



# Guided Capstone Project

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# Problem Identification

**Context:**

Big Mountain Resort (Montana) with 350,000 visitors annually

The resort has new equipment installed: cost \$1,540,000

The company may not be fully capitalizing its facilities (tickets underpriced)

**Success Criteria:**

The company can make changes to facilities that will either:

- Cut costs by \$1,540,000, or
- Support even a higher ticket price (increase revenue at least \$1,540,000)

**Scope of Solution Space:**

Facilities (e.g., chairlifts, chairs, t-bars, run, elevation, terrain parks, area that uses snow making, etc.)

**Constraints within the Solution Space:**


Limited time to cut costs

**Stakeholders:**

- Director of Operations, Jimmy Blackburn
- Alesha Eisen, the Database Manager.

**Data Source:**

CSV file provided by Alesha Eisen



# Recommendation and Key Findings

Developed and deployed Machine Learning models.

- Random forest performed better than regression

RF models predicted and supported an increase in ticket prices.

Tested four scenarios.

**The recommended scenario:** adding a run, increasing the vertical drop by 150 feet, and installing an additional chair lift.

- Supports \$1.99 increase in ticket prices (current price \$81)
- Revenue increase of \$3,474,638



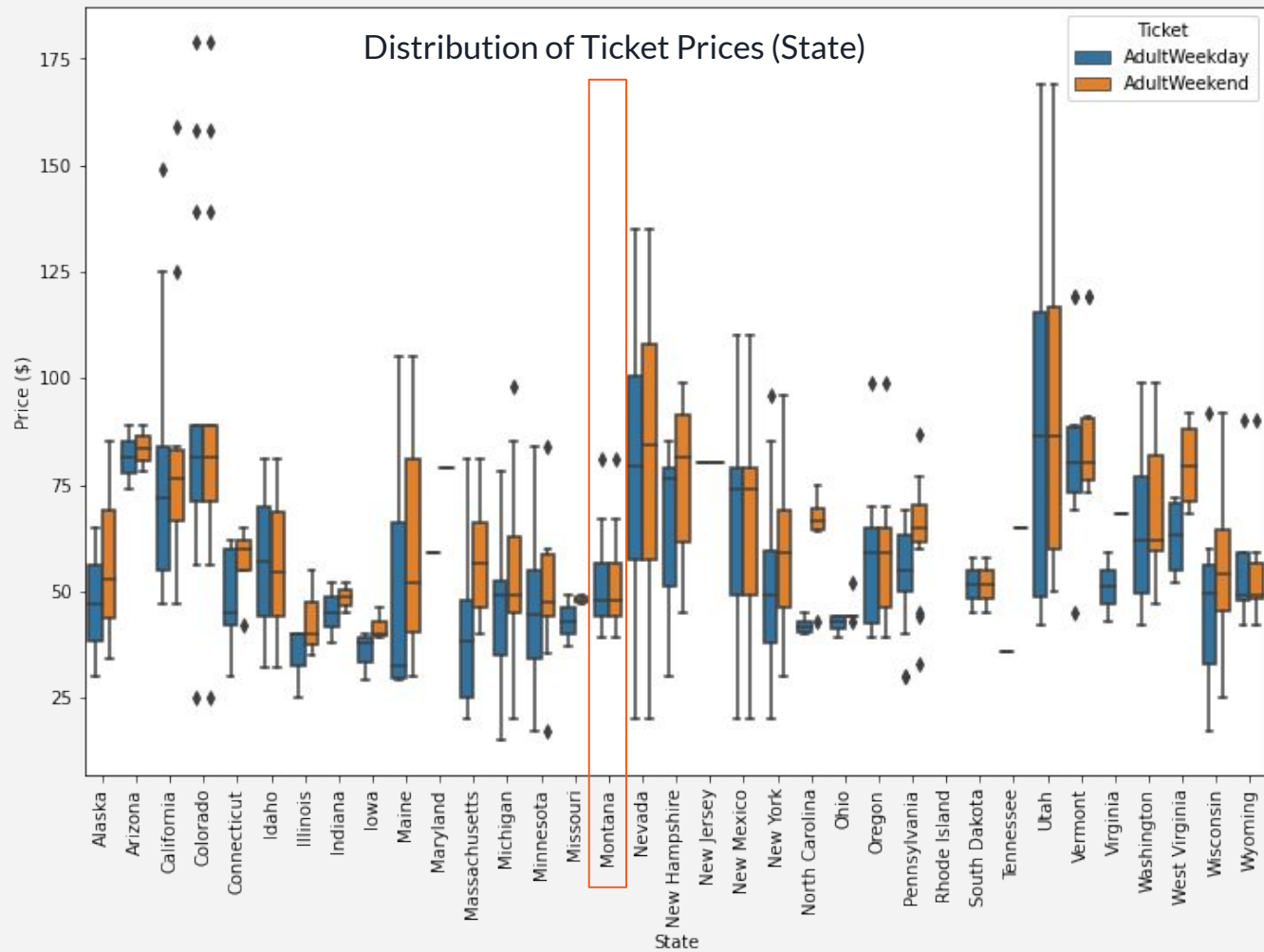
# Modeling Results and Analyses: Target Feature, Missing Values

