# Prediction Model on a Revolving Credit Line

**Group Name: GROUP-6** 

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#### **Business Problem:**



- Revolving Credit means you're borrowing against a Line of Credit.
- > Users can borrow the amount of credit allowed to use each month known as "Credit line or Credit Limit".
- > Similar to a **Credit card** with the only difference being "Lesser Interest Rates and Secured Business Assets".
- At the end of each statement period, a Bill gets generated.

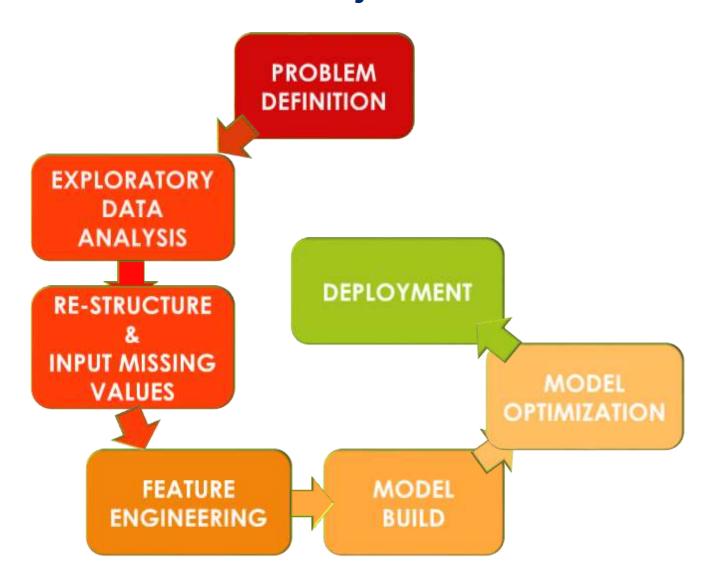
  If not paid fully, the balance is carried over, or revolved over to the next month along with Interest incurred on the remaining balance.
- >As you pay down the balance, more of your credit line becomes available.

#### **Objective:**

"To predict the revolving balance maintained by each customer to derive market strategies for a bank or investment firm".



### **Project Architecture / Project Flow**





## **Exploratory Data Analysis (EDA)**

"Exploratory data analysis can never be the whole story, but nothing else can serve as the foundation stone" — John Tukey

#### **Steps for EDA:**

- Describe Dataset: Rows/Columns, Missing Data, Data Types, Preview
- Clean corrupted data; Handle missing data, Invalid data types, Incorrect values.
- ➤ Visualize data distributions; Bar charts, Histograms, Box plots.
- ➤ Calculate and Visualize Correlations (Relationships) between variables; Heat map.

## **Data set details**



Dataset info		Variables types	
Number of variables	34	Numeric	21
Number of observations	887379	Categorical	13
Total Missing (%)	9.9%	Boolean	0
Total size in memory	230.2 MiB	Date	0
Average record size in memory	272.0 B	Text (Unique)	0
		Rejected	0
		Unsupported	0

	member_id	loan_amnt	terms	batch_ID	Rate_of_intrst	grade	sub_grade	Emp_designation	Experience	home_ownership	 recoveries	collection
0	58189336	14350	36 months		19.19	Е	E3	clerk	9 years	OWN	 0.0	
1	70011223	4800	36 months	BAT1586599	10.99	В	B4	Human Resources Specialist	< 1 year	MORTGAGE	 0.0	
2	70255675	10000	36 months	BAT1586599	7.26	Α	A4	Driver	2 years	OWN	 0.0	
3	1893936	15000	36 months	BAT4808022	19.72	D	D5	Us office of Personnel Management	10+ years	RENT	 0.0	
4	7652106	16000	36 months	BAT2833642	10.64	В	B2	LAUSD- HOLLYWOOD HIGH SCHOOL	10+ years	RENT	 0.0	
5	10247268	15000	36 months	BAT2575549	8.90	Α	A5	Design Consultant	2 years	MORTGAGE	 0.0	

## **Exploratory Data Analysis (EDA)**



#### Numerical Type

mths_since_last_record	0.845553
mths_since_last_major_derog	0.750160
mths_since_last_delinq	0.511971
tot_curr_bal	0.079195
tot_colle_amt	0.079195
collections_12_mths_ex_med	0.000163
inq_last_6mths	0.000033
acc_now_delinq	0.000033
delinq_2yrs	0.000033
total_credits	0.000033
pub_rec	0.000033
numb_credit	0.000033
annual_inc	0.000005

MALUES???

