

# 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.

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## 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	<b>Maintenance:</b> Maintain current price or apply +5% "Premium Guard" surge.
Underpriced Value	High Occupancy / Low Price	<b>Surge:</b> Apply +15% Surge Multiplier. Target for immediate yield growth.
Price Sensitive	Low Occupancy / High Price	<b>Re-evaluation:</b> Monitor for 2 weeks; if occupancy remains <30%, apply -5% correction.
Low Interest	Low Occupancy / Low Price	<b>Incentivize:</b> 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:

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

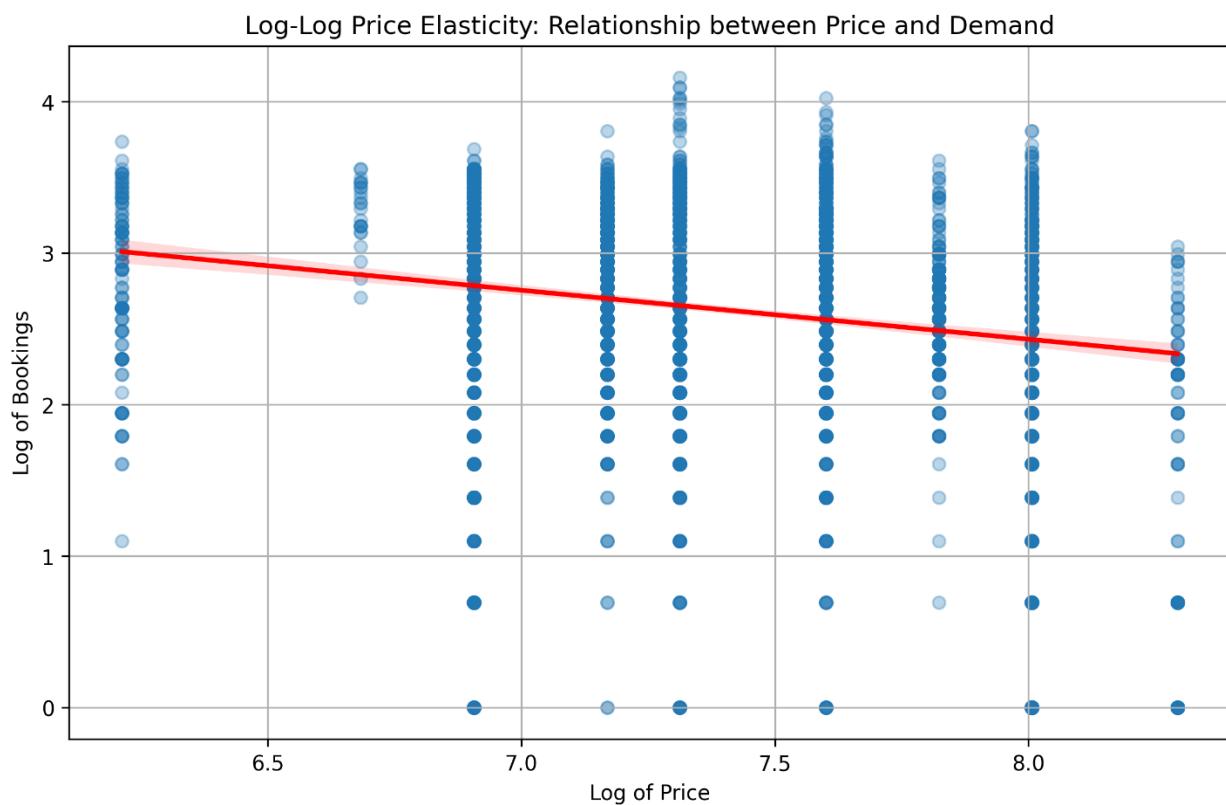
### 6. Verification and Performance Metrics

