

**DEPARTMENT OF COMPUTER SCIENCE  
BANGALORE YESWANTHPUR CAMPUS**

**M.Sc. DATA SCIENCE**

Project Report  
on

**Predictive Loan Approval System using K-Nearest  
Neighbours Algorithm: A Java Implementation**

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## **CHAPTER - 01**

### **ABSTRACT**

This project endeavours to create a sophisticated loan approval prediction system utilizing the K-Nearest Neighbours (KNN) algorithm. The primary objective is to develop a model capable of accurately predicting whether a loan application will be approved or denied based on various applicant attributes, including loan amount, loan amount term, and credit history. The project utilizes a comprehensive dataset comprising historical loan applicant information, including their application status. Through rigorous analysis and machine learning techniques, we aim to construct a reliable system capable of automating the loan approval process, thus streamlining operations for financial institutions.

#### **1.1 AIM**

The overarching aim of this project is to design and implement an efficient loan approval prediction system leveraging the power of the K-Nearest Neighbors algorithm. This system is intended to offer financial institutions a valuable tool for swiftly and accurately assessing loan applications, thereby enhancing decision-making processes and optimizing resource allocation. By harnessing the capabilities of machine learning, we seek to create a robust solution capable of addressing the complexities and challenges inherent in the loan approval process.

#### **1.2 Objective**

The specific objectives of this project are delineated as follows: Firstly, to preprocess the loan applicant dataset, incorporating essential steps such as data normalization to ensure uniformity and consistency. Subsequently, to implement the KNN algorithm for loan approval prediction, leveraging its simplicity and effectiveness in classification tasks. Thirdly, to evaluate the performance of the KNN model using a range of metrics including accuracy, R-squared score, mean squared error (MSE), and mean absolute error (MAE), thus ensuring the model's reliability and efficacy. Additionally, to visualize the results through various graphical representations such as scatter plots, histograms, and pie charts, facilitating a comprehensive understanding of the model's

predictions. Finally, to develop a user-friendly interface enabling stakeholders to input loan applicant details and obtain instantaneous predictions, thereby enhancing accessibility and usability.

## CHAPTER - 02

### INTRODUCTION

The loan approval process is a crucial aspect of the financial industry, providing funding for various purposes. However, traditional methods often lead to inefficiencies and biases due to cumbersome paperwork, prolonged processing times, and subjective decision-making. The increasing complexity of financial markets and digital lending platforms have further complicated these challenges. Automated systems for loan approval prediction have emerged as transformative tools, offering expedited decision-making, enhanced accuracy, and improved risk management.

The K-Nearest Neighbours (KNN) algorithm is a popular machine learning algorithm for its simplicity, versatility, and effectiveness in classification tasks. By leveraging proximity-based learning principles, KNN can provide reliable predictions for loan approval outcomes. Financial institutions have vast volumes of historical loan applicant data, which can be harnessed to develop predictive models capable of discerning patterns, trends, and risk factors inherent in the lending process.

Incorporating machine learning algorithms into loan approval processes offers several benefits beyond efficiency and accuracy. These include enhanced customer experience, reduced operational costs, improved compliance with regulatory requirements, and faster access to credit for creditworthy individuals and businesses. This project aims to harness the power of the K-Nearest Neighbours algorithm to develop a sophisticated loan approval prediction system. By leveraging a comprehensive dataset of historical loan applicant information, the project aims to construct a robust model capable of accurately assessing loan applications based on key attributes.

