

data_wrangling

June 22, 2021

```
[ ]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import os
```

0.0.1 Introduction

In this project, we aim to find a more efficient pricing model for Big Mountain Resort using a dataset with information on 330 different ski resorts across the United States.

Big Mountain Resort, a ski resort located in Montana, has recently installed an additional chair lift, which increases their operating costs by \$1,540,000 this season. Big Mountain Resort are looking for guidance on how to select a better value for their ticket price (currently charged based only on market average.)

To find a better pricing strategy for Big Mountain Resort, we will apply the techniques of Linear Regression and Random Forest to make prediction of prices using selected features, and evaluate each model using cross validation.

```
[ ]: ski_data = pd.read_csv('ski_resort_data.csv')
```

```
[ ]: ski_data.info
```

```
[ ]: <bound method DataFrame.info of
state  summit_elev \
0          Alyeska Resort    Alaska    Alaska    3939
1      Eaglecrest Ski Area    Alaska    Alaska    2600
2      Hilltop Ski Area      Alaska    Alaska    2090
3      Arizona Snowbowl     Arizona    Arizona    11500
4      Sunrise Park Resort   Arizona    Arizona    11100
..          ...            ...      ...      ...
325      Meadowlark Ski Lodge  Wyoming    Wyoming    9500
326      Sleeping Giant Ski Resort  Wyoming    Wyoming    7428
327      Snow King Resort      Wyoming    Wyoming    7808
328  Snowy Range Ski & Recreation Area  Wyoming    Wyoming    9663
329      White Pine Ski Area    Wyoming    Wyoming    9500

vertical_drop  base_elev  trams  fastEight  fastSixes  fastQuads  ... \
```

0	2500	250	1	0.0	0	2 ...
1	1540	1200	0	0.0	0	0 ...
2	294	1796	0	0.0	0	0 ...
3	2300	9200	0	0.0	1	0 ...
4	1800	9200	0	NaN	0	1 ...
..
325	1000	8500	0	NaN	0	0 ...
326	810	6619	0	0.0	0	0 ...
327	1571	6237	0	NaN	0	0 ...
328	990	8798	0	0.0	0	0 ...
329	1100	8400	0	NaN	0	0 ...

	LongestRun_mi	SkiableTerrain_ac	Snow Making_ac	daysOpenLastYear	\
0	1.0	1610.0	113.0	150.0	
1	2.0	640.0	60.0	45.0	
2	1.0	30.0	30.0	150.0	
3	2.0	777.0	104.0	122.0	
4	1.2	800.0	80.0	115.0	
..	
325	1.5	300.0	NaN	NaN	
326	1.0	184.0	18.0	61.0	
327	1.0	400.0	250.0	121.0	
328	0.7	75.0	30.0	131.0	
329	0.4	370.0	NaN	NaN	

	yearsOpen	averageSnowfall	AdultWeekday	AdultWeekend	\
0	60.0	669.0	65.0	85.0	
1	44.0	350.0	47.0	53.0	
2	36.0	69.0	30.0	34.0	
3	81.0	260.0	89.0	89.0	
4	49.0	250.0	74.0	78.0	
..	
325	9.0	NaN	NaN	NaN	
326	81.0	310.0	42.0	42.0	
327	80.0	300.0	59.0	59.0	
328	59.0	250.0	49.0	49.0	
329	81.0	150.0	NaN	49.0	

	projectedDaysOpen	NightSkiing_ac
0	150.0	550.0
1	90.0	NaN
2	152.0	30.0
3	122.0	NaN
4	104.0	80.0
..
325	NaN	NaN
326	77.0	NaN

```

327          123.0          110.0
328          NaN          NaN
329          NaN          NaN

