# Model 1

This report explains the reasoning behind the selection of features, functions, and values used in Model 1 for dynamic parking price prediction using streaming data. Model 1 serves as the baseline for dynamic parking pricing, employing a simple, explainable linear relationship between parking lot occupancy and price.

### **Rationale for Feature Selection**

### 1. Occupancy Rate

- Definition: The ratio of the number of parked vehicles to the total parking capacity at a given time.
- Reason for Inclusion: Occupancy rate directly reflects demand.

### 2. Base Price

• Definition: The starting price for parking, set at \$10 as required by the project guidelines.

### 3. Price Function

a. Formula Used: 10.0 + 2.0 \* occ\_rate \* tod\_weight \* wd\_weight

where.

tod\_weight = time-of-day multiplier

wd\_weight = weekday/weekend multiplier

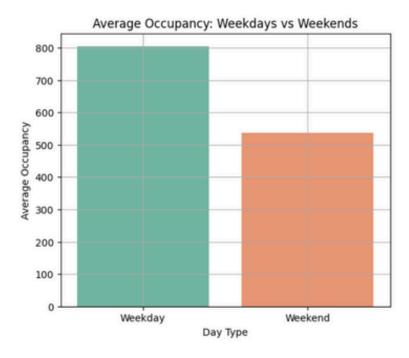
alpha=2.0 ---> to make price changes noticeable but not erratic.

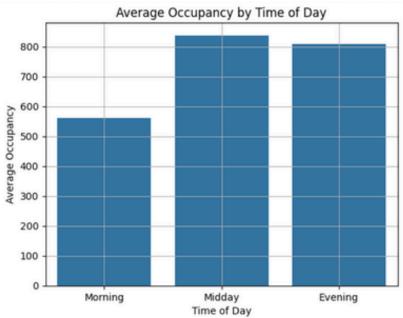
## Why This Formula?

Simplicity: Easy to implement and explain.

Direct Demand Response: Price increases smoothly with higher occupancy.

Transparency: Stakeholders can easily understand and trust the mechanism.





#### b. Time of Day Weight:

Midday (11–2 PM) is peak urban activity

Evening (2-5 PM) sees tapering demand

Morning is typically low demand

#### c.Weekday Weight:

Weekdays usually see more routine vehicle movement (commutes, business).

**Weekends** may have lower, less predictable demand → reduce price to encourage usage.

# Why These Parameters?

- **Simplicity**: This model serves as a baseline. It's interpretable, traceable, and easy to debug.
- Scalability: These weights can be easily modified or extended.
- Smooth Pricing: Using a multiplier design ensures gradual, explainable variations in price.

## **Data Processing Steps**

- Timestamp Creation: Combined date and time for accurate sequencing.
- Lot-wise Calculation: Each lot handled separately to reflect individual demand.
- Sorting: Data sorted by time to ensure correct price updates.

## **Assumptions**

- Only Occupancy Drives Price: Other factors (queue, traffic, vehicle type) are ignored in Model 1 for clarity.
- No Negative Prices: Prices never fall below the base price.
- Smooth Adjustments: Linear function prevents abrupt price jumps.

# **Visual Justification (Bokeh Plots)**

Each parking lot has:

- A real-time price graph with timestamps.
- Plot reflects:
  - Time-varying occupancy
  - Price variation patterns over hours
  - o Prices peak around midday.

### **Summary**

This model acts as a baseline dynamic pricing engine, using only real-time capacity-related data and basic time context to drive pricing decisions. It lays the foundation for more complex demand and competition-aware models in subsequent stages.