



**Decoding Danger – A Predictive Analysis of the Communities and Crime  
(UCI) Dataset**

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**Context: Final Project**

**Project GitHub Repository:** <https://github.com/Deepi-boobi02/communities-crime-prediction.git>

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## 1. Executive Summary

The goal of this project was to analyse how socio-economic factors (such as income, education, and family structure) influence violent crime rates in U.S. communities.

Using the Communities and Crime dataset from the UCI Machine Learning Repository, I cleaned raw data, performed exploratory data analysis, and built machine learning models to predict crime rates.

**Key Finding:** My analysis revealed that family structure—specifically the percentage of children living in two-parent households—is the single strongest predictor of community safety, significantly outranking economic variables like median rent or employment rates.

## 2. Methodology

### Data Source

- Dataset Name: Communities and Crime (UCI)
- Source: UCI Machine Learning Repository
- Description: The dataset combines socio-economic data from the 1990 US Census, law enforcement data from the 1990 LEMAS survey, and crime data from the 1995 FBI UCR.

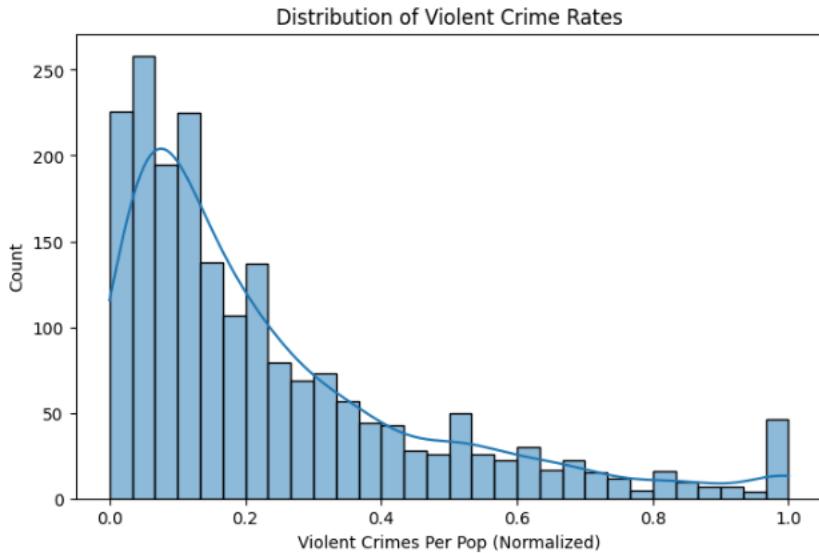
### Data Cleaning

The original dataset contained 128 variables and 1,994 communities.

- **Missing Data:** I identified that 27 columns related to police department reporting (e.g., LemmasSwornFT) were missing over 80% of their data. These columns were removed to prevent model errors.
- **Imputation:** Minor missing values in demographic columns were filled using the column mean.
- **Final Shape:** The clean dataset used for modelling contained 1,994 rows and 96 feature columns.

### Exploratory Data Analysis (EDA)

I examined the distribution of the target variable, ViolentCrimesPerPop. The data is right-skewed, indicating that while most communities have low-to-moderate crime rates, a small subset of communities' experiences disproportionately high crime.



*Figure 1: Distribution of Violent Crime Rates (Normalized).*

### 3. Model Performance

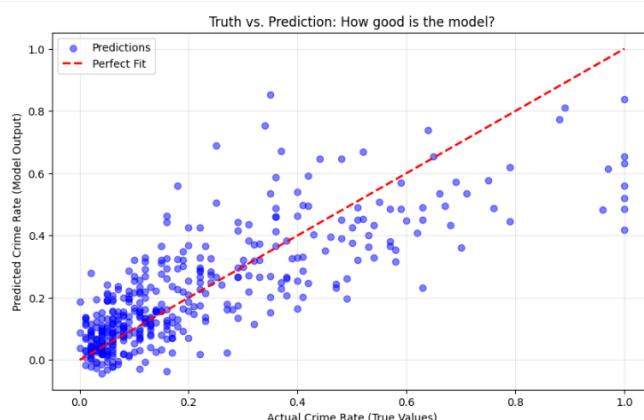
I trained two separate machine learning models to predict the crime rate.

