

Big Mountain Resort — Value-Based Pricing Optimization

1. Problem Statement

Big Mountain Resort required a **data-driven ticket pricing model** that reflects its facilities and market value. Competitor benchmarking limited revenue and undervalued its premium offerings.

Goal: Build a predictive model using machine learning to determine optimal pricing and test how operational investments (e.g., adding a new lift) affect profitability.

2. Data Preparation

- **Source:** 330 ski resorts (North America) with 27 attributes.
- **Cleaned:** 277 × 25 after removing outliers and imputing missing values.
- **Steps:**
 - Standardized state names and numeric entries.
 - Encoded categorical variables.
 - Validated consistency across units.
- **Target variable:** *AdultWeekend* (ticket price).

3. Exploratory Data Analysis (EDA)

Ticket price correlates strongly with **vertical drop**, **number of runs**, and **snow-making acreage**. State influence was mostly indirect through these features.

Figure 1: Correlation Heatmap of Resort Features



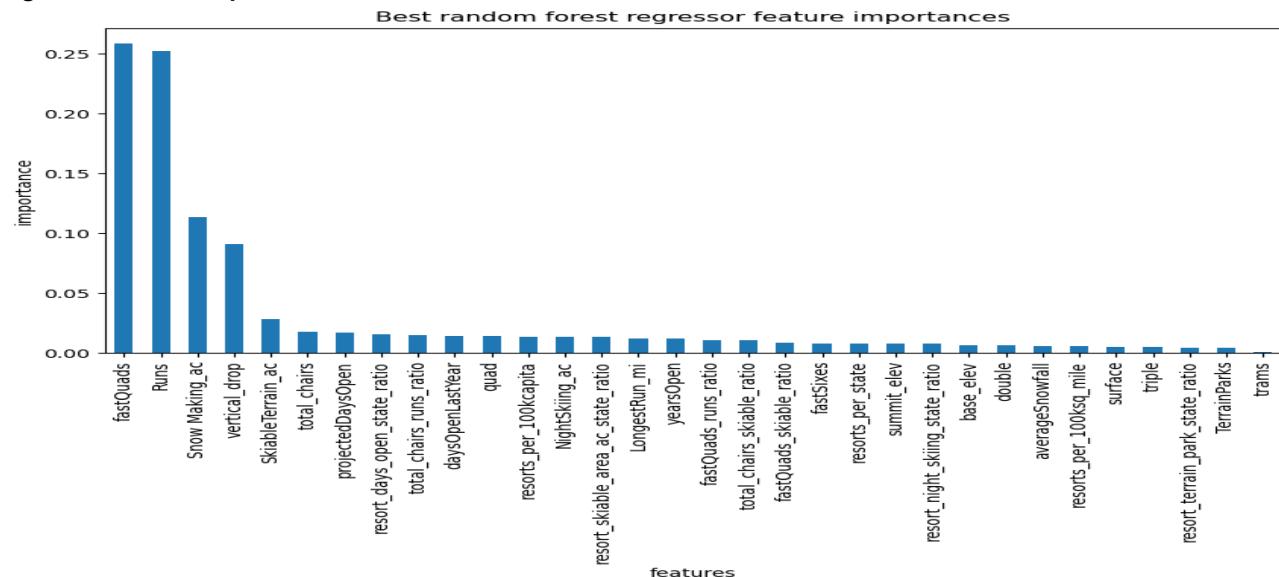
Shows strong relationships between price (*AdultWeekend*), vertical drop, runs, and snow-making capacity.

4. Model Development & Evaluation

Model	R ² (Train/Test)	MAE (\$)	Remarks
Dummy Regressor	0 / 0	19.1	Baseline
Linear Regression	0.82 / 0.72	9.4	Good linear fit
Random Forest Regressor	—	9.5	Most stable & robust

✓ **Selected Model:** Random Forest Regressor — lowest error, strong interpretability, and ideal for non-linear patterns.

Figure 2: Feature Importance Plot



Top drivers: fastQuads, Runs, Snow Making, Vertical Drop.

5. Scenario Modelling & Price Simulation

Predicted market-based price: $\approx \$95.74$, while current price = $\$81 \rightarrow \sim 15\%$ undervalued.

Scenario	Model Result	Business Implication
<input checked="" type="checkbox"/> Close runs	Decrease in supported price	Lowers perceived value
<input checked="" type="checkbox"/> Add run + new lift	$+\$1.99$ predicted price increase	Strong ROI impact
<input type="radio"/> Add snow area	Minimal change	Minimal effect
<input type="radio"/> Extend longest run	No measurable impact	Negligible impact

Adding a lift and run produces the greatest positive price effect.

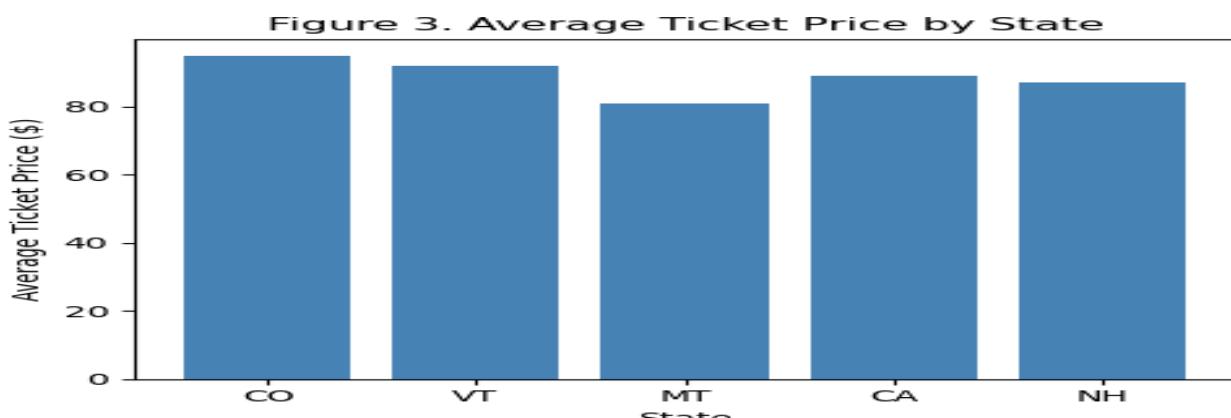
💰 **Revenue Projection:** 350K visitors $\times +\$2 \approx \$3.5M$ seasonal gain — offsets infrastructure costs.

6. Geographic Context

Regional averages indicate Big Mountain's under-pricing relative to competitors.

CO (\$95) > VT (\$92) > CA (\$89) > NH (\$87) > MT (\$81)

Figure 3: Average Ticket Price by State



Highlights Big Mountain's lower price position compared to similar resorts.

7. Pricing Recommendation

- Raise weekend ticket from **\$81 → \$92**, then **\$95** post-lift addition.
 - Emphasize **lift capacity** and **snow reliability** in marketing.
 - Introduce **dynamic pricing** during peak weekends.
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8. Conclusion & Future Scope

The **Random Forest model** predicts accurate, interpretable pricing outcomes ($R^2 \approx 0.72$). Big Mountain is undervalued by ~15%, but infrastructure investment offers high ROI.

Future Enhancements:

- Apply **dynamic pricing** tied to demand and weather.
 - Include **cost & profit** modeling for ROI optimization.
 - Build a **real-time dashboard** for price forecasting and scenario testing.
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