

# Big Mountain Resort

OpEx Cost Recovery Plan - July 2020

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# Problem Statement

How can Big Mountain Resort increase revenue to cover cost of new chair lift and maintain 9.2% profit margin in this financial year?

## Context

- ▶ The resort has recently installed a new chair lift to increase the distribution of visitors across the mountain, adding an additional \$1.54m to OPEX this year. The business has a profit margin of 9.2% and wants to maintain this going forward.

## Criteria for success

- ▶ Identify a way to increase revenue to level that allows company to maintain 9.2% business profit margin for the upcoming season

## Project Scope

- ▶ Understand if there is an opportunity for the resort to increase ticket prices by developing a model to predict optimal ticket price, and therefore drive additional revenue

## Constraints

- ▶ Data availability
- ▶ Ability to increase ticket price may not be viable for other commercial or operation reasons

# Recommendation

Use available data to model and predict Big Mountain price for upcoming year

## Steps:

### 1. Access and understand data

- ▶ Gain access to required data
- ▶ Explore data to understand trends and issues in data
- ▶ Adjust dataset to put in shape for analysis / modelling

### 2. Model data

- ▶ Model data with response variable of ticket price to better predict optimal pricing for Big Mountain Resort
- ▶ Iterate through models to identify best fit for data

### 3. Predict price

- ▶ Use best model to predict optimal price point for Big Mountain Resort going forward

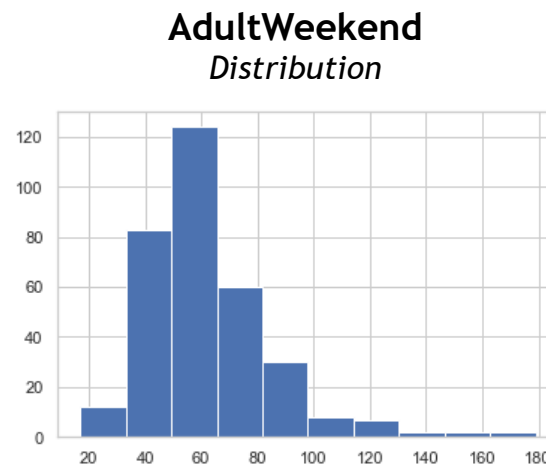
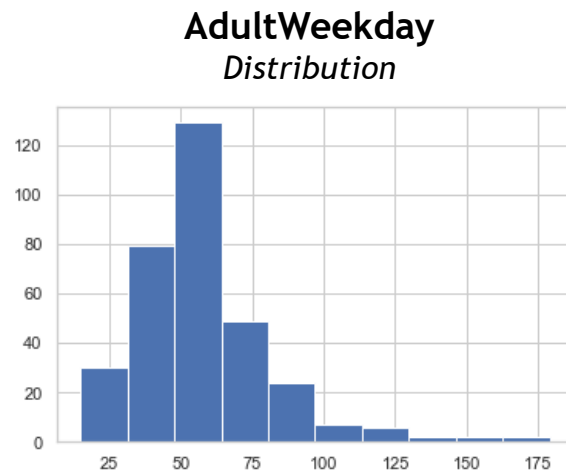
# Model - Dataset

## ► Data

- Acquired dataset with data on 330 resorts across the USA
- 27 variables (see table with mean value, if applicable)
- File updated to handle missing values
- File checked to remove duplicates

## ► Response Variable

- Focus on AdultWeekday and AdultWeekend as response variable to support problem statement



Variable Name	Mean
Name	n/a
Region	n/a
State	n/a
summit_elev	4591.82
vertical_drop	1215.43
base_elev	3374.00
trams	0.17
fastEight	0.01
fastSixes	0.18
fastQuads	1.09
quad	0.93
triple	1.50
double	1.83
surface	2.62
total_chairs	8.27
Runs	48.21
TerrainParks	2.82
LongestRun_mi	1.43
SkiableTerrain_ac	739.80
Snow Making_ac	174.87
daysOpenLastYear	115.10
yearsOpen	63.66
averageSnowfall	185.31
AdultWeekday	57.92
AdultWeekend	64.17
projectedDaysOpen	120.05
NightSkiing_ac	100.40

