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Crime Type and Occurrence Prediction Using Machine Learning Algorithm

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Abstract: This project is entitled as “Crime Type and Occurrence Prediction Using Machine Learning Algorithm” Crime is still a major worry and a serious problem in our society. As a result, crime prevention is a significant issue that needs to be examined methodically. Detecting and preventing crime requires effective crime analytics, which is also crucial for assessing how well criminal investigations are working. To create precise forecasts, inferences are drawn from trained data. a system that uses user input to forecast different characteristics of crime. Users select the year, offense type, and city name. The method forecasts the population, expected number of occurrences, and expected crime rate for the year in 100,000 units based on this data. The dataset used in this investigation was produced on its own. The system makes use of machine learning techniques, particularly Random Forest, SVM, Logistic Regression, and Linear Regression. To guarantee precise forecasts and dependable results, it has undergone testing and training.

I. INTRODUCTION

A crime is an act that qualifies as an offense. It is governed by the law. For the police, identifying and analyzing hidden crimes is an extremely challenging task. Furthermore, there is a wealth of criminal data available, therefore methods are required to make investigation easier. Consequently, the approach ought to be helpful in resolving criminal cases. Crime analysis and prediction can be further aided by machine learning techniques. Regression techniques are provided by the machine learning methodology. The classification methods aid in achieving the goal of the inquiry. Regression techniques like logistic regression and linear regression. This technique aids in determining how two quantitative values or variables relate to one another. This method uses the independent variables to predict the value of the dependent variable. classifier techniques like Random Forest, SVM, etc. Multi-class target variables are classified using these classifiers. Accuracy is increased by using a neural network. The neural network consists of an output layer and a dense input layer. The perpetrator's description, including prediction rate, estimated crime rate, estimated number of cases, and population in lakhs, is based on the algorithms mentioned above. As a result, she will assist in solving murder cases, which should lessen the load on police investigations.

II. OBJECTIVES OF THE PROJECT

- 1) To increase law enforcement's capacity to proactively prevent and reduce criminal activity, machine learning techniques are being used to forecast the types and incidences of crimes. The objective is to accurately anticipate the kind of crime that is likely to occur in a specific location and at a specific time by utilizing machine learning algorithms and examining current crime data.
- 2) By detecting crime hotspots and preventing crimes before they happen, this predictive power can help government better allocate resources and lower crime rates. The objective is to use technology to strengthen public safety and crime prevention tactics.

III. STUDY ON EXISTING SYSTEM

Models for predicting crime frequently use historical data, which may have built-in biases based on geography, socioeconomic position, or race. For minorities in particular, this might result in skewed projections and sustain structural injustices. Incomplete or noisy data may decrease the accuracy of crime prediction models. Predictions may not be accurate since crime is influenced by intricate social and psychological elements that are hard to record in data. Confidential data is frequently processed by crime predicting systems, making them possible targets for cyberattacks. Both public security and individual confidentiality may be jeopardized by violations of these systems.

A. Drawbacks

- 1) The classifier's use of a categorical value biases the output for nominal qualities with higher values, contributing to low accuracy in previous efforts.
- 2) The classification technique is not suitable for regions with inaccurate data and real-valued attributes
- 3) To optimize the classifier's performance, its value must be modified
- 4) Identifying crime patterns and extracting information from big amounts of data is comprehensive.
- 5) Many crime prediction systems involve complicated algorithms, also known as "black-box" models, that are difficult to explain.
- 6) The incorporation of personal data in crime prediction models, such as social media activity, surveillance footage, or mobile phone data, poses substantial privacy concerns.

IV. PROPOSED SYSTEM

A machine learning model that predicts the complex model and trend of a large dataset. This can identify crimes across geographies and acts. This leads to a more accurate prediction of crime in a historical database, including time, location, and kind of crime. Analyzing behavioral patterns and suspicious conduct can aid in the early detection and prevention of criminal activity. For example, ML models can be used to anticipate specific crimes, such as robbery or fraud, based on predetermined triggers. Predictive data enables law enforcement organizations to better allocate resources to high-risk regions, thereby lowering crime and increasing public safety.