```

```
[330 rows x 27 columns]>
```

‘AdultWeekday’ is the price of an adult weekday ticket. ‘AdultWeekend’ is the price of an adult weekend ticket. They are the target of our project. The other columns are potential features that could be used to fit our model and predict outcomes.

```
[ ]: ski_data.head()
```

```

[ ]:
      Name      Region  state  summit_elev  vertical_drop  \
0  Alyeska Resort   Alaska  Alaska        3939          2500
1  Eaglecrest Ski Area  Alaska  Alaska        2600          1540
2  Hilltop Ski Area   Alaska  Alaska        2090           294
3  Arizona Snowbowl  Arizona  Arizona       11500          2300
4  Sunrise Park Resort  Arizona  Arizona       11100          1800

      base_elev  trams  fastEight  fastSixes  fastQuads  ...  LongestRun_mi  \
0          250      1         0.0          0          2  ...           1.0
1         1200      0         0.0          0          0  ...           2.0
2         1796      0         0.0          0          0  ...           1.0
3         9200      0         0.0          1          0  ...           2.0
4         9200      0         NaN          0          1  ...           1.2

      SkiableTerrain_ac  Snow Making_ac  daysOpenLastYear  yearsOpen  \
0          1610.0          113.0          150.0          60.0
1           640.0           60.0           45.0          44.0
2           30.0           30.0          150.0          36.0
3           777.0          104.0          122.0          81.0
4           800.0           80.0          115.0          49.0

      averageSnowfall  AdultWeekday  AdultWeekend  projectedDaysOpen  \
0          669.0          65.0          85.0          150.0
1          350.0          47.0          53.0           90.0
2           69.0          30.0          34.0          152.0
3          260.0          89.0          89.0          122.0
4          250.0          74.0          78.0          104.0

      NightSkiing_ac
0          550.0
1           NaN
2           30.0
3           NaN
4           80.0

```

[5 rows x 27 columns]

Information on Big Mountain Resort

```
[ ]: ski_data[ski_data.Name == 'Big Mountain Resort'].T
```

```
[ ]:
```

Name	Big Mountain Resort
Region	Montana
state	Montana
summit_elev	6817
vertical_drop	2353
base_elev	4464
trams	0
fastEight	0.0
fastSixes	0
fastQuads	3
quad	2
triple	6
double	0
surface	3
total_chairs	14
Runs	105.0
TerrainParks	4.0
LongestRun_mi	3.3
SkiableTerrain_ac	3000.0
Snow Making_ac	600.0
daysOpenLastYear	123.0
yearsOpen	72.0
averageSnowfall	333.0
AdultWeekday	81.0
AdultWeekend	81.0
projectedDaysOpen	123.0
NightSkiing_ac	600.0

```
[ ]: missing = pd.concat([ski_data.isnull().sum(), 100 * ski_data.isnull().mean()],  
    ↪axis=1)  
missing.columns=['count', '%']  
missing.sort_values(by='count', ascending=False)
```

```
[ ]:
```

	count	%
fastEight	166	50.303030
NightSkiing_ac	143	43.333333
AdultWeekday	54	16.363636
AdultWeekend	51	15.454545
daysOpenLastYear	51	15.454545
TerrainParks	51	15.454545

projectedDaysOpen	47	14.242424
Snow Making_ac	46	13.939394
averageSnowfall	14	4.242424
LongestRun_mi	5	1.515152
Runs	4	1.212121
SkiableTerrain_ac	3	0.909091
yearsOpen	1	0.303030
total_chairs	0	0.000000
Name	0	0.000000
Region	0	0.000000
double	0	0.000000
triple	0	0.000000
quad	0	0.000000
fastQuads	0	0.000000
fastSixes	0	0.000000
trams	0	0.000000
base_elev	0	0.000000
vertical_drop	0	0.000000
summit_elev	0	0.000000
state	0	0.000000
surface	0	0.000000

```
[ ]: ski_data.select_dtypes('object')
```

```
[ ]:
```

