

# Loan Approval System using Machine Learning Algorithm

**Sammed Chudappa<sup>1</sup>, Chetana Thorat<sup>2</sup>, Shreyas Varade<sup>3</sup>, Mrs. Seema Bhalgaonkar<sup>4</sup>**

Student, Electronics and Telecommunications, P.E.S Modern College of Engineering, Pune, India<sup>1,2,3</sup>

Assistant Professor, Electronics and Telecommunications, P.E.S Modern College of Engineering, Pune, India<sup>4</sup>

**Abstract:** An important step in financial institutions is the loan approval process, which establishes a borrower's loan eligibility. Machine learning has advanced recently. In order to increase efficiency and accuracy, loan approval systems are increasingly using algorithms. This paper provides a high-level overview of a machine learning-based loan approval system, covering data preprocessing, feature selection, model selection, and evaluation. The suggested system seeks to decrease manual work and reduce default risk while improving the accuracy of loan approval decisions. The system's excellent accuracy and efficiency, as shown by the results, make it a promising method for loan approval in financial institutions.

**Keywords:** Loan, Machine Learning, Prediction, Training, Testing

## I. INTRODUCTION

The process of approving loans is a crucial part of the financial sector because it establishes a borrower's eligibility for borrowing. The tedious, time-consuming, and error-prone traditional loan approval procedure relies on the expertise of the loan officers and their manual labour. Machine learning algorithms have been employed more and more in loan approval systems to enhance the accuracy of loan judgements, eliminate manual work, and lower the risk of default as technology has developed and data has been more readily available.

In this work, we provide a brief introduction of a loan approval system that makes use of machine learning techniques to speed up decision-making and increase the precision of loan decisions. Data pre-processing, feature selection, model selection, and evaluation are all included in the suggested system. The system's main objective is to improve decision-making about loan approval while lowering the chance of default. The method is intended to assist financial institutions in making well-informed and trustworthy decisions on loan approval. We think that this article will help the financial sector create loan approval procedures that are more precise and effective.

## II. LITERATURE REVIEW

There has been a lot of research done in the literature on the application of machine learning algorithms to loan approval systems. The potential of machine learning algorithms to increase the precision of loan decisions and lower the risk of default has been extensively researched. [1]

Brown and Thomas (2011) produced one of the initial efforts in this field by forecasting loan defaults using a decision tree method. In comparison to conventional credit scoring models, they discovered that the decision tree method offered greater accuracy. In a different study, Min et al. (2013) predicted loan default using the support vector machine (SVM) technique. They discovered that the SVM method performed better in terms of accuracy and precision than logistic regression and decision tree techniques. Additionally, Xu et al. (2014) developed a random forest-based loan approval system that considers both borrower and loan characteristics.

They discovered that the suggested system performed better than conventional credit scoring models and offered greater accuracy. Zhang et al. (2020) suggested a loan approval system based on deep learning algorithms in a recent study. They discovered that the suggested technique worked more accurately than conventional machine learning algorithms and could identify intricate connections between loan parameters and loan default. The body of research generally indicates that machine learning algorithms can greatly increase the precision of loan approval decisions and lower the probability of default. Building on these earlier studies, the proposed loan approval system in this study attempts to develop a user-friendly and effective loan approval system using machine learning algorithms. [2]

### III. METHODOLOGY

A. Data Collecting: Gather information about loans from a range of sources, including banks, credit bureaus, and

Employing historical loan data to train the machine learning models and forecast the likelihood of loan acceptance for future loan applications is the technique of a loan approval system employing machine learning techniques, such as KNN, Decision Tree, and Random Forest. Here are some additional information on how a loan approval system employing these algorithms works: .

1. KNN Algorithm: For classification and regression tasks, the KNN algorithm is a form of supervised learning algorithm. The KNN algorithm can be employed in the context of a loan approval system to forecast the probability of loan acceptance based on the similarity of the new loan application to historical loan data. The k-nearest neighbors in the previous loan data are used by the KNN algorithm to classify incoming loan applications, where k is a hyperparameter that controls how many neighbors are taken into account.