A. Advantages

- 1) Optimal values do not need to be initialized.
- 2) High accuracy compared to other machine learning prediction models.
- 3) Eliminates the need to analyze independent attributes.
- 4) The suggested method accurately detects criminal patterns based on prediction rate, anticipated crime rate, and number of instances. It also eliminates the need to analyze independent effects.

V. METHODOLOGY

- 1) Data collection
- 2) Data preprocessing
- 3) Analysis
- 4) Training and Testing
- 5) Validation

A. Data Collection

The data for the implementation is sourced from Kaggle. The record collection totals over 3000. This dataset should preferably include information such as city, type, year, population, predicted rate, and estimated rate.

1	City	Crime Type	Year	Prediction Rate (%)	Estimated Crime Rate (%)	Estimated Number of Cases	Population (in lakhs)
2	6	7	2018	71	10	582	77
3	3	8	2024	83	8	259	100
4	7	5	2011	95	12	999	85
5	4	4	2001	98	15	104	45
6	6	5	2002	50	9	305	45
7	9	1	2017	58	10	193	79
8	2	3	2000	69	20	269	84
9	6	2	2019	68	9	808	84
10	7	6	2020	85	15	676	45
11	4	8	2005	89	1	195	55
12	3	4	2004	94	2	418	59
13	7	5	2000	53	19	78	5
14	7	9	2018	58	14	847	32
15	2	1	2000	90	8	782	6
16	5	2	2020	71	6	636	94
17	4	2	2020	67	12	868	26

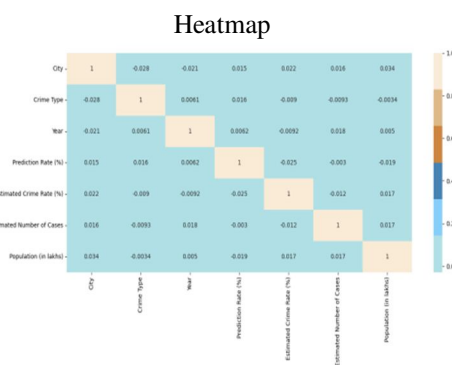
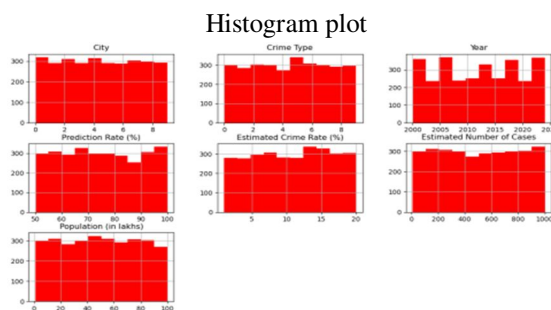
B. Data preprocessing

The dataset has been collected; however, it needs be preprocessed in order to provide a clean dataset. Python includes libraries such as pandas and numpy. One-hot encoding techniques can be used to encode categorical variables (such as city, crime type, and year) into numerical representation.

