

Crime Analysis Using DBSCAN

Course: Data Mining Techniques.

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Introduction & Problem Statement

1 Introduction

Crime analysis is crucial for law enforcement & public safety.

Problem

Traditional crime mapping lacks efficiency in detecting patterns.

3 Solution

DBSCAN offers an advanced approach to identify crime hotspots.



Literature Review & Why DBSCAN?

Crime Data Analysis

- Data mining is widely used for crime pattern detection.
- Various clustering techniques are applied in crime analysis.

Why DBSCAN?

- Detects irregularly shaped crime clusters.
- Automatically identifies noise (outliers).
- No need to specify the number of clusters.

Methodology

1

Data Collection

Crime dataset with location, crime type, time.

2

Data Preprocessing

Handling missing values, scaling, encoding.

3

Applying DBSCAN

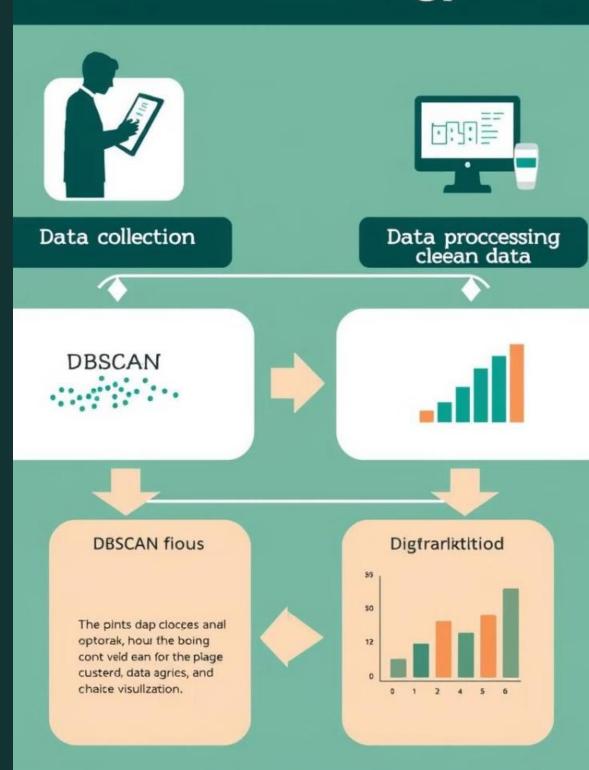
Identifying high-density crime regions.

4

Visualization

Scatter plots & heatmaps for better insights.

Methodology





Data Collection & Preprocessing

Data Sources & Features

- Public crime datasets, police records.
- Features: Latitude, Longitude, Crime Type.

Preprocessing Steps

- · Handling missing values.
- Feature normalization & encoding.
- Optimizing DBSCAN parameters.

DBSCAN Algorithm & Implementation

How DBSCAN Works?

Density-based clustering groups nearby points, detects outliers.

Epsilon (ε)

Defines the neighborhood radius.

MinPts

Minimum points required to form a dense region.

Implementation (Python Libraries Used)

- pandas, NumPy Data processing.
- scikit-learn DBSCAN algorithm.
- matplotlib, seaborn Visualization.

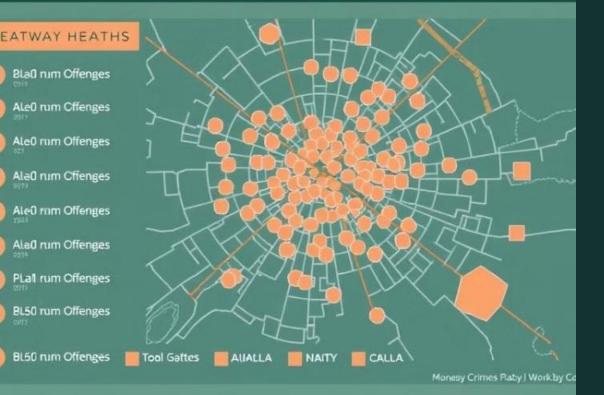


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Results & Visualizations



Outliers Identified

Isolated crimes needing investigation.

Visualization

Scatter plots & heatmaps enhance interpretation.

Challenges & Limitations

Challenges Faced

- · Data Quality Issues Missing/incomplete records.
- Parameter Sensitivity Choosing optimal ε & MinPts.
- · Computational Complexity Processing large datasets efficiently.

Limitations

- Focuses on spatial clustering only, does not analyze temporal trends.
- Lack of real-time crime data integration.



Future Enhancements

Predictive Analysis

Integrating Machine Learning.

Real-Time Data

Dynamic Clustering.

Interactive Dashboard

For Law Enforcement.

Hybrid Approach

Combining DBSCAN with Other Techniques.

Conclusion

- 1 Final Takeaways

 DBSCAN effectively
 identifies crime hotspots
 & outliers.
- Crime VisualizationHelps law enforcement in decision-making.

3 Further ImprovementsCan enhance crime prevention strategies.



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YOU!!