	Name	Region	state
0	Alyeska Resort	Alaska	Alaska
1	Eaglecrest Ski Area	Alaska	Alaska
2	Hilltop Ski Area	Alaska	Alaska
3	Arizona Snowbowl	Arizona	Arizona
4	Sunrise Park Resort	Arizona	Arizona
..
325	Meadowlark Ski Lodge	Wyoming	Wyoming
326	Sleeping Giant Ski Resort	Wyoming	Wyoming
327	Snow King Resort	Wyoming	Wyoming
328	Snowy Range Ski & Recreation Area	Wyoming	Wyoming
329	White Pine Ski Area	Wyoming	Wyoming

[330 rows x 3 columns]

```
[ ]: ski_data['Name'].value_counts().head()
```

```
[ ]:
```

Crystal Mountain	2
Mount Bohemia	1
Anthony Lakes Mountain Resort	1
Hunt Hollow Ski Club	1
Ski Granby Ranch	1

Name: Name, dtype: int64

There is a duplicated resort name: Crystal Mountain.

```
[ ]: (ski_data['Name'] + ', ' + ski_data['Region']).value_counts().head()
```

```
[ ]: Ski Apache, New Mexico          1
      Mount Pleasant of Edinboro, Pennsylvania  1
      Bolton Valley, Vermont          1
      Mulligan's Hollow Ski Bowl, Michigan  1
      Mt. Bachelor, Oregon            1
      dtype: int64
```

```
[ ]: (ski_data['Name'] + ', ' + ski_data['state']).value_counts().head()
```

```
[ ]: Kirkwood, California            1
      Mulligan's Hollow Ski Bowl, Michigan  1
      Cranmore Mountain Resort, New Hampshire  1
      Sundown Mountain, Iowa            1
      Marquette Mountain, Michigan      1
      dtype: int64
```

```
[ ]: ski_data[ski_data['Name'] == 'Crystal Mountain']
```

```
[ ]:
      Name      Region      state  summit_elev  vertical_drop \
104  Crystal Mountain  Michigan  Michigan      1132          375
295  Crystal Mountain  Washington Washington      7012         3100

      base_elev  trams  fastEight  fastSixes  fastQuads  ...  LongestRun_mi \
104         757      0         0.0         0         1  ...          0.3
295        4400      1         NaN         2         2  ...          2.5

      SkiableTerrain_ac  Snow Making_ac  daysOpenLastYear  yearsOpen \
104             102.0         96.0         120.0         63.0
295             2600.0         10.0          NaN         57.0

      averageSnowfall  AdultWeekday  AdultWeekend  projectedDaysOpen \
104             132.0         54.0         64.0         135.0
295             486.0         99.0         99.0          NaN

      NightSkiing_ac
104             56.0
295             NaN

[2 rows x 27 columns]
```

There are two different Crystal Mountain Resort

```
[ ]: (ski_data.Region != ski_data.state).count()
```

```
[ ]: 330
```

```
[ ]: ski_data['Region'].value_counts()
```

```
[ ]: New York          33
      Michigan         29
      Colorado         22
      Sierra Nevada    22
      Pennsylvania     19
      New Hampshire    16
      Wisconsin        16
      Vermont          15
      Minnesota        14
      Montana          12
      Idaho            12
      Massachusetts    11
      Washington       10
      Maine            9
      New Mexico       9
      Wyoming          8
      Utah             7
      Salt Lake City   6
      North Carolina   6
      Oregon           6
      Connecticut      5
      Ohio             5
      Illinois         4
      Virginia         4
      West Virginia    4
      Mt. Hood         4
      Alaska           3
      Iowa             3
      Nevada           2
      South Dakota     2
      Arizona          2
      Indiana          2
      Missouri         2
      New Jersey       2
      Tennessee        1
      Maryland         1
      Northern California 1
      Rhode Island     1
      Name: Region, dtype: int64
```

