

# **BANK LOAN PREDICTION**



PRESENTED BY:

TEAM 1

**MENTOR:** 

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DATE: 15/03/2020

## **Business Problem:**



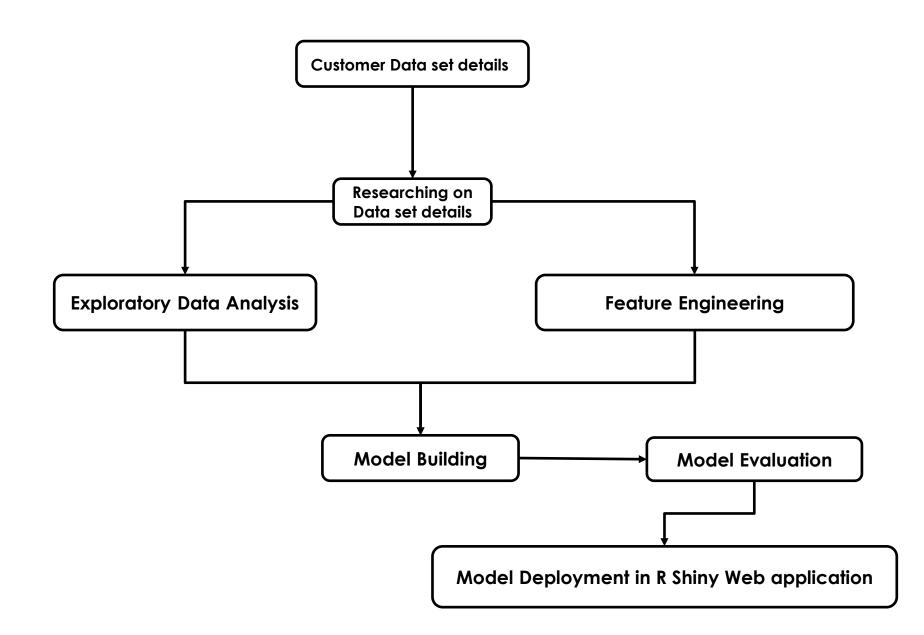
The customers have taken loans from the banks to start up business if there will clear the loan amount to the banks or not.

#### Objective:

To predict the Whether the customer will fall under default or not.



# **Project Architecture / Project Flow**





# **Exploratory Data Analysis (EDA) and Feature Engineering**

### Data set details

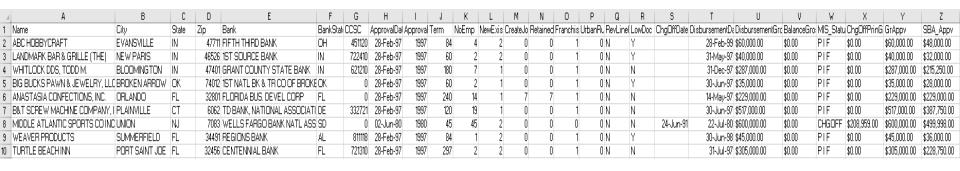


The given dataset is a "Bank\_final.csv" .The dataset is having 26 input variables where the dependent variable is to predict customer is a default or not. This dataset is having 150000 observations.

Date ranges

Data set Details	1 Week
EDA	1 ½ week
Model Building	1 Week
Model Evaluation	
Feedback	1½ week
Deployment	10 days
Final presentation	1 day

Missing values (NA's): 11,0951 (2.73 % Missing values in a Percentage in the whole data frame)



# **Exploratory Data Analysis (EDA)**



- MIS\_Status variable as the dependent variable. Where CHGOFF is default on loan i.e. Charged off and PIF is paid in full i.e. no default and CHGOFF having 26% and PIF having 74% in the whole dataset.
- The maximum amount of customers loan applied from the Bank Of America NATL ASSOC, state CA (California) and the city Los Angeles.
- Basic summary of the variables given below.