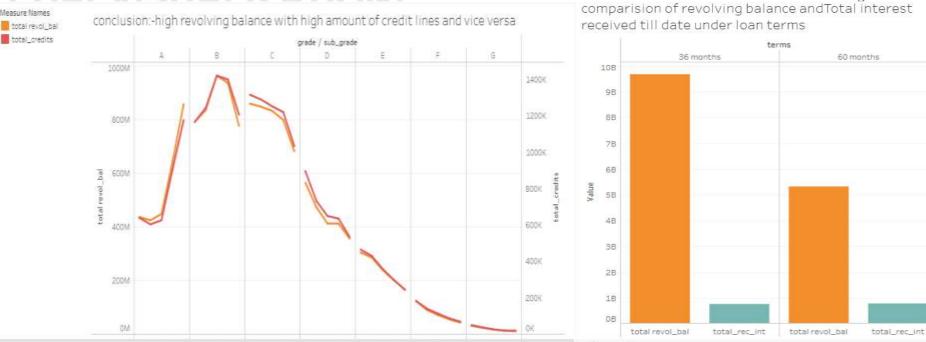
#### Categorical Type

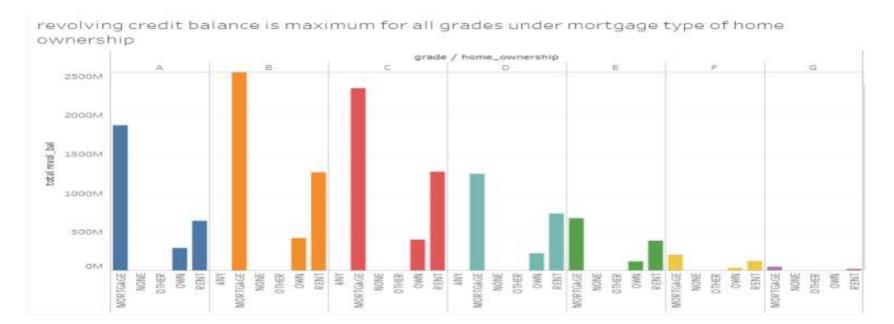
Emp_designation 0	.057993
Experience 0	.050514

- > 36 months in 'terms' column has 70% data compared to 60 months.
- > 33% of the customers have over 10 years of Experience.
- > 50% of 'home\_ownership' have MORTGAGED while 40% of them are on RENT.
- ➤ Almost 60% of loans taken cater the *purpose* of 'debt\_consolidation'.
- > 15% (MAX) of customers are based out of California State while only 12 users(MIN) are from Idaho state.
- ➤ 80% of the customers **DON'T HAVE** delinquent accounts for a span of 2 years while **56%** customers **DON'T HAVE** delinquent accounts since 6 months.
- **➢Almost 80%** of 'tot colle amt '(total collection amount ever owed) is **ZERO** dollars.
- ➤ 'mths\_since\_last\_record','mths\_since\_last\_major\_derog' and 'verification\_status\_joint' have LARGEST amount of <u>missing values</u> – 84%, 75% and 99.9% respectively.
- > 'annual\_inc', 'debt\_income\_ratio', 'collection\_recovery\_fee', 'acc\_now\_delinq' and 'tot\_colle\_amt' are **HIGHLY SKEWED**.
- 'State' and 'last\_week\_pay' have HIGH CARDINALITY.