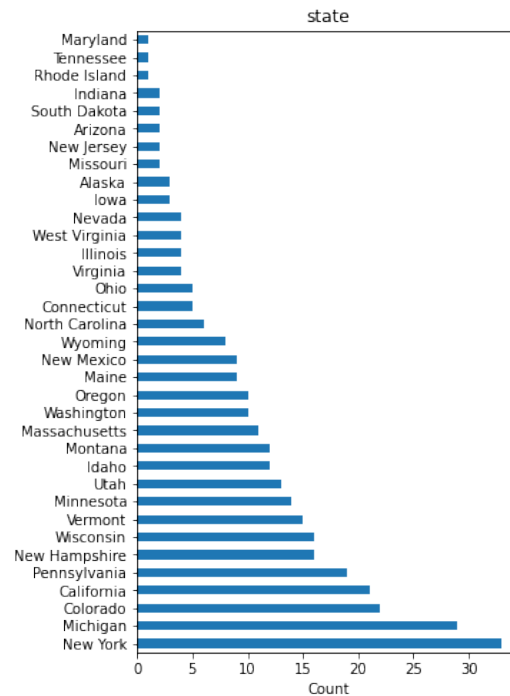
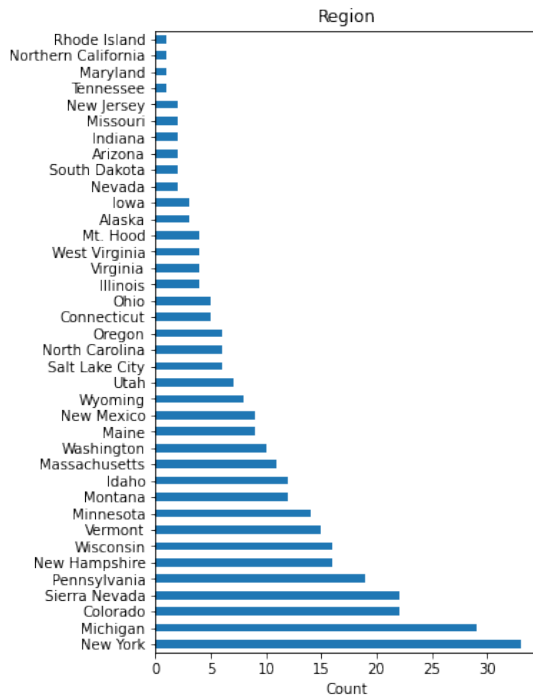
```
[ ]: (ski_data[ski_data.Region != ski_data.state]
      .groupby('state')['Region']
      .value_counts())
```

```
[ ]: state      Region
      California Sierra Nevada      20
           Northern California      1
      Nevada     Sierra Nevada      2
      Oregon     Mt. Hood           4
      Utah       Salt Lake City      6
      Name: Region, dtype: int64
```

```
[ ]: ski_data[['Region', 'state']].nunique()
```

```
[ ]: Region      38
      state       35
      dtype: int64
```