# CHAPTER - 03

## CODE

```

1 import java.util.List;
2 import java.util.ArrayList;
3 import java.util.Map;
4 import java.util.HashMap;
5 import java.util.Random;
6 import java.util.Scanner;
7 import java.io.File;
8 import java.io.FileNotFoundException;
9 import org.jfree.chart.ChartFactory;
10 import org.jfree.chart.ChartPanel;
11 import org.jfree.chart.JFreeChart;
12 import org.jfree.data.general.DefaultTitleDataset;
13 import org.jfree.chart.plot.PlotOrientation;
14 import org.jfree.data.category.DefaultCategoryDataset;
15 import org.jfree.ui.ApplicationFrame;
16 import org.jfree.ui.RefineryUtilities;
17 import javax.swing.JFrame;
18 import java.awt.BorderLayout;
19
20
21 public class KNN_loan {
22     // Define a class to represent a loan applicant
23     public static class LoanApplicant {
24         double loanAmount;
25         double loanAmountTerm;
26         double creditHistory;
27         String loanStatus;
28
29         public LoanApplicant(double loanAmount, double loanAmountTerm, double creditHistory, String loanStatus) {
30             this.loanAmount = loanAmount;
31             this.loanAmountTerm = loanAmountTerm;
32             this.creditHistory = creditHistory;
33         }
34     }
35
36     // Method to load data from CSV file starting from line 2
37     public static List<LoanApplicant> loadData(String filename) throws FileNotFoundException {
38         List<LoanApplicant> data = new ArrayList<>();
39         Scanner scanner = new Scanner(new File(filename));
40         // Skip the first line (header)
41         if (scanner.hasNextLine()) {
42             scanner.nextLine();
43         }
44         while (scanner.hasNextLine()) {
45             String line = scanner.nextLine();
46             String[] parts = line.split(",");
47             if (parts.length >= 13) { // Ensure that there are enough fields in the line
48                 String loanAmountStr = parts[8].trim(); // Index of loanAmount
49                 String loanAmountTermStr = parts[9].trim(); // Index of loanAmountTerm
50                 String creditHistoryStr = parts[10].trim(); // Index of Credit_History
51                 String loanStatus = parts[12].trim(); // Index of loanStatus
52
53                 // Check if any of the fields are empty
54                 if ((loanAmountStr.isEmpty() && loanAmountTermStr.isEmpty()) || creditHistoryStr.isEmpty()) {
55                     continue;
56                 }
57                 double loanAmountParsed = Double.parseDouble(loanAmountStr);
58                 double loanAmountTermParsed = Double.parseDouble(loanAmountTermStr);
59             }
60         }
61     }
62 }

```

```

26 double loanAmountTerm;
27 double creditHistory;
28 String loanStatus;
29
30 public LoanApplicant(double loanAmount, double loanAmountTerm, double creditHistory, String loanStatus) {
31     this.loanAmount = loanAmount;
32     this.loanAmountTerm = loanAmountTerm;
33     this.creditHistory = creditHistory;
34     this.loanStatus = loanStatus;
35 }
36
37 // Method to load data from CSV file starting from line 2
38 public static List<LoanApplicant> loadData(String filename) throws FileNotFoundException {
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54             // Check if any of the fields are empty
55             if ((loanAmountStr.isEmpty() && loanAmountTermStr.isEmpty()) || creditHistoryStr.isEmpty()) {
56                 continue;
57             }
58             double loanAmountParsed = Double.parseDouble(loanAmountStr);
59             double loanAmountTermParsed = Double.parseDouble(loanAmountTermStr);
60         }
61     }
62 }

```

```

67 // Method to normalize data
68 public static void normalizeData(List<LoanApplicant> data) {
69     // Find min and max values for each feature
70     double maxLoanAmount = Double.MAX_VALUE;
71     double minLoanAmount = Double.MIN_VALUE;
72     double maxLoanAmountTerm = Double.MAX_VALUE;
73     double minLoanAmountTerm = Double.MIN_VALUE;
74     double maxCreditHistory = Double.MAX_VALUE;
75     double minCreditHistory = Double.MIN_VALUE;
76
77     for (LoanApplicant applicant : data) {
78         maxLoanAmount = Math.max(maxLoanAmount, applicant.loanAmount);
79         minLoanAmount = Math.min(minLoanAmount, applicant.loanAmount);
80         maxLoanAmountTerm = Math.max(maxLoanAmountTerm, applicant.loanAmountTerm);
81         minLoanAmountTerm = Math.min(minLoanAmountTerm, applicant.loanAmountTerm);
82         maxCreditHistory = Math.max(maxCreditHistory, applicant.creditHistory);
83         minCreditHistory = Math.min(minCreditHistory, applicant.creditHistory);
84     }
85
86     // Normalize each feature
87     for (LoanApplicant applicant : data) {
88         applicant.loanAmount = (applicant.loanAmount - minLoanAmount) / (maxLoanAmount - minLoanAmount);
89         applicant.loanAmountTerm = (applicant.loanAmountTerm - minLoanAmountTerm) / (maxLoanAmountTerm - minLoanAmountTerm);
90         applicant.creditHistory = (applicant.creditHistory - minCreditHistory) / (maxCreditHistory - minCreditHistory);
91     }
92
93     // Method to split data into train and test sets
94     public static void splitData(List<LoanApplicant> data, List<LoanApplicant> trainSet, List<LoanApplicant> testSet, double splitRatio) {
95         Random random = new Random();
96         for (LoanApplicant applicant : data) {
97             if (random.nextDouble() < splitRatio) {
98                 trainSet.add(applicant);
99             } else {
100                 testSet.add(applicant);
101             }
102         }
103     }
104 }