-	ïName	City	State	Zip		Bank	BankState	
SUBWAY	: 78	LOS ANGELES: 1902	CA :19314	Min. : 0	BANK OF AMERICA NATL ASSOC	:28023	NC :26847	
SCHLOTZSKY'S DEL:	I : 39	NEW YORK : 1583	NY :12236	1st Qu.:20854	CITIZENS BANK NATL ASSOC	:13052	RI :14731	
QUIZNO'S CLASSIC	SUBS: 37	MIAMI : 1515	TX : 9822	Median :48053	CAPITAL ONE NATL ASSOC	:10611	IL :12672	
DOMINO'S PIZZA	: 36	CHICAGO : 1401	FL : 9269	Mean :49849	JPMORGAN CHASE BANK NATL AS	SSOC:10381	CA :12449	
DAIRY QUEEN	: 32	HOUSTON : 1351	OH : 7560	3rd Qu.:80003	WELLS FARGO BANK NATL ASSOC	: 6373	VA :11564	
(Other)	:149773	(Other) :142246	(Other):91796	Max. :99999	(Other)	:81412	(Other):71588	
NA's	: 4	NA's : 1	NA's : 2		NA's	: 147	NA's : 148	
CCSC	ApprovalDa	ite ApprovalFY	Term	NoEmp	NewExist Create	lob	RetainedJob	
Min. : 0	30-Sep-97:	466 Min. :1962	Min. : 0.00	Min. : 0.00	0 Min. :0.00 Min. :	0.000 M	in. : 0.000	
1st Qu.:236118	17-Mar-06:	425 1st Qu.:1998	1st Qu.: 57.00	1st Qu.: 2.00	0 1st Qu.:1.00 1st Qu.:	0.000 1	st Qu.: 0.000	
Median :447110	24-Mar-06:	422 Median :2005	Median : 84.00	Median: 4.00	0 Median:1.00 Median:	0.000 M	edian : 1.000	
Mean :401568	01-Apr-97:	418 Mean :2002	Mean : 93.01	Mean : 9.31	4 Mean :1.32 Mean :	1.278 M	ean : 3.686	
3rd Qu.:561612	31-Mar-06:	412 3rd Qu.:2006	3rd Qu.: 84.00	3rd Qu.: 8.00	0 3rd Qu.:2.00 3rd Qu.:	0.000 3	rd Qu.: 4.000	
Max. :928120	18-Apr-05:	402 Max. :2007	Max. :480.00	Max. :9999.00	0 Max. :2.00 Max. :3	3000.000 M	ax. :9500.000	
	(Other) :147	454						