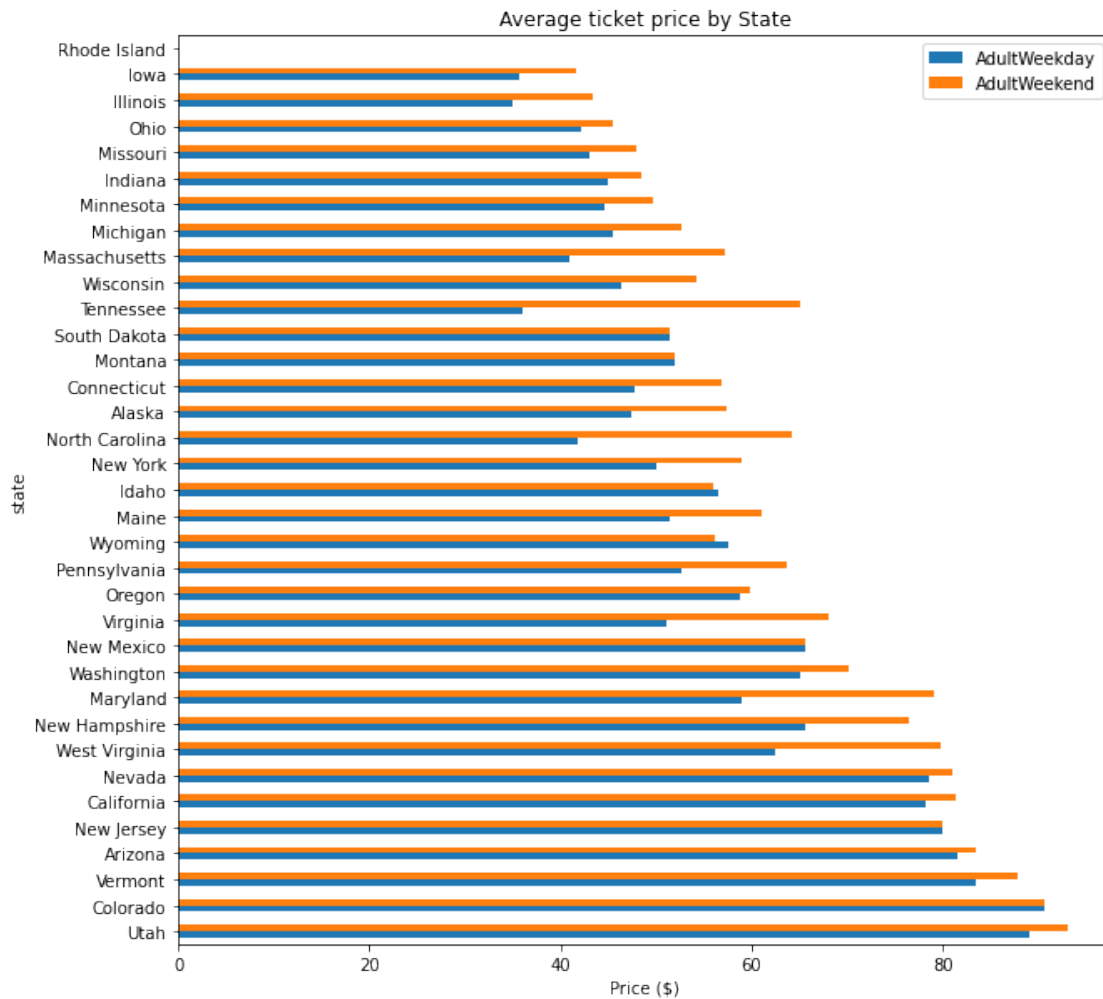
```
[ ]: fig, ax = plt.subplots(nrows=1, ncols=2, figsize=(12, 8))
      #Specify a horizontal barplot ('barh') as kind of plot (kind=)
      ski_data.Region.value_counts().plot(kind='barh', ax=ax[0])
      #Give the plot a helpful title of 'Region'
      ax[0].set_title('Region')
      #Label the xaxis 'Count'
      ax[0].set_xlabel('Count')
      #Specify a horizontal barplot ('barh') as kind of plot (kind=)
      ski_data.state.value_counts().plot(kind='barh', ax=ax[1])
      #Give the plot a helpful title of 'state'
      ax[1].set_title('state')
      #Label the xaxis 'Count'
      ax[1].set_xlabel('Count')
      #Give the subplots a little "breathing room" with a wspace of 0.5
      plt.subplots_adjust(wspace=0.5);
```

```
[ ]: state_price_means = ski_data.groupby('state')[['AdultWeekday', 'AdultWeekend']].
      ↪mean()
state_price_means.head()
```

```
[ ]:
state      AdultWeekday  AdultWeekend
Alaska      47.333333      57.333333
Arizona      81.500000      83.500000
California    78.214286      81.416667
Colorado     90.714286      90.714286
Connecticut   47.800000      56.800000
```

```
[ ]: (state_price_means.reindex(index=state_price_means.mean(axis=1)
      .sort_values(ascending=False)
      .index)
      .plot(kind='barh', figsize=(10, 10), title='Average ticket price by State'))
plt.xlabel('Price ($)');
```