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

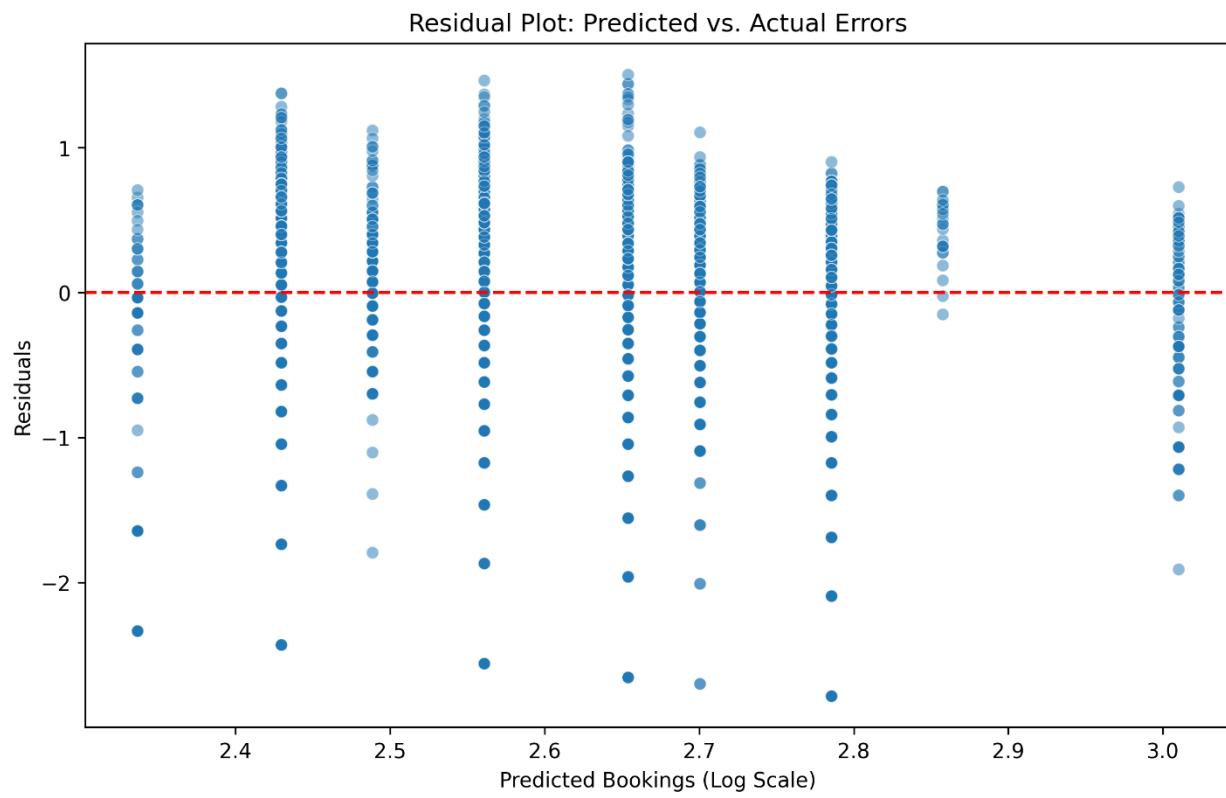
### Evidence 1: Price-Demand Sensitivity (Log-Log Regression)



