

Libraries Requirements

Flask==2.0.1

Flask-Cors==3.0.10

Jinja2==3.0.1

numpy==1.19.5

pandas==1.1.5

scikit-learn==0.24.2

sklearn==0.0

seaborn==0.0

warning=0.0

Architecture

Contents

		Abstract	4
		Introduction	4
		Why this Architecture Design Documentation?	4
1		Architecture	4
2		Architecture Design	5
	2.1	Data gathering from main source	5
	2.2	Data Description	5
	2.3	Data-Preprocessing	5
	2.4	Modelling	5
	2.5	UI Integration	6
	2.6	Data From User	6
	2.7	Data Validation	6
	2.8	Rendering the results	6
	2.9	Deployment	6

Architecture

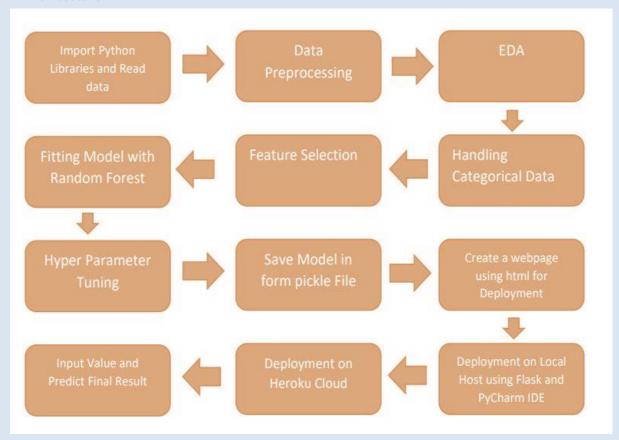
Abstract

Introduction

Why this Architecture Design documentation?

The main objective of the Architecture design documentation is to provide the internal logic understanding of the flight fare prediction code. The Architecture design documentation is designed in such a way that the programmer can directly code after reading each module description in the documentation.

1 Architecture



Architecture

2 Architecture design

This project is to create an interface for the user to predict if the person would default on credit card next month, in addition to this, in need of getting the real time project experience we are importing the gathered data into our own database and then start the project from the scratch.

2.1 Data gathering from main source

The data for the current project is being gathered from Kaggle dataset, the link to the data is:https://www.kaggle.com/datasets/uciml/default-of-credit-card-clients-dataset

2.2 Data description

There are about 30k+ records of credit card default with 25 columns consisting of information like payment balance, payment of September to April etc

2.3 Data pre-processing

Steps performed in pre-processing are:

- Data set is divided in 70:30 ratio for train and test respectively.
- ID column was dropped as its unnecessary for our modeling.
- We have also dropped columns 'PAY_2'. 'PAY_3', 'PAY_4', 'Pay_5', 'PAY_6', 'BILL_AMT2', 'BILL_AMT3', 'BILL_AMT4', 'BILL_AMT5', 'BILL_AMT6' as they possess multi-collinearity with columns 'PAY 0' and 'BILL AMT0' respectively
- The attribute name 'default.payment.next.month' was coverted to 'Payment_default' for naming convenience.
- Pay_0:No consumption of credit card=-2,Pay duly(paid on time)=-1,payment delay for one mouth=1, payment delay for two months=2,payment delay for nine months and above=-9.
- No Null values in dataset

2.4 Modelling

The pre-processed data is then visualized and all the required insights are being drawn. Although from the drawn insights, the data is randomly spread but still modelling is performed with different machine learning algorithms to make sure we cover all the possibilities. And finally, as expected random forest regression performed well and further hyperparameter tuning is done to increase the model's accuracy.

2.5 UI integration

HTML file is being created and are being integrated with the created machine learning model. All the required files are then integrated to the app.py file and tested locally. Note: We have not make the HTML File.

2.6 Data from user

The data from the user is retrieved from the created HTML web page.

2.7 Data validation

The data provided by the user is then being processed by app.py file and validated. The validated data is then sent for the prediction.

2.8 Rendering the results

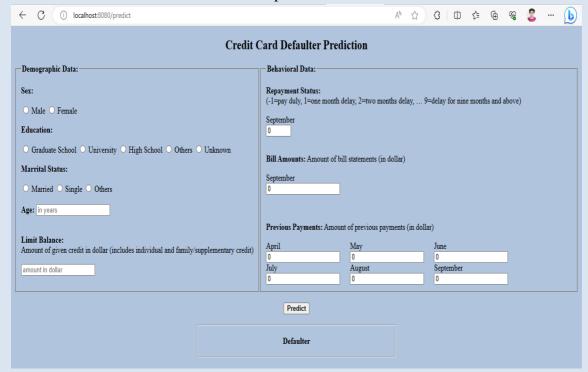
The data sent for the prediction is then rendered to the web page.

2.9 Deployment

The tested model is then deployed to NETLIFY. So, users can access the project from any internet devices.