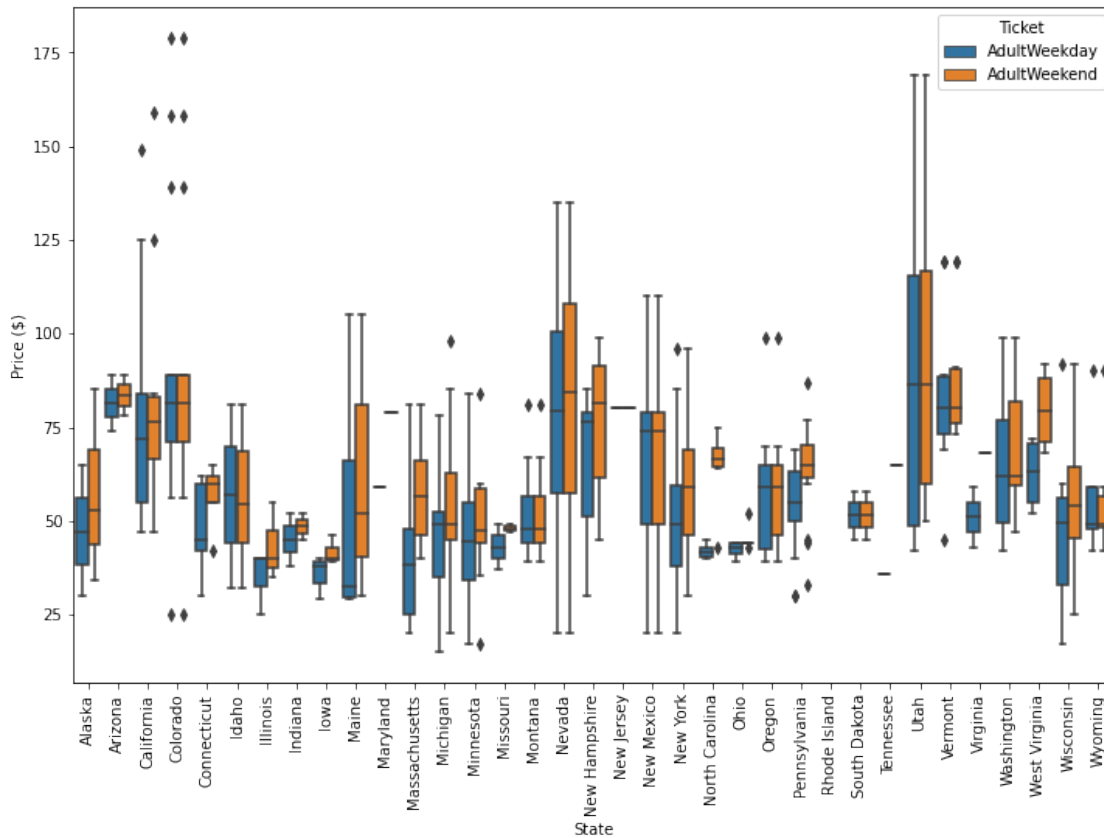


```
[ ]: ticket_prices = pd.melt(ski_data[['state', 'AdultWeekday', 'AdultWeekend']],
                             id_vars=['state'],
                             var_name= 'Ticket',
                             value_vars=['AdultWeekday', 'AdultWeekend'],
                             value_name='Price')
```

```
[ ]: ticket_prices.head()
```

```
[ ]:
   state      Ticket  Price
0  Alaska  AdultWeekday   65.0
1  Alaska  AdultWeekday   47.0
2  Alaska  AdultWeekday   30.0
3  Arizona  AdultWeekday   89.0
4  Arizona  AdultWeekday   74.0
```

```
[ ]: plt.subplots(figsize=(12, 8))
sns.boxplot(x='state', y='Price', hue='Ticket', data=ticket_prices)
plt.xticks(rotation='vertical')
plt.ylabel('Price ($)')
plt.xlabel('State');
```



```
[ ]: ski_data.describe().transpose()
```

```
[ ]:
```

