

Crime Hotspots Forecasting via Deep Learning Methodology

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Introduction

Crime remains a significant social issue in the United States, with a violent crime committed every 24.6 seconds and a property crime offense occurring every 4.1 seconds, as per the 2017 FBI data. This poses a serious threat to public safety and has profound implications for the economy. The need for effective prediction and comprehension of criminal activity patterns is crucial, particularly in identifying future high-risk areas, commonly referred to as "crime hot spots."

Methods

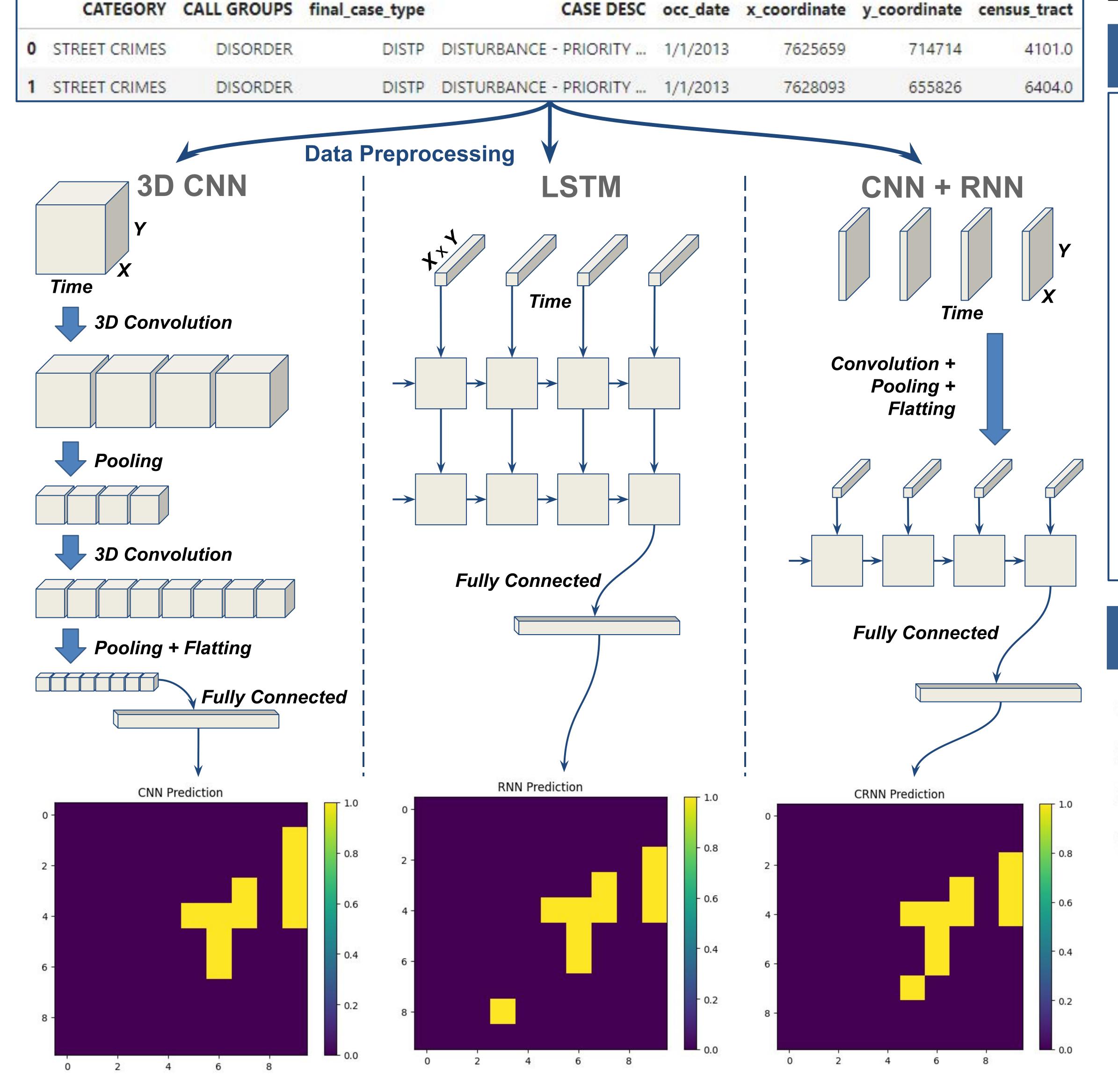
In this study, we leverage three deep neural network architectures to forecast future crime hotspots:

- 3D Convolutional Neural Networks (CNNs): use 3D convolutions to identify spatio-temporal patterns, capturing the complex dynamics of crime events over time and space.
- Long Short-Term Memory (LSTM): excels in analyzing sequential data, making it ideal for predicting patterns that evolve.
- CNN + RNN Hybrid: By combining CNN and RNN architectures, this hybrid approach harnesses CNN's ability to detect spatial patterns and RNN's strength in sequence prediction, offering a comprehensive model for crime hotspot forecasting.

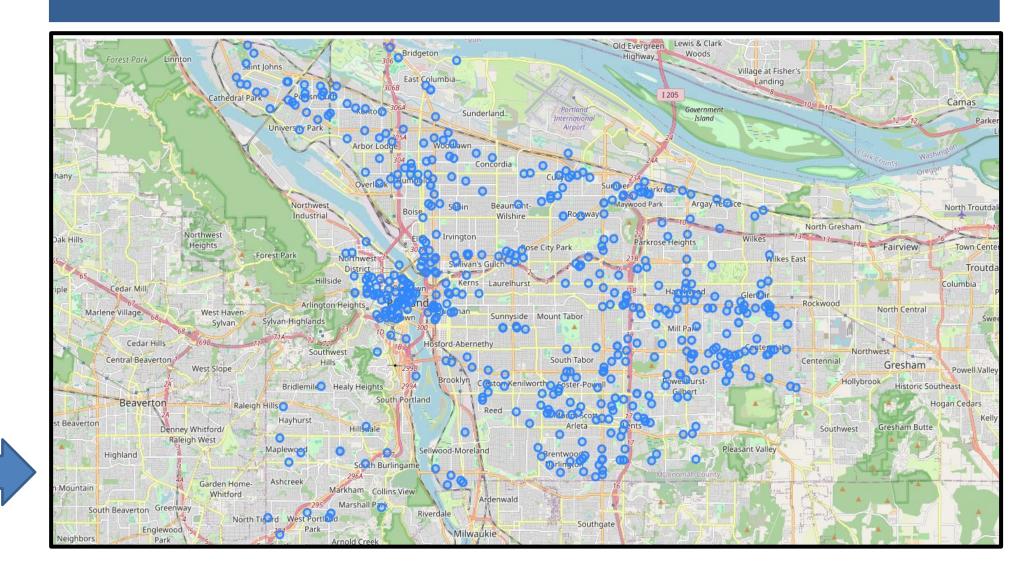
Experiments

To validate our approach, we utilized comprehensive call-for-service data provided by the Portland, Oregon Police Bureau (PPB). Comparative analyses were conducted with various deep neural network architectures. Experiment workflow as shown below:

Call For Service Data:



Data on the map



Future work

- Enhancing Spatial Resolution:
 We plan to refine our models
 to achieve higher resolution in
 hotspot map predictions,
 allowing for more detailed and
 precise forecasting.
- Incorporating Additional Data:
 We plan to integrate
 additional knowledge(e.g.
 census data) into our deep
 neural network architecture to
 improve its predictive
 accuracy.

Ground Truth