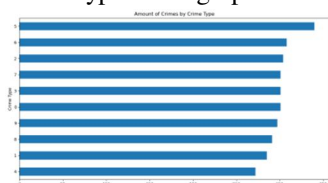
City	Crime Type	Year	Prediction Rate (%)	Estimated Crime Rate (%)	Estimated Crime Rate (%)	Estimated Number of Cases	Population (in lakhs)
1	1	2018	1.1	1.0	1.0	2.5	7.3
2	2	2018	1.1	1.0	1.0	2.5	7.3
3	3	2018	1.1	1.0	1.0	2.5	7.3
4	4	2018	1.1	1.0	1.0	2.5	7.3
5	5	2018	1.1	1.0	1.0	2.5	7.3
6	6	2018	1.1	1.0	1.0	2.5	7.3
7	7	2018	1.1	1.0	1.0	2.5	7.3
8	8	2018	1.1	1.0	1.0	2.5	7.3
9	9	2018	1.1	1.0	1.0	2.5	7.3
10	10	2018	1.1	1.0	1.0	2.5	7.3
11	11	2018	1.1	1.0	1.0	2.5	7.3
12	12	2018	1.1	1.0	1.0	2.5	7.3
13	13	2018	1.1	1.0	1.0	2.5	7.3
14	14	2018	1.1	1.0	1.0	2.5	7.3
15	15	2018	1.1	1.0	1.0	2.5	7.3
16	16	2018	1.1	1.0	1.0	2.5	7.3
17	17	2018	1.1	1.0	1.0	2.5	7.3
18	18	2018	1.1	1.0	1.0	2.5	7.3
19	19	2018	1.1	1.0	1.0	2.5	7.3
20	20	2018	1.1	1.0	1.0	2.5	7.3
21	21	2018	1.1	1.0	1.0	2.5	7.3
22	22	2018	1.1	1.0	1.0	2.5	7.3
23	23	2018	1.1	1.0	1.0	2.5	7.3
24	24	2018	1.1	1.0	1.0	2.5	7.3
25	25	2018	1.1	1.0	1.0	2.5	7.3
26	26	2018	1.1	1.0	1.0	2.5	7.3
27	27	2018	1.1	1.0	1.0	2.5	7.3
28	28	2018	1.1	1.0	1.0	2.5	7.3
29	29	2018	1.1	1.0	1.0	2.5	7.3
30	30	2018	1.1	1.0	1.0	2.5	7.3
31	31	2018	1.1	1.0	1.0	2.5	7.3
32	32	2018	1.1	1.0	1.0	2.5	7.3
33	33	2018	1.1	1.0	1.0	2.5	7.3
34	34	2018	1.1	1.0	1.0	2.5	7.3
35	35	2018	1.1	1.0	1.0	2.5	7.3
36	36	2018	1.1	1.0	1.0	2.5	7.3
37	37	2018	1.1	1.0	1.0	2.5	7.3
38	38	2018	1.1	1.0	1.0	2.5	7.3
39	39	2018	1.1	1.0	1.0	2.5	7.3
40	40	2018	1.1	1.0	1.0	2.5	7.3
41	41	2018	1.1	1.0	1.0	2.5	7.3
42	42	2018	1.1	1.0	1.0	2.5	7.3
43	43	2018	1.1	1.0	1.0	2.5	7.3
44	44	2018	1.1	1.0	1.0	2.5	7.3
45	45	2018	1.1	1.0	1.0	2.5	7.3
46	46	2018	1.1	1.0	1.0	2.5	7.3
47	47	2018	1.1	1.0	1.0	2.5	7.3
48	48	2018	1.1	1.0	1.0	2.5	7.3
49	49	2018	1.1	1.0	1.0	2.5	7.3
50	50	2018	1.1	1.0	1.0	2.5	7.3
51	51	2018	1.1	1.0	1.0	2.5	7.3
52	52	2018	1.1	1.0	1.0	2.5	7.3

C. Analysis

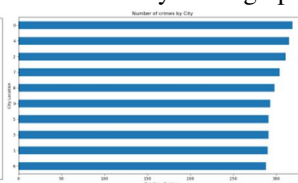
The analysis contains a graphical depiction of several variables used to analyze the dataset's properties. The various graphs are created using Matplotlib packages. Graphs include histogram plots, heatmaps, and crime type count graphs.



