

# **CRIME PREDICTION USING LINEAR REGRESSION MODEL**

**A PROJECT REPORT**

*Submitted by*

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**BONAFIDE CERTIFICATE**

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## **ABSTRACT**

Crime is one of the biggest and dominating problem in our society and its prevention is an Important Task. Daily there are huge numbers of crimes committed frequently. This requires keeping track of all the crimes and maintaining a database for same which may be used for future reference. We demonstrate the utility of the platform by visualizing spatial and temporal relationships in a set of real-world crime datasets. Predictive capabilities of the platform are demonstrated by predicting crime categories, for which a machine learning approach is used. To construct a model by using a algorithms as Random Forest Classifier, Logistic Regression, Support Vector Machine, Linear Regression. The current problems faced are maintaining of proper dataset of crime and analyzing this data to help in predicting and solving crimes. We propose a system which can analysis, detect, and predict various crime probability in given region. This project highlights the potential benefits of crime prediction for law enforcement agencies and policy makers, and discusses the challenges and limitations of this approach. Overall, this project emphasizes the importance of developing effective crime prediction models to improve public safety and reduce crime rates.

**KEYWORDS:** crime prediction, linear regression, public safety, future prediction, crime, Machine learning.

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## **LIST OF ABBREVIATION**

| <b>WORD</b> | <b>ABBREVIATION</b>      |
|-------------|--------------------------|
| <b>KNN</b>  | K Nearest Neighbor       |
| <b>SVM</b>  | Support Vector Machine   |
| <b>RFR</b>  | Random Forest Regression |



# CHAPTER 1

## INTRODUCTION

### **Outline of the project:**

Crimes are the significant threat to the human kind. There are many crimes that happen at regular interval of time. Perhaps it is increasing and spreading at a fast and vast rate. Crimes happen from small village, town to big cities. Crimes are of different type – robbery, murder, rape, assault, battery, false imprisonment, kidnapping, homicide. Since crimes are increasing there is a need to solve the cases in a much faster way. The crime activities have been increased at a faster rate and it is the responsibility of police department to control and reduce the crime activities. Crime prediction and criminal identification are the major problems to the police department as there are tremendous amount of crime data that exist. There is a need of technology through which the case solving could be faster. The aim of this project is to make crime prediction using the features present in the dataset.

The dataset is extracted from the official sites. With the help of machine learning algorithm, using python as core we can predict the type of crime which will occur in a particular area. Through many documentation and cases, it came out that machine learning and data science can make the work easier and faster. The inputs to our algorithms are year, state, district and class of crime:

- Act 379 - Robbery
- Act 13 - Gambling
- Act 279 - Accident
- Act 323 - Violence
- Act 302 - Murder
- Act 363 – Kidnapping

The output is the class of crime that is likely to have occurred. We try out multiple classification algorithms, such as Linear regression, Support Vector Machine (SVM), Logistic Regression and Random Forests.

Crime rate forecast is a scheme that uses different algorithms to determine the crime rate based on prior information. For our daily purposes we must go many places every day and many times in our daily we face numerous security issues such as hijacking, kidnapping, harassment, etc. In general, we see that we are searching for Google Maps when we need to go anywhere at first, Google Maps show that one, two or more ways to get to the destination, but we always choose the shortcut route, but we do not comprehend the path condition properly. Is it safe or not that's why we are faced with many unpleasant circumstances; this research introduces the design and execution of a strategy based on past crime data and analyzes the crime rate in past areas at distinct moments; for this work, we use primary data those are collected from the people based on their previous crime problem.

In our train information collection, we used different algorithms to figure out the highest precision between the Linear Regression algorithms that provides the greatest precision. In this paper, we use different models and table to show the different types of crime rate, mostly working data from last 7 years of crime and showing the level of crime prediction in different issues described in Sect.

**Dataset platforms:**

- Kaggle
- Open.gov.in
- India crime record bureau

**Objective:**

The prediction using machine learning techniques that is prediction rules. Frequent patterns are extracted based on the criteria's like crime type. Prediction is done based on the previous year datasets. The prediction report consists of all the datasets from the year 2015-2022. the year wise comparison is shown based on the state wise datasets. The clustering algorithm can be perform based on every datasets based on each year comparison is made.

## **CHAPTER 2**

### **LITERATURE SURVEY**

**2.1 Project Title:** Crime Analysis and Prediction Using Machine Learning

**Authors:** Roopa, Prof. Thouseef Ulla Khan

**Year:** 09 sep 2022 International Research Journal of Engineering and Technology (IJREAT).

**Description:**

This paper uses data mining approach known as “Prediction Rules” for crime pattern detection as well as automation for early crime prediction. Predicts crime trends based on past crime information, date and location

**Advantage:**

- Conducting criminal analysis and identifying trends in crime.
- Disseminate knowledge to help with the creation of crime reduction and preventive measures.

**Limitations:**

- The accuracy and reliability of machine learning models heavily depend on the quality and representativeness of the input data.

## **2.2 Title : Crime Analysis Using K-Means Clustering**

**Authors:** Khushabu A. Bokde, Tiksha P. Kakade, Dnyaneshwari S. Tumsare, Chetan G. Wadhai

**Published On :** International Journal of Engineering Research & Technology (IJERT) ISSN: 2278-0181 vol.7 Issue 04, April-2018

### **Description :**

In this paper crime analysis is done by performing k-means clustering on crime dataset using rapid miner tool and deploy on the web server.

### **Advantages:**

- Visual and intuitive criminal and intelligence investigation techniques can be developed for crime pattern.
- It can process the data with large number of variables or dimensions making it suitable for analyzing.

### **Limitations:**

- Crime data may have different types of features, such as time series or spatial data, which may require different distance metric

### **2.3 Title : Crime Data Analysis Using Machine Learning**

**Authors :** P. Bhanumathi, S. Greeshma

**Published On:** International Journal of Research in Engineering,  
Volume 12 Issue 07, July 2022

#### **Description:**

This paper uses the data of past 18 years that is collected from various trusted sources. Data pre-processing is as important as final prediction, this paper used feature selection, removing null values and label encoding to clean and nourish the data. This research gives an efficient machine learning model for predicting the next criminal case

#### **Advantage:**

- high accuracy in this model prediction methodology.

#### **Limitations:**

- Over fitting occurs when a model becomes too specialized in the training data and fails to generalize well unseen data.
- The accuracy and reliability of machine learning models heavily depend on the quality and representativeness of the input data.

**2.4 Title:** Empirical Analysis for Crime Prediction and Forecasting  
Using Machine Learning and Dee Learning Techniques

**Authors :** Wajiha Safat, Sohail Asghar and Saira Andleeb Gillani

**Published On :** May 6, 2021 IEEE Conference

**Description:**

In this paper crime forecasting results were further identified in the main regions for both cities, overall these results provide early identification of crime, hot-spots with higher crime rate and future trends with improved predictive accuracy than with other methods and are useful for directing police practice and strategies.

**Advantage:**

- It can handle large data sets with high number of variables efficiently.
- Facilitating scalability and faster computation.

**Limitations:**

- Typically requires a large amount of labeled data for training.

## **2.5 Title:** Crime Data Analysis Using Machine Learning

**Authors :** P.Kiran, B.Mounika, P.Naveen, N.Tejaswini and B.Karthik

**Published On :** Dec 2022, South Asian Journal of Engineering and Technology (SAJET).

### **Description:**

In this project, Machine Learning and data science Techniques are used for crime prediction of Chicago crime data set. For this supervised classification Random Forest method used.

### **Advantage:**

- The model predicts the type of crime with accuracy of 0.789.
- This can lead to more accurate crime predictions compared to traditional methods.

### **Limitations:**

While Random Forest provides accurate predictions, it can be challenging to interpret the underlying decision making process.



**2.6 Title:** Crime Prediction and Analysis Using Machine Learning

**Authors :** S.K. Senthil KUmair, Gampa Adarsh, Jessu Shashank and MD. Abdul Smaeer.

**Published On :** June 2021, International Journal of Techo Engineering (IJTE).

**Description:**

In this project, using the techniques of Machine Learning and data science for crime prediction of Chicago crime data set. The K-Nearest Neighbour and various Algorithms are tested for crime prediction.

**Advantage:**

- This allows it to adapt quickly to new data without requiring retaining the model.

**Limitations:**

- If the majority class dominates the neighbourhood, it may result in misclassification of the minority class.
- It can be slow and memory intensive for large datasets.

## **CHAPTER 3**

### **METHODOLOGY**

#### **3.1 Existing system:**

In the existing system of crime prediction using the K-nearest neighbors (KNN) algorithm is the "Chicago Crime Prediction" system developed by researchers at the University of Chicago. This system uses historical crime data from the Chicago Police Department to predict the type of crime that is likely to occur in a particular location and time period.

In crime prediction, there can be a large number of features to consider, such as location, time, type of crime, weather, and demographics. The KNN algorithm may struggle to make accurate predictions with a high-dimensional dataset.

#### **Limitation:**

- By this methodology they had less accuracy in prediction.
- The prediction results are not perfect.
- This algorithm can be sensitive to noisy data, which can lead to poor performance or incorrect predictions.

#### **3.2 Proposed system:**

The proposed system is made on the basis of the research work that is done by going through various such documentations. Nearly all of the crimes are predicting based on the location and the types of crimes that are occurring in those areas. The crime data is extracted from the official portal of police. It consists of crime information like location description, type of crime, year,

month, state district. Before training of the model data preprocessing will be done following this feature selection and scaling will be done so that accuracy obtain will be high. The Linear Regression classification and various other algorithms (Logistic Regression, Support Vector Machine and Random Forest) will be tested for crime prediction and one with better accuracy will be used for training. Visualization of dataset will be done in terms of graphical representation of many cases for example at which time the criminal rates are high or at which state of the criminal activities are high. The whole purpose of this project is to give a just idea of how machine learning can be used by the law enforcement agencies to detect, predict and solve crimes at a much faster rate and thus reduces the crime rate. This can be used in other states or countries depending upon the availability of the dataset.

The data set contains different types of crimes that being committed in India according to the state and year respectively .This paper takes types of crimes as input and gives the area in which crimes are committed as output. The data pre-processing involves data cleaning, feature selection, dropping null values, data scaling by normalizing and standardizing.

**Advantages:**

- Helps to prevent crime in society
- System will keep historical record of crime.
- System is user friendly
- Saves time
- Gives high accuracy
- Reduce crime rates

## **Machine Learning Algorithms:**

In our project we used Supervised Learning of Algorithms to analyze and predict the crime.

## **Supervised Learning:**

Supervised learning is a type of machine learning algorithm where the model is trained using a labeled dataset. In other words, the data used for training the model includes both input features and their corresponding output labels. The goal of supervised learning is to create a function that can accurately predict output labels for new input data based on patterns learned from the training data.

## **Types of Supervised Learning:**

There are two main types of supervised learning algorithms.

- Regression
- Classification

## **Regression:**

Regression is used when the output variable is a continuous value. The goal is to create a function that can predict a numerical value for a given set of input features. Regression includes some type of algorithms are,

- Linear Regression
- Regression Trees
- Non-Linear Regression
- Bayesian Linear Regression
- Polynomial Regression

**Classification:**

Classification algorithms are used when the output variable is categorical, which

means there are two classes such as Yes-No, Male-Female, True-false, etc.

Spam Filtering, It includes,

- Random Forest
- Logistic Regression
- Support vector Machines

**3.3 Proposed Algorithms:****3.3.1 Linear Regression Algorithm:**

Is a relationship between a dependent variable and one or more independent variables. In other words, linear regression is used to predict a continuous output variable based on one or more input features that have a linear relationship with the output variable.

The linear regression model assumes that there is a linear relationship between the independent variables (also known as features or predictors) and the dependent variable (also known as the response variable or target). This linear relationship is represented by a straight line equation of the form  $y = mx + b$ , where  $y$  is the dependent variable,  $x$  is the independent variable,  $m$  is the slope of the line, and  $b$  is the y-intercept.

Multi-linear regression is a sort of mathematical approach to finding a relation between the dependent variables (Type of crime) and a set of

independent variables those input values gathered from the crime spot. This methodology predicts the Era of the victims district values based on the input characteristics indicate in the metadata column. The multi-linear regression is:

$$Y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_p x_p \quad (1)$$

Here,

Y performs as the reliant variable

X performs the independent variable

$\beta$  represent the coefficient formula function of regression.

### 3.3.2 Random Forest Algorithm:

It is very popular ensemble learning method which builds a number of classifiers on the training data and combines all their outputs to make the best predictions on the test data. Thus, the Random Forests algorithm is a variance minimizing algorithm that uses randomness when making split decisions to help avoid over fitting on the training data. A random forests(Fig 4.5) classifier is an ensemble classifier, which aggregates a family of classifiers  $h(x|\theta_1), h(x|\theta_2), \dots, h(x|\theta_k)$ . Each member of the family,  $h(x|\theta)$ , is a classification tree and  $k$  is the number of trees chosen from a model random vector. Also, each  $\theta_k$  is a randomly chosen parameter vector. If  $D(x,y)$  denotes the training dataset, each classification tree in the ensemble is built using a different subset  $D_{\theta_k}(x,y) \subset D(x,y)$  of the training dataset.

$$y = \operatorname{argmax}_{p \in \{h(x_1) \dots h(x_k)\}} \left\{ \sum_{j=1}^k (I(h(x|\theta_j) = p)) \right\}$$

Thus,  $h(x|\theta_k)$  is the  $k$ th classification tree which uses a subset of features  $x_{\theta_k} \subset x$  to build a classification model. Each tree then works like regular decision trees: it partitions the data based on the value of a particular feature (which is selected randomly from the subset), until the data is fully partitioned, or the maximum allowed depth is reached. The final output  $y$  is obtained by aggregating the results thus: where  $I$  denote the indicator function.

### **3.3.3 Support Vector Machine Algorithm (SVM):**

It works by finding the best possible decision boundary (or hyperplane) that separates the different classes in the data. In the context of classification, SVM seeks to find the hyperplane that maximizes the margin between the different classes, which is the distance between the hyperplane and the closest data points from each class. These closest points are called support vectors, and they play a crucial role in defining the decision boundary.

The basic idea behind SVM is to transform the input data into a higher dimensional space, where a linear decision boundary can be found to separate the classes. This transformation is performed by a function called the kernel function, which calculates the similarity between each pair of data points in the input space.

There are several types of kernel functions that can be used with SVM, such as linear, polynomial, radial basis function (RBF), and sigmoid. The choice of kernel function depends on the nature of the data and the problem at hand.

### **3.3.4 Logistic Regression Algorithm:**

Logistic regression is one of the most popular Machine Learning algorithms, which comes under the Supervised Learning technique. It is used for predicting the categorical dependent variable using a given set of independent variables.