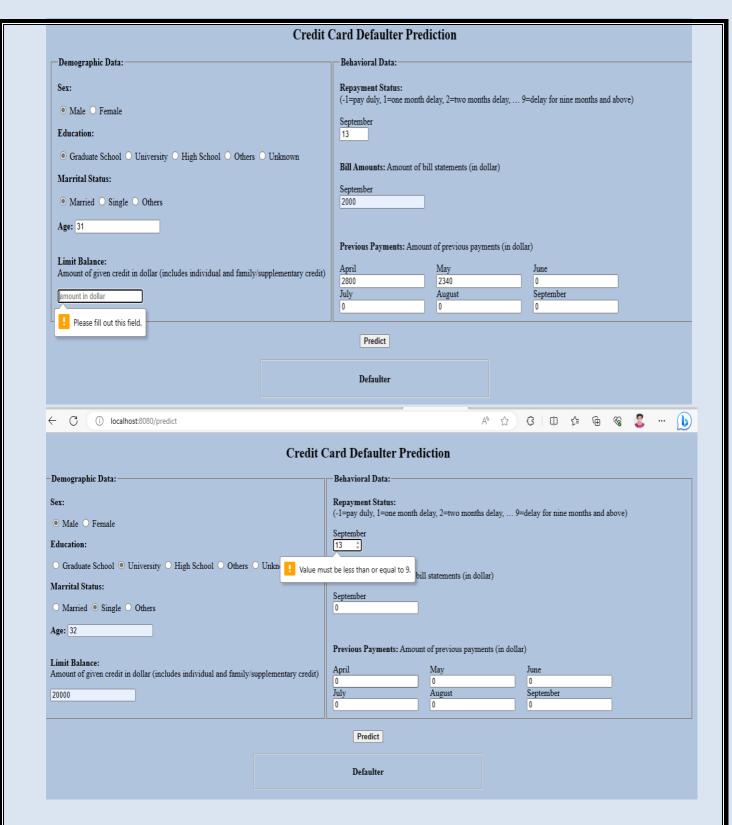
ScreenShot of the App Interface

- 1. The first page displays the pop up window where the user has to provide specific details such as Gender, Repayment status, Bill amount etc
 - All the information will be used to predict the results.

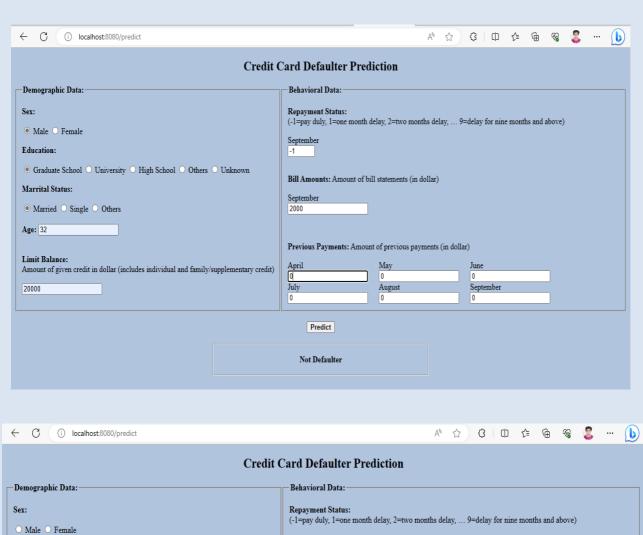


2. Incase you miss out one input the webpage will alert you for the same or if the entered value is more than

the set value it will alert you for the same.



- 2. After entering all the data we will get the predictions at the bottom page, i.e if theborrower will default or not.
 - a) In this case the Borrower will not be a Defaulter in next month.



Credit Card Defaulter Prediction							
Demographic Data: Behavioral Data:							
Sex:	Repayment Status: (-1=pay duly, 1=one month delay, 2=two months delay, 9=delay for nine months and above) September						
○ Male ○ Female							
Education:	0						
○ Graduate School ○ University ○ High School ○ Others ○ Unknown Bill Amounts: Amount of bill statements (in dollar)							
Marrital Status:	September						
○ Married ○ Single ○ Others	0						
Age: in years							
Limit Balance:	Previous Payments: Amount of previous payments (in dollar)						
Amount of given credit in dollar (includes individual and family/supplementary credit)	April May June 0 0 0						
amount in dollar	July August September 0 0 0						
Predict							
Not Defaulter							

b) In this case the Borrower will be a Defaulter in next month.

← (j (i) localhost:8080/predict		A" A) 4 D 4 G G W W W		
← O ① localhost:8080/predict A 公 公 ① localhost:8080/predict Card Defaulter Prediction				
Demographic Data:	Behavioral Data:			
Sex:	Repayment Status: (-1=pay duly, 1=one month delay, 2=two	o months delay, 9=delay for nine months and above)		
Male O Female	September			
Education:	3			
	own Bill Amounts: Amount of bill statement	ts (in dollar)		
Marrital Status:	September			
● Married ○ Single ○ Others	20000			
Age: 32				
Limit Balance:	Previous Payments: Amount of previous			
Amount of given credit in dollar (includes individual and family/supplement)	4000	June 0		
20000	July August 0	September 0		
	Predict			
	Not Defaulter			