Crime type count graph

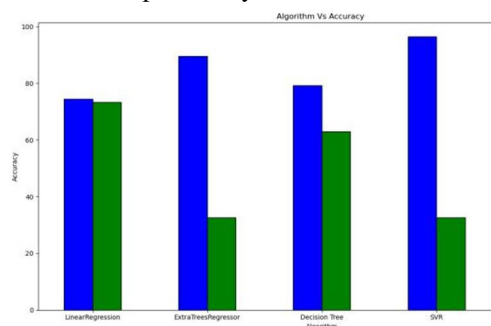


Crime city count graph



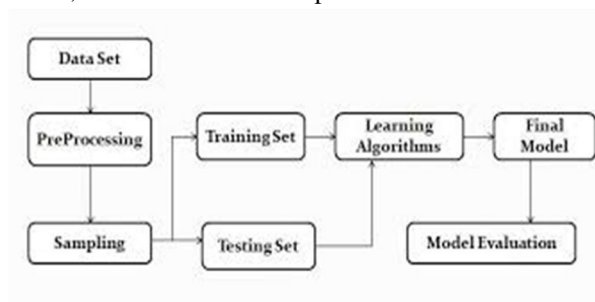
D. Training and Testing

The dataset is separated into two parts: training and testing. In general, 80% of the dataset is preserved for training, and 20% for testing. This divide allows the model to be assessed on previously unseen data.



E. Evaluation

After the model is constructed, it should be evaluated against real-time data values. This process is referred to as validation. The validation is nothing but the projected value, also known as the output value.



VI. ALGORITHMS AND DESCRIPTION:

A. Linear Regression

One method of supervised learning is linear regression. Based on a given independent variable (X), it is in charge of forecasting the value of a dependent variable (Y). The relationship between the input (X) and the output (Y) is what it is. It is among the most well-known and comprehended algorithms for machine learning. The linear regression models are Regularization, Gradient Descent, Ordinary Least Squares, and Simple Linear Regression.

B. Logistic Regression

A supervised machine learning approach called logistic regression is used for classification problems in which predicting the likelihood that an instance belongs to a particular class or not is the aim. A statistical procedure called logistic regression examines the connection between two data points. The basics of logistic regression are examined in this article.

C. SVM

By carrying out optimal data transformations that establish boundaries between data points based on predefined classes, labels, or outputs, supervised learning models enable support vector machines (SVMs), a type of machine learning algorithm, to solve challenging classification, regression, and outlier detection problems. SVMs are extensively used in a variety of sectors, including speech and picture identification, natural language processing, healthcare, and signal processing applications.

D. Random Forest

One well-known machine learning method that is a part of the supervised learning approach is Random Forest. In machine learning, it can be applied to both classification and regression issues. It is predicated on the idea of ensemble learning, which is the act of merging several classifiers to solve a challenging issue and enhance the model's functionality. Compared to previous algorithms, it requires less training time; it predicts output with high accuracy, especially for huge datasets; and it can retain accuracy even when a significant amount of the data is missing.

E. Result

Home page

Crime Prediction

Select City:

Select Crime Type:

Select Year:

Select Algorithm:

[Predict](#)

Results Page

Linear regression

Crime Prediction

Select City:

Select Crime Type:

Select Year:

Select Algorithm:

[Predict](#)

Prediction Results:

Prediction Rate: 74.54 %
 Estimated Crime Rate: 15.57 %
 Estimated Number of Cases: 512.78
 Population: 49.69 lakhs

Random forest

Crime Prediction

Select City:

Select Crime Type:

Select Year:

Select Algorithm:

[Predict](#)

Prediction Results:

Prediction Rate: 72.0 %
 Estimated Crime Rate: 13.95 %
 Estimated Number of Cases: 626.17
 Population: 48.21 lakhs

Logistic

Crime Prediction

Select City:
 New York

Select Crime Type:
 Theft

Select Year:
 2018

Select Algorithm:
 Logistic

Predict

Prediction Results:

Prediction Rate: 86.75 %
 Estimated Crime Rate: 12.27 %
 Estimated Number of Cases: 329.82
 Population: 25.52 lakhs