```

```

// K-Nearest Neighbors (K-NN) algorithm
// Returns the predicted class for a given instance
// Parameters: trainset (List<LoanApplication>), testInstance (LoanApplication), k (int)
// Returns: String (predicted class)

public static String knn(LoanApplication trainset, LoanApplication testInstance, int k) {
    // Calculate distances to all data points in train set
    Map<LoanApplication, Double> distances = new HashMap<>();
    for (LoanApplication trainInstance : trainset) {
        double distance = Math.sqrt(Math.pow(testInstance.getLoanAmount() - trainInstance.getLoanAmount(), 2) +
            Math.pow(testInstance.getLoanTerm() - trainInstance.getLoanTerm(), 2) +
            Math.pow(testInstance.getLoanInterestRate() - trainInstance.getLoanInterestRate(), 2));
        distances.put(trainInstance, distance);
    }

    // Sort distances and get top k neighbors
    List<LoanApplication> nearestNeighbors = new ArrayList<>();
    distances.keySet().stream().sorted().limit(k).forEach(d -> nearestNeighbors.add(d.getKey().getLoanAmount()));

    // Count the votes for each class
    int voteCount = 0;
    for (LoanApplication neighbor : nearestNeighbors) {
        if (neighbor.getLoanAmount().equals(testInstance.getLoanAmount())) {
            voteCount++;
        }
    }

    // Make prediction based on majority vote
    if (voteCount > 0) {
        return "APPROVED";
    } else {
        return "REJECTED";
    }
}

```

```

// K-Nearest Neighbors (K-NN) algorithm
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// Returns: String (predicted class)

public static String knn(LoanApplication trainset, LoanApplication testInstance, int k) {
    // Calculate distances to all data points in train set
    Map<LoanApplication, Double> distances = new HashMap<>();
    for (LoanApplication trainInstance : trainset) {
        double distance = Math.sqrt(Math.pow(testInstance.getLoanAmount() - trainInstance.getLoanAmount(), 2) +
            Math.pow(testInstance.getLoanTerm() - trainInstance.getLoanTerm(), 2) +
            Math.pow(testInstance.getLoanInterestRate() - trainInstance.getLoanInterestRate(), 2));
        distances.put(trainInstance, distance);
    }

    // Sort distances and get top k neighbors
    List<LoanApplication> nearestNeighbors = new ArrayList<>();
    distances.keySet().stream().sorted().limit(k).forEach(d -> nearestNeighbors.add(d.getKey().getLoanAmount()));

    // Count the votes for each class
    int truePositives = 0;
    int falsePositives = 0;
    for (int i = 0; i < actual.size(); i++) {
        if (predicted.get(i).equals(testInstance.getLoanAmount()) {
            if (actual.get(i).equals(testInstance.getLoanAmount())) {
                truePositives++;
            } else {
                falsePositives++;
            }
        }
    }

    if (truePositives + falsePositives == 0) {
        return 0.0;
    }
    return (double) truePositives / (truePositives + falsePositives);
}

// Method to calculate R-squared score
public static double calculateRSquared(List<Double> actual, List<Double> predicted) {
    double sumSquaredErrors = 0.0;
    double mean = 0.0;
    for (double value : actual) {
        mean += value;
    }
    mean /= actual.size();
    for (int i = 0; i < actual.size(); i++) {
        double error = actual.get(i) - predicted.get(i);
        sumSquaredErrors += Math.pow(error, 2);
    }
    double sumSquaredDeviations = 0.0;
    for (double value : actual) {
        sumSquaredDeviations += Math.pow(value - mean, 2);
    }
    return 1 - (sumSquaredErrors / sumSquaredDeviations);
}