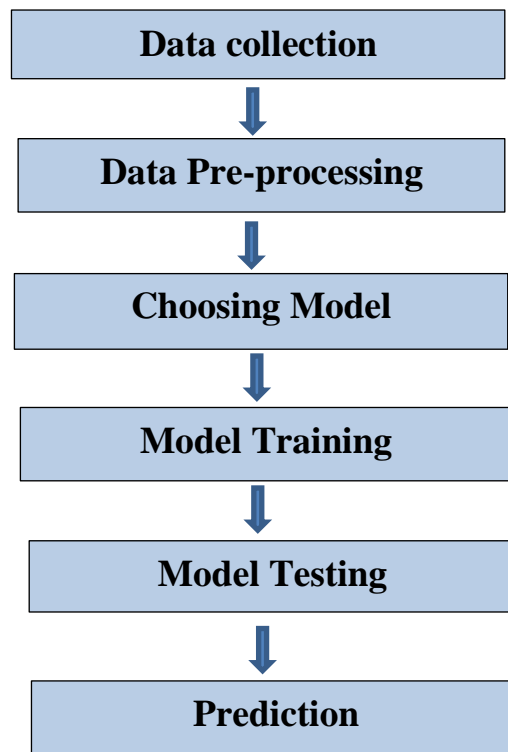
Logistic regression predicts the output of a categorical dependent variable. Therefore the outcome must be a categorical or discrete value. It can be either Yes or No, 0 or 1, true or False, etc. but instead of giving the exact value as 0 and 1, it gives the probabilistic values which lie between 0 and 1. Logistic Regression is much similar to the Linear Regression except that how they are used. Linear Regression is used for solving Regression problems, whereas Logistic regression is used for solving the classification problems. In Logistic regression, instead of fitting a regression line, we fit an "S" shaped logistic function, which predicts two maximum values (0 or 1).



## CHAPTER 4

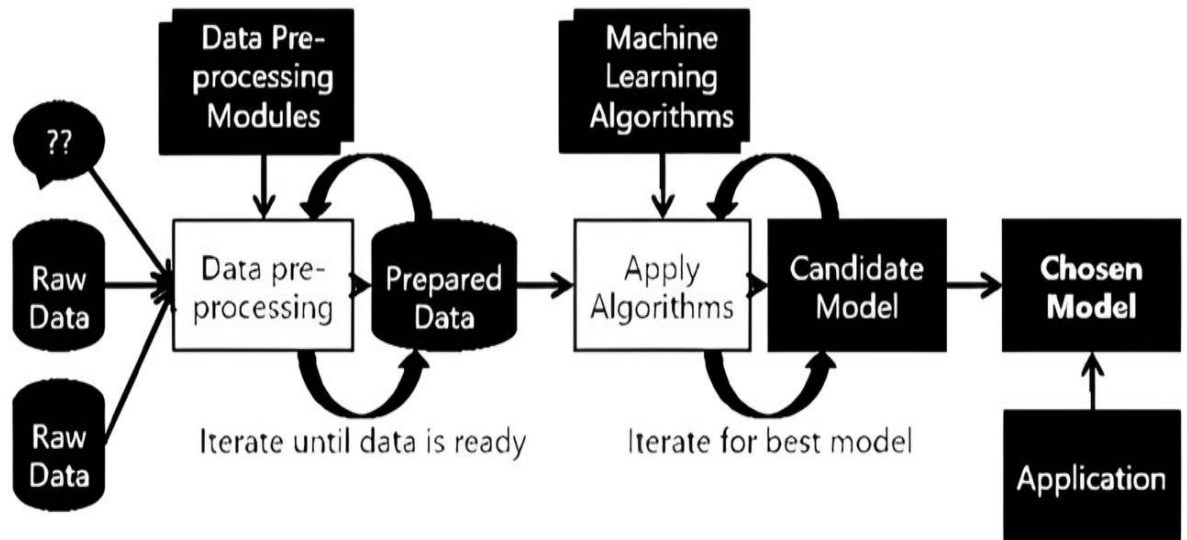
### DIAGRAM REPRESENTATION

#### 4.1 Data Flow Diagram



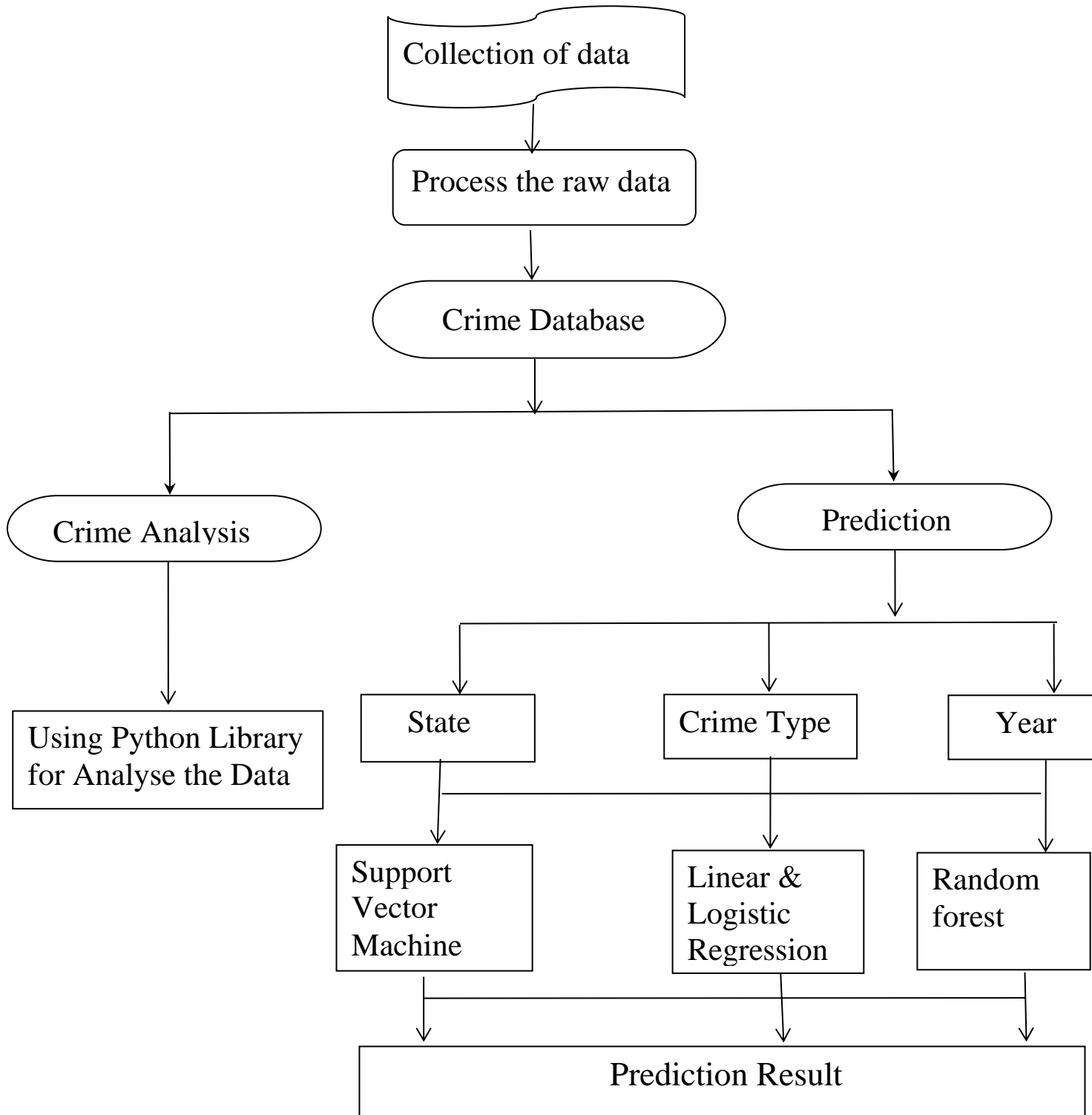
*Fig 4.1. Data flow diagram*

## 4.2 Simple Architecture Diagram



*Fig 4.2. Simple Architecture Diagram*

### 4.3 Detailed Architecture Diagram



*Fig 4.3. Detailed Architecture Diagram*

## **CHAPTER 5**

### **SYSTEM SPECIFICATION**

#### **5.1 Hardware Requirements:**

- Operating system : windows, linux
- Ram : minimum 4 gb
- Hard disk : minimum 250gb

#### **5.2 Software Requirements:**

- Python idel 3.7 version (or)
- Jupiter (or)
- Google colab (or)
- Visual studio

#### **Visual studio:**

Visual Studio is a powerful developer tool that you can use to complete the entire development cycle in one place. It is a comprehensive integrated development environment (IDE) that you can use to write, edit, debug, and build code, and then deploy your app. Beyond code editing and debugging, Visual Studio includes compilers, code completion tools, source control, extensions, and many more features to enhance every stage of the software development process.

## **Python:**

Here we use Python programming Language for a back-end process. Python is an interpreted high-level general-purpose programming language. Its design philosophy emphasizes code readability with its use of significant indentation. Its language constructs as well as its object-oriented approach aim to help programmers write clear, logical code for small and large-scale projects. Python is dynamically-typed and garbage-collected. It supports multiple programming paradigms, including structured (particularly, procedural), object-oriented and functional programming.

## **Creating Environment in VS:**

A conda environment is a Python environment that's managed using the conda package manager (see Getting started with conda). Whether to use a conda environment or a virtual one will depend on your packaging needs, what your team has standardized on, etc.

From within VS Code, you can create local environments, using virtual environments or Anaconda, by opening the Command Palette (Ctrl+Shift+P), start typing the Python: Create Environment command to search, and then select the command. The command presents a list of environment types: Venv or Conda.

A virtual environment is a built-in way to create an environment to isolate the packages you install per workspace. A virtual environment creates a folder that contains a copy (or symlink) to a specific interpreter. When you install packages into a virtual environment it will end up in this new folder so that they are not interspersed with other packages used or needed by other workspaces.

The Python extension automatically detects existing conda environments. We recommend you install a Python interpreter into your conda environment, otherwise one will be installed for you after you select the environment. For example, the following command creates a conda environment named env-01 with a Python 3.9 interpreter and several libraries.

```
conda create -n env-01 python=3.9 scipy=0.15.0 num.
```

### **Data Manipulation Packages:**

#### **Pandas:**

Pandas is a Python package providing fast, flexible, and expressive data structures designed to make working with structured (tabular, multidimensional, potentially heterogeneous) and time series data both easy and intuitive. It aims to be the fundamental high-level building block for doing practical, real world data analysis in Python. Additionally, it has the broader goal of becoming the most powerful and flexible open-source data analysis / manipulation tool available in any language. It is already well on its way toward this goal.

#### **NumPy:**

NumPy is a library for Python, adding support for large, multidimensional arrays and matrices, along with a large collection of high-level mathematical functions to operate on these arrays. NumPy is an open-source software and has many contributors. In comparison, MATLAB boasts a large number of additional toolboxes, notably Simulink, whereas NumPy is intrinsically integrated with Python, a more modern and complete programming language. Moreover, complementary Python packages are available; SciPy is a library that adds more MATLAB-like functionality and Matplotlib is a plotting package that provides MATLAB-like plotting functionality.

## **Model Building Package:**

### **Scikit-plot:**

Scikit-plot is the result of an unartistic data scientist's dreadful realization that visualization is one of the most crucial components in the data science process, not just a mere afterthought. Gaining insights is simply a lot easier when you're looking at a colored heatmap of a confusion matrix complete with class labels rather than a single-line dump of numbers enclosed in brackets. Besides, if you ever need to present your results to someone 15 (virtually any time anybody hires you to do data science), you show them visualizations, not a bunch of numbers in Excel. All in all, it is an intuitive library used to add plotting functionality to a scikit-learn object

### **Scikit-learn:**

Scikit-learn (formerly scikits. learn) is a free software machine learning library for the Python programming language. It features various classification, regression and clustering algorithms including support vector machines(svm), random forest, gradient boosting, k means and DBSCAN, and is designed to interoperate with the Python numerical and scientific libraries NumPy and SciPy. It has been built on top of NumPy, SciPy and matplotlib.

### **Matplotlib's pyplot:**

Matplotlib is a Python 2D plotting library which produces publication quality figures in a variety of hardcopy formats and interactive environments across platforms. Matplotlib can be used in Python scripts, the Python and IPython shells, the Jupyter notebook, web application servers, and four graphical user interface toolkits.

It provides an object-oriented API for embedding plots into applications using general-purpose GUI toolkits like Tkinter, wxPython, Qt or GTK.matplotlib. Pyplot provides a MATLAB-like plotting framework pylab combines pyplot with NumPy into a single namespace. This is convenient for interactive work, but for programming it is recommended that the namespaces be kept separate.

\



## **CHAPTER 6**

### **SOFTWARE TESTING**

The test scenario is a detailed document of test cases that cover end to end functionality of a software application in liner statements. The liner statement is considered as a scenario. The test scenario is a high-level classification of testable requirements. These requirements are grouped on the basis of the functionality of a module and obtained from the use cases. In the test scenario, there is a detailed testing process due to many associated test cases. Before performing the test scenario, the tester has to consider the test cases for each scenario. Documentation testing can start at the very beginning of the software process and hence save large amounts of money, since the earlier a defect is found the less it will cost to be fixed. The most popular testing documentation files are test reports, plans, and checklists. These documents are used to outline the team's workload and keep track of the process. Let's take a look at the key requirements for these files and see how they contribute to the process. Test strategy. An outline of the full approach to product testing. As the project moves along, developers, designers, product owners can come back to the document and see if the actual performance corresponds to the planned activities. Test data. The data that testers enter into the software to verify certain features and their outputs. Examples of such data can be fake user profiles, statistics, media content, similar to files that would be uploaded by an end-user in a ready solution. Test plans. A file that describes the strategy, resources, environment, limitations, and schedule of the testing process. It's the fullest testing document, essential for informed planning. Such a document is distributed between team members and shared with all stakeholders.

Test scenarios. In scenarios, testers break down the product's functionality and interface by modules and provide real-time status updates at all testing stages. A module can be described by a single statement, or require hundreds of statuses, depending on its size and scope.

### **Test Driven Development:**

Test Driven Development, or TDD, is a code design technique where the programmer writes a test before any production code, and then writes the code that will make that test pass. The idea is that with a tiny bit of assurance from that initial test, the programmer can feel free to refactor and refactor some more to get the cleanest code they know how to write. The idea is simple, but like most simple things, the execution is hard. TDD requires a completely different mind set from what most people are used to and the tenacity to deal with a learning curve that may slow you down at first.

### **Unit Testing:**

Unit testing focuses on verifying the effort on the smallest unit of software-module. The local data structure is examined to ensure that the data stored temporarily maintains its integrity during all steps in the algorithm's execution. Boundary conditions are tested to ensure that the module operates properly at boundaries established to limit or restrict processing.

**Integration Testing:**

Data can be tested across an interface. One module can have an inadvertent, adverse effect on the other. Integration testing is a systematic technique for constructing a program structure while conducting tests to uncover errors associated with interring.

**Validation Testing:**

It begins after the integration testing is successfully assembled. Validation succeeds when

the software functions in a manner that can be reasonably accepted by the client. In this the majority of the validation is done during the data entry operation where there is a maximum possibility of entering wrong data. Other validation will be performed in all process where correct details and data should be entered to get the required results.

**Recovery Testing:**

Recovery Testing is a system that forces the software to fail in variety of ways and verifies that the recovery is properly performed. If recovery is automatic, re-initialization, and data recovery are each evaluated for correctness.

### **Security Testing:**

Security testing attempts to verify that protection mechanism built into system will in fact protect it from improper penetration. The tester may attempt to acquire password through external clerical means, may attack the system with custom software design to break down any defenses to others, and may purposely cause errors.

### **Performance Testing:**

Performance Testing is used to test runtime performance of software within the context of an integrated system. Performance test are often coupled with stress testing and require both software instrumentation.

### **Specification Testing:**

We can set with, what program should do and how it should perform under various condition. This testing is a comparative study of evolution of system performance and system requirements.

