

Technical Document: Dynamic Pricing Algorithms and Rules

Project: Dynamic Pricing Strategies for Fitness Classes based on Demand, Time, and Location

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Objective: To transition fitness class pricing from a fixed-fee model to a demand-responsive system using Price Elasticity and Time-Series Forecasting.

1. The Core Pricing Algorithm

The pricing engine utilizes a **Heuristic Optimization Algorithm**. It does not predict price in a vacuum; instead, it calculates a **Price Adjustment Factor** by analyzing three distinct data dimensions:

- 1. **Demand Elasticity Factor:** Derived from Log-Log OLS Regression.
- 2. **Temporal Factor:** Derived from Prophet Time-Series Seasonality.
- 3. **Inventory Factor:** Derived from real-time occupancy rates.

2. Segment-Based Pricing Rules

Classes are categorized into four segments based on historical performance. Each segment triggers a specific pricing logic:

| Segment | Logic | Action Rule |
|--------------------|-----------------------------|--|
| Premium Powerhouse | High Occupancy / High Price | Maintenance: Maintain current price or apply +5% "Premium Guard" surge. |
| Underpriced Value | High Occupancy / Low Price | Surge: Apply +15% Surge Multiplier . Target for immediate yield growth. |
| Price Sensitive | Low Occupancy / High Price | Re-evaluation: Monitor for 2 weeks; if occupancy remains <30%, apply -5% correction. |
| Low Interest | Low Occupancy / Low Price | Incentivize: Apply -10% " Fill-the-Room " Discount to cover variable costs. |

3. Temporal Surge Rules (Prophet Integration)

The algorithm overrides standard pricing during "Power Hours" identified by the Prophet model:

- **Rule 3.1 (Peak Day Surge):** Apply an additional **2% multiplier** on Tuesdays and Wednesdays (statistically identified as the highest demand days).
- **Rule 3.2 (Rush Hour Surge):** Classes starting between **06:00–09:00** and **17:00–20:00** are subject to an immediate surge if occupancy is predicted to exceed 80%.

4. Revenue Protection Logic (The Elasticity Guardrail)

To ensure that price increases do not lead to a total loss of customers, the algorithm is capped by the **Elasticity Coefficient**:

- **Model Coefficient:** -0.3235 (Inelastic).
- **The Guardrail Rule:** No single price increase shall exceed **20%** in a single period. According to our model, a 20% increase results in a maximum projected volume drop of only **6.4%**, ensuring that Total Revenue ($P \times Q$) always increases.

5. Final Calculation Formula

The dynamic price for any specific class is calculated as follows:

$$\text{Price}_{\text{Dynamic}} = \text{Price}_{\text{Base}} \times (1 + \text{Segment Multiplier} + \text{Temporal Multiplier})$$

6. Verification and Performance Metrics

The success of these rules is verified through back-testing against the test dataset using:

- **Revenue Lift:** Target > 5% (Achieved: **6.91%**)
- **Yield per Slot:** Measures the average revenue earned per available mat.
- **Occupancy Variance:** Target a reduction in the gap between peak and off-peak **attendance**.

Appendix: Statistical & Empirical Evidence

Evidence 1: Price-Demand Sensitivity (Log-Log Regression)

