

# **TITLE:**

## **CREDIT CARD DEFAULT PREDICTION**

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# Introduction

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- *Overview of the project*
- *Importance of predicting credit card defaults*
- *Objectives of the project*



# Project Objective

## Objective 01

Predict whether a credit card holder will default on their payment

## Objective 02

Aid financial institutions in assessing and managing credit risk



# Data Source

- Dataset: Default of Credit Card Clients Dataset
- Source: Kaggle
- Link: [Dataset](#)



# Data Structure

- Features: LIMIT\_BAL, SEX, EDUCATION, MARRIAGE, AGE, PAY\_0, PAY\_2, PAY\_3, PAY\_4, PAY\_5, PAY\_6, BILL\_AMT1, BILL\_AMT2, BILL\_AMT3, BILL\_AMT4, BILL\_AMT5, BILL\_AMT6, PAY\_AMT1, PAY\_AMT2, PAY\_AMT3, PAY\_AMT4, PAY\_AMT5, PAY\_AMT6
- Target: default.payment.next.month

# Data Preprocessing

- Handling missing values
- Scaling features
- Handling Imbalanced Data

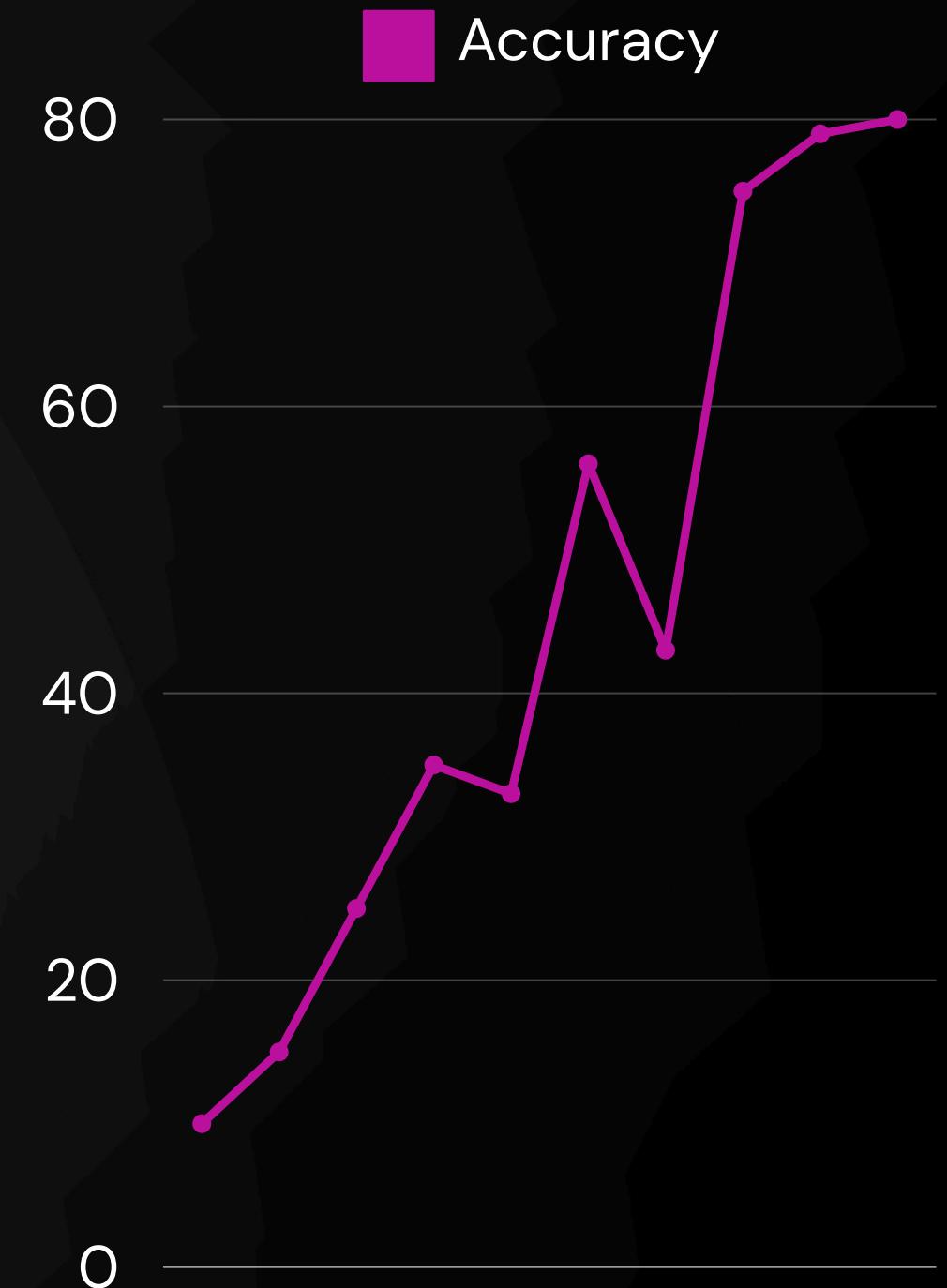


# Model Selection

- Algorithm used: XGBoost Classifier
- Reason for selection: High performance, robustness, and interpretability

# Model Training

- Training process
- Model tuning and optimization
- Achieved accuracy: 80%



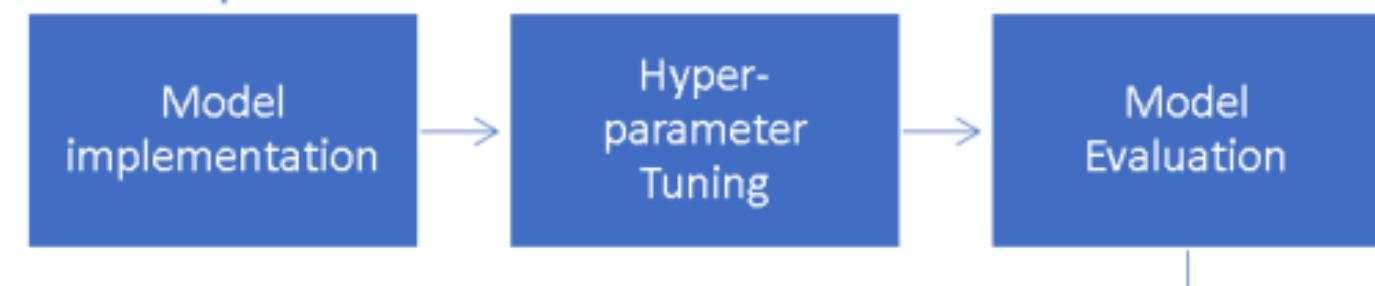
# System Architecture

## Architecture

### Data Preparation



### Model development



### Deployment



# UI Design

## Web Application

- Developed using Flask