```

```

// K-Nearest Neighbors (K-NN) algorithm
// Returns the predicted class for a given instance
// Parameters: trainset (List<LoanApplication>), testInstance (LoanApplication), k (int)
// Returns: String (predicted class)

// Method to create histogram
public static void createHistogram(List<Double> predictedLabels) {
    // Create a dataset
    DefaultCategoryDataset dataset = new DefaultCategoryDataset();

    // Count the frequency of each predicted value
    Map<Double, Integer> frequencyMap = new HashMap<>();
    for (Double predictedValue : predictedLabels) {
        frequencyMap.put(predictedValue, frequencyMap.getOrDefault(predictedValue, 0) + 1);
    }

    // Add the frequency of each predicted value to the dataset
    for (Map.Entry<Double, Integer> entry : frequencyMap.entrySet()) {
        dataset.addValue(entry.getKey(), "Frequency", String.valueOf(entry.getValue()));
    }

    // Create the histogram
    JFreeChart chart = ChartFactory.createBarChart(
        "Predicted Value Distribution", // chart title
        "predicted value", // domain axis label
        "Frequency", // range axis label
        dataset, // data
        PlotOrientation.VERTICAL, // orientation
        false, // include legend
        true, // tooltips
        false // URLs
    );

    // Display the chart in a frame
    ChartPanel chartPanel = new ChartPanel(chart);
    JFrame frame = new JFrame("Histogram");
    frame.add(chartPanel);
    frame.setVisible(true);
    frame.pack();
    frame.setLocationRelativeTo(null);
    frame.setDefaultCloseOperation(JFrame.EXIT_ON_CLOSE);
    frame.setVisible(true);
}

