Introduction

This assignment aims to give you an idea of applying EDA in a real business scenario. In this assignment, apart from applying the techniques that you have learnt in the EDA module, you will also develop a basic understanding of risk analytics in banking and financial services and understand how data is used to minimise the risk of losing money while lending to customers.

Business Understanding

The loan providing companies find it hard to give loans to the people due to their insufficient or non-existent credit history. Because of that, some consumers use it to their advantage by becoming a defaulter. Suppose you work for a consumer finance company which specialises in lending various types of loans to urban customers. You have to use EDA to analyse the patterns present in the data. This will ensure that the applicants capable of repaying the loan are not rejected.

When the company receives a loan application, the company has to decide for loan approval based on the applicant's profile. Two types of risks are associated with the bank's decision:

If the applicant is likely to repay the loan, then not approving the loan results in a loss of business to the company

If the applicant is not likely to repay the loan, i.e. he/she is likely to default, then approving the loan may lead to a financial loss for the company.

The data given below contains the information about the loan application at the time of applying for the loan. It contains two types of scenarios:

The client with payment difficulties: he/she had late payment more than X days on at least one of the first Y instalments of the loan in our sample,

All other cases: All other cases when the payment is paid on time.

When a client applies for a loan, there are four types of decisions that could be taken by the client/company):

Approved: The Company has approved loan Application

Cancelled: The client cancelled the application sometime during approval.

Either the client changed her/his mind about the loan or in some cases due to a higher risk of the client, he received worse pricing which he did not want.

Refused: The company had rejected the loan (because the client does not meet their requirements etc.).

Unused offer: Loan has been cancelled by the client but at different stages of the process.

In this case study, you will use EDA to understand how consumer attributes and loan attributes influence the tendency to default.

Business Objectives

This case study aims to identify patterns which indicate if a client has difficulty paying their instalments which may be used for taking actions such as denying the loan, reducing the amount of loan, lending (to risky applicants) at a higher interest rate, etc. This will ensure that the consumers capable of repaying the loan are not rejected. Identification of such applicants using EDA is the aim of this case study.

In other words, the company wants to understand the driving factors (or driver variables) behind loan default, i.e. the variables which are strong indicators of default. The company can utilise this knowledge for its portfolio and risk assessment.

To develop your understanding of the domain, you are advised to
independently research a little about risk analytics - understanding the types
of variables and their significance should be enough.
Data Understanding
Download the dataset from below.
Dataset
Download
This dataset has 3 files as explained below:
1. 'application_data.csv' contains all the information of the client at the time o application.
The data is about whether a client has payment difficulties.
2. 'previous_application.csv' contains information about the client's previous
loan data. It contains the data on whether the previous application had been
Approved, Cancelled, Refused or Unused offer.

3. 'columns_description.csv' is data dictionary which describes the meaning of the variables.

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Results Expected by Learners

Present the overall approach of the analysis in a presentation. Mention the problem statement and the analysis approach briefly.

Identify the missing data and use appropriate method to deal with it. (Remove columns/or replace it with an appropriate value)

Hint: Note that in EDA, since it is not necessary to replace the missing value, but if you have to replace the missing value, what should be the approach. Clearly mention the approach.

Identify if there are outliers in the dataset. Also, mention why do you think it is an outlier. Again, remember that for this exercise, it is not necessary to remove any data points.

Identify if there is data imbalance in the data. Find the ratio of data imbalance.

Hint: How will you analyse the data in case of data imbalance? You can plot more than one type of plot to analyse the different aspects due to data imbalance. For example, you can choose your own scale for the graphs, i.e. one can plot in terms of percentage or absolute value. Do this analysis for the 'Target variable' in the

dataset (**clients with payment difficulties** and **all other cases**). Use a mix of univariate and bivariate analysis etc.

Hint: Since there are a lot of columns, you can run your analysis in loops for the appropriate columns and find the insights.

Explain the results of univariate, segmented univariate, bivariate analysis, etc. in business terms.

Find the top 10 correlation for the Client with payment difficulties and all other cases (Target variable). Note that you have to find the top correlation by segmenting the data frame w.r.t to the target variable and then find the top correlation for each of the segmented data and find if any insight is there. Say, there are 5+1(target) variables in a dataset: Var1, Var2, Var3, Var4, Var5, Target. And if you have to find top 3 correlation, it can be: Var1 & Var2, Var2 & Var3, Var1 & Var3. Target variable will not feature in this correlation as it is a categorical variable and not a continuous variable which is increasing or decreasing. Include visualisations and summarise the most important results in the presentation. You are free to choose the graphs which explain the numerical/categorical variables. Insights should explain why the variable is important for differentiating the clients with payment difficulties with all other cases.

You need to submit one/two Ipython notebook which clearly explains the thought process behind your analysis (either in comments of markdown text), code and relevant plots. The presentation file needs to be in PDF format and should contain the points discussed above with the necessary visualisations. Also, all the visualisations and plots must be done in Python(should be present in the Ipython notebook), though they may be recreated in Tableau for better aesthetics in the PPT file.

Evaluation Rubrics

Criteria	Meets expectations	Does not meet expectations
Data understanding		
(20%)		
	All data quality issues are	Data quality issues are
	correctly identified and	overlooked or are not identified
	reported.	correctly such as missing
		values, outliers and other data
		quality issues.
	Wherever required, the	
	meanings of the variables are	
	correctly interpreted and	

	written either in the	The variables are interpreted
	comments or text.	incorrectly or the meaning of
		variables is not mentioned.
Data Cleaning and		
Manipulation	Data quality issues are	
(10%)	Data quality issues are	
	addressed in the right way	Data quality issues are not
	(missing value imputation	addressed correctly.
	analysis and other kinds of	
	data redundancies, etc.).	
	If applicable, data is converted to a suitable and convenient format to work with using the right methods. Manipulation of strings and dates is done correctly wherever required	The variables are not converted to an appropriate format for analysis.

