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Scientific Research and Methodology

Course Code: CSE 418 (Dual)

Course Teacher:

Ratri Datta

Lecturer, UITs

Md. Moradul Siddique

Lecturer, UITs



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Title of the Research: Enhancing Crime Prediction Accuracy
Through Machine Learning: A Predictive Policing Approach

Team Name: Digital_Warrior

Team Member:

Md. Shakibul Islam Ramim (2125051063)

MD.Nahian Islam Emon (2125051114)

Fazlay Rabbi (2125051070)

Mst. Sumi Akter (2125051037)

Title: Enhancing Crime Prediction Accuracy Through Machine Learning: A Predictive Policing Approach

Abstract: Crime and violation are the threat to justice and meant to be controlled. Accurate crime prediction and future forecasting trends can assist to enhance metropolitan safety computationally. The limited ability of humans to process complex information from big data hinders the early and accurate prediction and forecasting of crime. The accurate estimation of the crime rate, types and hot spots from past patterns creates many computational challenges and opportunities. This research paper investigates the application of machine learning algorithms to enhance crime prediction accuracy within urban environments, addressing the urgent need for effective predictive policing strategies. As crime rates continue to escalate in various regions, traditional analytical methods are proving insufficient in identifying and mitigating safety risks. We employ a comprehensive analysis of crime data using several supervised machines learning techniques, including Random Forest, K-Nearest Neighbors, AdaBoost, and Neural Networks, to predict crime occurrence based on location and time variables. Our methodology includes rigorous data preprocessing, validation, and visualization to ensure the reliability of our findings. The study highlights significant trends in crime patterns and the potential for these algorithms to inform law enforcement resource allocation. Notably, the Neural Network model demonstrated superior predictive accuracy, achieving an impressive accuracy rate of 90.77%. This research contributes to the ongoing discourse on utilizing advanced analytics in policing, offering practical insights for law enforcement agencies to enhance public safety and effectively respond to emerging crime hotspots.

Keywords: Crime prediction, Machine learning, Predictive policing, Urban safety, Supervised learning, Crime data analysis, Random Forest, K-Nearest Neighbors, AdaBoost, Neural Networks, Crime patterns, Resource allocation, Data visualization, public safety, Crime hotspots

Introduction: In today's rapidly urbanizing world, the rise in population density and city expansion presents significant challenges for urban management, particularly concerning public safety. As metropolitan areas grow, they often experience higher crime rates, which not only threaten individual security but also adversely impact social well-being and economic stability. Crime prediction has emerged as a crucial strategy to combat this issue, leveraging advanced algorithms to forecast criminal activities based on historical data.

The necessity for effective crime prediction is underscored by the prevalence of various criminal acts, including robbery, assault, and kidnapping, which pose a constant threat to community safety. As individuals navigate their daily lives, they frequently encounter potential dangers, making the ability to predict crime essential for informed decision-making and enhanced personal security. Just as one might consult Google Maps for optimal routes, understanding crime risk in different areas can help individuals choose safer paths.

This research aims to design and implement a predictive model that analyzes crime data to forecast crime rates at specific times and locations. We utilize primary data collected from communities regarding their past crime experiences, coupled with sophisticated machine learning algorithms, to derive accurate predictions. By employing techniques such as K-Nearest Neighbors (KNN) and other supervised learning models, we assess the predictive accuracy of various algorithms, providing insights into crime trends over the last three years.

The structure of this paper is as follows: Section 2 reviews the existing literature on crime prediction methodologies, highlighting previous research efforts and their findings. Section 3 details our methodology, including data collection, processing, and analysis techniques used to predict crime rates. Section 4 presents results and discussions on the effectiveness of different models and

their implications for resource allocation in law enforcement. Finally, Section 5 concludes with reflections on our findings and recommendations for future research in the domain of predictive policing. Through this work, we aim to contribute to the development of safer urban environments and enhance the effectiveness of policing strategies in crime prevention.

2. Literature Review:

Working requires more time

3. Research Methodology:

3.1. Dataset

3.2. Processing

3.3. Crime Dataset

Working requires more time

4. Presents results and discussions on the effectiveness of different models and their implications:

Working requires more time

5. Result and Discussion:

Working requires more time

6. Conclusion and acknowledgment and Reference: Working requires more time

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