Data: CSV file with 330 rows and 27 columns.

Target feature: Weekend ticket prices ('AdultWeekend')

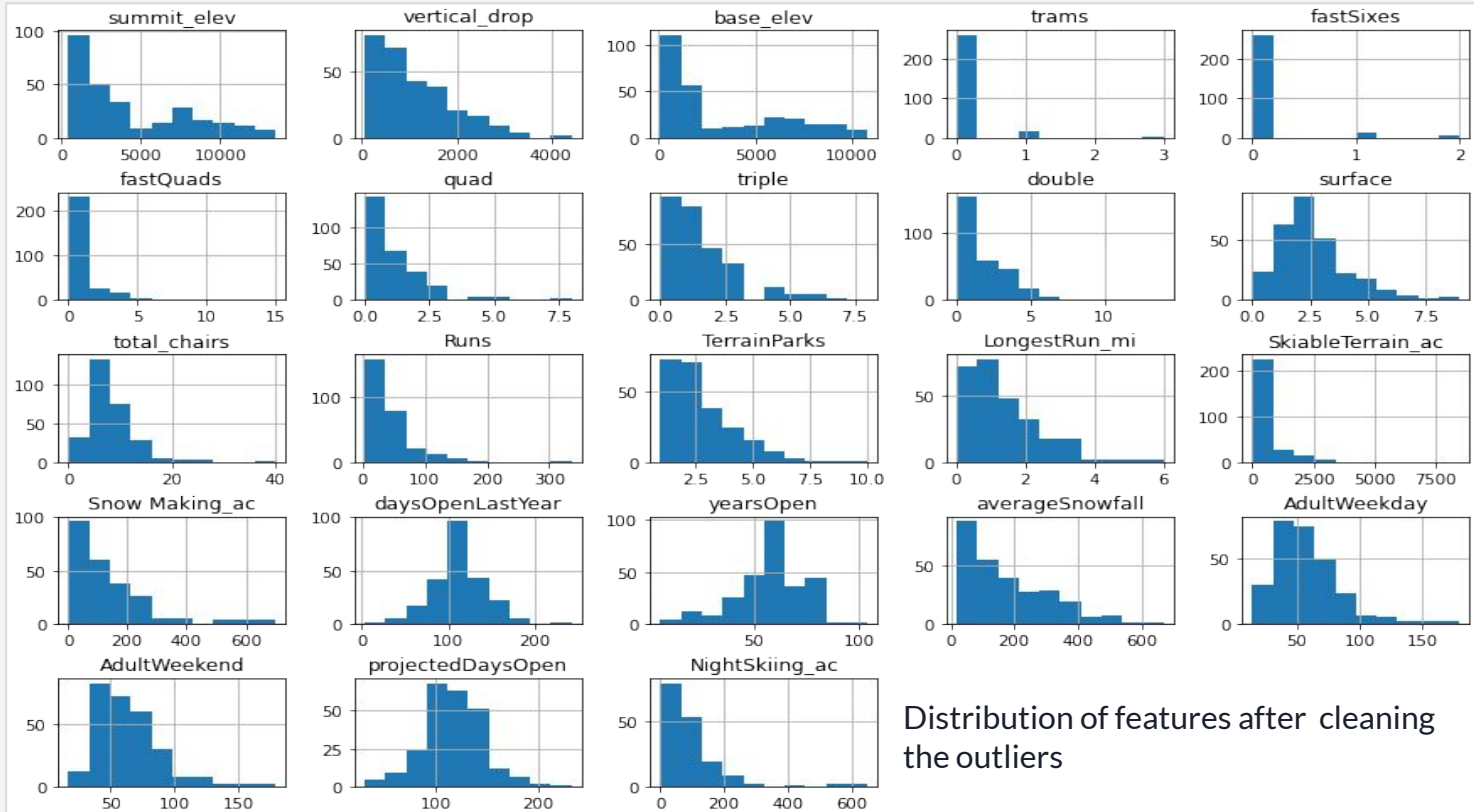
Missing Values: feature 'fastEight' dropped for 50% missing values

