

EDA Patrol - Crime in India (2001-2014)

Exploratory Data Analysis Report

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Technology Used: Python and Streamlit

[Link for Interactive Dashboard](#) - Hosted In Streamlit

Introduction

This report presents an in-depth exploratory data analysis (EDA) of the Indian Penal Code (IPC) crime dataset from 2001 to 2014. The dataset includes district-wise crime records across Indian states. The goal is to extract key insights, identify crime trends, evaluate high-risk areas, and apply machine learning models to assist in predictive analysis and policymaking.

Data Cleaning and Preparation

- Merged district-level crime data with geo-boundary shapefiles using fuzzy string matching.
- Removed rows with labels such as "Total" which represented aggregate values.
- Normalized district and state names for consistency across datasets.
- Created new columns such as `crime_risk_index`, `high_crime` binary labels, and `most_common_crime`.

Exploratory Data Analysis

Total Crimes and Murders

- **Total IPC Crimes Recorded (2001-2014):** Over 29 million.
- **Average Murders per District:** 43 murders.

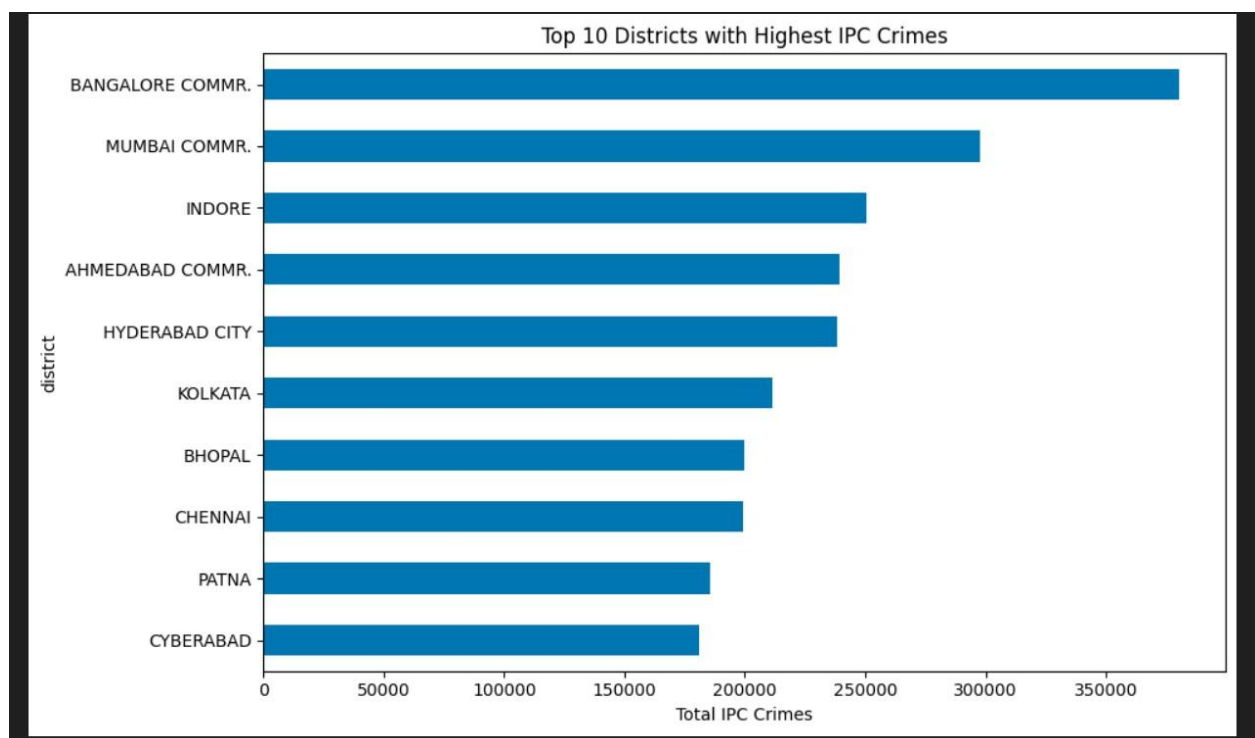
Crime Distribution Across States

States with highest total IPC crimes:

1. Madhya Pradesh
2. Maharashtra
3. Tamil Nadu
4. Andhra Pradesh
5. Uttar Pradesh

Top Crime-Intensive Districts

1. Bangalore Commr.
2. Mumbai Commr.
3. Indore
4. Ahmedabad Commr.
5. Hyderabad City



Urban vs. Rural Crime Patterns

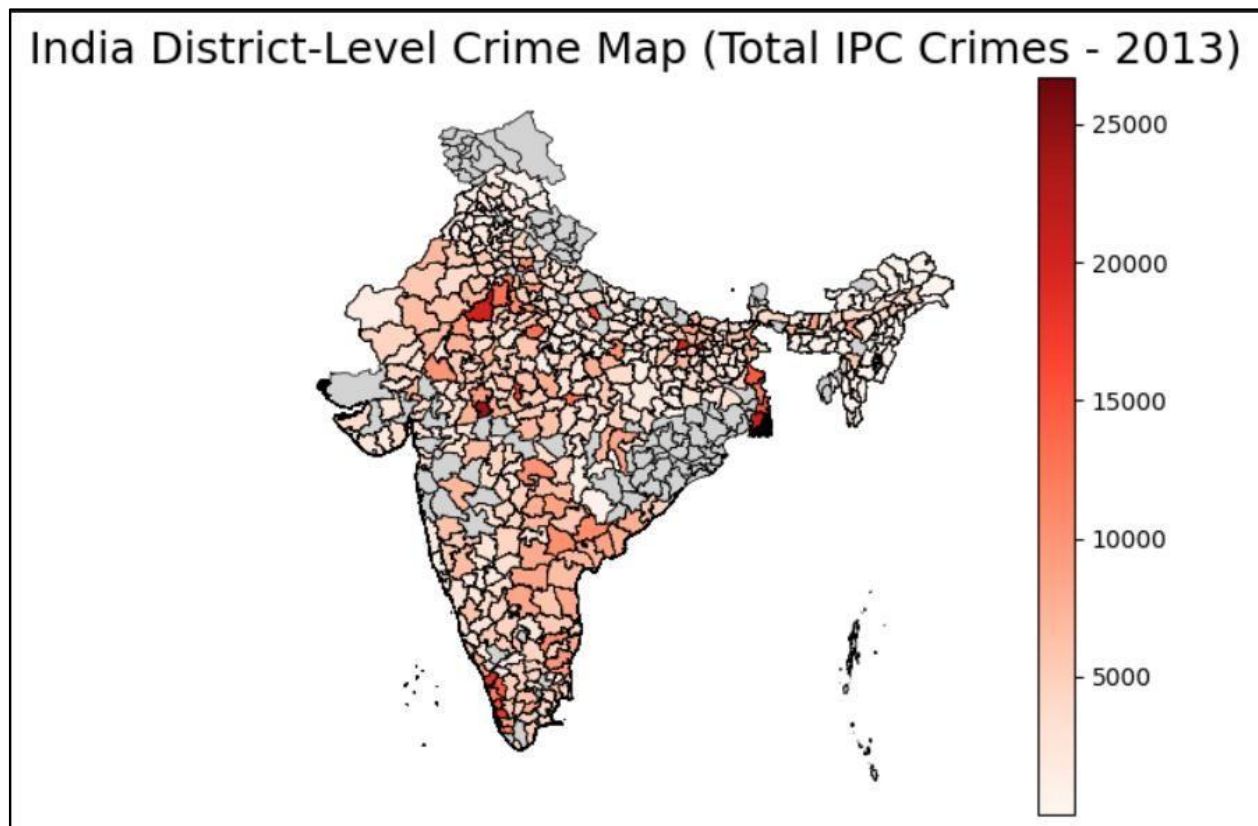
Due to the absence of an explicit urban/rural flag, districts containing city names (e.g., "commr", "urban") were used as a proxy for urban areas.

- **Average IPC crimes in urban areas:** ~7346
- **Average IPC crimes in non-urban areas:** ~270

Visualization and Interpretation

Heatmaps and Geospatial Plots

- Crime density is highest in central and southern India.
- Geospatial heatmaps highlighted major hotspots like Delhi, Mumbai, and Bangalore.
- Choropleth maps were generated by merging shapefile data with district crime aggregates.



Advanced Statistical and Machine Learning Analysis

Most Common Crimes by District

- For each district, the most reported crime was determined.
- Most common: "hurt/grevious hurt" and "theft".

