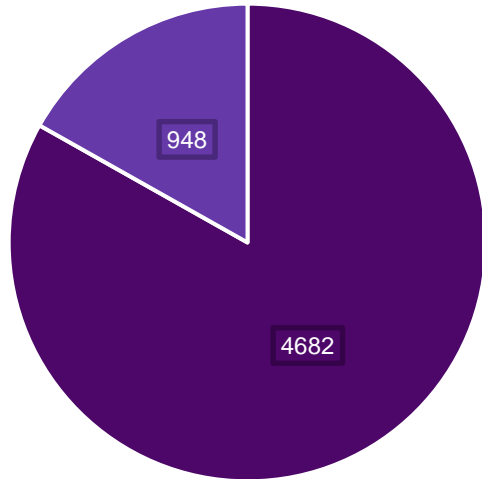
# PREPROCESSING

## Null values problem:

Creation of the two datasets

## Balance analysis:

Imbalanced Dataset



■ 0 (Did not left) ■ 1 (Left)



## Drop of useless features:

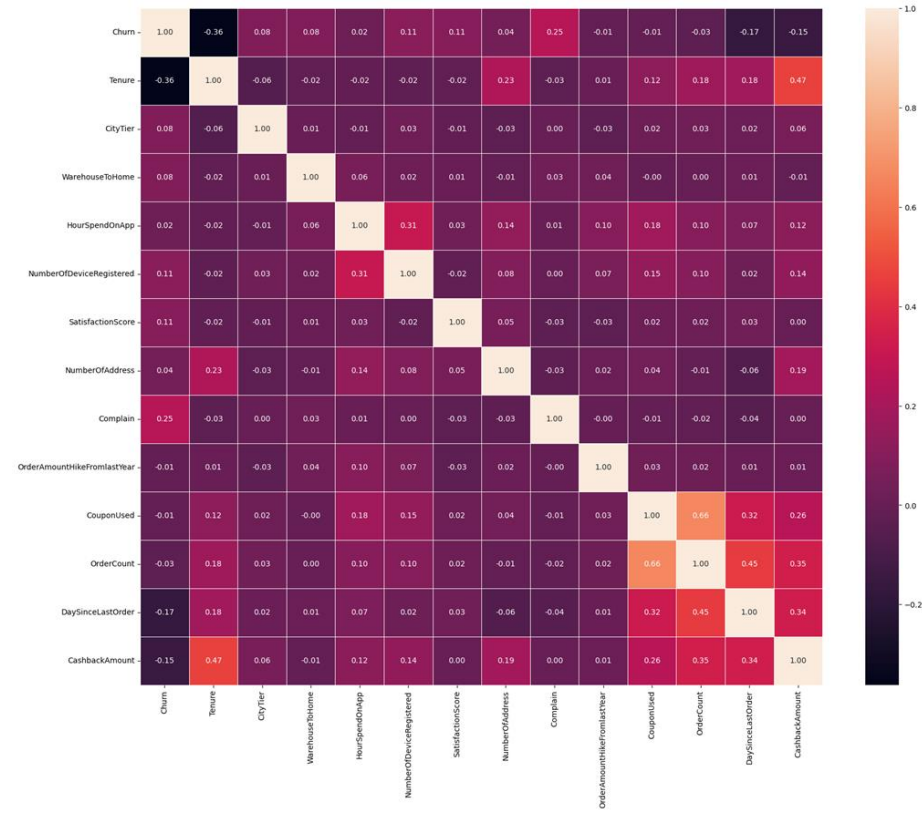
CustomerID

## Feature extraction:

Obtaining numerical representation of categorical features: `getDummies()`

## Correlation analysis:

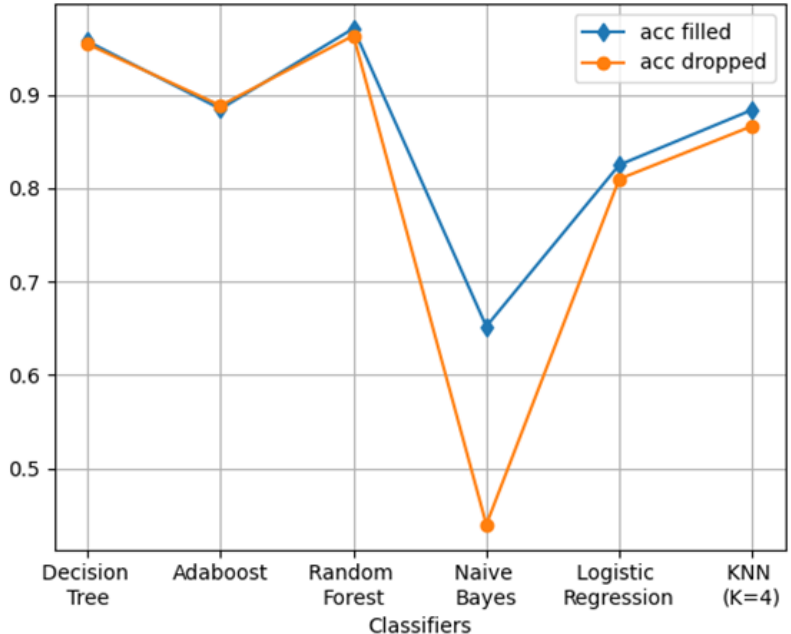
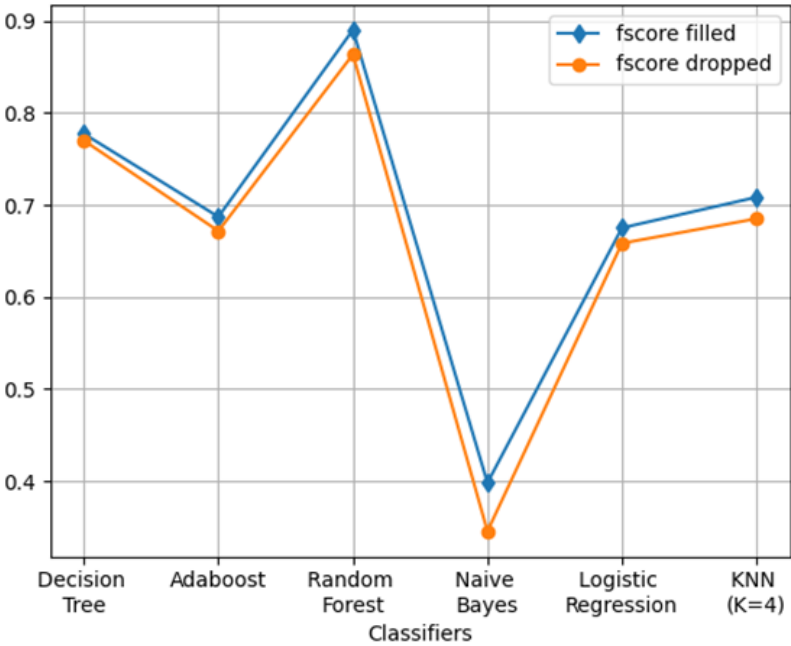
no correlated features



# CLASSIFICATION

Cross validation using Stratified K Fold (k=5)  
Rebalancing using SMOTE  
For each classifier I got f-score and Avg Accuracy

Classifiers		f-score	Avg accuracy
Decision Tree	Filled	0.777	0.957
	Dropped	0.770	0.954
AdaBoost	Filled	0.687	0.885
	Dropped	0.672	0.888
Random Forest	Filled	0.890	0.971
	Dropped	0.864	0.963
Gaussian Naive Bayes	Filled	0.398	0.652
	Dropped	0.345	0.440
Logistic Regression	Filled	0.675	0.825
	Dropped	0.658	0.810
KNN (K=4)	Filled	0.708	0.884
	Dropped	0.685	0.866



Wilcoxon test: pvalue=0.0625  
↓  
No statistical relevance

# APPLICATION IMPLEMENTATION


Input: The values of the features of the dataset  
Output: Presence or not of churn risk

ChurnPredict

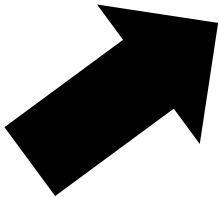
**ChurnPredict**

Tenure:	City tier:	Warehouse to home:	Hours spent on app:
<input type="text"/>	<input type="text"/>	<input type="text"/>	<input type="text"/>
Number of devices registered:	Satisfaction score:	Number of addresses:	Complained:
<input type="text"/>	<input type="text"/>	<input type="text"/>	<input type="text"/>
Order amount hike last year:	Coupon used:	Order count:	Day since last order:
<input type="text"/>	<input type="text"/>	<input type="text"/>	<input type="text"/>
Cashback amount:	Preferred login device:	Preferred payment mode:	Gender:
<input type="text"/>	<input type="text"/>	<input type="text"/>	<input type="text"/>
	Preferred order Category:	Marital Status:	
	<input type="text"/>	<input type="text"/>	

Confirm



NO RISK



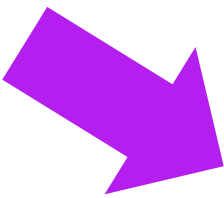

ChurnPredict

**ChurnPredict**

Tenure:	City tier:	Warehouse to home:	Hours spent on app:
<input type="text"/>	<input type="text"/>	<input type="text"/>	<input type="text"/>
Number of devices registered:	Satisfaction score:	Number of addresses:	Complained:
<input type="text"/>	<input type="text"/>	<input type="text"/>	<input type="text"/>
Order amount hike last year:	Coupon used:	Order count:	Day since last order:
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Cashback amount:	Preferred login device:	Preferred payment mode:	Gender:
<input type="text"/>	<input type="text"/>	<input type="text"/>	<input type="text"/>
	Preferred order Category:	Marital Status:	
	<input type="text"/>	<input type="text"/>	

Confirm

No churn risk



RISK


ChurnPredict

**ChurnPredict**

Tenure:	City tier:	Warehouse to home:	Hours spent on app:
<input type="text"/>	<input type="text"/>	<input type="text"/>	<input type="text"/>
Number of devices registered:	Satisfaction score:	Number of addresses:	Complained:
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<input type="text"/>	<input type="text"/>	<input type="text"/>	<input type="text"/>
	Preferred order Category:	Marital Status:	
	<input type="text"/>	<input type="text"/>	

Confirm

Churn risk detected!



THANK YOU FOR  
YOUR ATTENTION