**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.

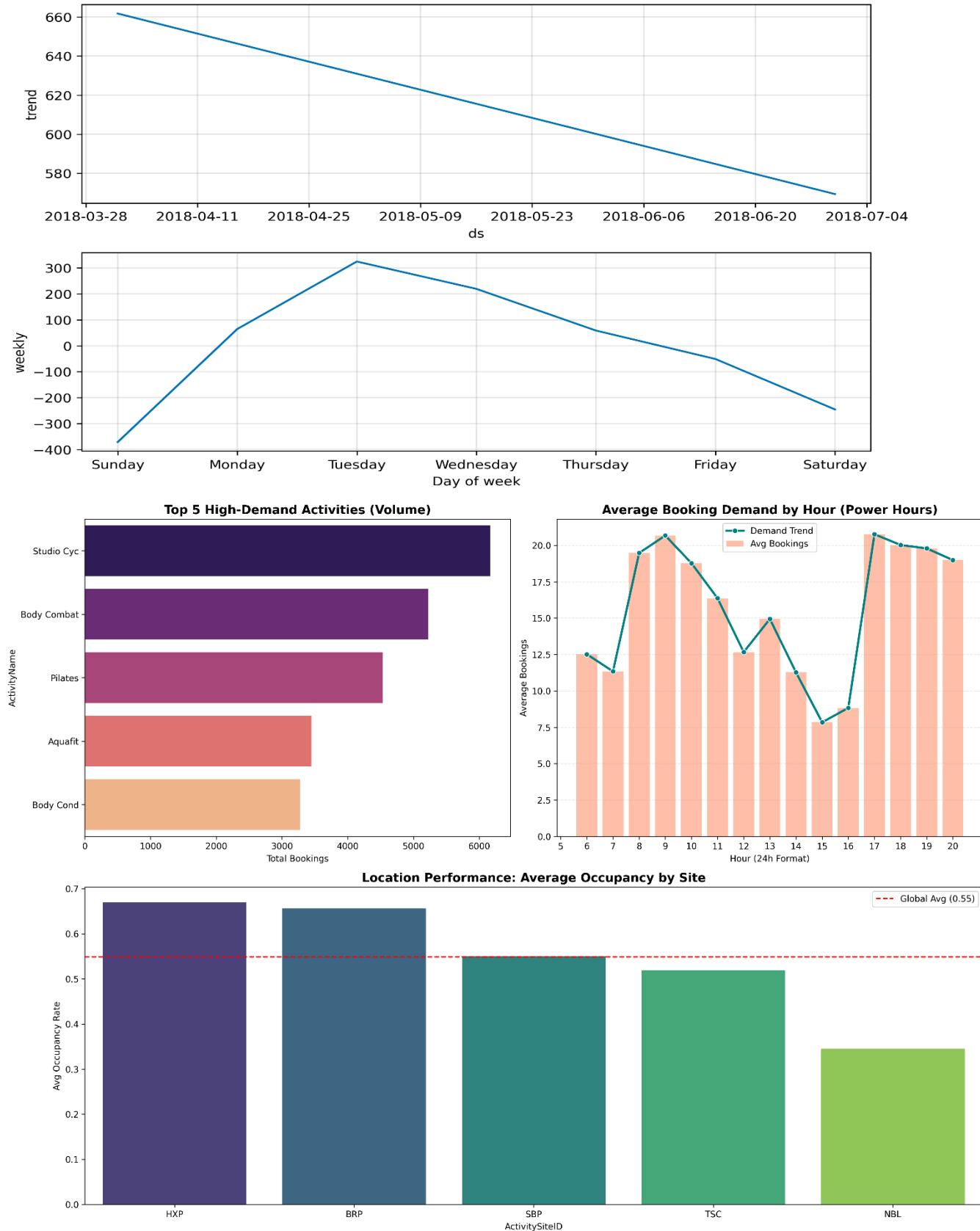
## Evidence 2: Model Diagnostic Check (Residual Plot)



**Figure A2:**

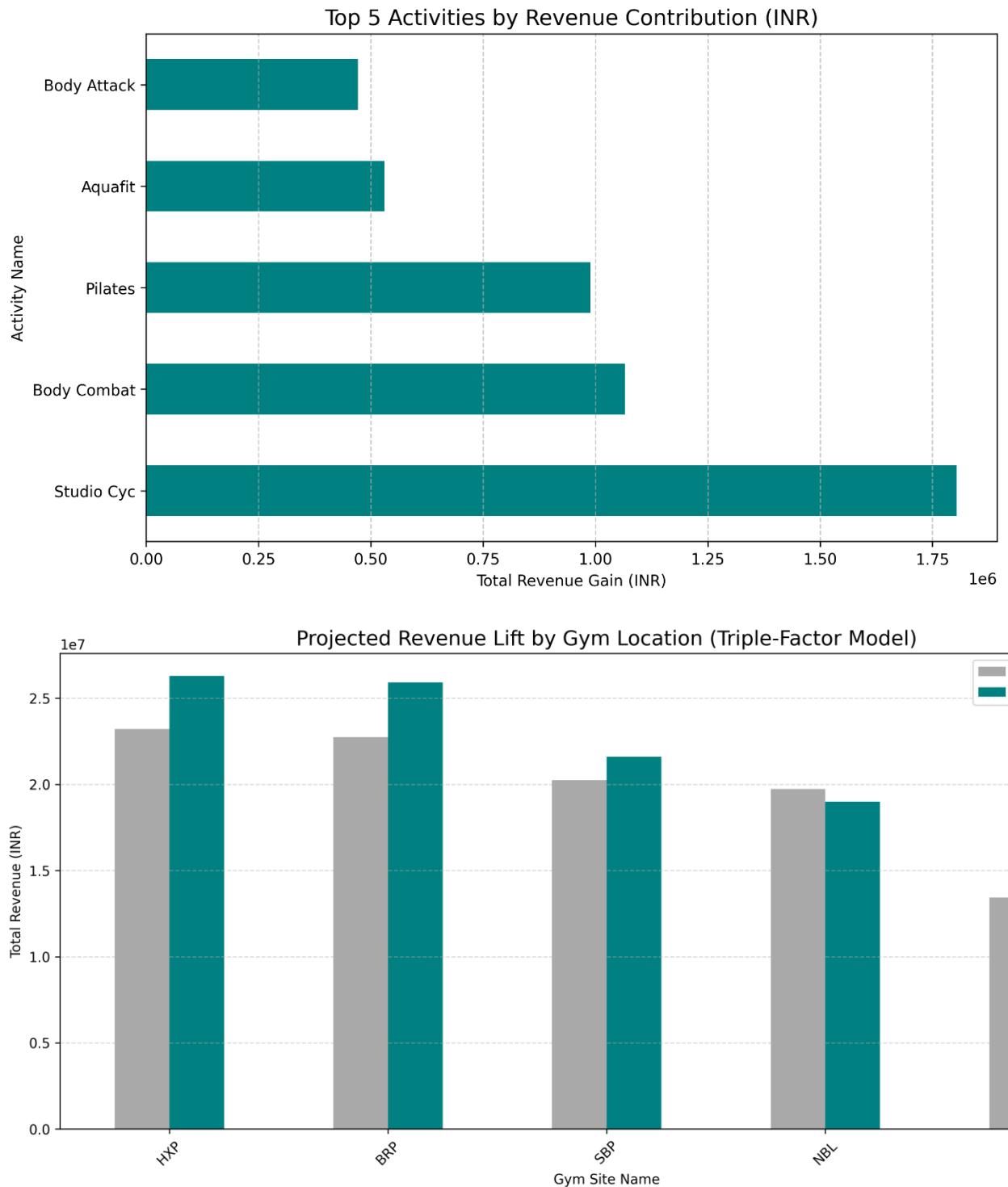
Residual Plot showing an unbiased distribution of errors around the zero-line, confirming the statistical validity of our elasticity coefficient