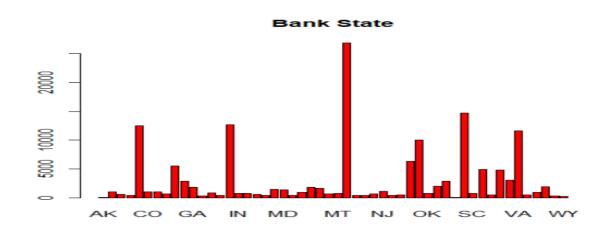


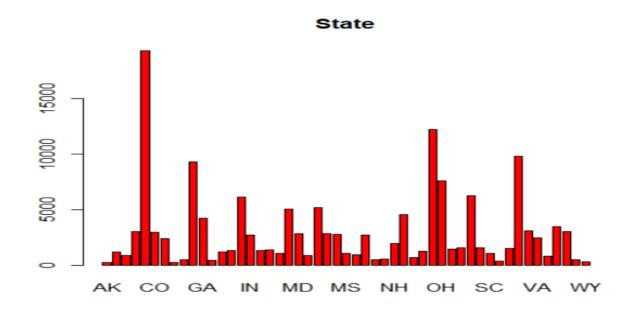
FranchiseCode	UrbanRu	ıral	RevLine	cr	LowDoc	Chg0ff	Date	Disburse	mentDate	Disburse	mentGross
Min. : 0	Min. :0	0.000	N :7	1611	1: 1	13-Mar-10:	201	31-May-06	: 6220	\$50,000.00	: 10214
1st Qu.: 0				9881	C: 83	30-Jan-10:		31-Mar-06		,	: 6992
Median: 1	Median :1	L.0000	0 :2	3659	N:137871	20-Feb-10:	175	30-Apr-06	5848	\$100,000.00	: 6978
Mean : 1656	Mean :0	).7679	T :	4819	Y: 12044	06-Feb-10:	158	30-Jun-06	: 5263	\$10,000.00	: 5612
3rd Qu.: 1	3rd Qu.:1	L.0000	1 :	3		06-Mar-10:	148	28-Feb-06	: 4902	\$35,000.00	: 3092
Max. :91999	Max. :2	2.0000	(Other):	3			39601	(Other)		\$20,000.00	: 2748
_			NA's :	23			109533	NA's	: 225	(Other)	:114363
Balan	ceGross	MIS_	Status		Chg0f	fPrinGr		GrAp	p∨	SBA	A_App∨
\$0.00	:149997		F: 39008		0.00	:109702	\$50,		17500	\$25,000.00	
\$12,750.00	: 1	ΡΙF	:11012	3 \$	10,000.00	: 956	\$25,	000.00 :	13268	\$5,000.00	:11900
\$827,875.00	: 1	NA's	: 868	3 \$	50,000.00	: 795	\$10,	000.00 :	12588	\$12,500.00	0:10836
				\$	25,000.00	: 437	\$100	,000.00:	11123	\$50,000.00	0:7764
				\$	100,000.00	: 400	\$20,	000.00 :	5451	\$10,000.00	0:4666
				9	35,000.00	: 266	\$35.	000.00 :	5067	\$17,500.00	0:4507
					(Other)	: 37443	(0th	er) :	85002	(Other)	:95263