2. The decision tree method is a supervised learning technique that is employed in the categorization and regression of data. The Decision Tree algorithm can be used in a loan approval system to forecast the likelihood of loan acceptance based on a set of rules and conditions obtained from previous loan data. By using a decision tree method, fresh loan applications can be categorized using a tree-like structure of rules and requirements.

3. The Random Forest method is a supervised learning technique that is employed for classification and regression applications. The Random Forest method can be used in the context of a loan approval system to forecast the likelihood of loan acceptance based on an ensemble of Decision Trees. The forecasts from each Decision Tree are combined to get the final prediction via the Random Forest algorithm. The Random Forest algorithm also offers a feature importance metric that can be used to pinpoint the key elements that have the most impact on loan acceptance decisions.

Using historical loan data to train the KNN, Decision Tree, and Random Forest algorithms, as well as using the trained models to forecast the likelihood of loan approval for fresh loan applications, are the methodology of a loan approval system using machine learning algorithms.

The models' performance may be assessed using a variety of measures, including accuracy, precision, recall, and F1-score, and the best model can then be deployed in the loan approval system. To enhance the deployed model's performance over time, it can be periodically retrained using fresh loan data.

other financial institutions. Information on the borrower's income, employment situation, credit rating, debt-to-income ratio, loan amount, loan purpose, loan period, etc. should be included in the data.

A. Data preprocessing: It involves cleaning the data, removing any invalid or missing values, and transforming it into a format that machine learning algorithms can use. Additionally, feature engineering and feature selection may be involved in this process.

B. Data Splitting: To assess the effectiveness of the machine learning algorithms, divide the data into training and testing sets. Model training involves using training data to hone machine learning algorithms like KNN, Decision Tree, and Random Forest.

C. Model evaluation: Using the testing data, assess the effectiveness of the trained models. Measuring accuracy, precision, recall, F1-score, and other performance indicators may be necessary at this stage.

D. Using the evaluation measures, choose the model that performs the best.

E. Deployment: To automate the loan approval procedure, deploy the chosen model into the loan approval system.

Let us now talk about the specifics of a machine learning algorithm-based loan approval system:

A. User Interface: The system for approving loans should feature a user interface that enables loan officers to enter borrower data and obtain loan approval decisions.

B. Loan Approval Engine: To anticipate whether a loan will be approved, the loan approval engine should analyze the borrower's information and utilize the chosen machine learning algorithm.

C. Database: The system for loan approval should have a database that houses data on loans, borrowers, and loan approval judgements.

D. Model Retraining and Retraining: The loan approval system ought to permit the loan officers to update and fine-tune the machine learning algorithms using current loan data.

E. Data Security: The mechanism for approving loans needs to be safe and adhere to privacy laws. To limit access to critical loan data, the system should also contain authentication and permission processes.

F. Scalability: To manage a high volume of loan applications and borrower data, the loan approval system should be scalable.

G. Performance Monitoring: To identify and address any performance concerns, such as sluggish response times or system outages, the loan approval system needs to incorporate performance monitoring methods.