Missing Values: Missing values in some rows were multiples of 4 (e.g., 4%, 8%, 12%, 16%, 20%)



# Modeling Results and Analyses:

## Distribution of Features



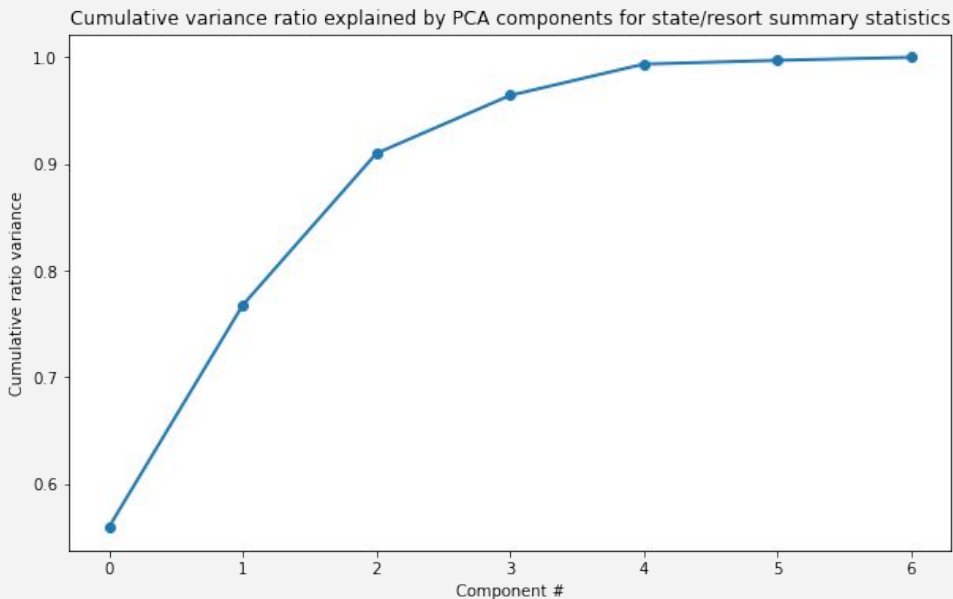
Distribution of features after cleaning the outliers


# Modeling Results and Analyses: State Data, PCA,

Created a new dataframe: state\_summary

- state population
- state area,
- number of resorts per state,
- total skiable area in the state,
- total number of days that resorts are open in the state,
- state total terrain parks, and state total night skiing area

Ran PCA: Two components explain 90% variation in the state\_summary





# Modeling Results and Analyses: ML Models

## ML Model 1: Linear Regression

### Model Metrics (MAE=Mean Absolute Error)

- MAE mean: 10.49
- MAE std: 1.62
- MAE for X\_test: 14.10

## ML Mode 2: Random Forest

### Model Metrics: (MAE=Mean Absolute Error)

- MAE mean: 9.58
- MAE std: 1.38
- MAE for X\_test: 9.5

Random Forest Model has better model metrics

- Lower MAE
- Lower Std Dev for MAE
- Lower MAE for X test





# Tested Four Scenarios using Random Forest ML Model

**Scenario 1:** Close up to 10 of the least used runs. The number of runs is the only parameter varying.

**Prediction:** Closing runs lead to drops in ticket prices (see Figure 5).

Best Scenario

**Scenario 2:** In this scenario, Big Mountain is adding a run, increasing the vertical drop by 150 feet, and installing an additional chair lift.

**Prediction:** This scenario increases support for ticket price by \$1.99 (revenue increase of \$3,474,638).

**Scenario 3:** Same as Scenario 2 but adding 2 acres of snowmaking.

**Prediction:** This scenario increases support for ticket price by \$1.99 (revenue increase of \$3,474,638).

**Scenario 4:** This scenario calls for increasing the longest run by .2 miles and guaranteeing its snow coverage by adding 4 acres of snow-making capability.

**Prediction:** No difference whatsoever.



# Summary and Conclusion

ML Models predicted that Scenario 2 is the best option

Scenario 2 supports increasing ticket prices by \$1.99

Requires

- adding a run,
- increasing the vertical drop by 150 feet,
- installing an additional chair lift.

Scenario 2 brings minimal increase in operational costs

Can generate an additional \$3,474,638 revenue annually