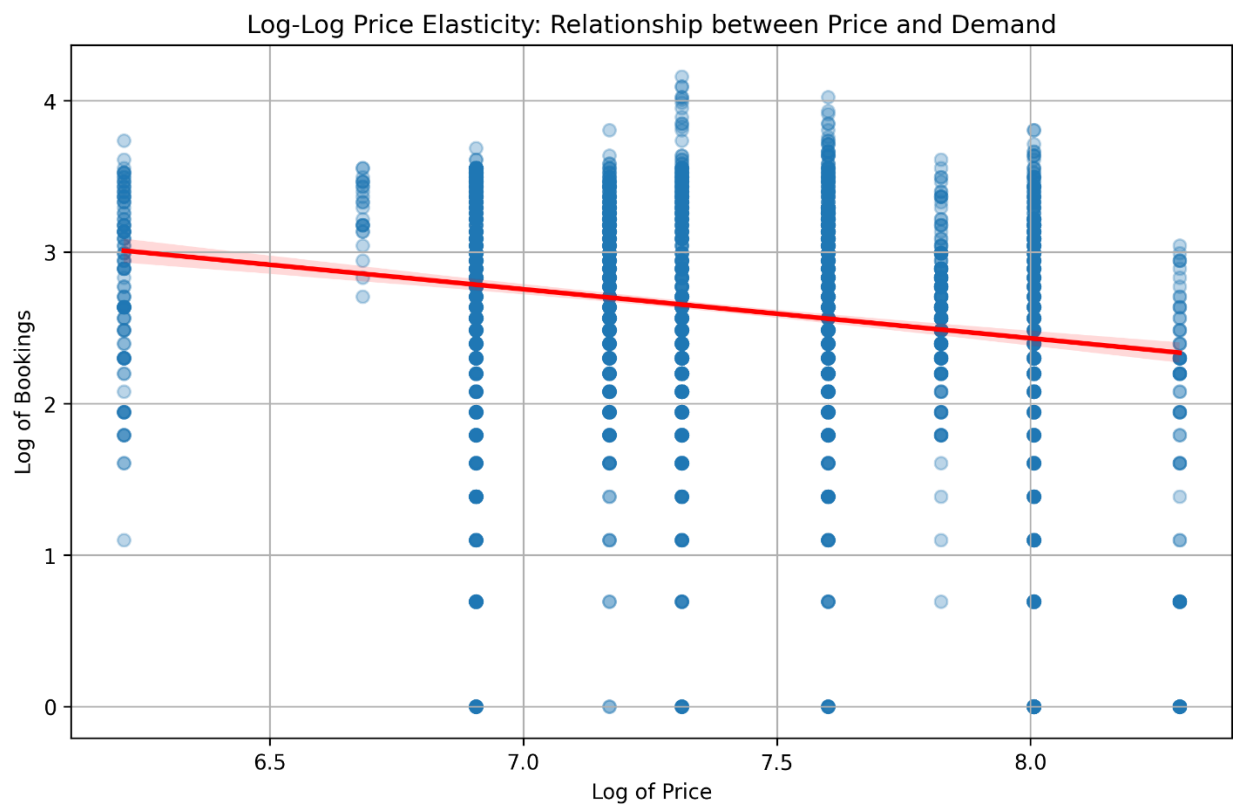


Figure A1: Log-Log Regression analysis demonstrating a negative price elasticity of -0.3235. This provides the mathematical basis for surge pricing during high-occupancy windows.

Evidence 2: Model Diagnostic Check (Residual Plot)