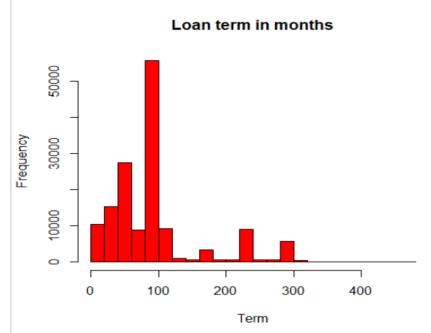


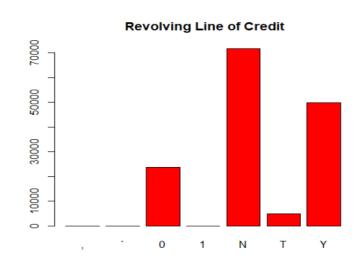
#### **Basic Plots:**

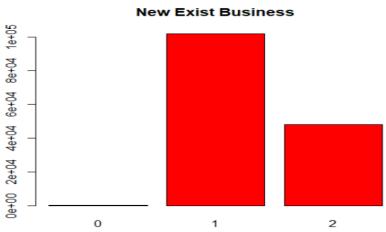


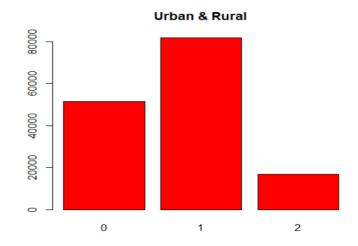




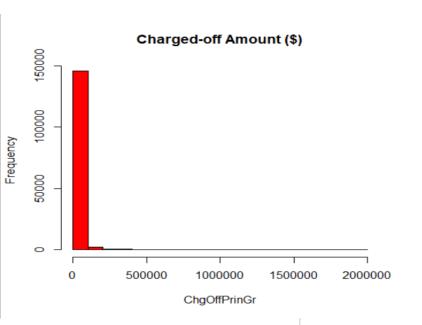


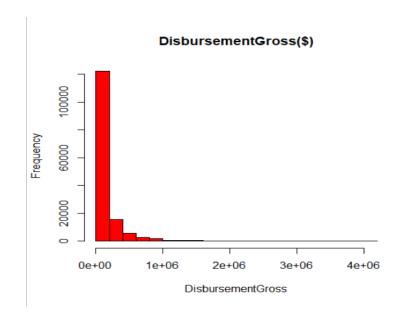


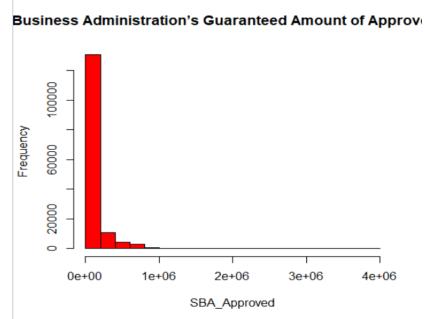




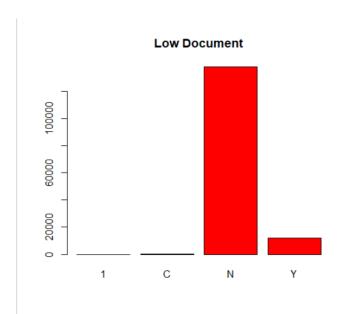


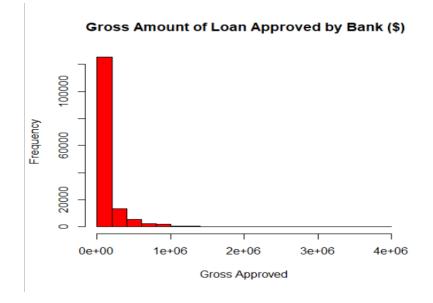


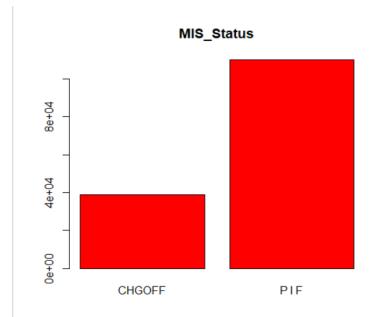












## **Feature Engineering**



- Few variables such as Name, City, State, Zip, Bank, Bank State, CCSC, Approval Date, Approval FY, ChgOff Date, Disbursement Date, Balance Gross were eliminated due to considering Weight of Evidence and Information Value they variables were irrelevant for prediction.
- Variables such as Gross Approval, SBA Approval, Charge off amount, Disbursement Gross were changed from currency format to numeric format.
- If there having any Franchise considered as (0) and No Franchise code as (1).
- There having any existing business considered as (1) and New business as (2).
- In MIS Status charge off considered as (0) and paid in full as (1).
- In low document process considered as LowDoc (1) and No lowDoc as (0).
- In there Revolving Line of Credit considered as yes (1) and No as (0).
- Out of dictionary values of few variables were changed accordingly.
- To avoid overfitting ChgOffPrinGr variable was eliminated.



# **Model Building**

## **Template for Model results presentation**



#### Model – Logistic Regression

Data set details

After eliminating some variable in the dataset there having 14 input variables 150000 observations.

Data Partition details

Randomly dataset divided into 70% training dataset and 30% testing dataset.

#### Algorithms

**Logistic Regression** is a classification **algorithm**. It is used to predict a binary outcome (1 / 0, Yes / No, True / False) given a set of independent variables.

Algorithm details and configuration model <- glm(train\$MIS\_Status~.,data=train,family = "binomial")

glm = function for logistic regression.

Y = MIS\_Status variable.

Data = Training dataset.

Family = Binomial object are used.

#### **Confusion Matrix Details:**

Accuracy:: 0.8314 (83.14%) 95% CI: (0.8279, 0.8348)