```

```

// K-Nearest Neighbors (K-NN) algorithm
// Returns the predicted class for a given instance
// Parameters: trainset (List<LoanApplication>), testInstance (LoanApplication), k (int)
// Returns: String (predicted class)

// Count the frequency of each predicted value
Map<Double, Integer> frequencyMap = new HashMap<>();
for (Double predictedValue : predictedLabels) {
    frequencyMap.put(predictedValue, frequencyMap.getOrDefault(predictedValue, 0) + 1);
}

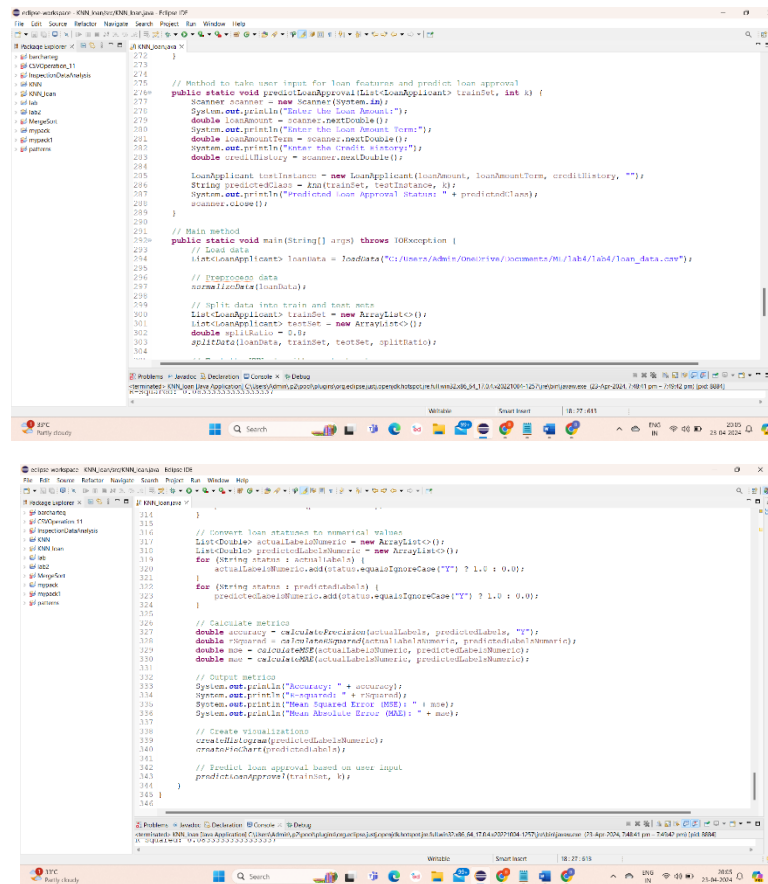
// Add the frequency of each predicted value to the dataset
for (Map.Entry<Double, Integer> entry : frequencyMap.entrySet()) {
    dataset.addValue(entry.getKey(), "Frequency", String.valueOf(entry.getValue()));
}

// Create the pie chart
JFreeChart chart = ChartFactory.createPieChart(
    "Predicted Value Distribution", // chart title
    dataset, // data
    true, // include legend
    true, // tooltips
    false // URLs
);

// Display the chart in a frame
ChartPanel chartPanel = new ChartPanel(chart);
JFrame frame = new JFrame("Pie Chart");
frame.add(chartPanel);
frame.setVisible(true);
frame.pack();
frame.setLocationRelativeTo(null);
frame.setDefaultCloseOperation(JFrame.EXIT_ON_CLOSE);
frame.setVisible(true);
}

// Method to take user input for loan features and predict loan approval
public static void predictLoanApproval(List<LoanApplication> trainset, List<LoanApplication> testset) {
    // Create a JFrame for user input
    JFrame frame = new JFrame("Loan Prediction");
    frame.add(new LoanPredictionForm());
    frame.setVisible(true);
    frame.pack();
    frame.setLocationRelativeTo(null);
    frame.setDefaultCloseOperation(JFrame.EXIT_ON_CLOSE);
    frame.setVisible(true);
}

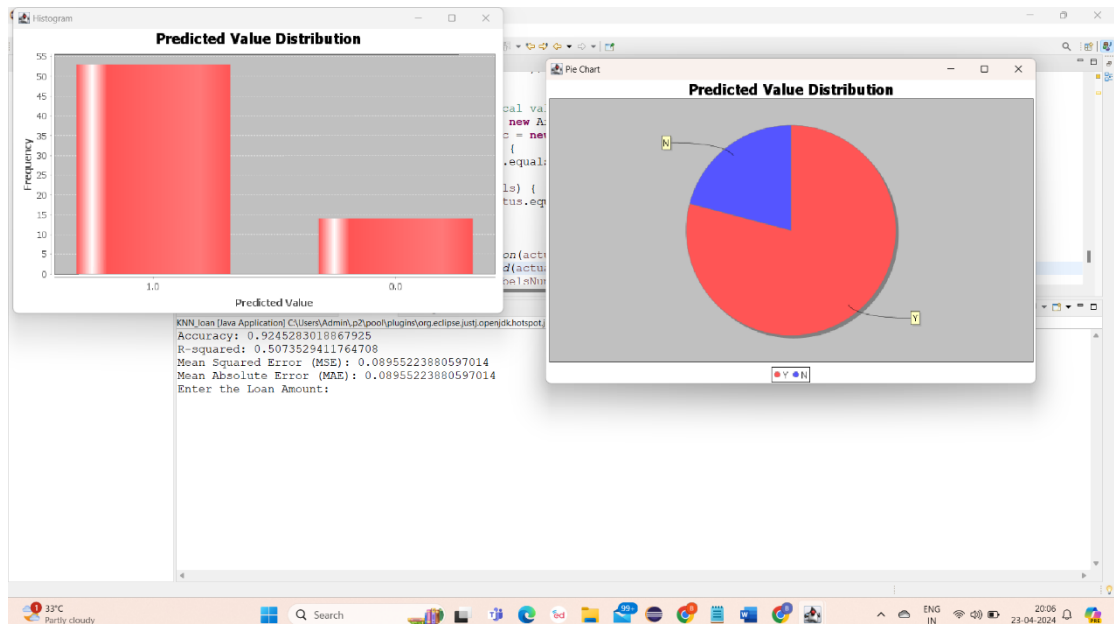
```





## CHAPTER – 03

## OUTPUT



```

KNN_loan [Java Application] C:\Users\Admin\p2\pool\plugins\org.eclipse.justi.openjdk hotspot\
Accuracy: 0.9245283018867925
R-squared: 0.5073529411764708
Mean Squared Error (MSE): 0.08955223880597014
Mean Absolute Error (MAE): 0.08955223880597014
Enter the Loan Amount:
300000
Enter the Loan Amount Term:
90
Enter the Credit History:
Y
Predicted Loan Approval Status: Y