### **Module Level Testing:**

In this the error will be found at each individual module, it encourages the programmer to find and rectify the errors without affecting the other modules

**Blackbox Testing:**

Black- box testing focuses on functional requirement of software. It enables to derive sets of input conditions that will fully exercise all functional requirements for a program

**Output Testing:**

After performing the validation testing, the next step is output testing of the proposed system since no system would be termed as useful until it does produce the required output in the specified format. Output format is considered in two ways, the screen format and the printer format.

**User Acceptance Testing:**

User Acceptance Testing is the key factor for the success of any system. The system under consideration is tested for user acceptance by constantly keeping in touch with prospective system users at the time of developing and making changes whenever required.

## **CHAPTER 7**

### **MODULES**

#### **List of Modules:**

- Data collection Module
- Data Preprocessing Module
- Feature selection Module
- Building and Training Model
- Prediction Module
- Visualization Module

#### **Data collection Module:**

Collection of Crime dataset from kaggle, open gov.in and India crime record bureau is used in CSV format. The crime dataset is extracted from primary data collection based on field work. This dataset consists of about 433 in 10 rows details. The key features such as State, Years, Crime Type, and it contains 2015 to 2022 years are selected from the dataset as the system input features.

#### **Data Preprocessing Module:**

There are 5000 entries are present in the dataset. The null values are removed using `df = df.dropna()` where `df` is the data frame. The categorical attributes (Location, Block, Crime Type, Community Area) are converted into numeric using Label Encoder. The date attribute is splitted into new attributes like month and hour which can be used as features for the model.

|    | A1               |            | fx   | STATE/UT |      |      |      |      |      |      |   |   |   |   |   |   |   |   |  |
|----|------------------|------------|------|----------|------|------|------|------|------|------|---|---|---|---|---|---|---|---|--|
|    | A                | B          | C    | D        | E    | F    | G    | H    | I    | J    | K | L | M | N | O | P | Q | R |  |
| 1  | STATE/UT         | CRIME HEAD | 2015 | 2016     | 2017 | 2018 | 2019 | 2020 | 2021 | 2022 |   |   |   |   |   |   |   |   |  |
| 2  | ANDHRA PRA       | MURDER     | 2685 | 2690     | 2449 | 2538 | 2808 | 2717 | 2484 | 1175 |   |   |   |   |   |   |   |   |  |
| 3  | ARUNACHAL        | MURDER     | 68   | 72       | 59   | 75   | 65   | 73   | 69   | 86   |   |   |   |   |   |   |   |   |  |
| 4  | ASSAM            | MURDER     | 1374 | 1426     | 1323 | 1223 | 1303 | 1368 | 1354 | 1451 |   |   |   |   |   |   |   |   |  |
| 5  | BIHAR            | MURDER     | 3034 | 3139     | 3152 | 3362 | 3198 | 3566 | 3441 | 3403 |   |   |   |   |   |   |   |   |  |
| 6  | CHHATTISGARH     | MURDER     | 1097 | 1169     | 1083 | 1065 | 1110 | 998  | 945  | 998  |   |   |   |   |   |   |   |   |  |
| 7  | GOA              | MURDER     | 33   | 49       | 53   | 35   | 48   | 45   | 36   | 34   |   |   |   |   |   |   |   |   |  |
| 8  | GUJARAT          | MURDER     | 1166 | 1106     | 1020 | 1048 | 1126 | 1126 | 1118 | 1124 |   |   |   |   |   |   |   |   |  |
| 9  | HARYANA          | MURDER     | 911  | 921      | 948  | 1005 | 1062 | 991  | 982  | 1106 |   |   |   |   |   |   |   |   |  |
| 10 | HIMACHAL PRADESH | MURDER     | 127  | 129      | 125  | 132  | 130  | 113  | 104  | 130  |   |   |   |   |   |   |   |   |  |
| 11 | JAMMU & KASHMIR  | MURDER     | 318  | 237      | 237  | 217  | 169  | 124  | 150  | 156  |   |   |   |   |   |   |   |   |  |
| 12 | JHARKHAND        | MURDER     | 1617 | 1697     | 1636 | 1689 | 1747 | 1694 | 1630 | 1658 |   |   |   |   |   |   |   |   |  |
| 13 | KARNATAKA        | MURDER     | 1538 | 1698     | 1702 | 1805 | 1820 | 1860 | 1601 | 1636 |   |   |   |   |   |   |   |   |  |
| 14 | KERALA           | MURDER     | 367  | 362      | 343  | 363  | 365  | 374  | 372  | 367  |   |   |   |   |   |   |   |   |  |
| 15 | MADHYA PRADESH   | MURDER     | 2244 | 2322     | 2386 | 2423 | 2511 | 2373 | 2112 | 2274 |   |   |   |   |   |   |   |   |  |
| 16 | MAHARASHTRA      | MURDER     | 2693 | 2795     | 2653 | 2744 | 2818 | 2712 | 2512 | 2670 |   |   |   |   |   |   |   |   |  |
| 17 | MANIPUR          | MURDER     | 240  | 187      | 131  | 92   | 78   | 83   | 93   | 96   |   |   |   |   |   |   |   |   |  |
| 18 | MEGHALAYA        | MURDER     | 114  | 126      | 128  | 134  | 170  | 137  | 166  | 159  |   |   |   |   |   |   |   |   |  |
| 19 | MIZORAM          | MURDER     | 43   | 35       | 31   | 48   | 26   | 30   | 27   | 42   |   |   |   |   |   |   |   |   |  |
| 20 | NAGALAND         | MURDER     | 111  | 143      | 46   | 45   | 46   | 75   | 78   | 58   |   |   |   |   |   |   |   |   |  |
| 21 | ODISHA           | MURDER     | 1210 | 1250     | 1250 | 1308 | 1477 | 1548 | 1454 | 1450 |   |   |   |   |   |   |   |   |  |
| 22 | PUNJAB           | MURDER     | 760  | 769      | 853  | 907  | 842  | 855  | 711  | 767  |   |   |   |   |   |   |   |   |  |
| 23 | RAJASTHAN        | MURDER     | 1303 | 1347     | 1395 | 1471 | 1461 | 1461 | 1573 | 1637 |   |   |   |   |   |   |   |   |  |

## Feature selection Module:

Features selection is done which can be used to build the model. The attributes used for feature selection are Block, Location, State, Year, Crime Type, X coordinate, Y coordinate, Latitude, Longitude, Hour and month.

## Building and Training Model:

After feature selection Crime Type and year attribute are used for training. The dataset is divided into pairs of xtrain, train and xtest, y test. The algorithm model is imported from sklearn. Building models is done using models. Fit (xtrain, ytrain).

**Prediction Module:**

After the model is built using the above process, prediction is done using `model.predict(xtest)`. The accuracy is calculated using `accuracy_score` imported from `metrics` - `metrics.accuracy_score (ytest, predicted)`.

**Visualization Module:**

Using `matplotlib` library from `sklearn`. Analysis of the crime dataset is done by plotting various graphs



## CHAPTER 8

### FLASK FRAMEWORK

Flask is a web application framework written in Python. Armin Ronacher, who leads an international group of Python enthusiasts named Pocco, develops it. Flask is based on the Werkzeug WSGI toolkit and Jinja2 template engine. Both are Pocco project Http protocol is the foundation of data communication on the world wide web. Different methods of data retrieval from specified URLs are defined in this protocol.

Save the following script as login.html

navi.html

```
<!DOCTYPE html>
<html lang="en" dir="ltr">
  <head>
    <meta charset="utf-8">
    <title>Responsive Navbar</title>
    <meta name="viewport" content="width=device-width, initial-
scale=1.0">
    <link rel="stylesheet" href="C:\Users\Rex\Desktop\crime prediction
website\navi.css">
    <script src="https://kit.fontawesome.com/a076d05399.js"></script>
  </head>
  <body>
    <nav>
      <input type="checkbox" id="check">
      <label for="check" class="checkbtn">
        <i class="fas fa-bars"></i>
      </label>
      <label class="logo">CRIME PREDICTION & ANALYSIS</label>
      <ul>
        <li><a class="active" href="C:\Users\Rex\Desktop\crime prediction
website\login.html">Home</a></li>
        <li><a href="C:\Users\Rex\Desktop\crime prediction
```

```

website\about.html">About</a></li>
    <li><a      href="C:\Users\Rex\Desktop\crime      prediction
website\crime.html">Services</a></li>
    <li><a      href="C:\Users\Rex\Desktop\crime      prediction
website\contact.html">Contact</a></li>
</ul>
</nav>
<section>

```

```

```

```

    <p>Crime is a term used to describe an unlawful or illegal act that
violates a set of established laws or regulations.

```

```

    It is an action that goes against the moral or ethical standards of a
society or community.

```

```

    Crimes can range from minor offenses,such as traffic violations or
petty theft, to more serious offenses, such as murder, robbery, and assault.

```

```

    To combat crime, law enforcement agencies and justice systems are
in place in most countries. These systems work to prevent and investigate
crimes,

```

```

    arrest and prosecute offenders, and ensure that justice is served. The
aim is to provide a safe and secure environment for individuals and
communities to live and work in.

```

```

    However, crime continues to be a significant challenge, and
understanding the root causes, such as poverty and social inequality, is
essential in developing effective strategies to prevent it. </p>

```

```

    </section>

```

```

</body>
</html>

```

login.html

```

<!DOCTYPE html>
<html lang="en">
<head>
    <meta charset="UTF-8">
    <meta http-equiv="X-UA-Compatible" content="IE=edge">

```

```
<meta name="viewport" content="width=device-width, initial-  
scale=1.0">
```

```
<title> Home Login and Signup Form </title>
```

```
<!-- CSS -->
```

```
<link rel="stylesheet" href="C:\Users\Rex\Desktop\crime prediction  
website\login.css">
```

```
<!-- Boxicons CSS -->
```

```
<link href='https://unpkg.com/boxicons@2.1.2/css/boxicons.min.css'  
rel='stylesheet'>
```

```
</head>
```

```
<body>
```

```
<section class="container forms">
```

```
<div class="form login">
```

```
<div class="form-content">
```

```
<header>Login</header>
```

```
<form action="#">
```

```
<div class="field input-field">
```

```
<input type="email" placeholder="Email"  
class="input">
```

```
</div>
```

```
<div class="field input-field">
```

```
<input type="password"  
placeholder="Password" class="password">
```

```
<i class='bx bx-hide eye-icon'></i>
```

```
</div>
```

```
<div class="form-link">
```

```
<a href="#" class="forgot-pass">Forgot  
password?</a>
```

```
</div>
```

```
<div class="field button-field">
```

```
<button>Login</button>
```

```
</div>
```

```
</form>
```

```

        <div class="form-link">
            <span>Don't have an account? <a href="#"
class="link signup-link">Signup</a></span>
        </div>
    </div>

```

```

<div class="line"></div>

```

```

<div class="media-options">
    <a href="#" class="field facebook">
        <i class='bx bxl-facebook facebook-icon'></i>
        <span>Login with Facebook</span>
    </a>
</div>

```

```

<div class="media-options">
    <a href="#" class="field google">
        
        <span>Login with Google</span>
    </a>
</div>

```

```

</div>

```

```

<!-- Signup Form -->

```

```

<div class="form signup">
    <div class="form-content">
        <header>Signup</header>
        <form action="#">
            <div class="field input-field">
                <input type="email" placeholder="Email"
class="input">
            </div>

            <div class="field input-field">
                <input type="password" placeholder="Create
password" class="password">
            </div>

```

```

        <div class="field input-field">
            <input type="password" placeholder="Confirm
password" class="password">
            <i class='bx bx-hide eye-icon'></i>
        </div>

        <div class="field button-field">
            <button>Signup</button>
        </div>
    </form>

    <div class="form-link">
        <span>Already have an account? <a href="#"
class="link login-link">Login</a></span>
    </div>
</div>

<div class="line"></div>

<div class="media-options">
    <a href="#" class="field facebook">
        <i class='bx bxl-facebook facebook-icon'></i>
        <span>Login with Facebook</span>
    </a>
</div>

<div class="media-options">
    <a href="#" class="field google">
        
        <span>Login with Google</span>
    </a>
</div>

</div>
</section>

<!-- JavaScript -->
<script src="C:\Users\Rex\Desktop\crime prediction
website\login.js"></script>

```

```

    </body>
</html>

```

crime.html

```

<!DOCTYPE html>
<html>
  <head>
    <title>CRIME PREDICTION & ANALYSIS</title>
    <script src="https://code.jquery.com/jquery-3.6.0.min.js"></script>

  </head>
  <body>
    <form action="/predict" method="POST" enctype="multipart/form-
data">
      <h2>CRIME PREDICTION & ANALYSIS</h2>
      <link rel="stylesheet" href="C:\Users\Rex\Desktop\crime prediction
website\crime.css">
      <label for="STATE/UT">STATE/UT:</label>
      <input type="text" id="STATE/UT" name="STATE/UT" required>
      <label for="CRIME TYPE ">CRIME TYPE:</label>
      <input type="text" id="CRIME TYPE" name="CRIME TYPE"
required>
      <Label for="YEAR">YEAR:</Label>
      <input type="number" id="year" name="year" min="1900"
max="2099" step="1" value="2022" required>
      <label for="dataset">Upload Dataset:</label>
      <input type="file" id="dataset" name="dataset" accept=".csv">
      <div class="button-wrapper">
        <button type="button" onclick="predict()">Predict Crime
Rate</button>
      </div>
    </form>
  </body>
</html>

```

contact.html

```

<!DOCTYPE html>

```

```

<!-- Created By CodingNepal -->
<html lang="en" dir="ltr">
  <head>
    <meta charset="utf-8">
    <title>Responsive Contact us Form | CodingNepal</title>
    <link rel="stylesheet" href="C:\Users\Rex\Desktop\crime prediction
website\contact.css">
    <meta name="viewport" content="width=device-width, initial-
scale=1.0">
  </head>
  <body>
    <div class="container">
      <div class="text"> Contact us Form</div>
      <form action="#">
        <div class="form-row">
          <div class="input-data">
            <input type="text" required>
            <div class="underline"></div>
            <label for="">First Name</label>
          </div>
          <div class="input-data">
            <input type="text" required>
            <div class="underline"></div>
            <label for="">Last Name</label>
          </div>
        </div>
        <div class="form-row">
          <div class="input-data">
            <input type="text" required>
            <div class="underline"></div>
            <label for="">Email Address</label>
          </div>
          <div class="input-data">
            <input type="text" required>
            <div class="underline"></div>
            <label for="">Website Name</label>
          </div>
        </div>
        <div class="form-row">
          <div class="input-data textarea">

```

```

        <textarea rows="8" cols="80" required></textarea>
        <br />
        <div class="underline"></div>
        <label for="">Write your message</label>
        <br />
        <div class="form-row submit-btn">
            <div class="input-data">
                <div class="inner"></div>
                <input type="submit" value="submit">
            </div>
        </div>
    </div>
</div>
</form>
</div>
</body>
</html>