Kappa : 0.5288 Sensitivity : 0.9229 Specificity : 0.5699

Pos Pred Value : 0.8598

Neg Pred Value : 0.7211

Prevalence : 0.7408

Detection Rate : 0.6837

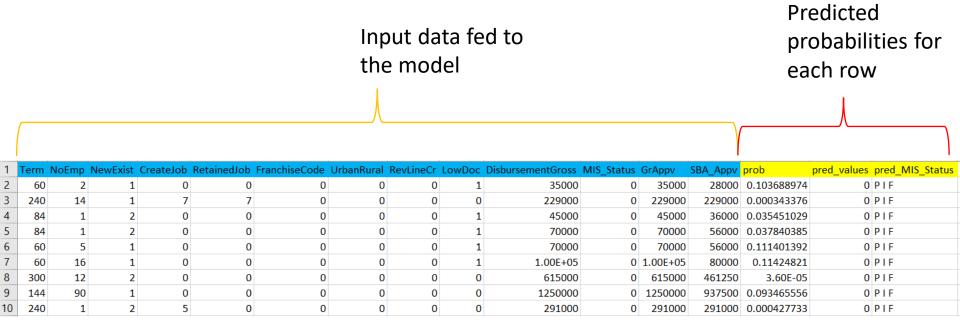
Detection Prevalence: 0.7951

Balanced Accuracy : 0.7464

## **Model Predictions**



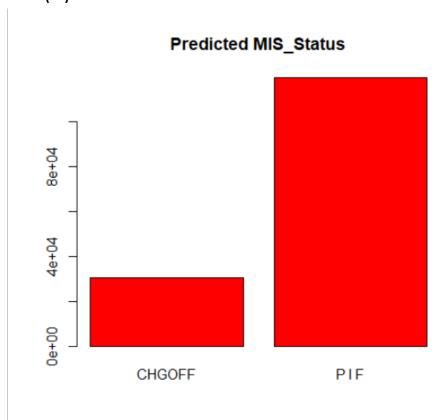
- The model builds with train data and after that predicted with test data to know the performance of the model.
- Other models were tried but their results were not better.



## **Model Results**



Model generated the output for the customers loan data set in which 79.51% customers are genuine(0)non default paid in full(P I F) and 20.48% Chargeoff (CHGOFF) are default (1) customers.



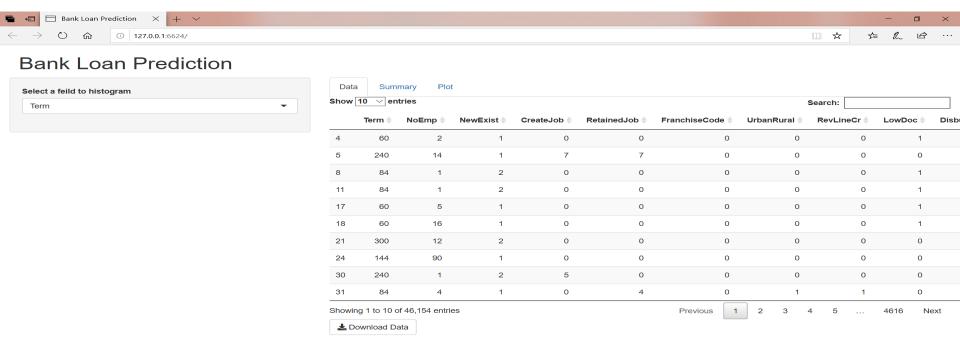


# **Model Deployment using R shiny**

#### **Model Deployment using R shiny**



- The model deployment was accomplished through Rshiny.
- We are using three tab panel, they are –
- 1. Data
- 2. Summary
- 3. Plot
- The bank loan customer data were invoked into Data tab panel which shown below. Their display total
  dataset and at the end there having a Download Data button.





• Click on Download Data button they're downloaded only predicted MIS Status in with CVS format.

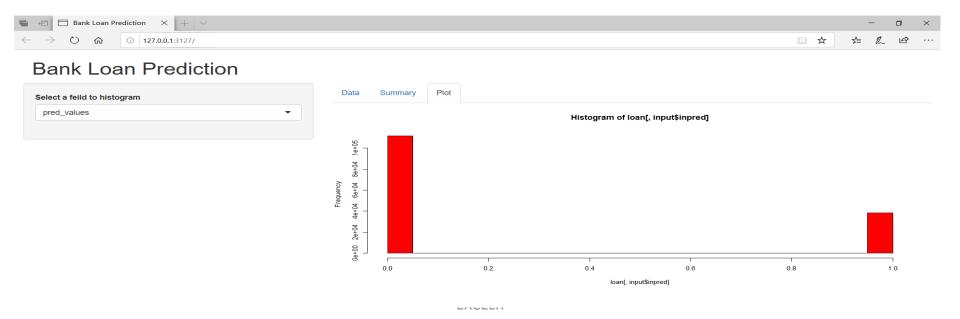
index	pred_MIS_Status
1	PIF
2	PIF
3	PIF
4	PIF
5	PIF
6	PIF
7	PIF
8	PIF
9	PIF
10	PIF
11	PIF
12	PIF

• The customer bank loan data summary were shown in second tab panel which given below.



