



Credit Card Fraud Detection using Machine Learning

Stage II Project Review

Project Guide – Prof. Ashwini Bhosale

Presenting –

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Credit Card Fraud Detection using Machine Learning



Problem Statement

With businesses moving online, fraud and abuse in online systems are constantly increasing as well. Traditionally, rule-based fraud detection systems are used to combat online fraud, but these rely on a static set of rules created by human experts. This project uses machine learning to create models for fraud detection that are dynamic, self-improving, and maintainable. Importantly, they can scale with the online business.



Study by RBI

The average daily transactions is 100 million now for a volume of Rs 5 trillion.

Payments through digital modes are expected to jump to 1.5 billion transactions, worth Rs 15 trillion a day in five years, the Reserve Bank of India (RBI) estimates.





Statistics

2020 – 389,845 Credit card fraud reports.

2021 – 1,862 data breaches

2022 – banking frauds in India amounted for 1.38 trillion Indian rupees

Main Objective of Proposed Work

Automate the detection of potentially fraudulent activity, and the flagging of that activity for review



overview

The project spans over different key stages.

These stages include **data collection**, **preprocessing**, **model development**, **feature engineering** and **selection**, **Streamlit** application development, and deployment.

The use of machine learning algorithms will enable accurate fraud detection, while the Streamlit platform will ensure a user-friendly experience.

Data Collection and Preprocessing

Data collection involves acquiring a dataset containing credit card transactions, including both fraudulent and legitimate ones.

Preprocessing the data is crucial to ensure its quality and reliability.

Challenges such as class imbalance, missing values, and duplicates need to be addressed through appropriate techniques.