```

about.html

```

<!DOCTYPE html>
<html lang="en" dir="ltr">
    <head>
        <meta charset="utf-8">
        <title>Project Guide and Team Members</title>
        <link rel="stylesheet" href="C:\Users\Rex\Desktop\crime prediction
website\about.css">
        <meta name="viewport" content="width=device-width, initial-
scale=1.0">
        <!-- Fontawesome Link for Icons -->
        <link rel="stylesheet" href="https://cdnjs.cloudflare.com/ajax/libs/font-
awesome/6.3.0/css/all.min.css">
        <script src="C:\Users\Rex\Desktop\crime prediction website\about.js"
defer></script>
    </head>
    <body>
        <h2>Project Guide and Team Members</h2>
        <div class="wrapper">
            <i id="left" class="fa-solid fa-angle-left"></i>
            <ul class="carousel">

```



```

        <li class="card">
            <div class="img"></div>
            <h2>MR. D. PRABHU - (B.E, M.E)</h2>
            <span>Head of the department,LIT</span>
        </li>
        <li class="card">
            <div class="img"></div>
            <h2>Amy Delphiya Mary R</h2>
            <span>Final year IT</span>
        </li>
        <li class="card">
            <div class="img"></div>
            <h2>Keerthana v</h2>
            <span>Final year IT</span>
        </li>
        <li class="card">
            <div class="img"></div>
            <h2>Thilagavathi S</h2>
            <span>Final year IT</span>
        </li>
    </ul>
    <i id="right" class="fa-solid fa-angle-right"></i>
</div>

```

```

</body>
</html>

```

css file

```

/* Import Google font - Poppins */
@import
url('https://fonts.googleapis.com/css2?family=Poppins:wght@400;500;600&
display=swap');
* {
    margin: 0;

```

```

padding: 0;
box-sizing: border-box;
font-family: "Poppins", sans-serif;
}
body {
display: flex;
padding: 0 35px;
align-items: center;
justify-content: center;
min-height: 100vh;
background: linear-gradient(to left top, #4f07a1, #580c7d);
}
h2 {
font-size: 24px;
font-weight: bold;
margin-bottom: 20px;
text-align: center;
color: white;
border-bottom: #080564;

}

.wrapper {
max-width: 1100px;
width: 100%;
position: relative;
}
.wrapper i {
top: 50%;
height: 50px;
width: 50px;
cursor: pointer;
font-size: 1.25rem;
position: absolute;
text-align: center;
line-height: 50px;
background: hsla(240, 33%, 99%, 0.839);
border-radius: 50%;
box-shadow: 0 3px 6px rgba(0,0,0,0.23);
transform: translateY(-50%);

```

```

    transition: transform 0.1s linear;
  }
.wrapper i:active{
  transform: translateY(-50%) scale(0.85);
}
.wrapper i:first-child{
  left: -22px;
}
.wrapper i:last-child{
  right: -22px;
}
.wrapper .carousel{
  display: grid;
  grid-auto-flow: column;
  grid-auto-columns: calc((100% / 3) - 12px);
  overflow-x: auto;
  scroll-snap-type: x mandatory;
  gap: 16px;
  border-radius: 8px;
  scroll-behavior: smooth;
  scrollbar-width: none;
}
.carousel::-webkit-scrollbar {
  display: none;
}
.carousel.no-transition {
  scroll-behavior: auto;
}
.carousel.dragging {
  scroll-snap-type: none;
  scroll-behavior: auto;
}
.carousel.dragging .card {
  cursor: grab;
  user-select: none;
}
.carousel :where(.card, .img) {
  display: flex;
  justify-content: center;
  align-items: center;

```

```

}
.carousel .card {
  scroll-snap-align: start;
  height: 342px;
  list-style: none;
  background: rgb(251, 250, 251);
  cursor: pointer;
  padding-bottom: 15px;
  flex-direction: column;
  border-radius: 8px;
}
.carousel .card .img {
  background: #6d079c,#363295;
  height: 148px;
  width: 148px;
  border-radius: 50%;
}
.carousel .card .img img {
  width: 140px;
  height: 140px;
  border-radius: 50%;
  object-fit: cover;
  border: 4px solid #73139f;
}
.carousel .card h2 {
  font-weight: 500;
  font-size: 1.56rem;
  margin: 30px 0 5px;
}
.carousel .card span {
  color: #680f95;
  font-size: 1.31rem;
}

@media screen and (max-width: 900px) {
  .wrapper .carousel {
    grid-auto-columns: calc((100% / 2) - 9px);
  }
}

```

```

@media screen and (max-width: 600px) {
  .wrapper .carousel {
    grid-auto-columns: 100%;
  }
}
contact.css
@import
url('https://fonts.googleapis.com/css?family=Poppins:400,500,600,700&dis
lay=swap');
*{
  margin: 0;
  padding: 0;
  outline: none;
  box-sizing: border-box;
  font-family: 'Poppins', sans-serif;
}
body{
  display: flex;
  align-items: center;
  justify-content: center;
  min-height: 100vh;
  padding: 40px;
  background: linear-gradient(115deg, #56d8e4 10%, #9f01ea 90%);
}
.container{
  max-width: 800px;
  background: #fff;
  width: 800px;
  padding: 25px 40px 10px 40px;
  box-shadow: 0px 0px 10px rgba(0,0,0,0.1);
}
.container .text{
  text-align: center;
  font-size: 35px;
  font-weight: 600;
  background: -webkit-linear-gradient(right, #56d8e4, #9f01ea, #56d8e4,
#9f01ea);
  -webkit-background-clip: text;
  -webkit-text-fill-color: transparent;
}

```

```

.container form{
  padding: 30px 0 0 0;
}
.container form .form-row{
  display: flex;
  margin: 32px 0;
}
form .form-row .input-data{
  width: 100%;
  height: 40px;
  margin: 0 20px;
  position: relative;
}
form .form-row .textarea{
  height: 70px;
}
.input-data input,
.textarea textarea{
  display: block;
  width: 100%;
  height: 100%;
  border: none;
  font-size: 17px;
  border-bottom: 2px solid rgba(0,0,0, 0.12);
}
.input-data input:focus ~ label, .textarea textarea:focus ~ label,
.input-data input:valid ~ label, .textarea textarea:valid ~ label{
  transform: translateY(-20px);
  font-size: 14px;
  color: #3498db;
}
.textarea textarea{
  resize: none;
  padding-top: 10px;
}
.input-data label{
  position: absolute;
  pointer-events: none;
  bottom: 10px;
  font-size: 16px;
}

```

```

    transition: all 0.3s ease;
}
.textarea label{
    width: 100%;
    bottom: 40px;
    background: #fff;
}
.input-data .underline{
    position: absolute;
    bottom: 0;
    height: 2px;
    width: 100%;
}
.input-data .underline:before{
    position: absolute;
    content: "";
    height: 2px;
    width: 100%;
    background: #3498db;
    transform: scaleX(0);
    transform-origin: center;
    transition: transform 0.3s ease;
}
.input-data input:focus ~ .underline:before,
.input-data input:valid ~ .underline:before,
.textarea textarea:focus ~ .underline:before,
.textarea textarea:valid ~ .underline:before{
    transform: scale(1);
}
.submit-btn .input-data{
    overflow: hidden;
    height: 45px !important;
    width: 25% !important;
}
.submit-btn .input-data .inner{
    height: 100%;
    width: 300%;
    position: absolute;
    left: -100%;
    background: -webkit-linear-gradient(right, #56d8e4, #9f01ea, #56d8e4,

```

```

#9f01ea);
    transition: all 0.4s;
}
.submit-btn .input-data:hover .inner{
    left: 0;
}
.submit-btn .input-data input{
    background: none;
    border: none;
    color: #fff;
    font-size: 17px;
    font-weight: 500;
    text-transform: uppercase;
    letter-spacing: 1px;
    cursor: pointer;
    position: relative;
    z-index: 2;
}
@media (max-width: 700px) {
    .container .text{
        font-size: 30px;
    }
    .container form{
        padding: 10px 0 0 0;
    }
    .container form .form-row{
        display: block;
    }
    form .form-row .input-data{
        margin: 35px 0 !important;
    }
    .submit-btn .input-data{
        width: 40% !important;
    }
}

```

crime.css

form {



```

display: flex;
flex-direction: column;
align-items: center;
margin-top: 50px;
}

h2 {
color: white;
font-size: 28px;
margin-bottom: 20px;
}

label {
display: block;
font-size: 18px;
margin-bottom: 10px;
color: white;
}

input[type="text"],
input[type="number"] {
padding: 10px;
font-size: 16px;
border: 5px solid #ccc;
border-radius: 4px;
margin-bottom: 20px;
width: 50%;
box-sizing: border-box;
}

button {
background-color: #030434;
color: #fff;
border: none;
padding: 10px 20px;
font-size: 16px;
border-radius: 4px;
cursor: pointer;
}

```

```
button:hover {  
    background-color: #174868;  
}
```

```
body {  
    background-color:hsl(224, 92%, 15%);  
}
```

```
input[type="file"] {  
    padding: 10px;  
    font-size: 16px;  
    border: 5px solid #ccc;  
    border-radius: 4px;  
    margin-bottom: 20px;  
    width: 50%;  
    box-sizing: border-box;  
    background-color: #fff; /* Optional: Set background color */  
    color: #be5408; /* Optional: Set text color */  
    outline: none; /* Optional: Remove outline on focus */  
    cursor: pointer;  
}
```

```
input[type="file"]:hover {  
    background-color: #174868; /* Optional: Set hover background color */  
}
```

```
input[type="file"]:focus {  
    box-shadow: 0 0 5px #a0d4ff; /* Optional: Add focus shadow */  
}
```

login.css

```
/* Google Fonts - Poppins */  
@import  
url('https://fonts.googleapis.com/css2?family=Poppins:wght@300;400;500;600&display=swap');
```

```
*{  
    margin: 0;
```

```

padding: 0;
box-sizing: border-box;
font-family: 'Poppins', sans-serif;
}
.container{
height: 100vh;
width: 100%;
display: flex;
align-items: center;
justify-content: center;
background-color: hsl(224, 92%, 15%);
column-gap: 30px;
}
.form{
position: absolute;
max-width: 430px;
width: 100%;
padding: 30px;
border-radius: 6px;
background: #FFF;
}
.form.signup{
opacity: 0;
pointer-events: none;
}
.forms.show-signup .form.signup{
opacity: 1;
pointer-events: auto;
}
.forms.show-signup .form.login{
opacity: 0;
pointer-events: none;
}
header{
font-size: 28px;
font-weight: 600;
color: #232836;
text-align: center;
}
form{

```

```

        margin-top: 30px;
    }
    .form .field{
        position: relative;
        height: 50px;
        width: 100%;
        margin-top: 20px;
        border-radius: 6px;
    }
    .field input,
    .field button{
        height: 100%;
        width: 100%;
        border: none;
        font-size: 16px;
        font-weight: 400;
        border-radius: 6px;
    }
    .field input{
        outline: none;
        padding: 0 15px;
        border: 1px solid#CACACA;
    }
    .field input:focus{
        border-bottom-width: 2px;
    }
    .eye-icon{
        position: absolute;
        top: 50%;
        right: 10px;
        transform: translateY(-50%);
        font-size: 18px;
        color: #8b8b8b;
        cursor: pointer;
        padding: 5px;
    }
    .field button{
        color: #fff;
        background-color: #0171d3;
        transition: all 0.3s ease;
    }

```

```

        cursor: pointer;
    }
    .field button:hover{
        background-color: #016dcb;
    }
    .form-link{
        text-align: center;
        margin-top: 10px;
    }
    .form-link span,
    .form-link a{
        font-size: 14px;
        font-weight: 400;
        color: #232836;
    }
    .form a{
        color: #0171d3;
        text-decoration: none;
    }
    .form-content a:hover{
        text-decoration: underline;
    }
    .line{
        position: relative;
        height: 1px;
        width: 100%;
        margin: 36px 0;
        background-color: #d4d4d4;
    }
    .line::before{
        content: 'Or';
        position: absolute;
        top: 50%;
        left: 50%;
        transform: translate(-50%, -50%);
        background-color: #FFF;
        color: #8b8b8b;
        padding: 0 15px;
    }
    .media-options a{

```

```

        display: flex;
        align-items: center;
        justify-content: center;
    }
    a.facebook{
        color: #fff;
        background-color: #4267b2;
    }
    a.facebook .facebook-icon{
        height: 28px;
        width: 28px;
        color: #0171d3;
        font-size: 20px;
        border-radius: 50%;
        display: flex;
        align-items: center;
        justify-content: center;
        background-color: #fff;
    }
    .facebook-icon,
    img.google-img{
        position: absolute;
        top: 50%;
        left: 15px;
        transform: translateY(-50%);
    }
    img.google-img{
        height: 20px;
        width: 20px;
        object-fit: cover;
    }
    a.google{
        border: 1px solid #CACACA;
    }
    a.google span{
        font-weight: 500;
        opacity: 0.6;
        color: #232836;
    }

```

```

@media screen and (max-width: 400px) {
  .form{
    padding: 20px 10px;
  }
}

```

javascript file

```

const wrapper = document.querySelector(".wrapper");
const carousel = document.querySelector(".carousel");
const firstCardWidth = carousel.querySelector(".card").offsetWidth;
const arrowBtns = document.querySelectorAll(".wrapper i");
const carouselChildrens = [...carousel.children];

let isDragging = false, isAutoPlay = true, startX, startScrollLeft, timeoutId;

// Get the number of cards that can fit in the carousel at once
let cardPerView = Math.round(carousel.offsetWidth / firstCardWidth);

// Insert copies of the last few cards to beginning of carousel for infinite scrolling
carouselChildrens.slice(-cardPerView).reverse().forEach(card => {
  carousel.insertAdjacentHTML("afterbegin", card.outerHTML);
});

// Insert copies of the first few cards to end of carousel for infinite scrolling
carouselChildrens.slice(0, cardPerView).forEach(card => {
  carousel.insertAdjacentHTML("beforeend", card.outerHTML);
});

// Scroll the carousel at appropriate postition to hide first few duplicate cards on Firefox
carousel.classList.add("no-transition");
carousel.scrollLeft = carousel.offsetWidth;
carousel.classList.remove("no-transition");

// Add event listeners for the arrow buttons to scroll the carousel left and right
arrowBtns.forEach(btn => {

```

```

    btn.addEventListener("click", () => {
        carousel.scrollLeft += btn.id === "left" ? -firstCardWidth :
firstCardWidth;
    });
});

```

```

const dragStart = (e) => {
    isDragging = true;
    carousel.classList.add("dragging");
    // Records the initial cursor and scroll position of the carousel
    startX = e.pageX;
    startScrollLeft = carousel.scrollLeft;
}

```

```

const dragging = (e) => {
    if(!isDragging) return; // if isDragging is false return from here
    // Updates the scroll position of the carousel based on the cursor
movement
    carousel.scrollLeft = startScrollLeft - (e.pageX - startX);
}

```

```

const dragStop = () => {
    isDragging = false;
    carousel.classList.remove("dragging");
}

```

```

const infiniteScroll = () => {
    // If the carousel is at the beginning, scroll to the end
    if(carousel.scrollLeft === 0) {
        carousel.classList.add("no-transition");
        carousel.scrollLeft = carousel.scrollWidth - (2 *
carousel.offsetWidth);
        carousel.classList.remove("no-transition");
    }
    // If the carousel is at the end, scroll to the beginning
    else if(Math.ceil(carousel.scrollLeft) === carousel.scrollWidth -
carousel.offsetWidth) {
        carousel.classList.add("no-transition");
        carousel.scrollLeft = carousel.offsetWidth;
        carousel.classList.remove("no-transition");
    }
}