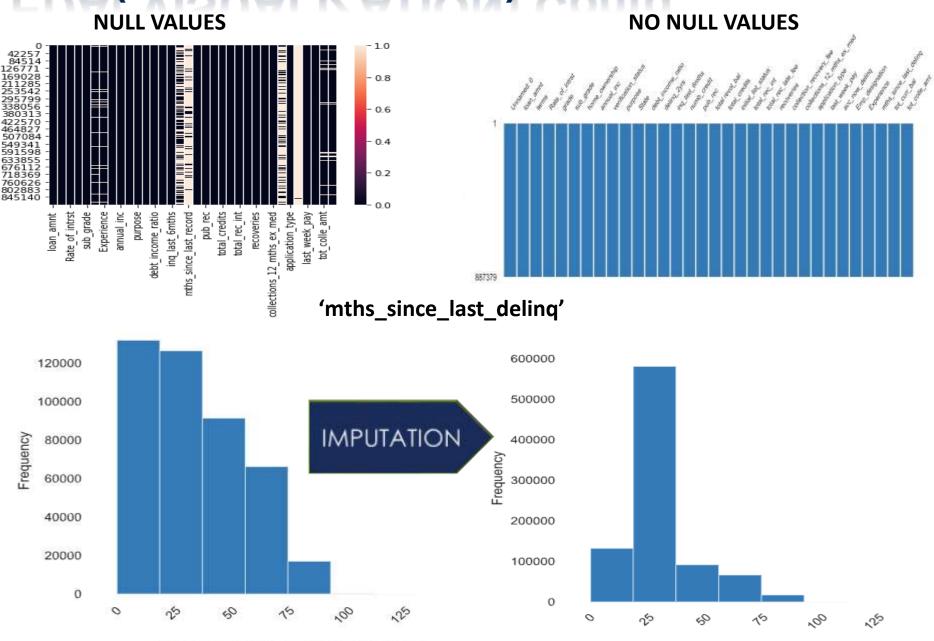
## **EDA( VISUALIZATION)**







## EDA( VISUALIZATION) Contd..

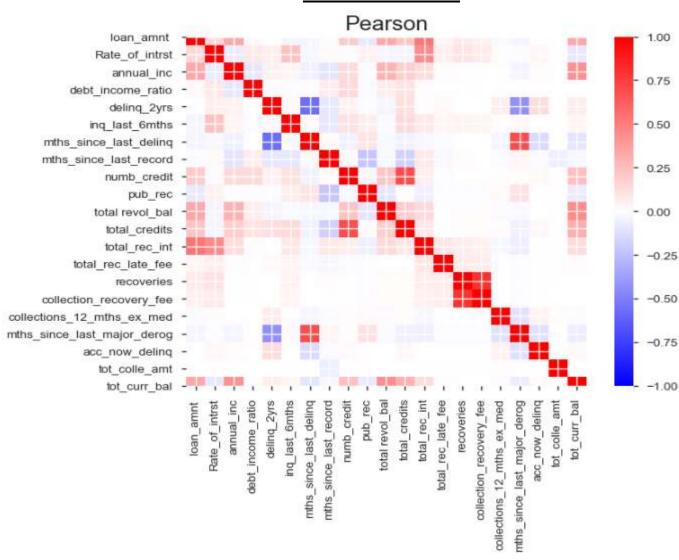


Histogram with fixed size bins (bins=10)

Histogram with fixed size bins (bins=10)

## **EDA( VISUALIZATION) Contd..**

#### **CORRELATION**







#### **Linear Regression**

Dataset Type	Test Size	Random State	R-squared	RMSE
NULL	0.3	0	0.261	18947.63
NOT NULL	0.3	0	0.261	18851.73

#### **Random Forest**

Data Type	Model Type	Test Size	Random State	R-squared (Train)	R-squared (Test)	RMSE (Train)	RMSE (Test)
NOT NULL	Only Numerical	0.3	0	0.66	0.33	13296.07	17496.04
NOT NULL	Only Numerical ( Feature Importance)	0.3	0	0.804	0.33	10092.98	17565.39
NOT NULL	Both Numerical& Categorical	0.3	0	0.805	0.33	10061.32	17574.13
NOT NULL	Both Numerical& Categorical (Feature Importance)	0.3	0	0.814	0.326	9842.85	17629.99

#### **XGBoost**

Dataset Type	Encoding	Min. Child Weight	Max Depth	Learning Rate	Gamma	R-squared (Train)	R-squared (Test)	RMSE (Train)	RMSE (Test)
NOT NULL	Label Encoding	3	6	0.15	0.4	0.55	0.35	15162.58	17242.74
NOT NULL	Frequency Encoding	5	3	0.08	0.1	0.43	0.35	17121.70	17260.07



## Model Deployment using R shiny / Flask or any other method



## **Challenges faced?**

How did you overcome?



## Thank you