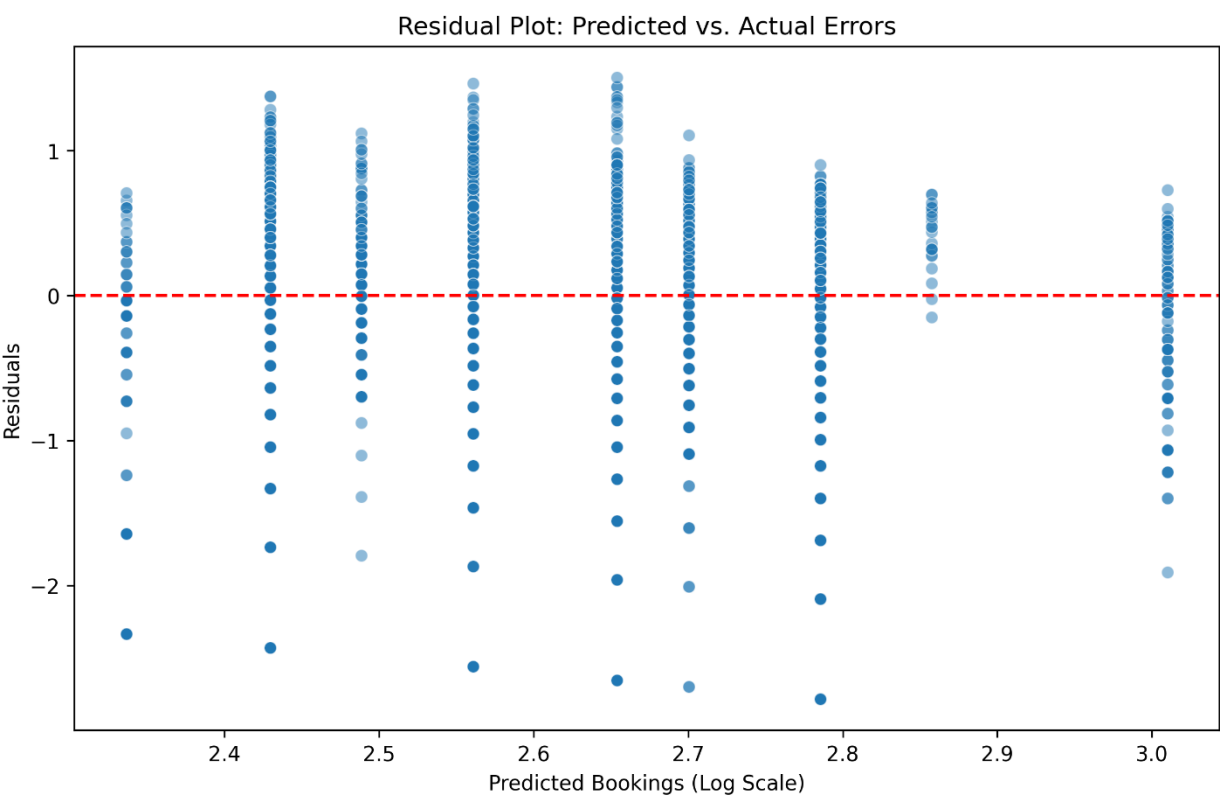


Figure A2:

Residual Plot showing an unbiased distribution of errors around the zero-line, confirming the statistical validity of our elasticity coefficient

Evidence 3: Temporal Demand Decomposition (Prophet Forecast)