```



```

    }

    // Clear existing timeout & start autoplay if mouse is not hovering over
    carousel
    clearTimeout(timeoutId);
    if(!wrapper.matches(":hover")) autoPlay();
  }

const autoPlay = () => {
  if(window.innerWidth < 800 || !isAutoPlay) return; // Return if window
  is smaller than 800 or isAutoPlay is false
  // Autoplay the carousel after every 2500 ms
  timeoutId = setTimeout(() => carousel.scrollLeft += firstCardWidth,
  2500);
}
autoPlay();

```

login.js

```

const forms = document.querySelector(".forms"),
    pwShowHide = document.querySelectorAll(".eye-icon"),
    links = document.querySelectorAll(".link");

pwShowHide.forEach(eyeIcon => {
  eyeIcon.addEventListener("click", () => {
    let pwFields = eyeIcon.parentElement.parentElement.querySelectorAll(".password");

    pwFields.forEach(password => {
      if(password.type === "password"){
        password.type = "text";
        eyeIcon.classList.replace("bx-hide", "bx-show");
        return;
      }
      password.type = "password";
      eyeIcon.classList.replace("bx-show", "bx-hide");
    })
  })
})

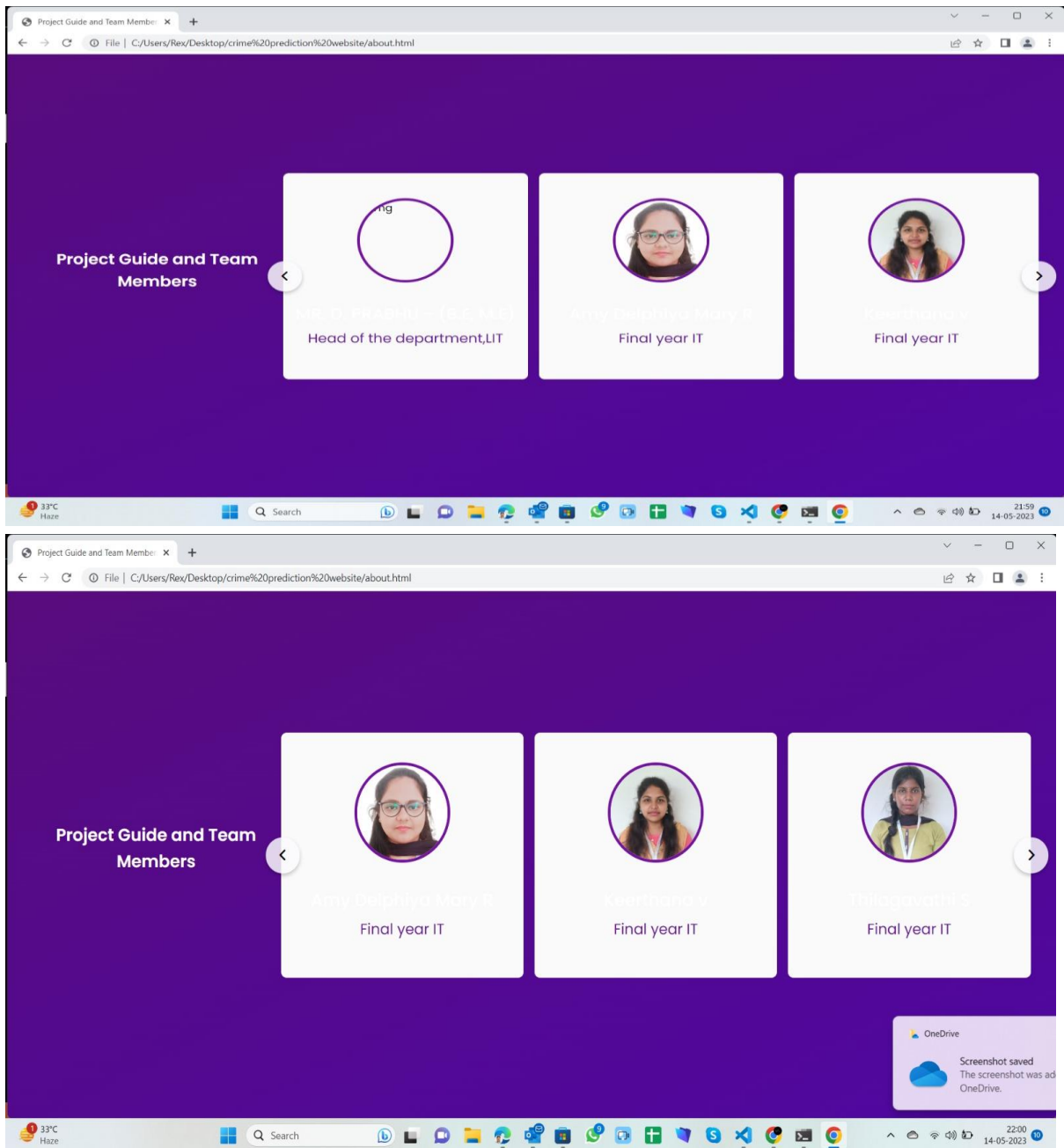
```

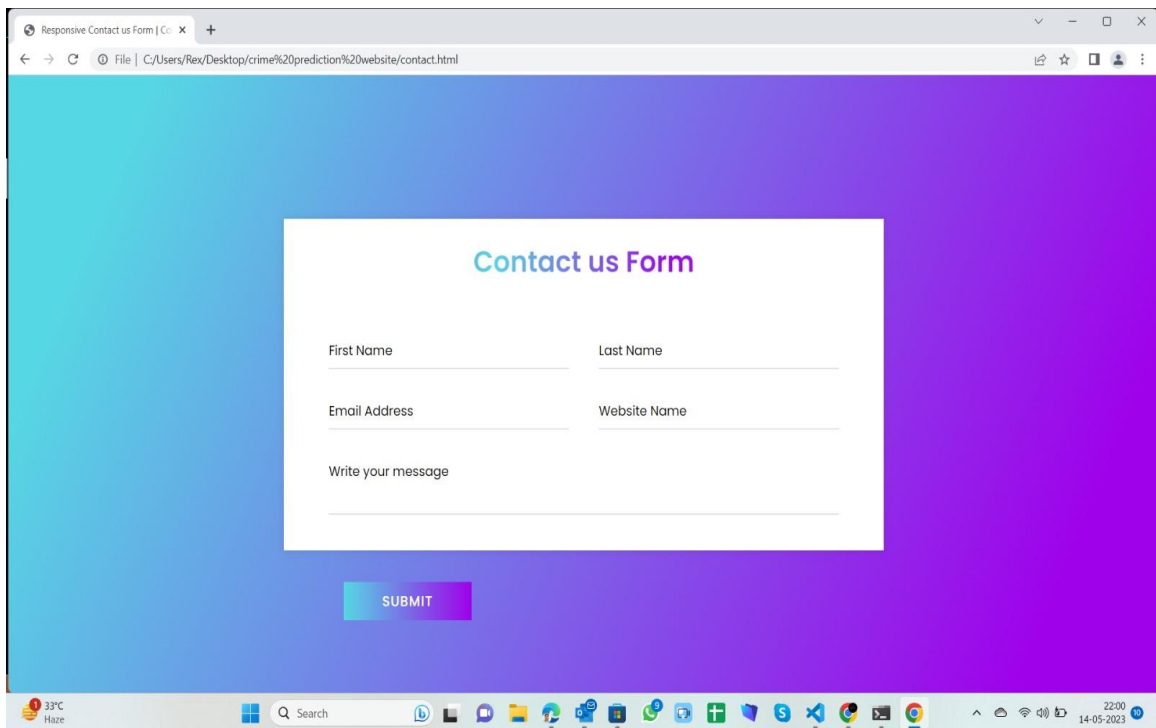
```

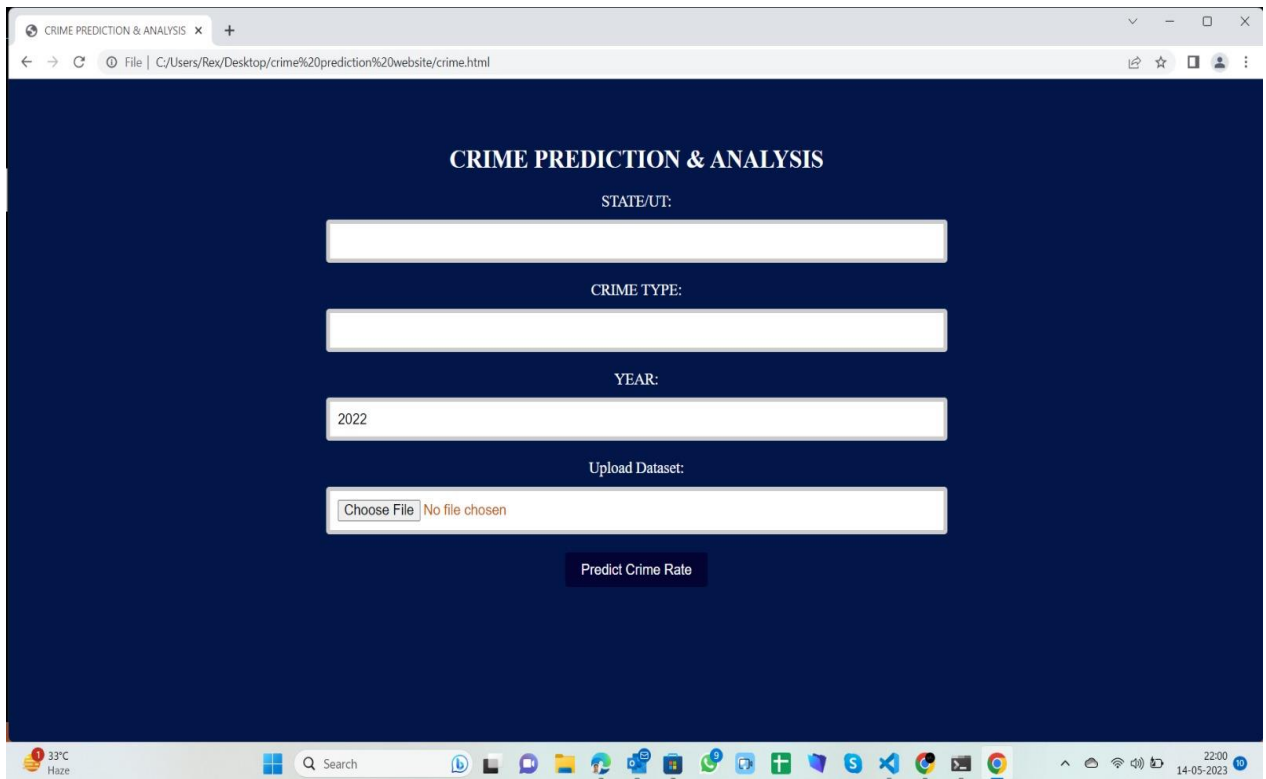
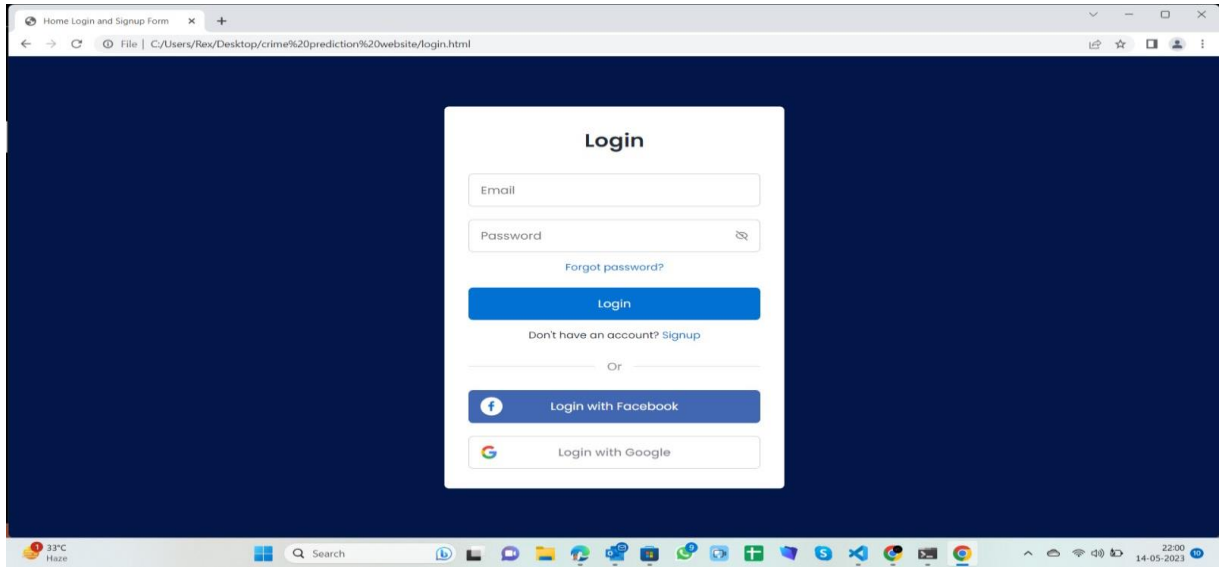
links.forEach(link => {
    link.addEventListener("click", e => {
        e.preventDefault(); //preventing form submit
        forms.classList.toggle("show-signup");
    })
})

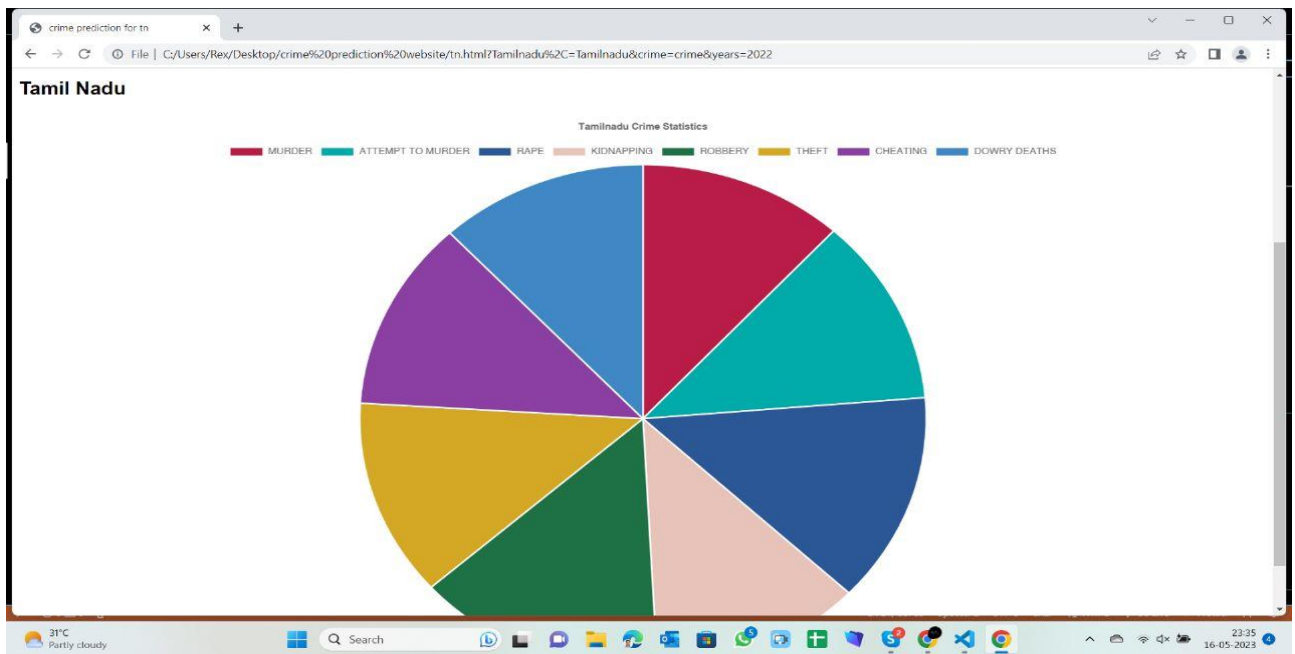
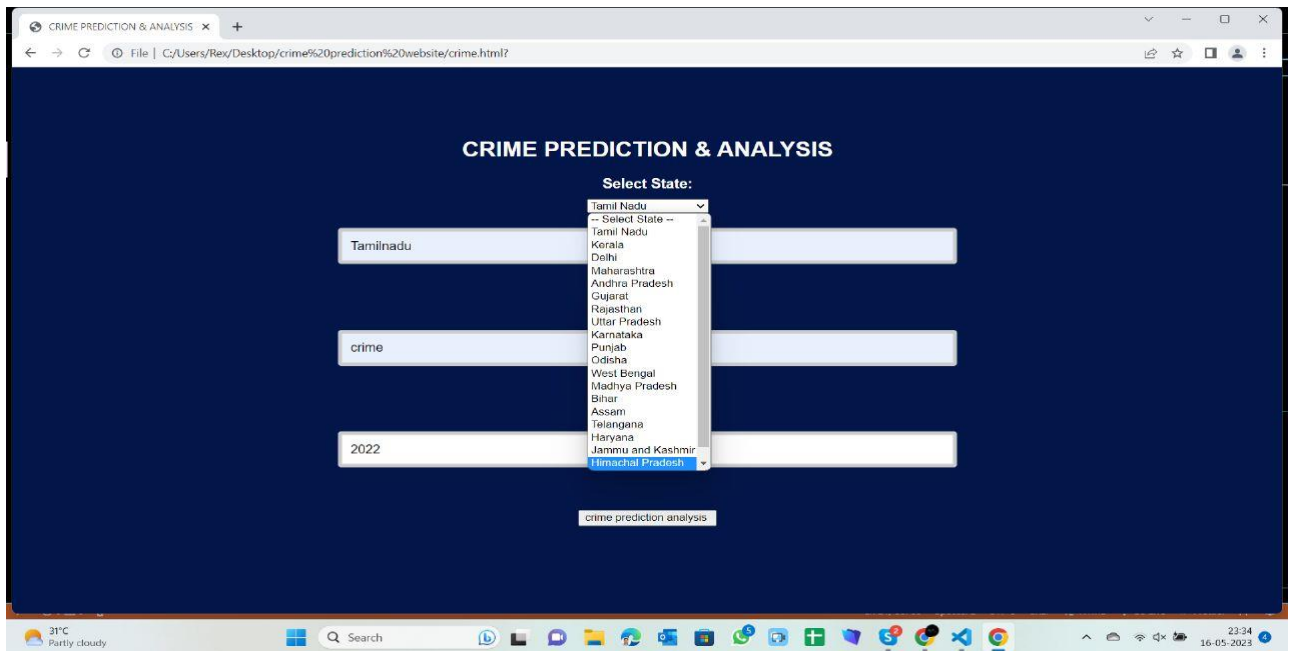
```