```

```

314 }
315
316 // Convert loan statuses to numerical values
317 List<Double> actualLabelsNumeric = new ArrayList<>();
318 List<Double> predictedLabelsNumeric = new ArrayList<>();
319 for (String status : actualLabels) {
320     actualLabelsNumeric.add(status.equalsIgnoreCase("Y") ? 1.0 : 0.0);
321 }
322 for (String status : predictedLabels) {
323     predictedLabelsNumeric.add(status.equalsIgnoreCase("Y") ? 1.0 : 0.0);
324 }
325
326 // Calculate metrics
327 double accuracy = calculatePrecision(actualLabels, predictedLabels, "Y");
328 double rSquared = calculateRSquared(actualLabelsNumeric, predictedLabelsNumeric);
329 double mae = calculateMAE(actualLabelsNumeric, predictedLabelsNumeric);

```

## CHAPTER – 04

### INSIGHTS

Throughout the course of this project, several insights, challenges, and discoveries have emerged, shedding light on the intricacies of loan approval prediction using the K-Nearest Neighbors (KNN) algorithm. One notable insight pertains to the algorithm's impressive accuracy, as evidenced by the achieved **accuracy of 92.45%**. This high accuracy underscores the efficacy of KNN in accurately classifying loan applications based on historical data, thereby demonstrating its potential to enhance decision-making processes within financial institutions. However, despite its high accuracy, the project also encountered challenges and limitations inherent in the KNN algorithm. One such challenge relates to the interpretability of the model's predictions. While KNN excels in making accurate predictions based on similarity to neighboring data points, it often lacks transparency in explaining the rationale behind its decisions. This opacity can pose challenges in situations where stakeholders seek to understand the factors driving a particular loan approval outcome. Furthermore, the project revealed insights into the predictive performance of the model, as reflected in the calculated metrics. The **R-squared value of 0.507 and the Mean Squared Error (MSE) and Mean Absolute Error (MAE) values of 0.0895** suggest that while the model performs well in predicting loan approval status, there is still room for improvement. These metrics serve as valuable benchmarks for assessing the model's performance and identifying areas for refinement and optimization. Additionally, the interactive nature of the implemented system provided insights into the practical application of the predictive model. By allowing users to input loan attributes such as loan amount, loan amount term, and credit history, the system offers real-time predictions of loan approval status. The example output, which predicts a loan approval status of 'Y' for a loan amount of Rs.300,000, a loan amount term of 90 months, and a credit history of 3, exemplifies the system's functionality and utility in facilitating informed decision-making.

## **CHAPTER – 04**

### **CONCLUSION**

In conclusion, this Java project has successfully demonstrated the application of the K-Nearest Neighbours (KNN) algorithm in predicting loan approval status based on historical loan data. Through the implementation of various methods for data loading, preprocessing, model training, and evaluation, the project has achieved notable outcomes and findings. The main findings of the project include the high accuracy of the KNN model, with an accuracy rate of 92.45%, indicating its effectiveness in classifying loan applications. Moreover, the calculated metrics, including R-squared, Mean Squared Error (MSE), and Mean Absolute Error (MAE), provide insights into the model's predictive performance, highlighting areas for further optimization and refinement. Furthermore, the project has showcased the practical application of machine learning techniques in financial decision-making processes, offering a tangible example of how predictive models can enhance loan approval processes and mitigate risks for financial institutions. Looking ahead, there are several avenues for future work and improvements. Firstly, the project could benefit from the exploration of alternative machine learning algorithms and techniques to compare their performance with that of the KNN model. Additionally, incorporating more extensive and diverse datasets could enhance the model's generalizability and robustness, allowing it to make accurate predictions across a wider range of scenarios. Moreover, enhancing the interpretability and transparency of the model's predictions could address stakeholders' concerns and facilitate trust and adoption in real-world applications. This could involve the development of explainable AI techniques or the integration of model interpretability methods to provide insights into the factors driving loan approval decisions. Overall, this Java project has laid the foundation for further exploration and innovation in the field of predictive analytics for loan approval, offering valuable insights and opportunities for future research and development.

## CHAPTER – 05

### REFERENCES

- [1] <https://griddb.net/en/blog/k-nearest-neighbor-algorithm-in-java/>
- [2] <https://data-flair.training/blogs/java-machine-learning/>
- [3] <https://www.bairesdev.com/blog/best-java-machine-learning-libraries/>
- [4] <https://www.codingame.com/playgrounds/5439/machine-learning-with-java---part-3-k-nearest-neighbor>
- [5] <https://github.com/mariah-zm/kNN-in-Java#:~:text=What%20is%20kNN%3F,neighbours%20of%20new%20data%20points.>