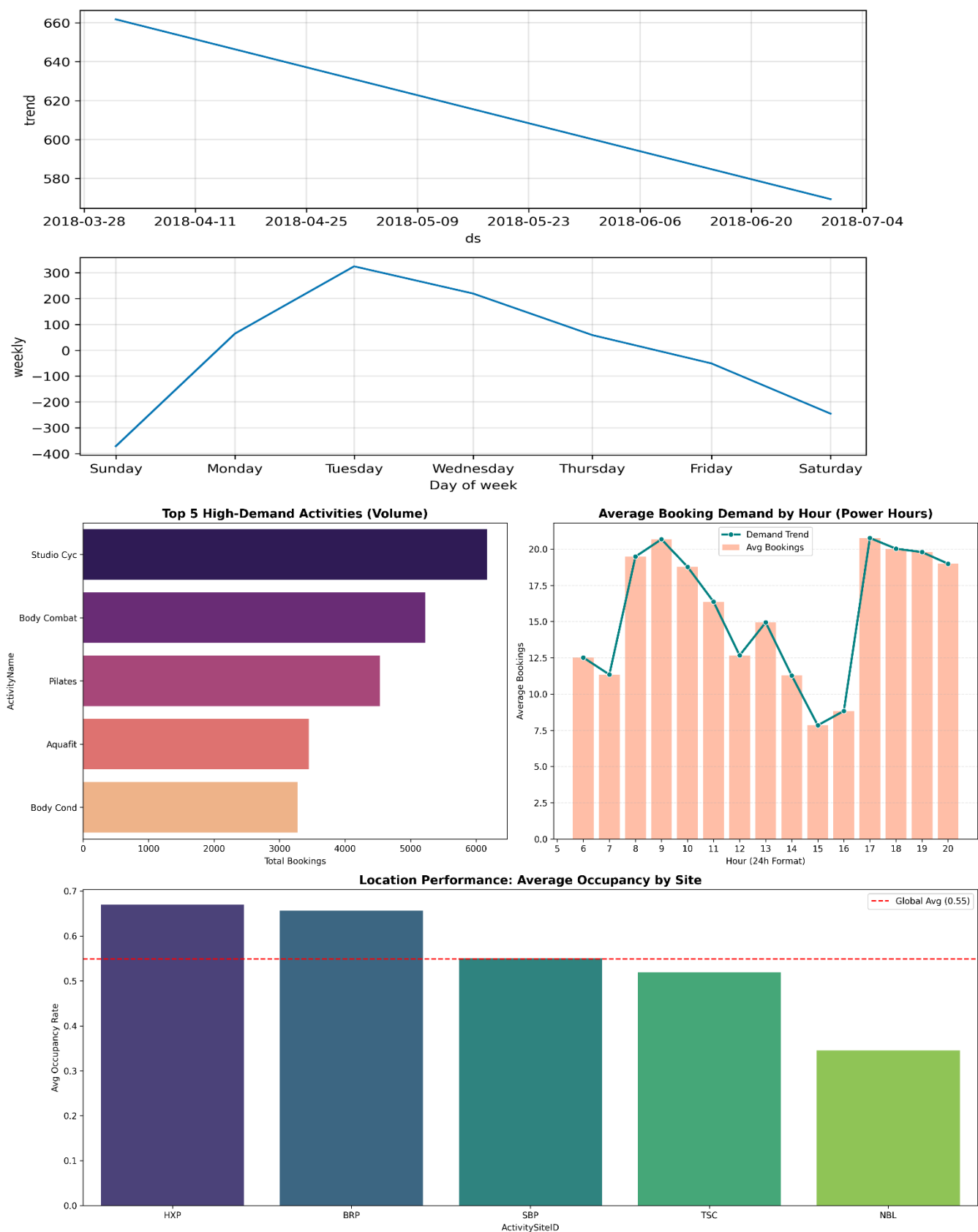


Figure A3: Seasonal decomposition of demand. The 'Daily' trend panel justifies the 'Power Hour' surge windows (06:00-09:00 and 17:00-20:00) used in the final pricing logic. Effect of Location and Activity decides final pricing