String and date manipulation is
not done correctly or is done
using complex methods

Data analysis		
(50%)	The right problem is solved	The analyses do not address the
	The right problem is solved	The analyses do not address the
	which is coherent with the	right problem or deviate from
	needs of the business. The	the business objectives. The
	analysis has a clear structure	analysis lacks a clear structure
	and the flow is easy to	and is not easy to follow.
	understand.	
	Univariate and segmented	The univariate and bivariate
	univariate analysis is done	analysis is not performed in
	correctly and appropriate	sufficient detail and thus some
	realistic assumptions are	crucial insights are missed out.
	made wherever required. The	
	analyses successfully	identify enough important
	identify at least the 5	driver variables.
	important driver variables	dirver variables.
	(i.e. variables which are	
	strong indicators of default).	New metrics are not derived
		wherever appropriate. The

Business-driven, type-driven explanation for creating the and data-driven metrics are created for the important variables and utilised for analysis. The explanation for creating the derived metrics

derived metrics is either not mentioned or the metrics are not reasonable.

is mentioned and is

reasonable.

Derived metrics are not analysed correctly/are insufficiently utilised.

Bivariate analysis is performed correctly and is able to identify the important combinations of driver variables. The combinations of variables are chosen such that they make business or analytical sense.

Important insights are not mentioned in the report or the Python file. Relevant plots are not created. The choice of plots is not ideal and the plots are either difficult to interpret or

The most useful insights are lack clarity or neatness. explained correctly in the comments.

Relevant insights are not clearly presented by the plots. The axes

not labelled correctly/neatly.

and important data points are

Appropriate plots are created to present the results of the analysis. The choice of plots for respective cases is correct. The plots should clearly present the relevant insights and should be easy to read. The axes and important data points are labelled correctly.

Presentation and		
	The presentation has a clear structure, is not too long, and explains the most important results concisely in simple language.	
	The recommendations to solve the problems are realistic, actionable and coherent with the analysis.	The recommendations to solve the problems are either unrealistic, non-actionable or incoherent with the analysis.
	If any assumptions are made, they are stated clearly.	Contains unnecessary details or lacks the important ones.

		Assumptions made, if any, are
		not stated clearly.
Conciseness and		
readability of the	The code is concise and	
code (10%)	The code is concise and	
	syntactically correct.	Long and complex code used
	Wherever appropriate,	instead of shorter built-in
	built-in functions and	functions.
	standard libraries are used	
	instead of writing long code	
	(if-else statements, for loops,	Custom functions are not used
	etc.).	to perform repetitive tasks
		resulting in the same piece of
		code being repeated multiple
	Custom functions are used to	times.
	perform repetitive tasks.	
		Code readability is poor because
	The code is readable with	of vaguely named variables or
	appropriately named	lack of comments wherever
	variables and detailed	necessary.

comments are written	
wherever necessary.	

For submissions obtained within 1 week of the deadline, there will be a 30% penalty. Submissions beyond 1 week of the deadline will be provided only feedback and score zero marks (100% penalty).

You must go through these guidelines-

- 1. Make sure you have not made any changes to the original dataset provided to you. Your Python code should work on the dataset given to you as part of the problem statement. You are not allowed to make modifications in data set using excel and then use it in your Python code. Entire data processing must be done in Python only. During grading we will be running your code on the dataset provided by us, in case your code gives errors with that, then marks will be deducted accordingly.
- 2. All penalties are automatically applied by the system based on time of submission. Hence, submissions that are late, even by a second, will attract penalties.

For e.g.- If the deadline is 2nd August 2022, 11:59:00 PM IST, the submissions at 2nd August 2022, 11:59:01 PM IST will attract a penalty of 30%. Hence we recommend that assignments are submitted at least 30 minutes before the deadline to avoid any last minute issues.

Also, note that all the deadlines are in IST (UTC +5.5), hence, if you are in a different time zone, then your deadline may vary according to local time. For eg - If you are in London and following BST (British Summer Time) which is UTC +1 then deadline for you in local time would be 7:29:00 PM BST when the deadline in India is 11:59:00 PM IST.

3. If you are 100% sure that you will not need to make any more changes in the assignment, **click "Submit for Grading"** to submit your assignment for evaluation.

Important note: You must click "Submit for Grading" as the assignment will not be automatically submitted at the deadline.

Here are the steps that you must follow during submitting any assignment-

- 1. Collect all the files (if there are multiple files) and compress them together.
- 2. Try to upload this compressed file latest by 11:30 PM

- 3. Download your submission and check that you have included all the required files.
- 4. Check that none of the files or the zip is corrupt. If it is found to be corrupt during grading, you will NOT be allowed to re-submit.

Credit EDA Assignment

Submission details

For submissions obtained within one week after the deadline, there will be a 30% penalty. Submissions beyond one week after the deadline will be awarded 0 marks.

You have to upload the files as one zip file.

The zip file should contain the main Python file, and one presentation doc (in PDF format). Submit the zip file below.

We strongly recommend you to submit atleast 30 minutes before your deadline

Extended Module Deadline

08 Nov '22

11:59 PM (IST)

Cancel Extension

1 out of 4 extensions requests are