Week 7 submission

Project: Bank Marketing Campaign

Group Name: Evolve Data Name: Dmitry Sharukhin Email: sharuhinda@gmail.com

Country: Russia

College/Company: Finval GC Specialization: Data Science

Problem description

ABC Bank wants to sell its term deposit product to customers and before launching the product they want to develop a model which help them in understanding whether a particular customer will buy their product or not (based on customer's past interaction with bank or other Financial Institution).

Bank wants to use ML model to shortlist customer whose chances of buying the product are more so that their marketing channel (tele marketing, SMS/email marketing etc.) can focus only to those customers.

This will save resource and their time (which is directly involved in the cost (resource billing)).

The task is to create binary classifier to forecast the probability of customer's agreement to open term deposit

Business understanding

To get an answer about new product bank is going to make a series of calls. Each call has its cost and Bank wants to make only reasonable number of calls to limit corresponding costs.

The results of such calls are considered final. Situations when potential customer changes his mind (for whatever reason) after the call are not considered.

Any client can be contacted several times if it raises chances on positive result.

Some social and economic indicators might be useful (features 16-20) as they reflect general situation that influences the propensity to use banking services.

Project lifecycle along with deadline (see annex A)

Data Intake report (see annex B)

Github repo link: https://github.com/sharuhinda/bank_marketing_campaign

Project lifecycle

No.	Section / Tasks	Deadline	Completion mark
	Initiation and planning s	stage	
1.	Week 7	Dec 19, 2022	
	Report form: github repo link (1) PDF document		
1.1.	Problem description		
1.2.	Business understanding		
1.3.	Project lifecycle along with deadline		
1.4.	Data Intake Report		
	Execution stage		
2.	Week 8	Dec 26, 2022	
	Report form: github repo link (1) PDF document	,	
	Problem description		
	Data understanding		
	Data types		
	Data problems (missing values, outliers, skeweness,		
	etc.)		
	What approaches you are trying to apply on your		
	dataset to overcome problems and why?		
3.	Week 9	Jan 02, 2023	
	Report form: github repo link (1) PDF document, (2) IPYNB notebook, (3) peers review comments		
3.1.	Problem description		
3.2.	Data cleansing and transformations done on the data	Dec 28, 2022	
3.3.	Each member should code and review peers work.		
	(Review comment should be present in the github		
	repo)		
	NOTES:		
	(1) Each team member should work on different data		
	cleansing approach		
	(2) Try at least 2 techniques to clean the data for NA		
	values: (mean/median/mode/Model based approach to handle NA value/WOE)		
	(3) Try different techniques to identify and handle		
	outliers as well		
	(4) You are <u>allowed to merge the code</u> of each		
	individual and work together to get good result		
	(5) If team decide to not merge the code, then code of		
	each team member should be placed at provided URL		
	(single repository for whole team)		
4.	Week 10	Jan 09, 2023	
	Report form: github repo link (1) PDF document, (2) IPYNB notebook with EDA		

4.1.	Problem description	
4.2.	EDA performed on the data	
4.3.	Final Recommendations	
1.5.	Time Recommendations	
5.	Week 11	Jan 16, 2023
	Report form: github repo link (1) PDF document	
	Problem description	
	EDA presentation for business users	
	Last slide of EDA should be dedicated to technical	
	user which should contain recommended models for	
	this dataset	
6.	Week 12	Jan 23, 2023
	Report form: github repo link	
6.1.	Select your base model and then explore 1 model of	
	each family (Linear models, Ensemble model,	
	Boosting model, other models if you have time (like	
	stacking))	
	NOTES:	
	(1) Selected model should fit in your business	
	requirement. For example: if your business does not	
	want black box model then select only those models	
	which can be used to explain the prediction	
	(2) You are allowed to merge the code of each	
	individual and work together to get good result	
	(3) If team decide to not merge the code, then <u>upload</u>	
	the code of each team member and other deliverables	
	in the single repo and share the URL of that repo	
	Closure stage	
7.	Week 13	Jan 30, 2023
	Report form: github repo link (1) Report, (2)	
	Power point presentation	
7.1.	As it was group assignment hence go far a call with	
	your team and discuss the solution of each member	
	and select that solution which is best and is per the	
	requirement	
	NOTE:	
	(1) You are allowed to merge the code of each	
	individual and work together to get good result	

Note: All PDF reports should contain:

- Team member's details : Group Name (give a name to your group)
- Name
- Email
- Country
- College/CompanySpecialization: Data Science

Data Intake Report

Name: Bank Marketing (Campaign)

Report date: Dec 18, 2022 Internship Batch: LISUM15

Version: 1.0

Data intake by: Dmitry Sharukhin

Data intake reviewer:

Data storage location: https://archive.ics.uci.edu/ml/datasets/Bank+Marketing

Tabular data details:

Total number of observations	41 188
Total number of files	3
Total number of features	21
Base format of the file	.csv
Size of the data	6.3 MB

bank-additional-full.csv

	11.100
Total number of observations	41 188
Total number of files	1
Total number of features	age (int), no missing values
	job (str), no missing values
	marital (str), no missing values
	education (str), no missing values
	default (str), no missing values
	housing (str), no missing values
	loan (str), no missing values
	contact (str), no missing values
	month (str), no missing values
	day_of_week (str), no missing values
	duration (int), no missing values
	campaign (int), no missing values
	pdays (int), no missing values
	previous (int), no missing values
	poutcome (str), no missing values
	emp.var.rate (float), no missing values
	cons.price.idx (float), no missing values
	cons.conf.idx (float), no missing values
	euribor3m (float), no missing values
	nr.employed (float), no missing values
	y (str), no missing values – target feature
	Total: 21 features
Base format of the file	.csv (';'-separated)
Size of the data	5,8 MB

bank-additional.csv

Total number of observations	4 119
Total number of files	1

Total number of features	age (int), no missing values
	job (str), no missing values
	marital (str), no missing values
	education (str), no missing values
	default (str), no missing values
	housing (str), no missing values
	loan (str), no missing values
	contact (str), no missing values
	month (str), no missing values
	day_of_week (str), no missing values
	duration (int), no missing values
	campaign (int), no missing values
	pdays (int), no missing values
	previous (int), no missing values
	poutcome (str), no missing values
	emp.var.rate (float), no missing values
	cons.price.idx (float), no missing values
	cons.conf.idx (float), no missing values
	euribor3m (float), no missing values
	nr.employed (float), no missing values
	y (str), no missing values – target feature
	Total: 21 features
Base format of the file	.csv (';'-separated)
Size of the data	584 KB

bank-additional-names.txt

~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~	
Total number of observations	-
Total number of files	1
Total number of features	Description of features
Base format of the file	.txt
Size of the data	5 KB

- 1. The most of the data is concentrated in file bank-additional-full.csv. Therefore primary key will be the default index (row #).
- 2. Bank-additional.csv contains 10% sample from full version of the dataset. It's not indended to be used in project.
- 3. There are no obvious missing values but some values mark absence of data ('unknown', 999)
- 4. Assume that 'default' feature means the presence of the loan in default in any of the banks on the moment of contact. The same applies to 'housing' and 'loan' features

### **Proposed Approach:**

- Use only bank-additional-full.csv as data source. Use bank-additional-names.txt to clarify meanings of features.
- There's no separate test dataset so we will have to split given data to train and test datasets before performing EDA
- EDA should be performed only on train dataset