Evidence 4: Revenue Contribution by Activity and Location

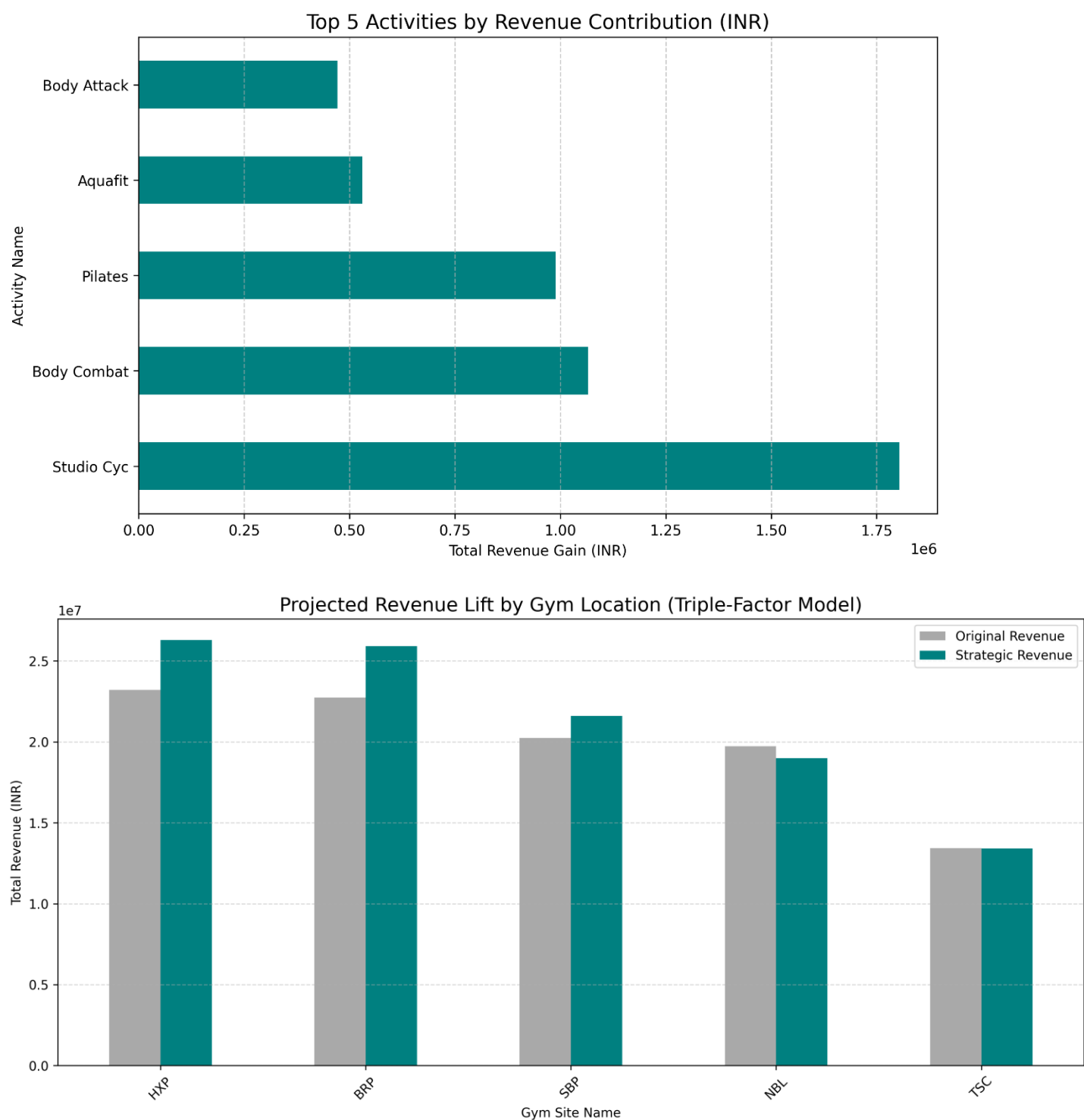


Figure A4:

Projected Revenue Gain by Activity type. This highlights that Studio Cyc, Body Bombat, and Hxp,BRP are the primary engines of the projected 6.91% total revenue lift.