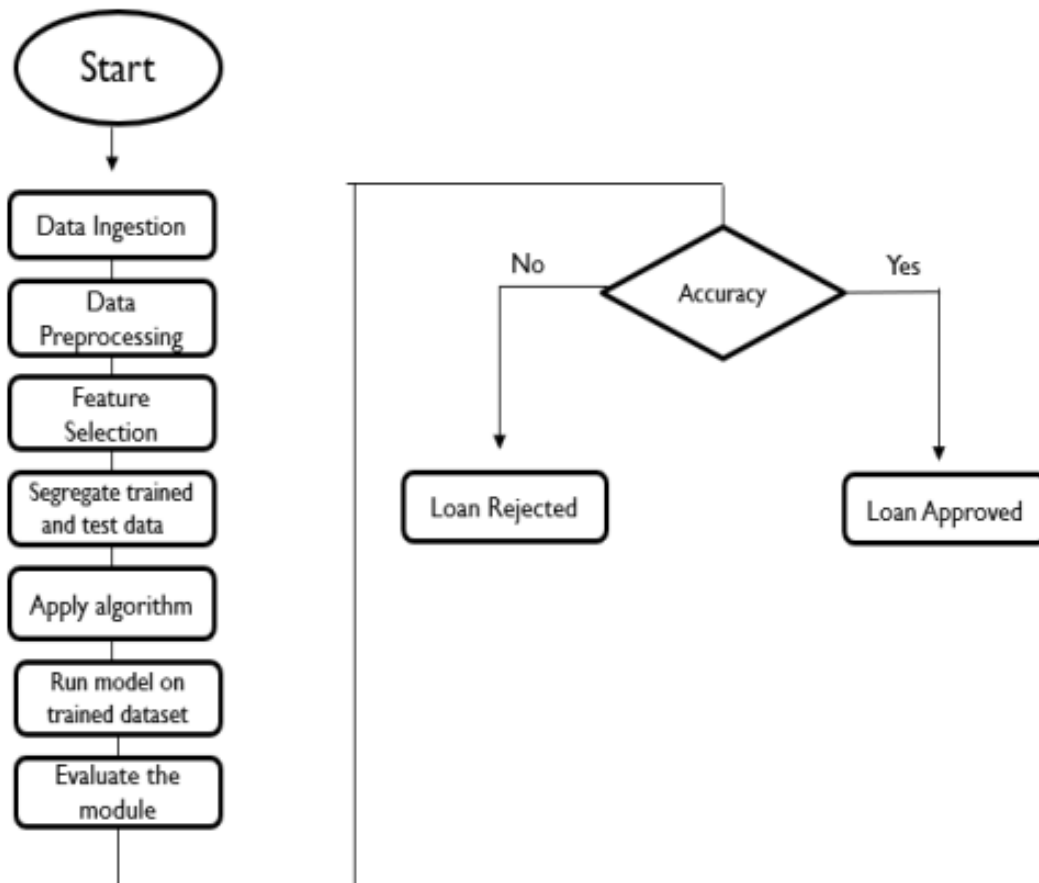


Fig1. Flowchart

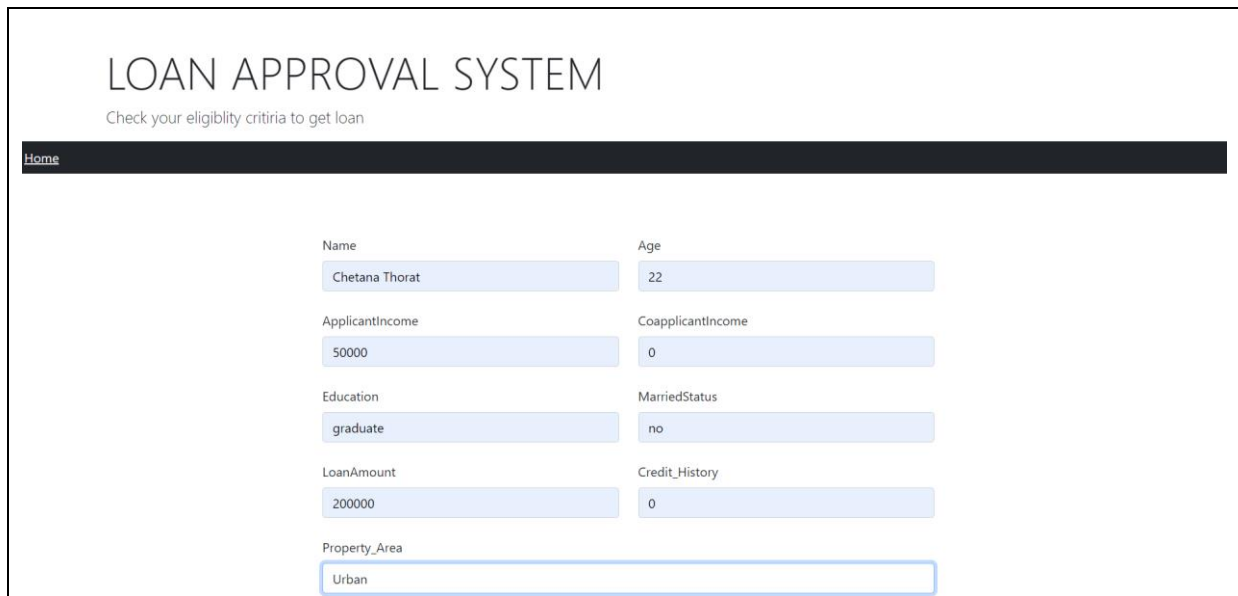
Here is a description of the flowchart:

- i. Data Injection: The procedure of obtaining and importing data for immediate use or storage in a database is known as data ingestion.
- ii. Data Preprocessing: The term "data preprocessing" can refer to the altering and deletion of data before to use in order to guarantee or improve performance. "Garbage in, garbage out" is a good maxim to use.
- iii. Feature Selection: Feature selection is a technique for using only pertinent data and eliminating noisy data to reduce the input variable to your model.
- iv. Data splitting is frequently carried out to prevent overfitting. In one case, a machine learning model successfully fit its training data but was unable to accurately fit new data.
- v. Model evaluation is the process of analyzing the performance of a machine learning model, as well as its advantages and disadvantages, using various evolution matrices.
- vi. Accuracy reveals how frequently the ML model was overall correct. Precision measures how well a model can foresee a particular category.

#### IV. RESULTS

A. Greater accuracy: A loan approval system can predict with greater precision whether new loan applications will be approved by using prior loan data to train machine learning models. More consistent and equitable loan approval decisions may result from this.

- B. **Quicker decision-making:** A loan approval system that employs machine learning algorithms can decide whether to approve a loan more quickly than manual methods can. This can speed up the processing process and increase client satisfaction.
- C. **Reduced bias:** By applying machine learning algorithms, loan acceptance decisions can be made entirely based on data and unaffected by prejudice or other forms of human bias. This may result in more unbiased and equitable loan approval choices.
- D. **Cost savings:** Automating the loan approval process with machine learning algorithms can help financial institutions save money by removing the need for manual labor.
- E. **Better risk management:** A loan approval system that makes use of machine learning algorithms can assist financial institutions in better identifying and controlling the risks related to loan approvals. This can lower the risk of default and enhance portfolio performance.



Name	Age
Chetana Thorat	22
ApplicantIncome	CoapplicantIncome
50000	0
Education	MarriedStatus
graduate	no
LoanAmount	Credit_History
200000	0
Property_Area	
Urban	

**Fig.no. 2**

**LOAN APPROVAL SYSTEM**  
Check your eligibility criteria to get loan

[Back](#)