## 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

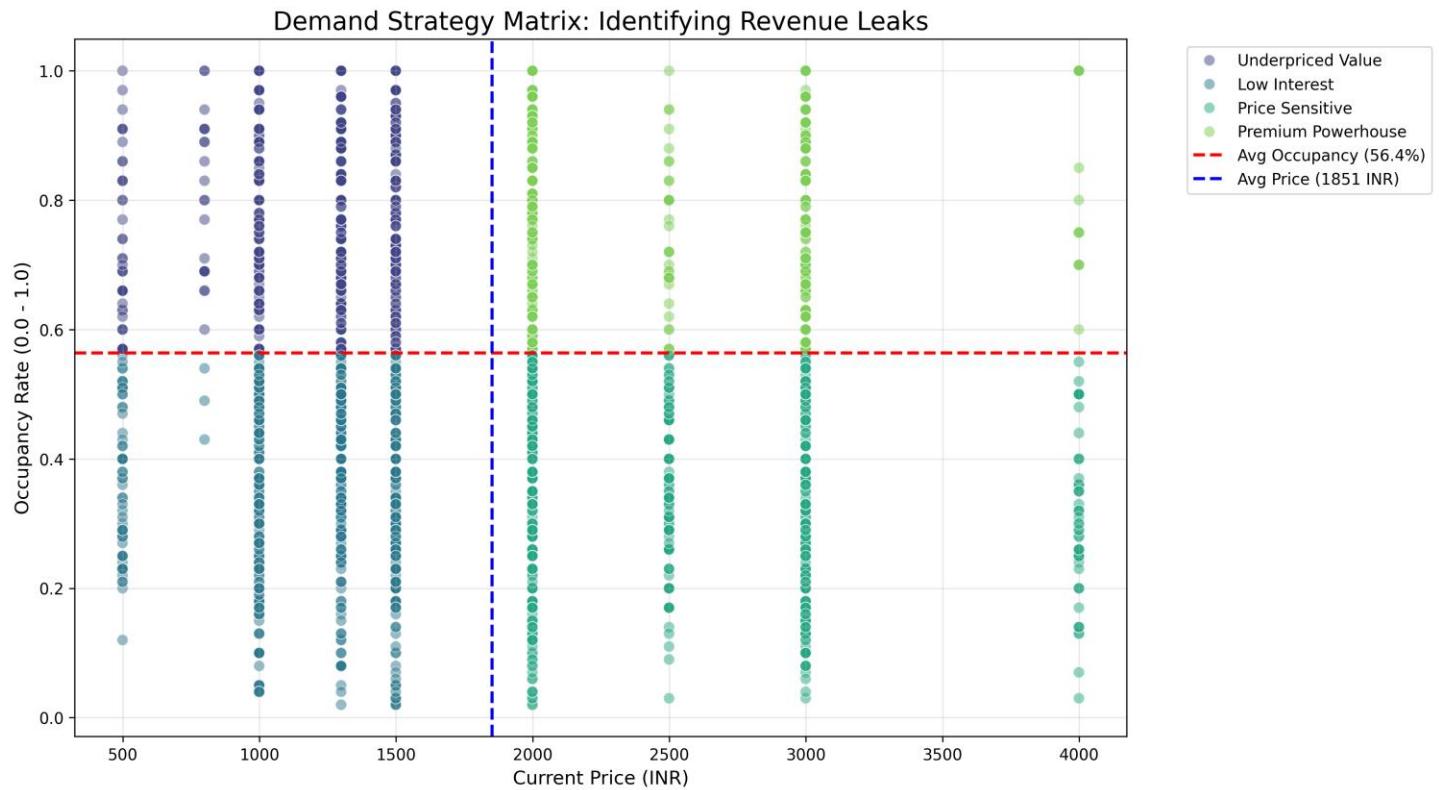
## Evidence 4: Revenue Contribution by Activity and Location