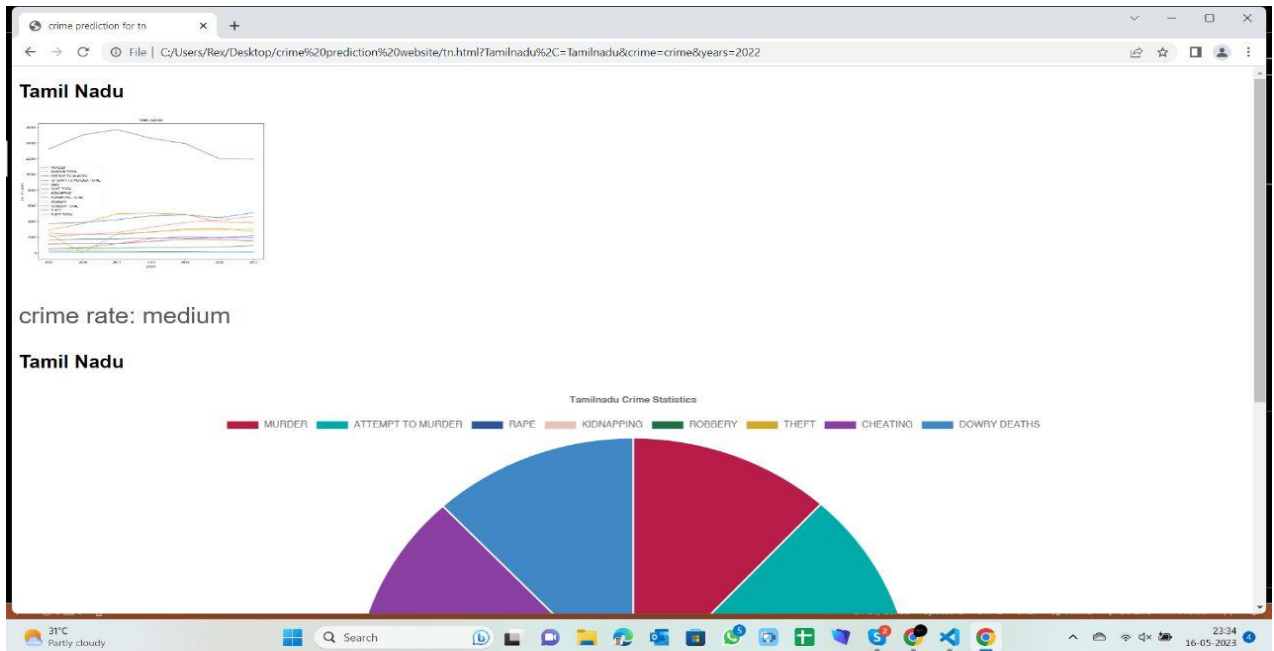
## WEBSITE OUTPUT:











CRIME PREDICTION & ANALYSIS

State/UT:

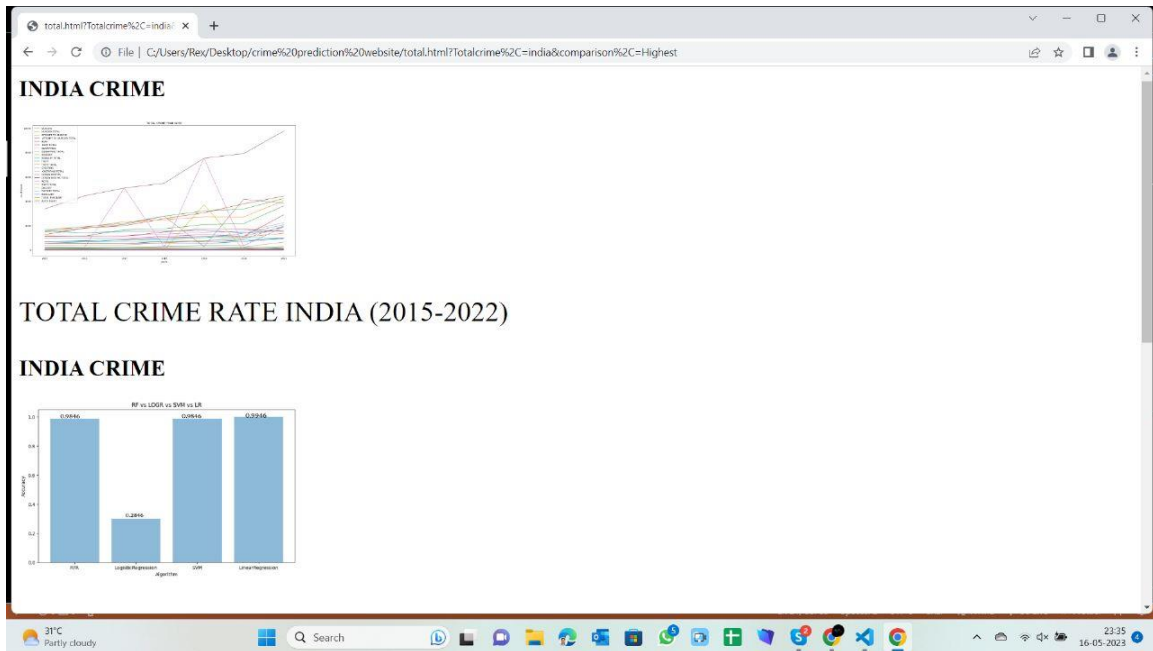
India

Algorithms:

Highest

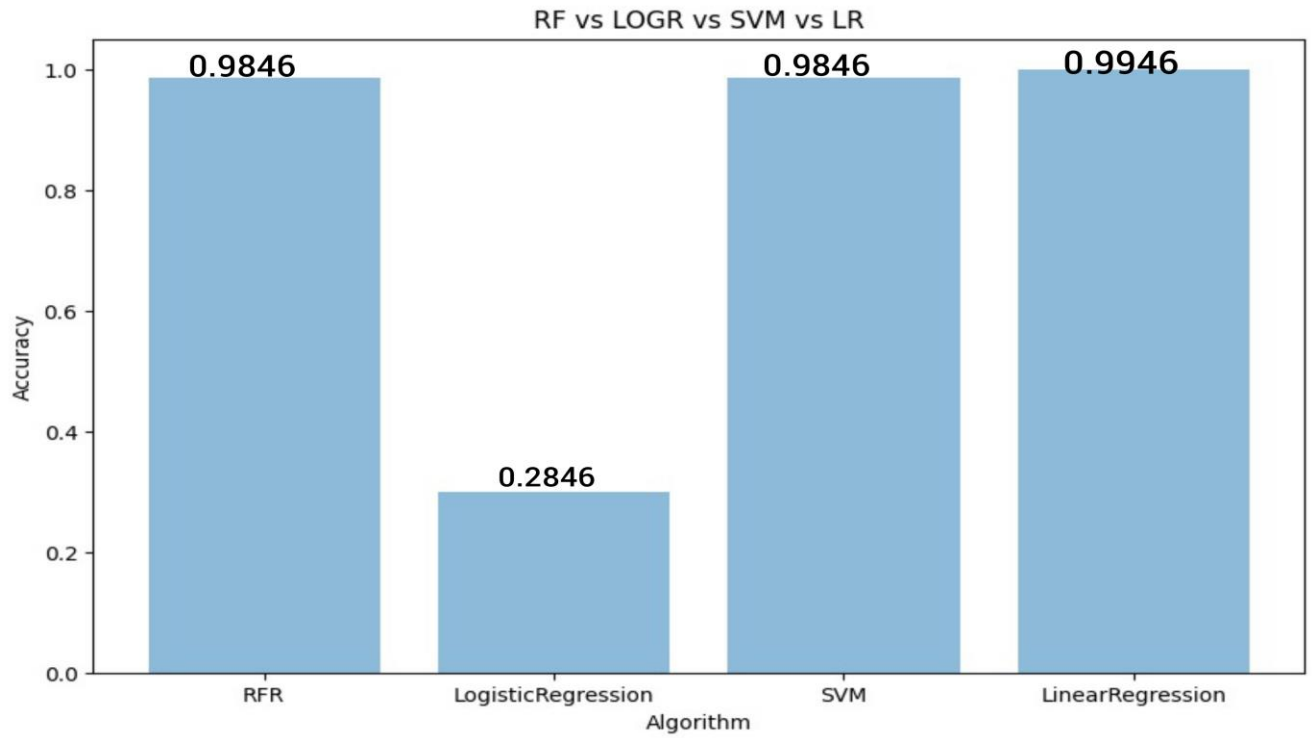
compare Algorithms

The figure shows a web interface for 'CRIME PREDICTION & ANALYSIS'. It features a dark blue background with white text and input fields. The interface includes a title 'CRIME PREDICTION & ANALYSIS', a 'State/UT:' label with a dropdown menu showing 'India', an 'Algorithms:' label with a dropdown menu showing 'Highest', and a 'compare Algorithms' button. The browser window title is 'CRIME PREDICTION & ANALYSIS' and the URL is 'File | C:/Users/Rex/Desktop/crime%20prediction%20website/prediction.html'.



## CHAPTER 9

### RESULT ANALYSIS



*7.1 Accuracy graph.*



## **CHAPTER 10**

### **CONCLUSION AND FUTURE ENHANCEMENT**

#### **CONCLUSION:**

In this paper focused on building predictive models for crime frequencies per crime type per month. The crime rates in India are increasing day by day due to many factors such as increase in poverty, implementation, corruption, etc. The proposed model is very useful for both the investigating agencies and the police official in taking necessary steps to reduce crime. The project helps the crime analysis to analysis these crime networks by means of various interactive visualization. The proposed method was analysed with different classification algorithm the accuracy of Linear Regression found to be good with the accuracy of 95.122%. We generated many graphs and found interesting statistics that helped in understanding India crimes datasets that can help in capturing the factors that can help in keeping society safe.

#### **FUTURE ENHANCEMENT:**

From the encouraging results, we believe that crime data mining has a promising future for increasing the effectiveness and efficiency of criminal and intelligence analysis. Visual and intuitive criminal and intelligence investigation techniques can be developed for crime pattern. Future enhancement of this research work on training bots to predict the crime prone areas by using machine learning techniques. Since, machine learning is similar to data mining advanced concept of machine learning can be used for better prediction.