#### Model 1: Linear Regression (Baseline)

- **RMSE:** 0.1337
- **R<sup>2</sup> Score:** 0.6269
- **Interpretation:** This model performed the best, explaining ~63% of the variance in crime rates. This suggests the relationships in the data are largely linear.

#### Model 2: Random Forest Regressor

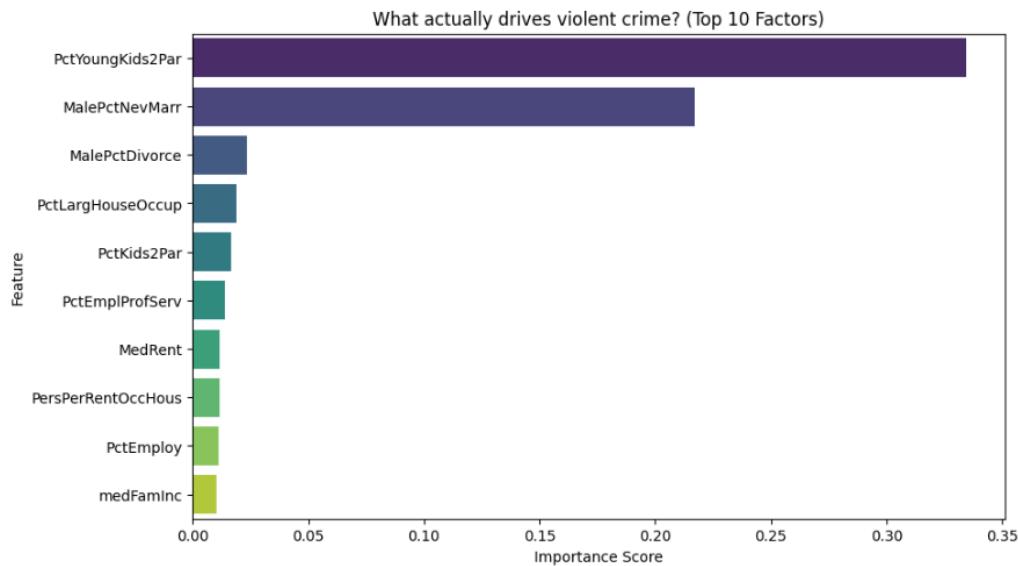
- **RMSE:** 0.1373
- **R<sup>2</sup> Score:** 0.6062
- **Interpretation:** The Random Forest model provided valuable feature importance data but slightly overfitted the noise in the training set compared to the linear model.



*Figure 2: Actual vs. Predicted Crime Rates. The proximity of the blue dots to the red dashed line demonstrates the model's accuracy.*

## 4. Key Insights: What Drives Crime?

Using the Random Forest model, I extracted the "Feature Importance" to understand which specific variables drove the predictions.



*Figure 3: Top 10 Factors Influencing Violent Crime.*

### Analysis of Drivers:

**Family Stability:** The top predictor was PctYoungKids2Par (Percentage of kids in two-parent households). The model indicates a strong correlation between family stability and lower crime rates.

**Marital Status:** The second strongest predictor was MalePctNevMarr (Percentage of males never married), reinforcing the family structure trend.

**Economic Impact:** While medIncome (Median Income) was a factor, it ranked lower than family dynamics, suggesting that economic wealth alone does not guarantee safety.

## 5. Conclusion

I successfully built a machine learning pipeline using the **Communities and Crime (UCI)** dataset that predicts community crime rates with **~63% accuracy**. The project highlights that social support structures (family units) may be more critical predictive signals than purely economic indicators.

## 6. References

1. **Dataset:** Redmond, M. A., & Baveja, A. (2002). *Communities and Crime Data Set*. UCI Machine Learning Repository.  
<https://archive.ics.uci.edu/ml/datasets/Communities+and+Crime>
2. **Original Research:** Redmond, M. A., & Baveja, A. (2002). A Data-Driven Software Tool for Enabling Cooperative Information Sharing Among Police Departments. *European Journal of Operational Research*, 141(3), 660-678.
3. **Tools Used:** Python (pandas, scikit-learn, matplotlib, seaborn).