- User interface for input and prediction
- Features and functionalities

## Wireframe Design

- Key components of the UI: Header, Input Form, Prediction Result Area

# Model Results

## Credit Card Fraud Detection

Enter 23 Feature Values (comma separated):

Enter the values in the order of columns: ['LIMIT\_BAL', 'SEX', 'EDUCATION', 'MARRIAGE',  
'AGE', 'PAY\_0', 'PAY\_2', 'PAY\_3', 'PAY\_4', 'PAY\_5', 'PAY\_6', 'BILL\_AMT1', 'BILL\_AMT2',  
'BILL\_AMT3', 'BILL\_AMT4', 'BILL\_AMT5', 'BILL\_AMT6', 'PAY\_AMT1', 'PAY\_AMT2',  
'PAY\_AMT3', 'PAY\_AMT4', 'PAY\_AMT5', 'PAY\_AMT6']

**Predict**

**Prediction:** Person is not Faulty

# Conclusion

- Summary:
  - The project successfully developed a predictive model using XGBoost to assess credit card default risk.
  - Data preprocessing and feature engineering were crucial steps to prepare the dataset for modeling.
- Key Findings:
  - Achieved an accuracy of 80% on the test data.
  - Important features influencing defaults included payment history and bill amounts.
- Future Work:
  - Improve model accuracy with more advanced techniques or additional data.
  - Integrate real-time data processing for dynamic predictions.
- Takeaways:
  - Predictive modeling can significantly aid in managing financial risks.
  - Effective data preprocessing and feature selection are essential for building robust models.

# THANK YOU

For watching this presentation



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