## CHAPTER 11

### APPENDIX

#### 11.1 APPENDIX 1-CODE

```
{
  "cells": [
    {
      "cell_type": "code",
      "source": [
        "import pandas as pd\n",
        "import numpy as np\n",
        "import seaborn as sns\n",
        "import matplotlib.pyplot as plt\n",
        "from sklearn.cluster import KMeans\n",
        "from sklearn.linear_model import LinearRegression\n",
        "from sklearn.model_selection import train_test_split\n",
        "from sklearn.metrics import mean_squared_error, r2_score\n",
        "from sklearn.ensemble import RandomForestRegressor\n"
      ],
      "metadata": {
        "id": "JqsbYgSc2QHS"
      },
      "execution_count": null,
      "outputs": []
    },
    {
      "cell_type": "code",
      "source": [
        "df=pd.read_csv('/content/CRIME DATA.csv')\n"
      ],
      "metadata": {
        "id": "rRqwY5Px2P6Z"
      },
      "execution_count": null,
      "outputs": []
    }
  ],
  "metadata": {
    "id": "rRqwY5Px2P6Z"
  },
  "execution_count": null,
  "outputs": []
}
```

```

"cell_type": "code",
"source": [
  "df.head()"
],
"metadata": {
  "id": "Ncy331dW2Prr",
  "outputId": "63f991f2-33bc-4779-de44-b0891dc515cf",
  "colab": {
    "base_uri": "https://localhost:8080/",
    "height": 206
  }
},
"execution_count": null,
"outputs": [
  {
    "output_type": "execute_result",
    "data": {
      "text/plain": [
        "
                                STATE/UT CRIME HEAD  2015  2016
2017  2018  2019  2020  2021  \\n",
        "0      ANDHRA PRADESH    MURDER    2665  2690
2449  2538  2808  2717  2484  \\n",
        "1  ARUNALACHALA PRADESH    MURDER        68
72    59    75    65    73    69  \\n",
        "2                                ASSAM    MURDER    1374  1426
1323  1223  1303  1368  1354  \\n",
        "3                                BIHAR    MURDER    3034  3139
3152  3362  3198  3566  3441  \\n",
        "4      CHHATTISGARH    MURDER    1097  1169
1083  1065  1110   998   945  \\n",
        "\\n",
        "  2022  \\n",
        "0  1175  \\n",
        "1    86  \\n",
        "2  1451  \\n",
        "3  3403  \\n",
        "4   998  "
      ],
      "text/html": [
        "\\n",

```

```

"          <div          id=\"df-939cdfa7-29df-428d-af8a-
53ca4ea35232\">\n",
"    <div class=\"colab-df-container\">\n",
"      <div>\n",
"    <style scoped>\n",
"      .dataframe tbody tr th:only-of-type {\n",
"        vertical-align: middle;\n",
"      }\n",
"    \n",
"      .dataframe tbody tr th {\n",
"        vertical-align: top;\n",
"      }\n",
"    \n",
"      .dataframe thead th {\n",
"        text-align: right;\n",
"      }\n",
"    </style>\n",
"    <table border=\"1\" class=\"dataframe\">\n",
"      <thead>\n",
"        <tr style=\"text-align: right;\">\n",
"          <th></th>\n",
"          <th>STATE/UT</th>\n",
"          <th>CRIME HEAD</th>\n",
"          <th>2015</th>\n",
"          <th>2016</th>\n",
"          <th>2017</th>\n",
"          <th>2018</th>\n",
"          <th>2019</th>\n",
"          <th>2020</th>\n",
"          <th>2021</th>\n",
"          <th>2022</th>\n",
"        </tr>\n",
"      </thead>\n",
"      <tbody>\n",
"        <tr>\n",
"          <th>0</th>\n",
"          <td>ANDHRA PRADESH</td>\n",
"          <td>MURDER</td>\n",
"          <td>2665</td>\n",
"          <td>2690</td>\n",

```

```

"      <td>2449</td>\n",
"      <td>2538</td>\n",
"      <td>2808</td>\n",
"      <td>2717</td>\n",
"      <td>2484</td>\n",
"      <td>1175</td>\n",
"    </tr>\n",
"    <tr>\n",
"      <th>1</th>\n",
"      <td>ARUNALACHALA PRADESH</td>\n",
"      <td>MURDER</td>\n",
"      <td>68</td>\n",
"      <td>72</td>\n",
"      <td>59</td>\n",
"      <td>75</td>\n",
"      <td>65</td>\n",
"      <td>73</td>\n",
"      <td>69</td>\n",
"      <td>86</td>\n",
"    </tr>\n",
"    <tr>\n",
"      <th>2</th>\n",
"      <td>ASSAM</td>\n",
"      <td>MURDER</td>\n",
"      <td>1374</td>\n",
"      <td>1426</td>\n",
"      <td>1323</td>\n",
"      <td>1223</td>\n",
"      <td>1303</td>\n",
"      <td>1368</td>\n",
"      <td>1354</td>\n",
"      <td>1451</td>\n",
"    </tr>\n",
"    <tr>\n",
"      <th>3</th>\n",
"      <td>BIHAR</td>\n",
"      <td>MURDER</td>\n",
"      <td>3034</td>\n",
"      <td>3139</td>\n",
"      <td>3152</td>\n",

```

```

"      <td>3362</td>\n",
"      <td>3198</td>\n",
"      <td>3566</td>\n",
"      <td>3441</td>\n",
"      <td>3403</td>\n",
"    </tr>\n",
"    <tr>\n",
"      <th>4</th>\n",
"      <td>CHHATTISGARH</td>\n",
"      <td>MURDER</td>\n",
"      <td>1097</td>\n",
"      <td>1169</td>\n",
"      <td>1083</td>\n",
"      <td>1065</td>\n",
"      <td>1110</td>\n",
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"      <td>945</td>\n",
"      <td>998</td>\n",
"    </tr>\n",
"  </tbody>\n",
"</table>\n",
"</div>\n",
"      <button      class=\"colab-df-convert\"
onclick=\"convertToInteractive('df-939cdfa7-29df-428d-af8a-
53ca4ea35232')\"\\n",
"      title=\"Convert this dataframe to an interactive
table.\\n",
"      style=\"display:none;\">\n",
"    \\n",
"    <svg      xmlns=\"http://www.w3.org/2000/svg\"
height=\"24px\"viewBox=\"0 0 24 24\"\\n",
"      width=\"24px\">\n",
"      <path d=\"M0 0h24v24H0V0z\" fill=\"none\"/>\n",
"      <path d=\"M18.56 5.44l.94 2.06.94-2.06 2.06-.94-2.06-
.94-.94-2.06-.94 2.06-2.06.94zm-11 1L8.5 8.5l.94-2.06 2.06-.94-2.06-
.94L8.5 2.5l-.94 2.06-2.06.94zm10 10l.94 2.06.94-2.06 2.06-.94-2.06-
.94-2.06-.94 2.06-2.06.94z\"/><path d=\"M17.41 7.96l-1.37-1.37c-.4-.4-.92-
.59-1.43-.59-.52 0-1.04.2-1.43.59L10.3 9.45l-7.72 7.72c-.78.78-.78 2.05 0
2.83L4 21.41c.39.39.95.59 1.41.59.51 0 1.02-.2 1.41-.59l7.78-7.78 2.81-
2.81c.8-.78.8-2.07 0-2.86zM5.41 20L4 18.59l7.72-7.72 1.47 1.35L5.41

```

```

20z\"/>\n",
    "</svg>\n",
    "</button>\n",
    "\n",
    "<style>\n",
    "  .colab-df-container {\n",
    "    display: flex;\n",
    "    flex-wrap: wrap;\n",
    "    gap: 12px;\n",
    "  }\n",
    "\n",
    "  .colab-df-convert {\n",
    "    background-color: #E8F0FE;\n",
    "    border: none;\n",
    "    border-radius: 50%;\n",
    "    cursor: pointer;\n",
    "    display: none;\n",
    "    fill: #1967D2;\n",
    "    height: 32px;\n",
    "    padding: 0 0 0 0;\n",
    "    width: 32px;\n",
    "  }\n",
    "\n",
    "  .colab-df-convert:hover {\n",
    "    background-color: #E2EBFA;\n",
    "    box-shadow: 0px 1px 2px rgba(60, 64, 67, 0.3), 0px
1px 3px 1px rgba(60, 64, 67, 0.15);\n",
    "    fill: #174EA6;\n",
    "  }\n",
    "\n",
    "  [theme=dark] .colab-df-convert {\n",
    "    background-color: #3B4455;\n",
    "    fill: #D2E3FC;\n",
    "  }\n",
    "\n",
    "  [theme=dark] .colab-df-convert:hover {\n",
    "    background-color: #434B5C;\n",
    "    box-shadow: 0px 1px 3px 1px rgba(0, 0, 0, 0.15);\n",
    "    filter: drop-shadow(0px 1px 2px rgba(0, 0, 0, 0.3));\n",
    "    fill: #FFFFFF;\n",

```

```

    }\n",
    "</style>\n",
    "\n",
    "<script>\n",
    "    const buttonEl =\n",
    "        document.querySelector('#df-939cdfa7-29df-428d-af8a-53ca4ea35232 button.colab-df-convert');\n",
    "        buttonEl.style.display =\n",
    "            google.colab.kernel.accessAllowed ? 'block' :
'none';\n",
    "\n",
    "        async function convertToInteractive(key) {\n",
    "            const element = document.querySelector('#df-939cdfa7-29df-428d-af8a-53ca4ea35232');\n",
    "            const dataTable =\n",
    "                await
google.colab.kernel.invokeFunction('convertToInteractive',\n",
    "                    [key],
    {});\n",
    "            if (!dataTable) return;\n",
    "\n",
    "            const docLinkHtml = 'Like what you see? Visit the
'+\n",
    "                '<a    target=\"_blank\"
href=https://colab.research.google.com/notebooks/data_table.ipynb>data
table notebook</a>'\n",
    "                + ' to learn more about interactive tables.';\n",
    "            element.innerHTML = \";\n",
    "            dataTable['output_type'] = 'display_data';\n",
    "            await google.colab.output.renderOutput(dataTable,
element);\n",
    "            const docLink = document.createElement('div');\n",
    "            docLink.innerHTML = docLinkHtml;\n",
    "            element.appendChild(docLink);\n",
    "        }\n",
    "    </script>\n",
    "</div>\n",
    "</div>\n",
    "

```



```
},  
"metadata": {}},
```

python code

```
# -*- coding: utf-8 -*-  
"""Crime prediction
```

Automatically generated by Colaboratory.

Original file is located at

```
https://colab.research.google.com/drive/18_55amsKyoAVWmR-  
OIpsrUVCLouGHqkF  
"""
```

```
import pandas as pd  
import numpy as np  
import seaborn as sns  
import matplotlib.pyplot as plt  
from sklearn.cluster import KMeans  
from sklearn.linear_model import LinearRegression  
from sklearn.model_selection import train_test_split  
from sklearn.metrics import mean_squared_error, r2_score  
from sklearn.ensemble import RandomForestRegressor
```

```
df=pd.read_csv('/content/CRIME DATA.csv')
```

```
df.head()
```

```
years_title=[str(i)for i in range(2015,2022)]
```

```
STATES_IN_INDIA=df['STATE/UT'].unique()  
STATES_IN_INDIA=STATES_IN_INDIA[:-4]  
STATES_IN_INDIA
```

```
TYPES_OF_CASES=df['CRIME HEAD'].unique()  
TYPES_OF_CASES=TYPES_OF_CASES[:-1]  
TYPES_OF_CASES
```

```
df.info()
```

```
sns.pairplot(df)
```

```
mask = np.tril(df.corr())
fig, ax = plt.subplots(figsize=(15, 15))
sns.heatmap(df.corr(), fmt='.1g', annot=True, cmap= 'coolwarm',
mask=mask,ax=ax)
```

```
for state in STATES_IN_INDIA:
    fig=plt.figure(figsize=(12,8), dpi=80, facecolor='w',edgecolor='k')
    plt.title(state)
    plt.xlabel('years')
    plt.ylabel('no of cases')
    for case in TYPES_OF_CASES:
        temp_df=df[(df['STATE/UT']==state) & (df['CRIME HEAD']== case )]
        if len(temp_df) > 0:
            N_cases=[temp_df[c].values[0]for c in years_title]
            plt.plot(years_title,N_cases)
    plt.legend(TYPES_OF_CASES)
```

```
fig=plt.figure(figsize=(20,10), dpi=80, facecolor='w',edgecolor='k')
plt.title('TOTAL CRIME YEAR WISE')
plt.xlabel('years')
plt.ylabel('no of cases')
for state in STATES_IN_INDIA:
    temp_df=df[(df['STATE/UT']==state) & (df['CRIME
HEAD']== 'KIDNAPPING' )]
    if len(temp_df) > 0:
        N_cases=[temp_df[c].values[0]for c in years_title]
        plt.plot(years_title,N_cases)
    plt.legend(TYPES_OF_CASES)
```

```
from sklearn import preprocessing
lab=preprocessing.LabelEncoder()
df['CRIME HEAD']=lab.fit_transform(df['CRIME HEAD'])
df.head()
```

```
from sklearn.cluster import KMeans
```

```

kmeans = KMeans(n_clusters=9)
kmeans.fit(df.iloc[:,1:])

kmeans.cluster_centers_

import numpy as np

unique, count=np.unique(kmeans.labels_, return_counts=True)

dict_data=dict(zip(unique, count))
dict_data

df["cluster"]=kmeans.labels_

kmeans.inertia_

kmeans.score

df

cust = [[13,2665,2690,2449,2538,2808,2717,2484,1175]]
kmeans.predict(cust)[0]

f, ax= plt.subplots(figsize=(24,15))

stats = df.sort_values(["cluster", "STATE/UT"], ascending=True)
sns.set_color_codes("pastel")
sns.barplot(y="STATE/UT",x="2018",data=stats)

sns.despine(left=True, bottom=True)

x = df.iloc[:,1:14]
y = df.iloc[:,df.columns=='cluster']

print(x.head())

y.head()

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.3,
random_state=0)

```

```

from sklearn.ensemble import RandomForestClassifier

rfc = RandomForestClassifier(n_estimators=100, )

rfc.fit(x_train, y_train)

y_pred = rfc.predict(x_test)
print(y_pred)

from sklearn.metrics import accuracy_score, classification_report
accuracy = accuracy_score(y_test, y_pred)
print('Accuracy:', accuracy)

clf=classification_report(y_pred,y_test)
print(clf)

from sklearn.linear_model import LogisticRegression

log_reg = LogisticRegression()

log_reg.fit(x_train, y_train)

y_pred2 = log_reg.predict(x_test)

accuracy2 = accuracy_score(y_test, y_pred2)
print("Accuracy:", accuracy2)

clf2=classification_report(y_pred2, y_test)
print(clf2)

from sklearn.svm import SVC
from sklearn.metrics import accuracy_score

svm = SVC(kernel='linear', random_state=0)

svm.fit(x_train, y_train)

y_pred3 = svm.predict(x_test)

```

```

accuracy3 = accuracy_score(y_test, y_pred3)
print('Accuracy:', accuracy3)

clf3=classification_report(y_pred3, y_test)
print(clf3)

from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error

print("Shape of X_test:", x_test.shape)
print("Shape of y_test:", y_test.shape)

lg = LinearRegression()

lg.fit(x, y)

y_pred4 = lg.predict(x)

mse = mean_squared_error(y, y_pred4)
print("Mean Squared Error:", mse)

linreg =LinearRegression()
linreg.fit(x_train, y_train)
y_pred4 = linreg.predict(x_test)
mse = np.mean((y_pred4 - y_test)**2)
accuracy4 = linreg.score(x_test, y_test)

import matplotlib.pyplot as plt; plt.rcdefaults()

objects = ('RFR', 'LogisticRegression', 'SVM', 'LinearRegression')
y_pos = np.arange(len(objects))
performance = (accuracy, accuracy2, accuracy3, accuracy4)

fig, ax = plt.subplots(figsize=(10, 6))
plt.bar(y_pos, performance, align='center', alpha=0.5, )
plt.xticks(y_pos, objects)
plt.xlabel('Algorithm')
plt.ylabel('Accuracy')
plt.title('RF vs LOGR vs SVM vs LR')
plt.show()

```

## APPENDIX 2-SCREENSHOTS

The screenshot shows a Google Colab notebook titled "Crime prediction". The dataset is loaded as a DataFrame with columns: STATE/UT, CRIME HEAD, 2015, 2016, 2017, 2018, 2019, 2020, 2021, 2022, and cluster. The data is displayed in a table format with 432 rows and 11 columns. The first few rows are:

|   | STATE/UT             | CRIME HEAD | 2015 | 2016 | 2017 | 2018 | 2019 | 2020 | 2021 | 2022 | cluster |
|---|----------------------|------------|------|------|------|------|------|------|------|------|---------|
| 0 | ANDHRA PRADESH       | 13         | 2665 | 2690 | 2449 | 2538 | 2808 | 2717 | 2484 | 1175 | 8       |
| 1 | ARUNALACHALA PRADESH | 13         | 68   | 72   | 59   | 75   | 65   | 73   | 69   | 86   | 0       |
| 2 | ASSAM                | 13         | 1374 | 1426 | 1323 | 1223 | 1303 | 1368 | 1354 | 1451 | 0       |
| 3 | BIHAR                | 13         | 3034 | 3139 | 3152 | 3362 | 3198 | 3566 | 3441 | 3403 | 8       |
| 4 | CHHATTISGARH         | 13         | 1097 | 1169 | 1083 | 1065 | 1110 | 998  | 945  | 998  | 0       |

The code cell shows the following execution:

```
[21] cust = [[13,2665,2690,2449,2538,2808,2717,2484,1175]]
      kmeans.predict(cust)[0]
```

The output shows a warning from sklearn: "UserWarning: X does not have valid feature names, but KMeans was fitted with feature names". The execution is completed at 10:20 PM.

