

# Dynamic Pricing for Urban Parking Lots

Capstone Project - Summer Analytics 2025

Parth Pardeshi

## 1. Introduction

Urban parking spaces are limited and demand fluctuates heavily throughout the day. This project implements a dynamic pricing strategy for 14 urban parking lots using real-time data, economic principles, and machine learning techniques.

## 2. Data Overview

The dataset spans 73 days of 30-minute interval data for 14 parking lots. Features include: Capacity, Occupancy, Queue Length, Vehicle Type, Traffic Condition, Special Event Indicator, etc

## 3. Model 1 - Linear Baseline

This simple model increases the price proportionally to occupancy:  $\text{Price}_{t+1} = \text{Price}_t + \alpha * (\text{Occupancy} / \text{Capacity})$  ; Base Price = \$10, Alpha = 2.0

## 4. Model 2 - Demand-Based Pricing

This model incorporates multiple factors into a demand function:

$\text{Demand} = \alpha * (\text{Occupancy} / \text{Capacity}) + \beta * \text{QueueLength} - \gamma * \text{Traffic} + \delta * \text{IsSpecialDay} + \epsilon * \text{VehicleTypeWeight}$  .

$\text{Price} = \text{Base} * (1 + \lambda * \text{Normalized Demand})$

Bounded between 0.5x and 2x the base price.

## 5. Model 3 - Competitive Pricing

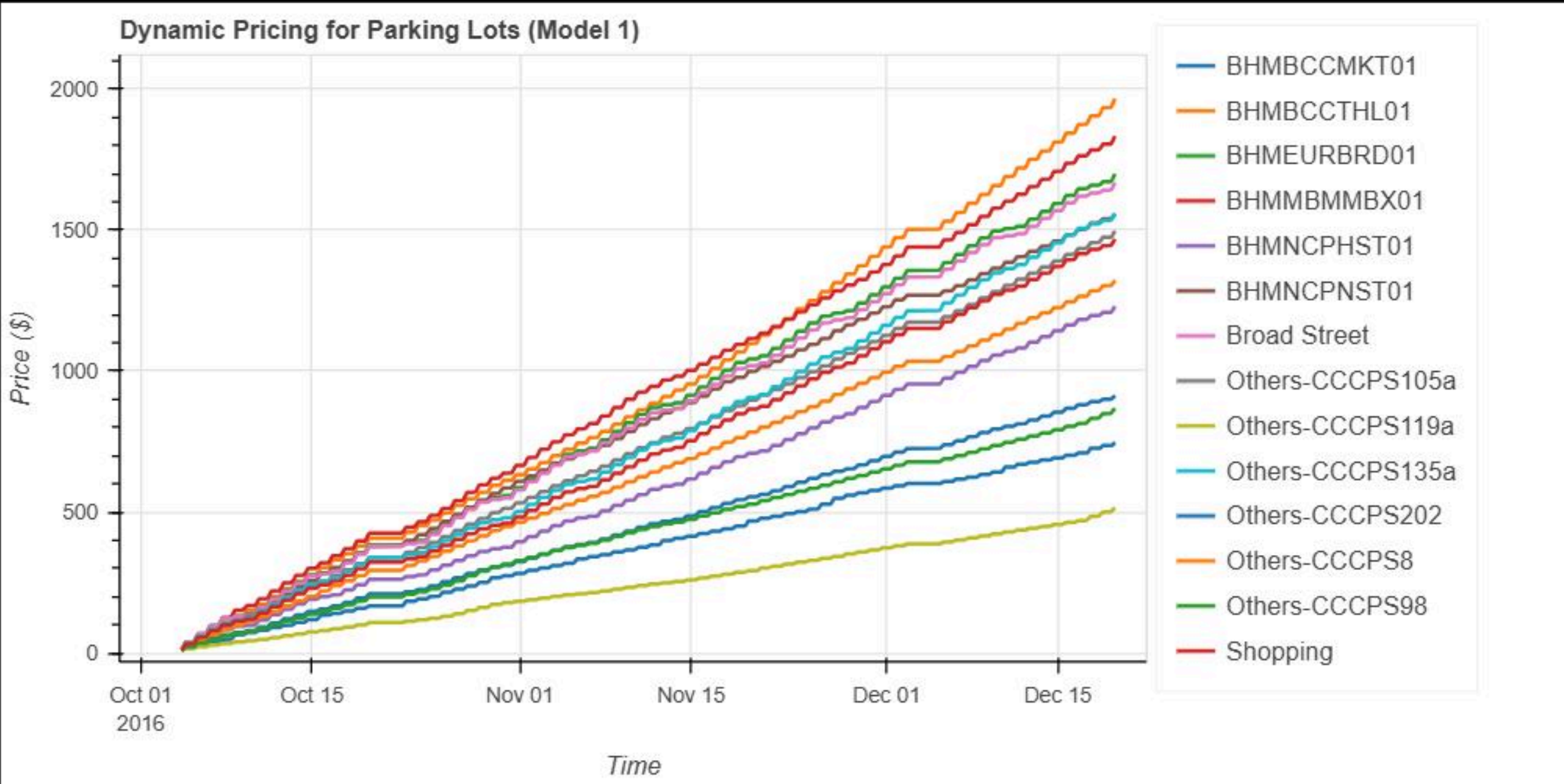
Uses geolocation to find nearby lots via haversine distance. If current lot is full and cheaper lots nearby, either reroute vehicles or reduce price to remain competitive..