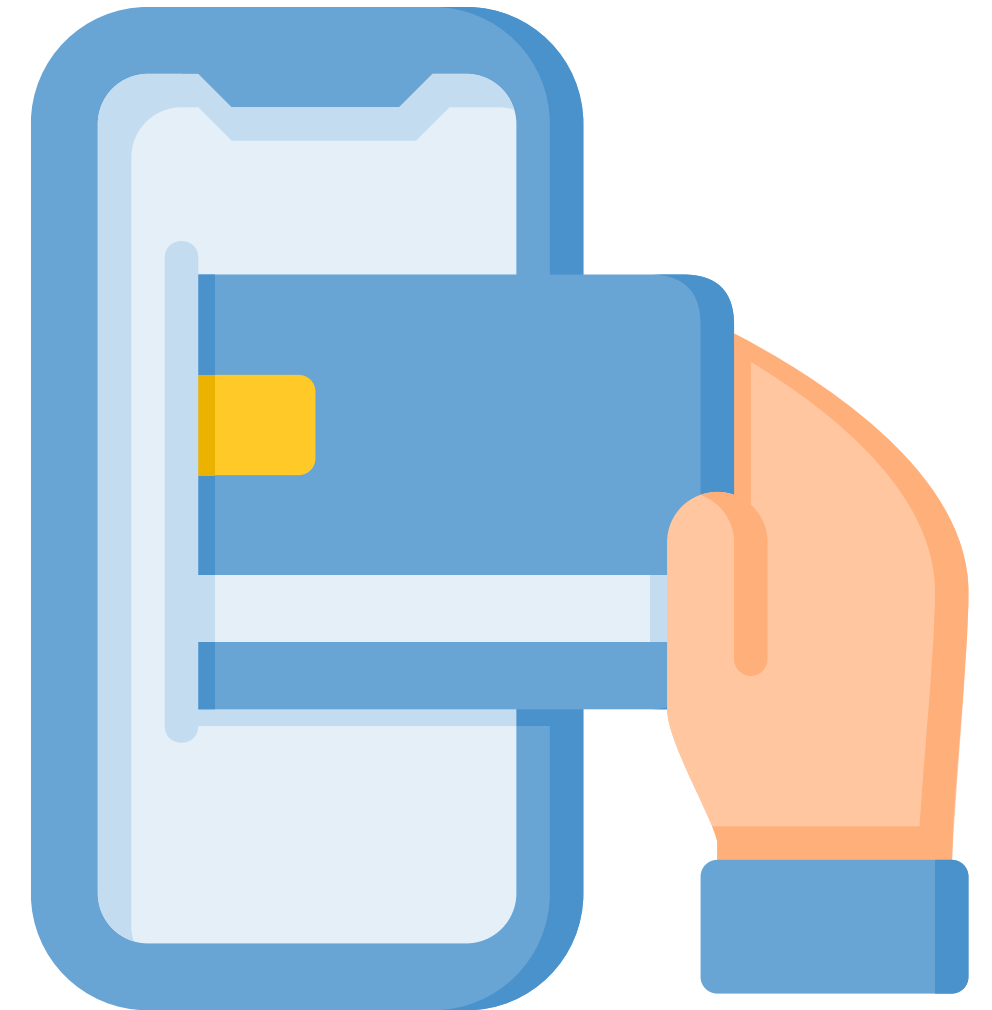


Model Development

Machine learning algorithms play a crucial role in detecting patterns and anomalies within the credit card transaction data.

logistic regression, decision trees, random forests, and gradient boosting, are explored and evaluated.

performance metrics and their ability to accurately detect fraud.



Feature Engineering and Selection

Feature engineering involves extracting meaningful features from the dataset that contribute to fraud detection.

Feature selection techniques, such as Principal Component Analysis (PCA) or Recursive Feature Elimination (RFE), will be employed to identify the most relevant features.

These processes aim to enhance the model's performance and improve its ability to differentiate between fraudulent and legitimate transactions.

Types of Fraudulent Behaviour

Layer 1

Endpoint Authentication

⇨ stolen card or machine

Layer 2

Anomaly within a session

⇨ Irregular behaviour within a session—e.g. transfer before balance

Layer 3

Anomaly within an account

⇨ Irregular transactions—e.g. spike in transfer and recipients

Layer 4

Anomaly within multiple channels of the same account

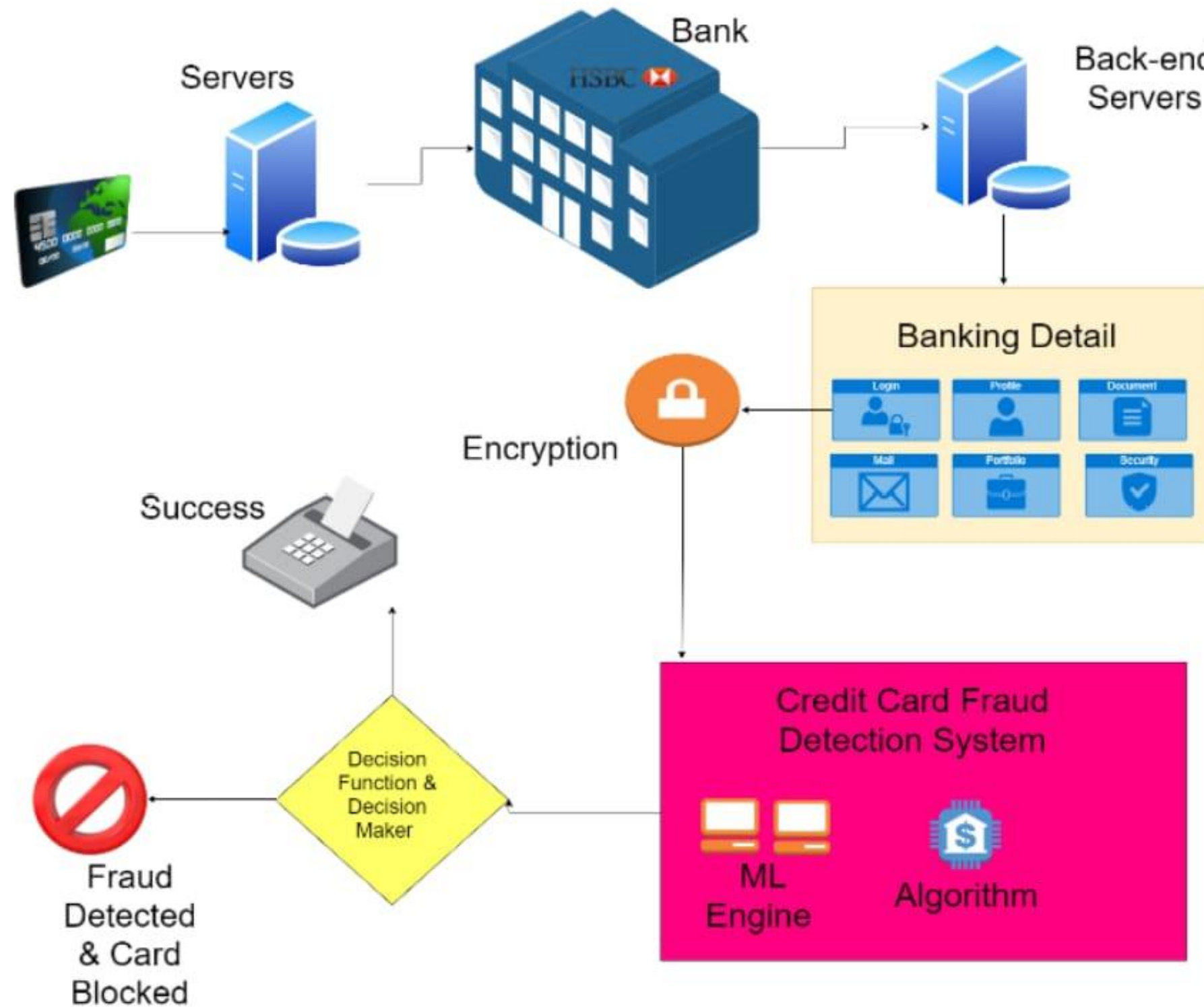
⇨ Irregular transactions across channels—e.g. spike in transfer and recipients