**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.

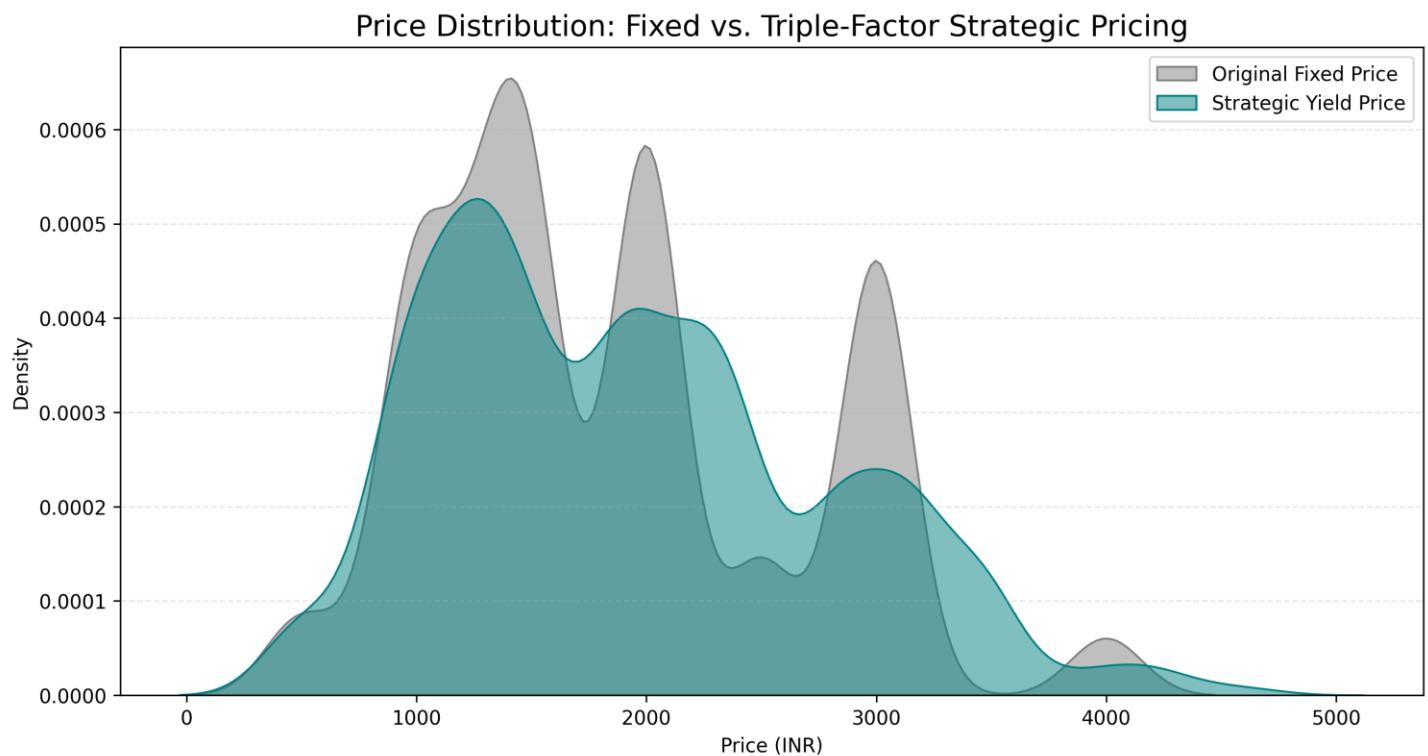
## Evidence 5: Demand Strategy Matrix



**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.