## 6. Real-Time Simulation using Pathway

Streaming ingestion is handled using Pathway's streaming CSV reader. The pipeline computes occupancy ratio and outputs price prediction in real-time.

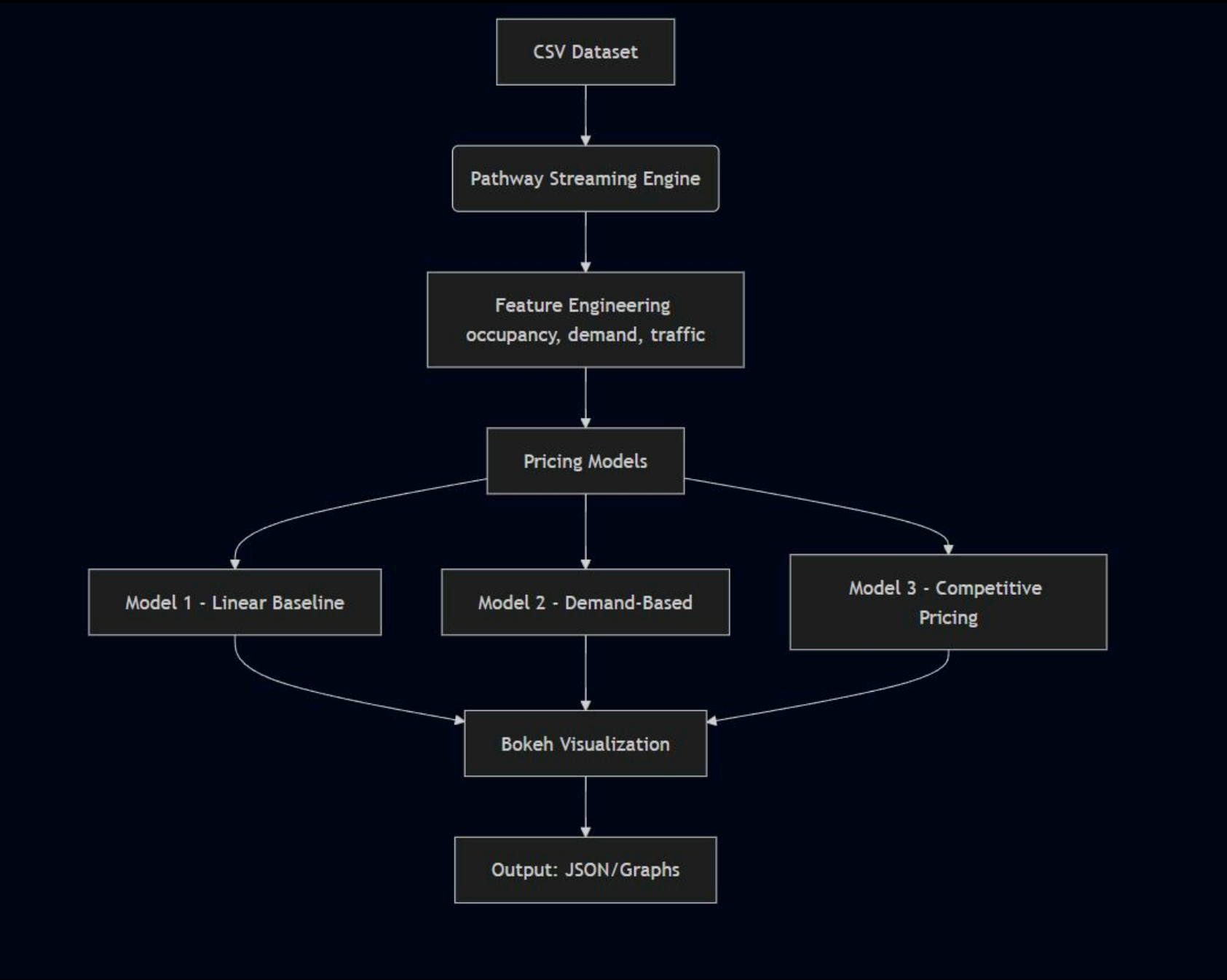
# 7. Visualization using Bokeh

Prices are visualized over time using Bokeh. Each lot has a separate colored line.



# 8. Assumptions

- Base price for all lots starts at \$10
- Vehicle type weights: car=1.0, bike=0.5, truck=1.5
- Demand coefficients are chosen heuristically for demonstration



# Conclusion

**This pricing system helps balance parking demand across multiple lots using live data.**

**Future improvements: fine-tuned coefficients, more real-time context (e.g., weather, bookings), feedback-based learning**

# Thankyou