Layer 5

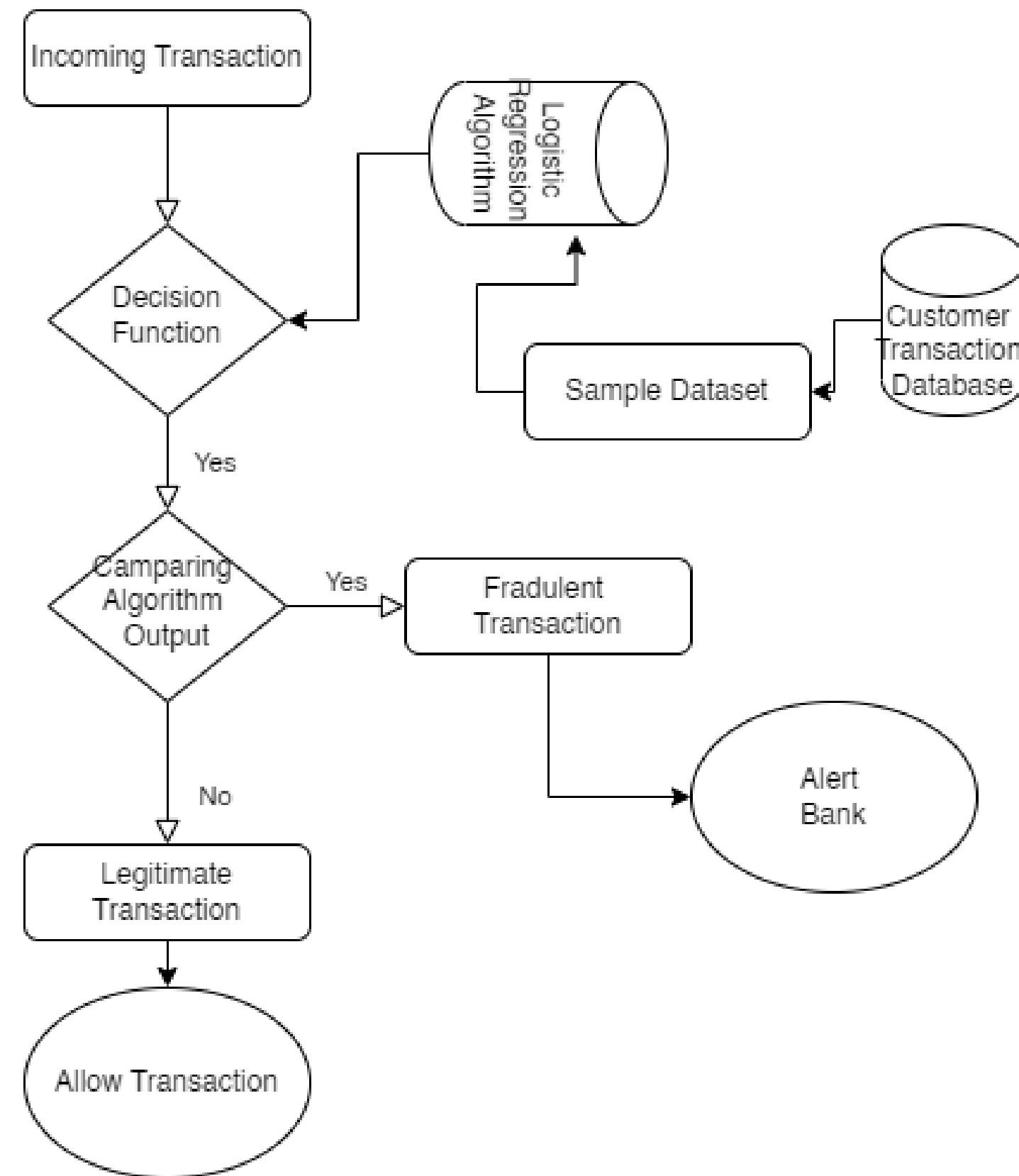
Anomaly within multiple channels of multiple accounts