The screenshot shows the same Google Colab notebook. The code cell shows the following execution:

```
[14] from sklearn.cluster import KMeans

      kmeans = KMeans(n_clusters=9)
      kmeans.fit(df.iloc[:,1:])
```

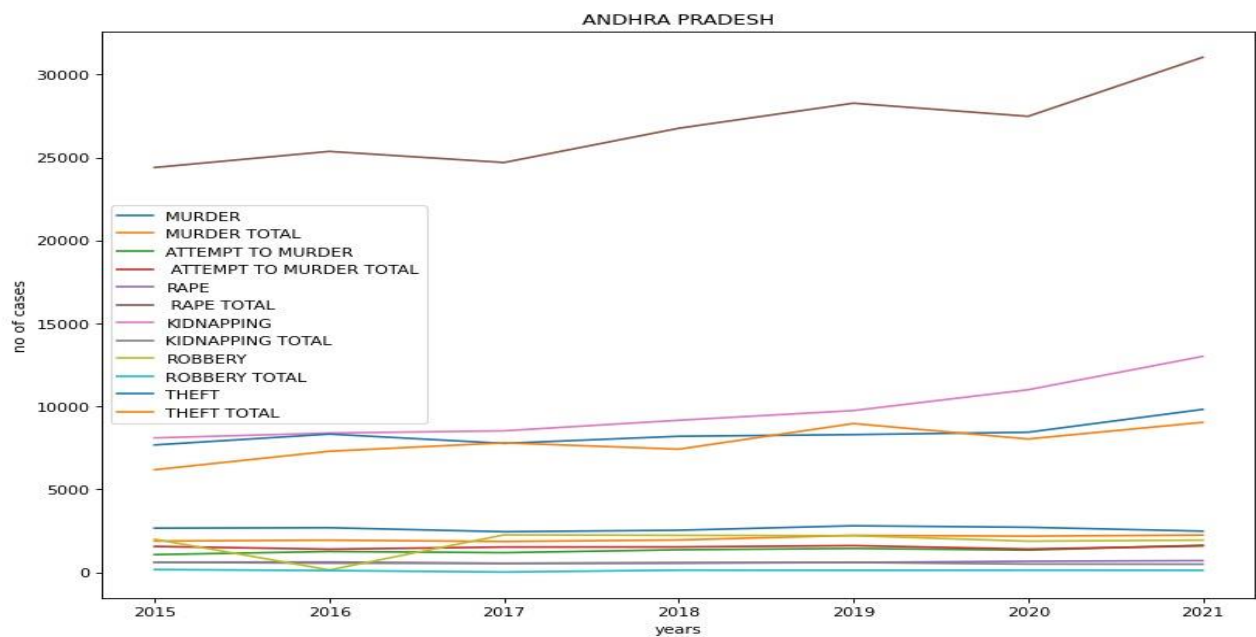
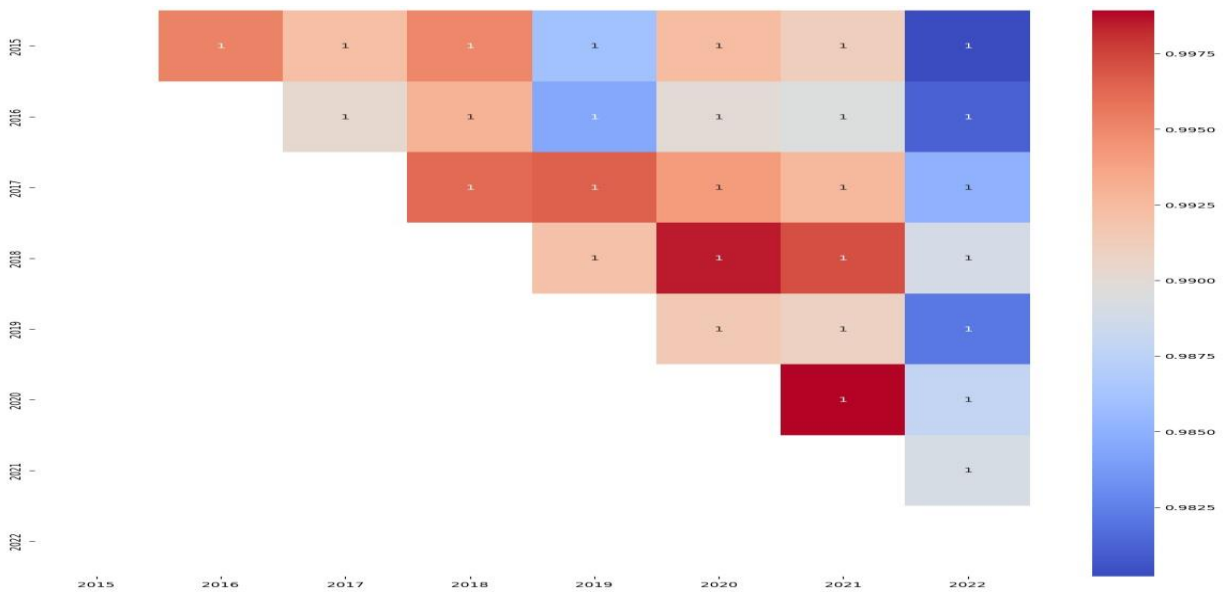
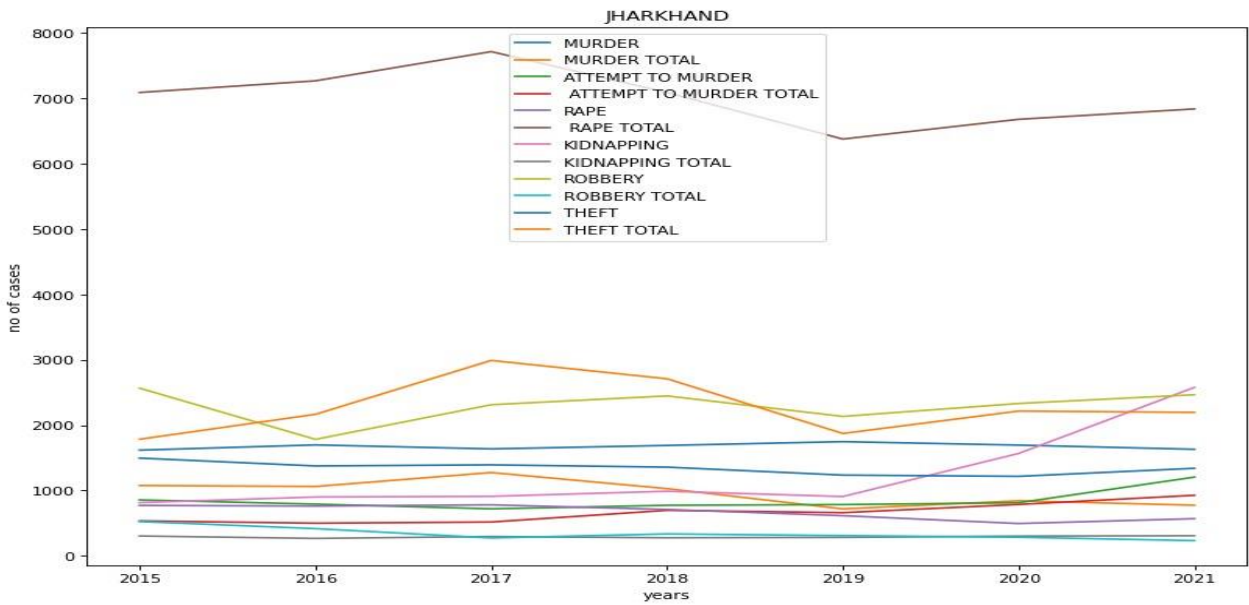
The output shows a warning from sklearn: "FutureWarning: The default value of 'n\_init' will change from 10 to 'auto' in 1.4. Set the value of 'n\_init' explicitly to suppress the warning". The execution is completed at 10:20 PM.

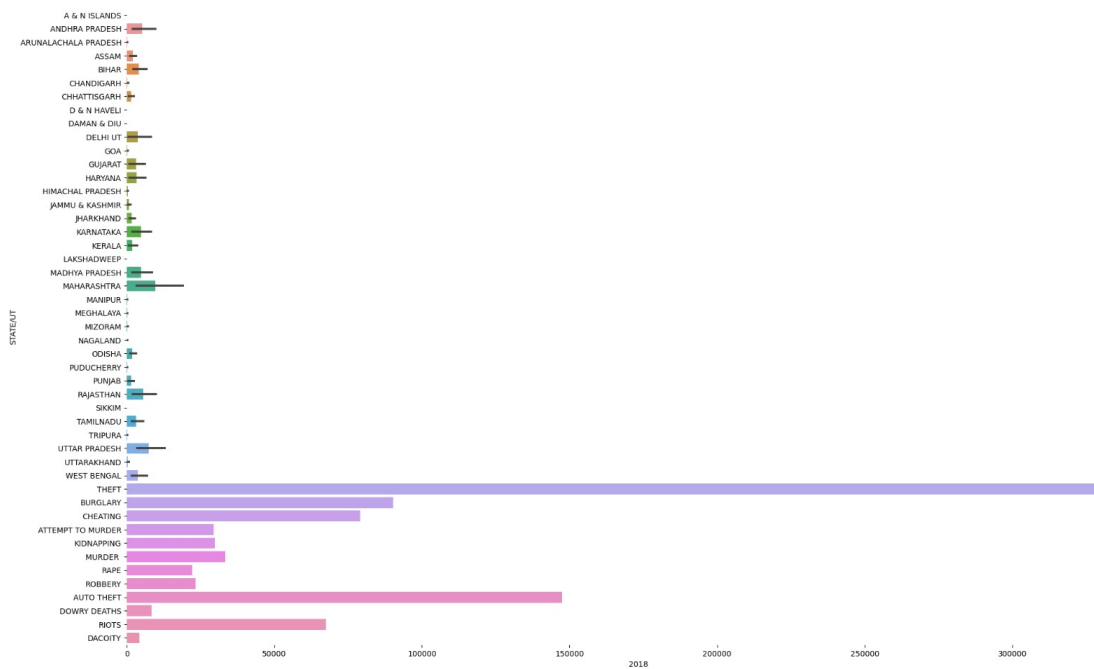
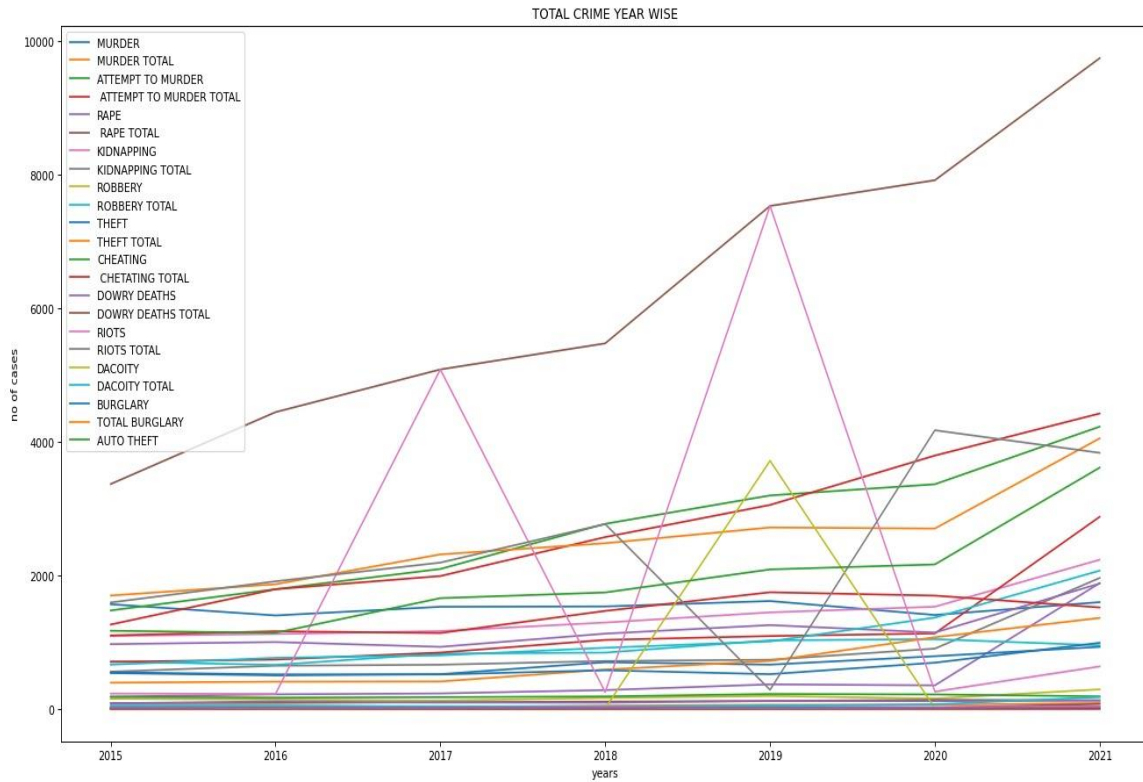
The next code cell shows the following execution:

```
[15] kmeans.cluster_centers_
```

The output shows the cluster centers as an array of 9 clusters, each with 11 values (STATE/UT, CRIME HEAD, and years 2015-2022). The first few clusters are:

- Cluster 0: [1.05746832e+01, 3.51468317e+02, 4.49882540e+02, 4.53834921e+02, 3.77047619e+02, 5.80469841e+02, 3.97542857e+02, 4.55749206e+02, 4.23338159e+02, 2.10000000e+01, 2.85843000e+05, 3.16761000e+05, 3.51820000e+05, 3.30312000e+05, 3.77395000e+05, 3.37407000e+05, 3.72622000e+05, 4.27122000e+05]
- Cluster 1: [1.20000000e+01, 7.82720000e+04, 8.01605000e+04, 8.89910000e+04, 8.45890000e+04, 9.78675000e+04, 9.35475000e+04, 1.05865000e+05, 9.12890000e+04]
- Cluster 2: [1.24285714e+01, 2.28372857e+04, 2.85290000e+04, 2.83987143e+04, 2.74795714e+04, 3.31451429e+04, 3.13582857e+04, 3.68865714e+04, 4.38875714e+04]
- Cluster 3: [2.20000000e+01, 1.00255000e+05, 1.20032000e+05, 1.55401000e+05, 1.47475000e+05, 1.72559000e+05, 1.54351000e+05, 1.65690000e+05, 1.81200000e+05]
- Cluster 4: [1.40769231e+01, 1.51069231e+04, 1.69403846e+04, 1.79966154e+04, 1.87510000e+04, 1.78006154e+04, 1.97616071e+04, 1.77330771e+04, 1.87510000e+04]







## **CHAPTER 12**

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