# Loan Tap Case Study

### **Problem Statement**

LoanTap deploys formal credit to salaried individuals and businesses 4 main financial instruments:

- 1. Personal Loan
- 2. EMI Free Loan
- 3. Personal Overdraft
- 4. Advance Salary Loan But the main focus is to interpret the underwriting process behind the Personal Loan only

Given a set of attributes for an Individual, determine if a credit line should be extended to them. If so, what should the repayment terms be in business recommendations?

- Additional views
  - We need to track the users previous credit line history and repayment status.
  - Analyzing the previous loans tenure and the total liability.
  - As we are focusing more on salaried individuals, we need to take the salary of the person into consideration.

# Solution - Colab Link

#### **Actionable Insights & Recommendations**

- 80% of customers have fully paid their loans, while 20% have defaulted.
- The trained model predicts a 94% f1-score for the negative class (Fully Paid) and a 62% f1-score for the positive class (Charged Off).
- Cross-validation accuracy and testing accuracy being nearly the same indicate the model's decent performance and reliability on unseen data.
- To enhance the model's performance, options like collecting more data, employing a more complex model, or fine-tuning hyperparameters can be considered.
- The ROC AUC curve, with an area of 0.73, suggests the model correctly classifies around 73% of instances, signifying a good but improvable performance.
- The precision-recall curve visualizes the trade-off between precision and recall based on varying thresholds, enabling identification of the optimal point for specific application needs.
- Post balancing the dataset, significant improvements were not observed in precision and recall scores for both classes.
- The Logistic Regression Classifier achieves an accuracy of 0.891 on the test set, reflecting a decent performance, not merely by chance.

# Questionnaire

- 1. What percentage of customers have fully paid their Loan Amount? Around 80.38% of customers have fully paid their Loan Amount.
- 2. Comment about the correlation between Loan Amount and Installment features. There is a very high correlation between loan amount and installment. pearson coefficient is ~ 0.95. This indicates high multicollinearity between these two features. Hence the installment column is dropped while creating the model.
- 3. The majority of people have home ownership as \_\_\_\_\_.

  Mortaged and then followed by rent. With 50.07% and 40.35% in order.
- **4.** People with grades 'A' are more likely to fully pay their loan. (T/F) True. Out of all people with grade 'A', ~ 92-93% got their loan approved.
- 5. Name the top 2 afforded job titles.

Teacher & Manager

- 6. Thinking from a bank's perspective, which metric should our primary focus be on..
  - a. ROC-AUC: Not a good metric to consider as we have highly imbalanced data.
  - b. Precision: Consider when only want to reduce NPA
  - c. Recall: Consider when we do not want to miss good opportunity for providing loan to customer
  - d. F1-score: Good metric for us, because we want to consider both Precision and Recall.

Answer - F1-score

#### 7. How does the gap in precision and recall affect the bank?

- The disparity between precision and recall impacts risk assessment for lending decisions.
- Higher precision ensures more accurate identification of potential defaulters, aiding in cautious decisions.
- Lower recall could result in missing actual defaulters, leading to missed opportunities.
- Balancing precision and recall affects credit denials and default rates.

- Striking a suitable balance is crucial for risk management aligned with business goals.

# 8. Which were the features that heavily affected the outcome?

Zip\_code ,grade ,dti , open\_acc ,revol\_util ,loan\_amnt ,purpose\_small\_business ,home\_ownership ,int\_rate

9. Will the results be affected by geographical location? (Yes/No)

Yes. Few states have high changes of getting loan compare to other states.