⇨ Irregular transactions across channels and accounts

Proposed Work Flow Chart



Flow Chart



Streamlit Application Development

Streamlit is a powerful platform for building interactive web applications.

The credit card fraud detection application has a user-friendly interface, allowing users to input transaction details.

The application processes the inputs using the deployed machine learning model and displays real-time fraud detection results.

How Application work?

Credit Card Fraud Detection Model

Enter the following features to check if the transaction is legitimate or fraudulent:

Input All features (comma-separated)

Submit

User Interface of the Application

How Application work?

Credit Card Fraud Detection Model

Enter the following features to check if the transaction is legitimate or fraudulent:

Input All features (comma-separated)

```
472,-1.10091965379217,1.02958757720055,1.34833301916397,-1.36208235210811,-0.3434649874989  
57,-0.671658951020721,0.291221712639042,0.37999435232058,0.338839349085901,-0.438420634526  
655,-0.511769707940766,-0.547060718137173,-1.24594591157821,0.263623706254318,0.9014134921  
17228,0.382382630623662,-0.267487897556997,-0.649733762760941,-1.06710915904496,0.00214726  
644040852,-0.110387981416818,-0.180427277105698,-0.0071968021446594,0.0688767023249156,-0.  
410540366822498,0.731643091704887,0.0840507900564247,-0.0572356291516184,0.92
```

Submit

Transaction details

How Application work?

Credit Card Fraud Detection Model

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472,-1.10091965379217,1.02958757720055,1.34833301916397,-1.36208235210811,-0.3434649874989
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410540366822498,0.731643091704887,0.0840507900564247,-0.0572356291516184,0.92
```

Submit

Input:

```
472,-1.10091965379217,1.02958757720055,1.34833301916397,-1.36208235210811,-0.343464987498957,-0.
671658951020721,0.291221712639042,0.37999435232058,0.338839349085901,-0.438420634526655,-0.511
769707940766,-0.547060718137173,-1.24594591157821,0.263623706254318,0.901413492117228,0.382382
630623662,-0.267487897556997,-0.649733762760941,-1.06710915904496,0.00214726644040852,-0.110387
981416818,-0.180427277105698,-0.0071968021446594,0.0688767023249156,-0.410540366822498,0.73164
3091704887,0.0840507900564247,-0.0572356291516184,0.92
```

Output: Legitimate transaction

Output

Future Scope

- **Incorporating Advanced ML Techniques**
- **Continuous Model Training**
- **Integration with External Data Sources**
- **Collaboration with Industry Stakeholders**
- **Training and Deployment using AWS SageMaker**

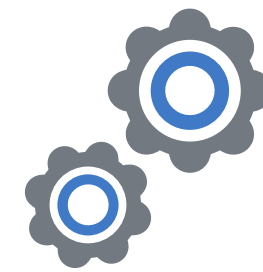


What we Used ?



Software Requirements

- Operating system – Windows 10/11
- Visual Studio Community - 2022
- Microsoft Office 2019
- Any Web Browser – Latest Version
- Python
- Jupyter Notebook
- Streamlit



Hardware Requirements

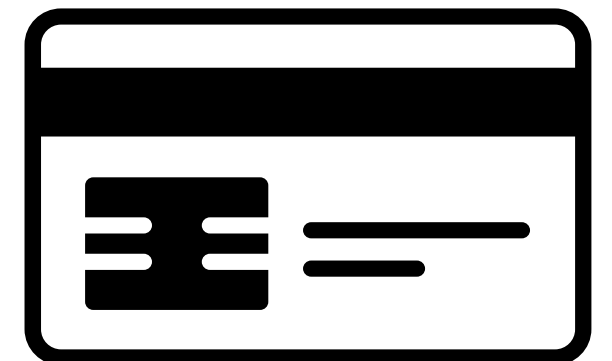
- Processor - 64-bit, four-core, 2.5 GHz minimum per core
- RAM– 8 GB Minimum
- Hard Disk – 256 GB Required

Work Carried Out from stage I

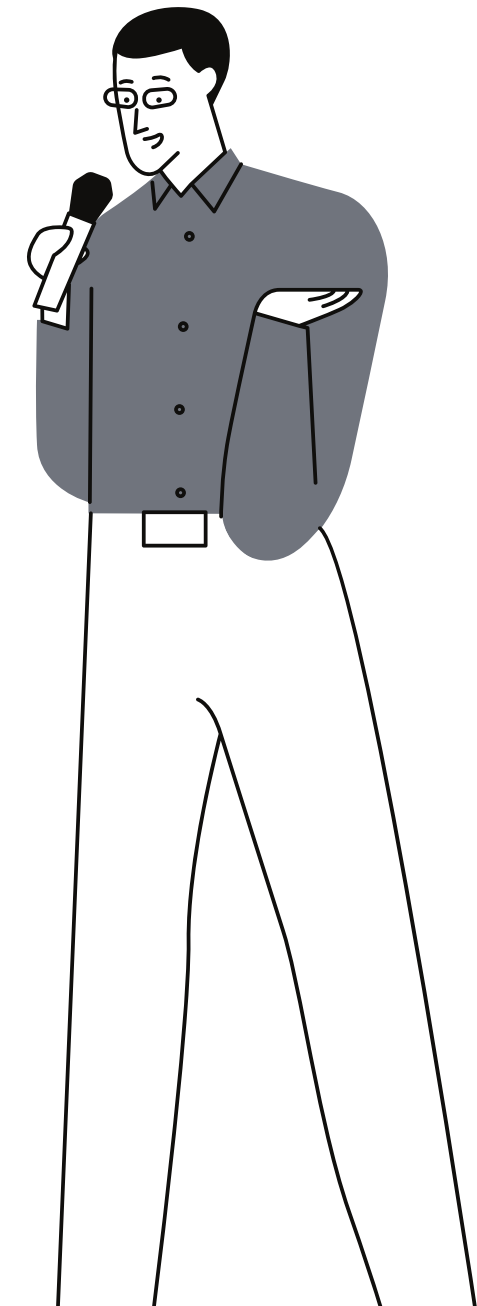


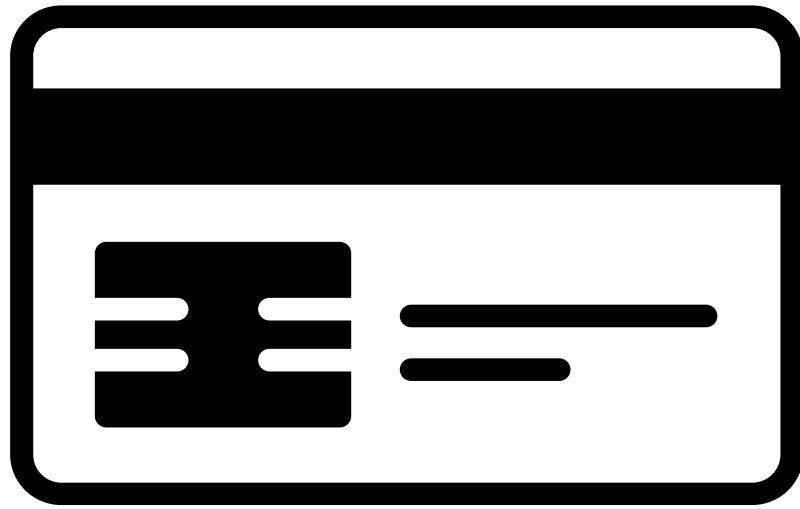
**Application Development
and Deployment**

**explore new Algorithms to
detect Fraudulent
Transactions**



Open for Suggestion!





Thank You!