Data	Summar	y Plot														R	aising .	Excellen
Te	erm	NoEmp		New	Exist	Cre	ateJo	b	Reta	ainedJ	ob	Franc	chiseCo	de	Urba	nRural	Re	vLineCr
Min.	: 0.00	Min. :	0.000	Min.	:1.00	Min.	:	0.000	Min.	:	0.000	Min.	:0.0	000	Min.	:0.000	Min.	:0.000
1st Qu.	: 57.00	1st Qu.:	2.000	<b>1</b> st Qu	.:1.00	1st Q	u.:	0.000	1st (	Qu.:	0.000	1st (	Qu.:0.0	000	1st Qu	.:0.000	1st	Qu.:0.000
Median	: 84.00	Median :	4.000	Median	:1.00	Media	n :	0.000	Media	an :	1.000	Media	an :0.0	000	Median	:1.0000	Medi	an :0.000
Mean	: 93.01	Mean :	9.314	Mean	:1.32	Mean	:	1.278	Mean	:	3.686	Mean	:0.4	238	Mean	:0.7679	9 Mear	:0.332
3rd Qu.	: 84.00	3rd Qu.:	8.000	3rd Qu	.:2.00	3rd Q	u.:	0.000	3rd (	Qu.:	4.000	3rd (	Qu.:1.0	000	3rd Qu	.:1.0000	3rd	Qu.:1.000
Max.	:480.00	Max. :9	999.000	Max.	:2.00	Max.	:30	00.000	Max.	:95	00.000	Max.	:1.0	000	Max.	:2.0000	Max.	:1.000
Low	Doc	Disburse	mentGross	MIS_St	atus Cl	ngOffPr	inGr		GrAppy	<i>y</i>	SE	BA_Appv	/		prob		pred_v	alues
Min.	:0.00000	Min. :	0	0:1109	91 Mi	n. :		0 Min	. :	200	Min.	:	100	Min.	:0.0	00000	Min.	:0.0000
1st Qu.	:0.00000	1st Qu.:	30000	1: 390	08 1st	t Qu.:		0 1st	Qu.:	25000	1st (	Qu.: 1	12500	1st (	Qu.:0.0	02912	1st Qu.	:0.0000
Median	:0.00000	Median :	65500		Med	dian :		0 Med	ian :	50000	Media	an : 2	25000	Media	an :0.0	18936	Median	:0.0000
Mean	:0.08029	Mean :	141294		Mea	an :	1505	7 Mea	n : :	127692	Mean	: 9	92500	Mean	:0.2	60336	Mean	:0.2551
3rd Qu.	:0.00000	3rd Qu.:	150000		3r	d Qu.:	600	3 3rd	Qu.: 3	120000	3rd (	2u.: 8	30000	3rd (	Qu.:0.5	84585	3rd Qu.	:1.0000
Max.	:1.00000	Max. :	4029520		Max	x. :1	99999	9 Max	. :40	900000	Max.	:400	90000	Max.	:1.0	00000	Max.	:1.0000
default	_pred																	
Length:	149999																	
Class :	character																	
Mode :	character																	

 Their can plot histogram with selecting each variable in the customer bank loan data and it was shown in the third tab panel.





# **Challenges faced?**

- Understanding the dataset it takes a lot of time for each variable.
- Selecting the right algorithm with the best results and it plays a key role in the project.
- New to model deployment part with the R shiny web app is too difficult to deployment process.

# How did you overcome?

- Researching on more part to understand the dataset of each variable.
- Try to build model with many algorithms, but the logistic regression model having the best performance.
- New to this part of model deployment with the R shiny web app so it's taken more time to deploy and help of mentor and online tutorial deployment part has been completed.



## **Conclusion**

EDA process was performed on the given bank full dataset of csv format. Some variables were dropped based on the Weight of Evidence and Information Value. The dataset was split into the train and test datasets and the model builds with logistic regression with train dataset and prediction model were developed with test dataset to predict the customers loan default status. The logistic regression model was selected based on accuracy, sensitivity and specificity which has better than other models. The Model deployment was achieved through R shiny.



# Thank you