Activate Windows
 Go to Settings to activate Windows

SVM

Crime Prediction

Select City:
 New York

Select Crime Type:
 Theft

Select Year:
 2018

Select Algorithm:
 SVM

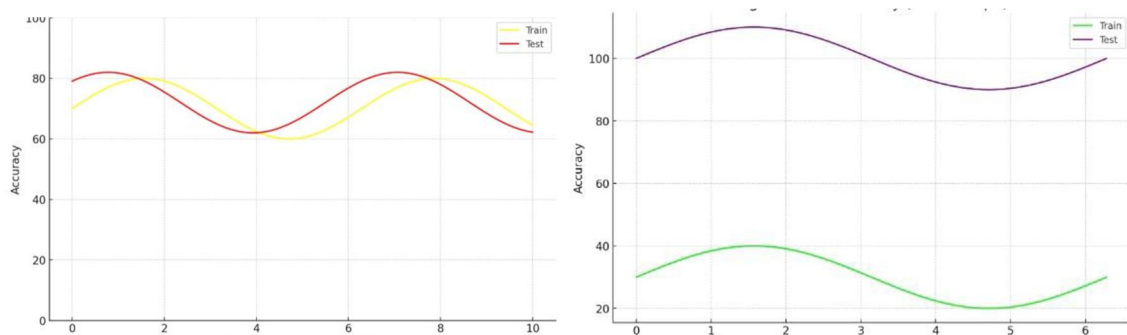
Predict

Prediction Results:

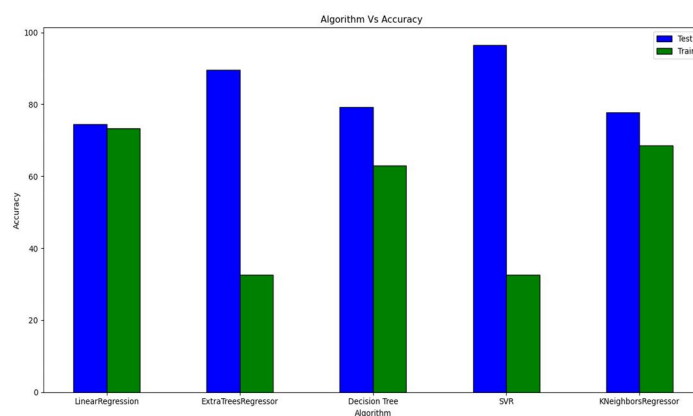
Prediction Rate: 73.9 %
 Estimated Crime Rate: 10.9 %
 Estimated Number of Cases: 514.0
 Population: 49.9 lakhs

Activate Windows
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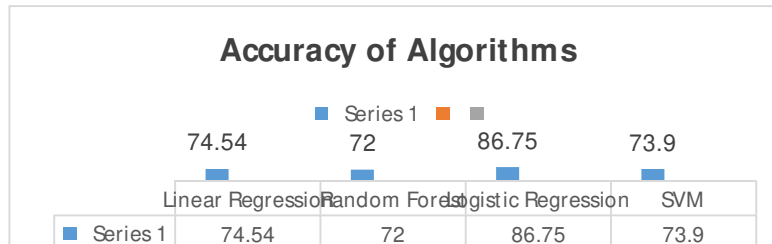
F. Accuracy Graph



G. Testing and Training



H. Comparison of Algorithms



VII. FUTURE WORK

In order to enhance overall performance and increase the accuracy of crime prediction, future research in this field should concentrate on using additional categorization models. Comparing and contrasting the accuracy of the different classification algorithms for forecasting the kind of crimes that would occur would be beneficial. By examining the relationship between neighborhood income level and crime rate, it may be possible to gain a better understanding of the root causes of crime and create focused strategies to prevent it in high-risk regions. This kind of study may be helpful to governments and law enforcement agencies in better allocating resources to reduce crime and improve safety.

VIII. CONCLUSION

Consequently, the machine learning model that makes use of SVM, random forest, linear regression, and logistical regression is highly effective in predicting criminal activity. India's crime rate rises daily due to a number of causes, including corruption, poverty, and poor execution. When taking the required actions to lower crime, the suggested model is very helpful to both police officials and investigating agencies. Through a variety of interactive visualizations, the project assists the criminal analysis in analyzing these crime networks. Future research in this field will focus on training robots to use automated learning techniques to forecast crime in a region. Since machine driving and data extraction are similar, autonomous learning expansion concepts can be utilized to improve predictions. To make better forecasts, increase the accuracy, dependability, and privacy of your data.

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