	count	mean	std	min	25%	50%	\
summit_elev	330.0	4591.818182	3735.535934	315.0	1403.75	3127.5	
vertical_drop	330.0	1215.427273	947.864557	60.0	461.25	964.5	
base_elev	330.0	3374.000000	3117.121621	70.0	869.00	1561.5	
trams	330.0	0.172727	0.559946	0.0	0.00	0.0	
fastEight	164.0	0.006098	0.078087	0.0	0.00	0.0	
fastSixes	330.0	0.184848	0.651685	0.0	0.00	0.0	
fastQuads	330.0	1.018182	2.198294	0.0	0.00	0.0	
quad	330.0	0.933333	1.312245	0.0	0.00	0.0	
triple	330.0	1.500000	1.619130	0.0	0.00	1.0	
double	330.0	1.833333	1.815028	0.0	1.00	1.0	
surface	330.0	2.621212	2.059636	0.0	1.00	2.0	

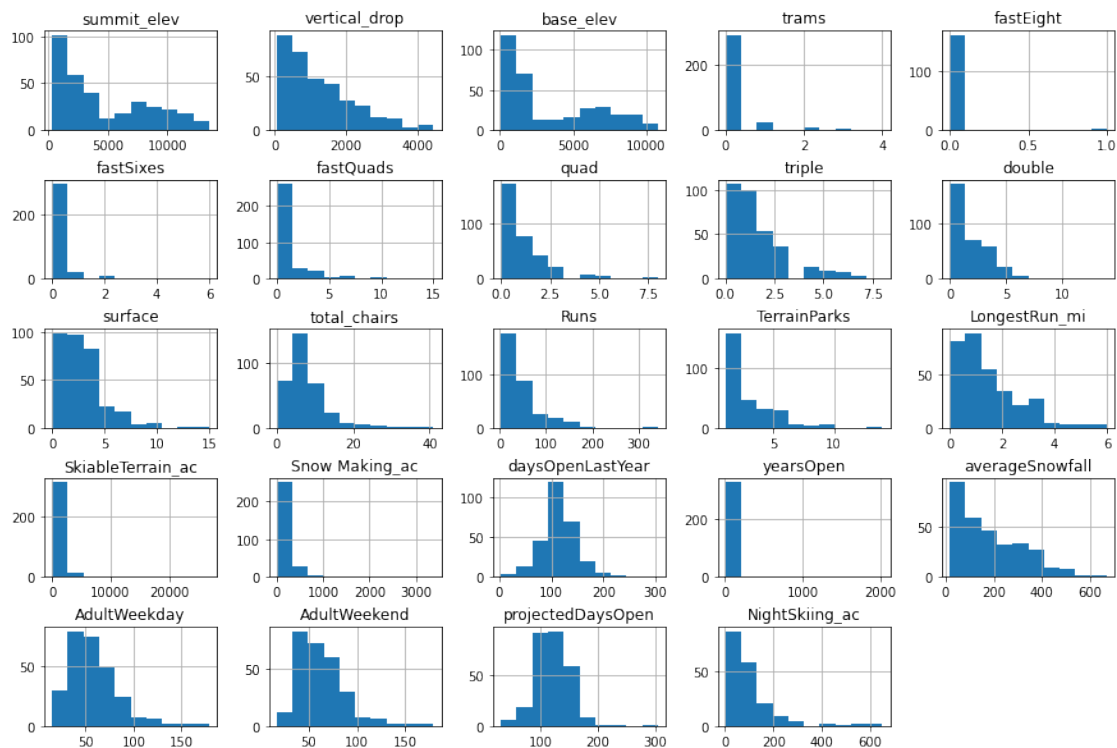
total_chairs	330.0	8.266667	5.798683	0.0	5.00	7.0
Runs	326.0	48.214724	46.364077	3.0	19.00	33.0
TerrainParks	279.0	2.820789	2.008113	1.0	1.00	2.0
LongestRun_mi	325.0	1.433231	1.156171	0.0	0.50	1.0
SkiableTerrain_ac	327.0	739.801223	1816.167441	8.0	85.00	200.0
Snow Making_ac	284.0	174.873239	261.336125	2.0	50.00	100.0
daysOpenLastYear	279.0	115.103943	35.063251	3.0	97.00	114.0
yearsOpen	329.0	63.656535	109.429928	6.0	50.00	58.0
averageSnowfall	316.0	185.316456	136.356842	18.0	69.00	150.0
AdultWeekday	276.0	57.916957	26.140126	15.0	40.00	50.0
AdultWeekend	279.0	64.166810	24.554584	17.0	47.00	60.0
projectedDaysOpen	283.0	120.053004	31.045963	30.0	100.00	120.0
NightSkiing_ac	187.0	100.395722	105.169620	2.0	40.00	72.0