# Model - Exploratory Analysis

## ► Anomalies & Outliers

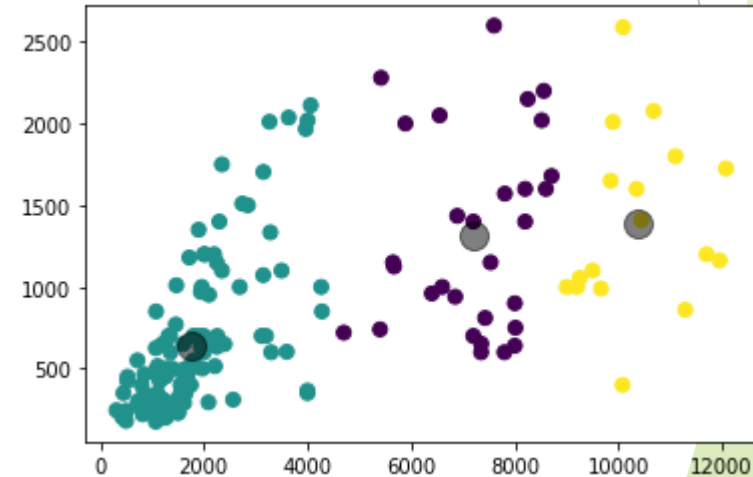
- Removed rows with values outside Inter-quartieral range (IQR) reducing data to 176 resorts

## ► Data Relationships

- Identify variables with high correlation ( $>0.95$ ) to remove from data set
- This included 'base\_elev' and 'sum\_chairs' that were dropped from data set

## ► Clusters

- Used k-means clustering technique to identify three clusters in dataset (see beside)



# Model – Modelling

- ▶ Three models performed
  - ▶ All included splitting data into training and test data (75/25)
  - ▶ Linear regression models
  - ▶ Performance of each model in table below
- ▶ **1<sup>st</sup> Model (including all features)**
  - ▶ Strong fit with 0.928 explained variance
  - ▶ States had the top 10 strongest coefficients of all features in this model
  - ▶ Provides little value to support action for Big Mountain Resort so removed from model
- ▶ **2<sup>nd</sup> Model (without ‘state’)**
  - ▶ Strong fit with 0.922 explained variance
  - ▶ Summit elevation and base elevation were in Top 3 strongest coefficients of all features in this model
  - ▶ Provides little value to support action for Big Mountain Resort so removed from model
- ▶ **3<sup>rd</sup> Model (without states or elevations)**
  - ▶ Strong fit with 0.924 explained variance
  - ▶ Provides model similar to previous but with different top 10 coefficients

Model	Explained Variance	Mean Absolute Error	Features Dropped
Model 1.	0.928	5.35	-
Model 2.	0.922	5.54	'state'
Model 3.	0.924	5.53	'state','summit_elev','base_elev'

# Model - Result

- ▶ **3<sup>rd</sup> Model selected as best model**
  - ▶ Maintained high explained variance and similar mean absolute error compared to alternative models
  - ▶ Contains more relevant and actionable features than alternative model (see coefficient model on right)
- ▶ **Big Mountain Resort Prediction**
  - ▶ Ran model on Big Mountain Resort data to allow it to predict suggested pricing based on features included in model
  - ▶ **Model suggested price of \$88.77 for Adult Weekend tickets**
  - ▶ This is +\$7.77 higher than existing price of \$81

Model 3 Coefficients

Feature	Coefficient
AdultWeekday	19.893397
averageSnowfall	1.982708
Runs	1.665804
quad	1.578662
triple	1.380822
vertical_drop	1.291189
surface	1.266572
daysOpenLastYear	0.829641
fastQuads	0.771395
clusters	0.766623



# Summary

## Objective

- ▶ Identify a way to increase revenue to level that allows company to maintain 9.2% business profit margin for the upcoming season
- ▶ Understand if there is an opportunity for the resort to increase ticket prices by developing a model to predict optimal ticket price, and therefore drive additional revenue

## Outcome

- ▶ Model suggests that Big Mountain Resort could increase its Weekend Ticket price from current \$81 to \$88.77

## Next Steps

- ▶ Develop forecast to understand if price increase will cover Opex cost
- ▶ Run similar model on AdultWeekday to understand if pricing opportunity exist
- ▶ Understand if any other commercial, operational or guest impacts to consider before implementing price changes