Crime Risk Index

Developed using a weighted sum of serious crimes:

- Murder (weight=3), Rape (2), Robbery (2), Theft (1), Dacoity (2), Kidnapping (2)

Top districts: Outer Delhi, Mumbai, North-East Delhi.

	crime_risk_index
district	
indore	70190
patna	69929
south	69910
mumbai	65462
west	65314
kolkata	60666
east	60230
lucknow	53446
cyberabad	52274
chennai	49069

Classification of High vs. Low Crime Districts

- Binary target created using median split on `total_ipc_crimes`.
- Model: Random Forest Classifier
- Accuracy: 95%

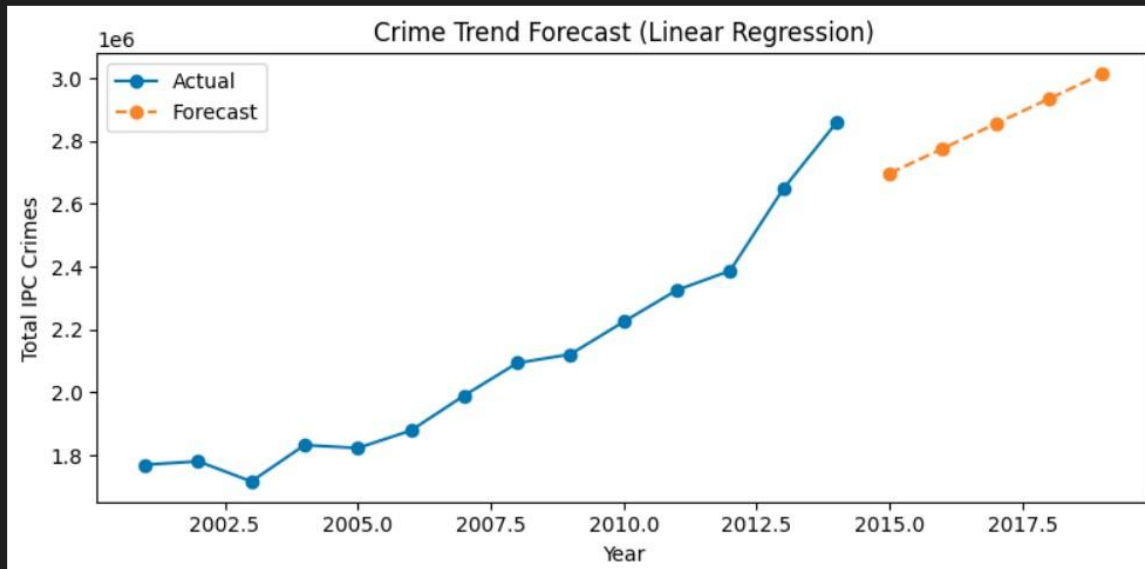
Precision/Recall

Balanced performance indicates strong predictability from raw crime type features.

	precision	recall	f1-score	support
0	0.95	0.95	0.95	1389
1	0.95	0.95	0.95	1406
accuracy			0.95	2795
macro avg	0.95	0.95	0.95	2795
weighted avg	0.95	0.95	0.95	2795

Time-Series Forecasting

- Used linear regression to forecast IPC crime totals for upcoming years.
- Projected crimes for 2019: ~3 million.



Bonus Analyses

Crimes Against Women

- Grouped crimes like rape, dowry deaths, assault, and trafficking.
- Women-targeted crimes formed ~10% of total IPC crimes.

✓ About 9.97% of all IPC crimes are against women.

We grouped the following crime columns under women_crimes because they specifically represent crimes committed against women and girls. This includes offenses like rape, dowry deaths, kidnapping, domestic violence, and assault on modesty.

By summing these columns, we calculated the total number of crimes against women, which was then used to determine what percentage of overall IPC crimes these cases represent — addressing Bonus Question 1.

Dowry Deaths

Uttar Pradesh reported the highest number of dowry deaths.

Crime and Cities

- Urban centers exhibited higher crime volumes.
- City-based crimes like auto theft, robbery, and assault had above-average frequencies.

Conclusion

This analysis revealed stark regional differences in crime intensity, uncovered deep correlations across IPC sections, and enabled risk indexing of Indian districts. The use of machine learning models like classification and forecasting demonstrated practical value in strategic planning and crime prevention. Future enhancements may include monthly-level data to enable seasonal trend analysis and deeper socio-economic feature integration.