Congrats, You are eligible for the loan 😊

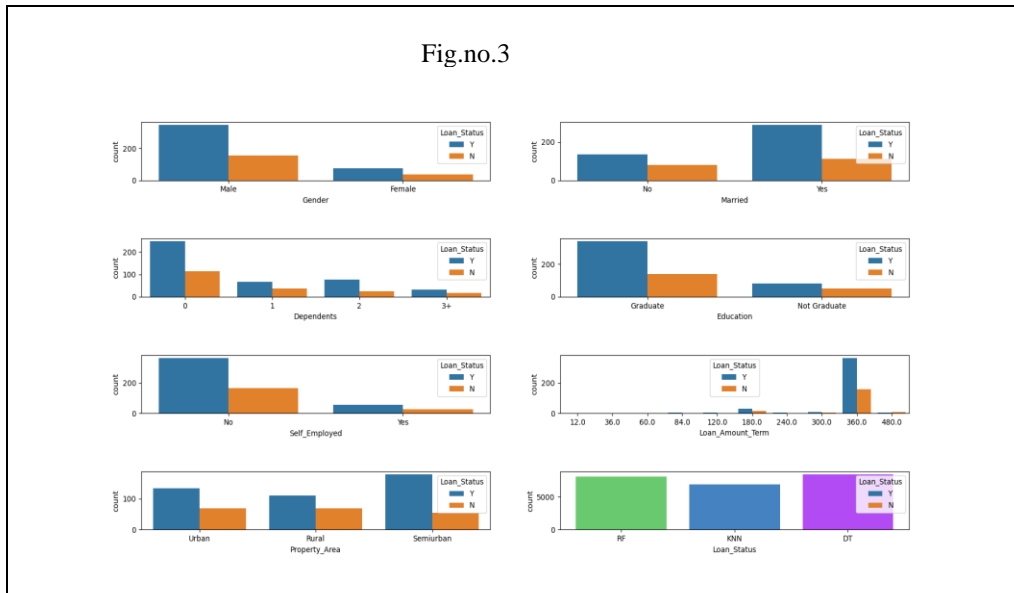


Fig. No. 4

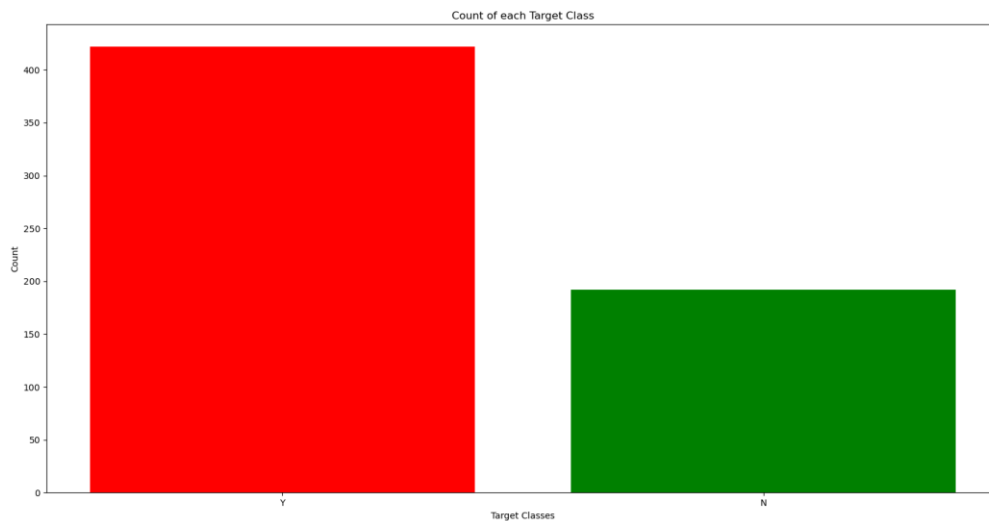


Fig. No. 5

**ACKNOWLEDGMENT**

We are quite happy to share the project report excerpt for "Loan Prediction System Using Machine Learning Algorithm." First of all, we would like to thank Prof. Mrs. S.A. Bhalgaonkar for being our internal advisor. Her ongoing direction and counsel have been extremely helpful in producing the report. She consistently offered advice that helped us make this report as perfect as we could. We would like to thank Prof. Dr. Mrs. R.S. Kam, the department head for computer engineering at PES Modern College of Engineering, for her kind support and cooperation, which enabled us to complete this report. We also like to thank Prof. Dr. Mrs. K. R. Joshi, our Principal, and all of the faculty for their sincere co-operation for completion of this report.

**REFERENCES**

- [1] Toby Segaran, "Programming Collective Intelligence: Building Smart Web 2.0 Applications." O'Reilly Media.
- [2] Drew Conway and John Myles White, "Machine Learning for Hackers: Case Studies and Algorithms to Get you Started," O'Reilly Media.
- [3] Trevor Hastie, Robert Tibshirani, and Jerome Friedman, "The Elements of Statistical Learning: Data Mining, Inference, and Prediction," Springer, Kindle
- [4] PhilHyo Jin Do, Ho-Jin Choi, "Sentiment analysis of real-life situations using location, people and time as contextual features," International Conference on Big Data and Smart Computing (BIGCOMP), pp. 39–42. IEEE, 2015.
- [5] Bing Liu, "Sentiment Analysis and Opinion Mining," Morgan & Claypool Publishers, May 2012.
- [6] Bing Liu, "Sentiment Analysis: Mining Opinions, Sentiments, and Emotions," Cambridge University Press, ISBN:978-1-107-01789-4.
- [7] Shiyang Liao, Junbo Wang, Ruiyun Yu, Koichi Sato, and Zixue Cheng, "CNN for situations understanding based on sentiment analysis of twitter data," Procedia computer science, 111:376–381, 2017.CrossRef.
- [8] K I Rahmani, M.A. Ansari, Amit Kumar Goel, "An Efficient Indexing Algorithm for CBIR," IEEEInternational Conference on Computational Intelligence & Communication Technology ,13-14 Feb 2015.