Evidence 5: Demand Strategy Matrix

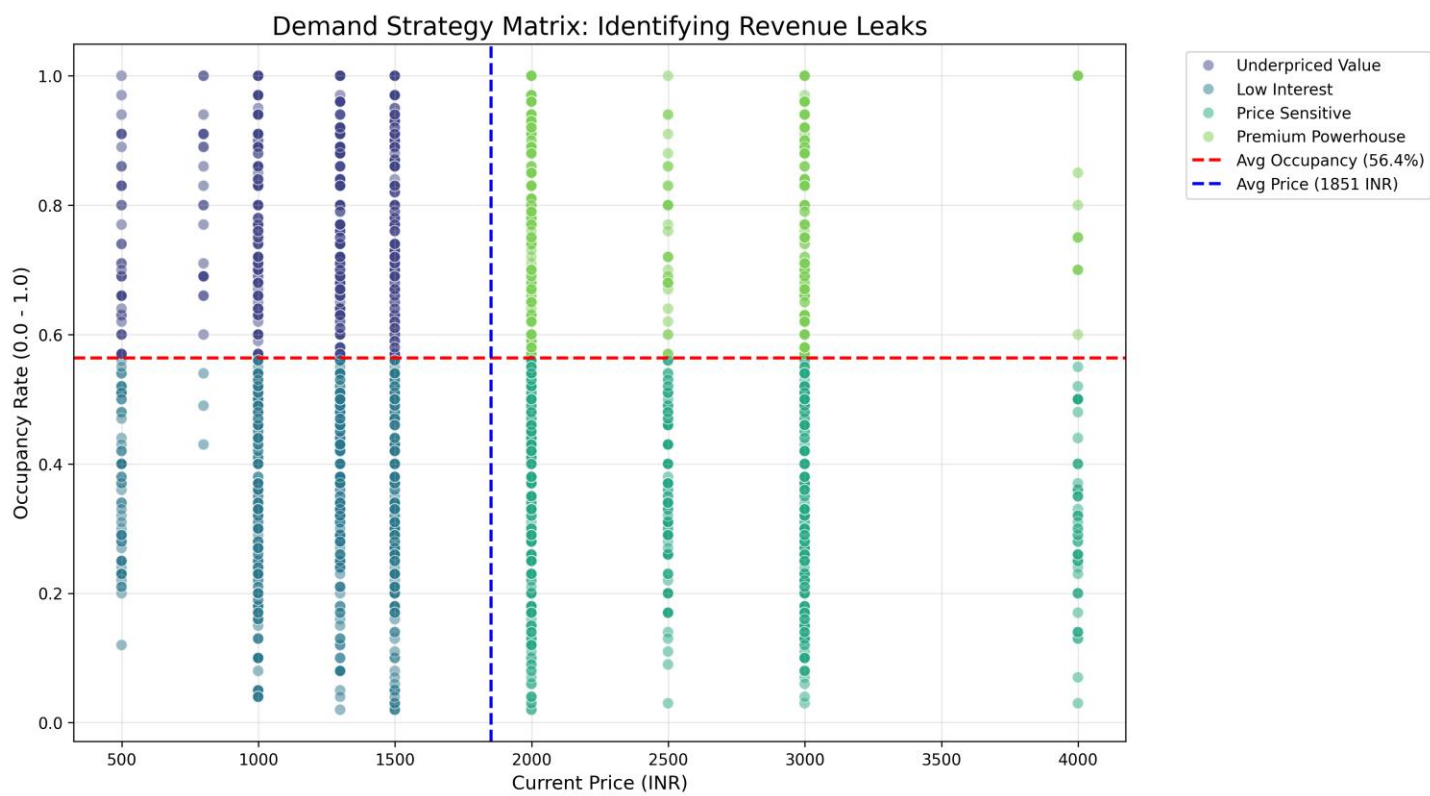


Figure A5:

This is used to identify Revenue Leaks

Evidence 6: Price Distribution: Fixed vs Triple Factor Strategic Pricing

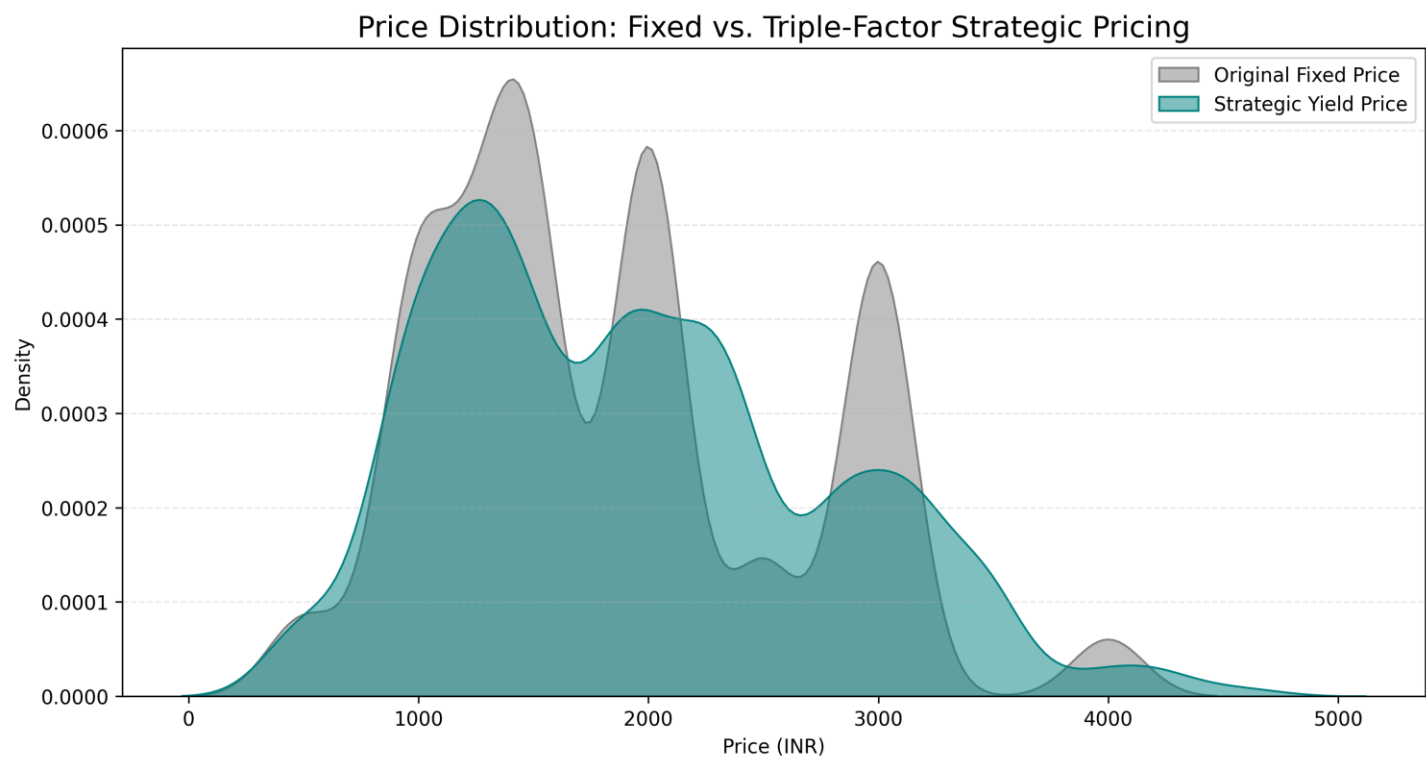


Figure A6:

Evidence 7: Statistical Summary Table

| OLS Regression Results | | | | | | |
|------------------------|------------------|---------|---------------------|----------|--------|--------|
| Dep. Variable: | log_bookings | | R-squared: | 0.035 | | |
| Model: | OLS | | Adj. R-squared: | 0.034 | | |
| Method: | Least Squares | | F-statistic: | 118.4 | | |
| Date: | Mon, 23 Feb 2026 | | Prob (F-statistic): | 4.03e-27 | | |
| Time: | 18:51:41 | | Log-Likelihood: | -3784.7 | | |
| No. Observations: | 3289 | | AIC: | 7573. | | |
| Df Residuals: | 3287 | | BIC: | 7586. | | |
| Df Model: | 1 | | | | | |
| Covariance Type: | nonrobust | | | | | |
| | coef | std err | t | P> t | [0.025 | 0.975] |
| const | 5.0198 | 0.221 | 22.691 | 0.000 | 4.586 | 5.454 |
| log_price | -0.3235 | 0.030 | -10.883 | 0.000 | -0.382 | -0.265 |
| Omnibus: | 742.512 | | Durbin-Watson: | 0.492 | | |
| Prob(Omnibus): | 0.000 | | Jarque-Bera (JB): | 1509.406 | | |
| Skew: | -1.331 | | Prob(JB): | 0.00 | | |
| Kurtosis: | 4.982 | | Cond. No. | 126. | | |

Table A1:

OLS Regression Summary. The P-value of 0.000 confirms that the impact of price on demand is highly significant and not due to random chance.