	75%	max
summit_elev	7806.00	13487.0
vertical_drop	1800.00	4425.0
base_elev	6325.25	10800.0
trams	0.00	4.0
fastEight	0.00	1.0
fastSixes	0.00	6.0
fastQuads	1.00	15.0
quad	1.00	8.0
triple	2.00	8.0
double	3.00	14.0
surface	3.00	15.0
total_chairs	10.00	41.0
Runs	60.00	341.0
TerrainParks	4.00	14.0
LongestRun_mi	2.00	6.0
SkiableTerrain_ac	690.00	26819.0
Snow Making_ac	200.50	3379.0
daysOpenLastYear	135.00	305.0
yearsOpen	69.00	2019.0
averageSnowfall	300.00	669.0
AdultWeekday	71.00	179.0
AdultWeekend	77.50	179.0
projectedDaysOpen	139.50	305.0
NightSkiing_ac	114.00	650.0

```
[ ]: missing_price = ski_data[['AdultWeekend', 'AdultWeekday']].isnull().sum(axis=1)
missing_price.value_counts()/len(missing_price) * 100
```

```
[ ]: 0    82.424242
      2    14.242424
      1     3.333333
      dtype: float64
```

```
[ ]: ski_data.hist(figsize=(15,10))
plt.subplots_adjust(hspace=0.5);
```



```
[ ]: ski_data.loc[ski_data.SkiableTerrain_ac > 10000]
```

```
[ ]:
```

	Name	Region	state	summit_elev	vertical_drop	base_elev	trams	fastEight	fastSixes	fastQuads	...	LongestRun_mi	SkiableTerrain_ac	Snow Making_ac	daysOpenLastYear	yearsOpen	averageSnowfall	AdultWeekday	AdultWeekend	projectedDaysOpen	NightSkiing_ac
39	Silverton Mountain	Colorado	Colorado	13487	3087	10400	0	0.0	0	0	...	1.5	26819.0	NaN	175.0	17.0	400.0	79.0	79.0	181.0	NaN

```
[1 rows x 27 columns]
```

```
[ ]: ski_data[ski_data.SkiableTerrain_ac > 10000].transpose()
```

```
[ ]:
Name          Silverton Mountain
Region        Colorado
state         Colorado
summit_elev   13487
vertical_drop 3087
base_elev     10400
trams         0
fastEight     0.0
fastSixes     0
fastQuads     0
quad          0
triple        0
double        1
surface       0
total_chairs  1
Runs          NaN
TerrainParks  NaN
LongestRun_mi 1.5
SkiableTerrain_ac 26819.0
Snow Making_ac NaN
daysOpenLastYear 175.0
yearsOpen     17.0
averageSnowfall 400.0
AdultWeekday  79.0
AdultWeekend  79.0
projectedDaysOpen 181.0
NightSkiing_ac NaN
```

```
[ ]: ski_data.loc[39, 'SkiableTerrain_ac']
```

```
[ ]: 26819.0
```

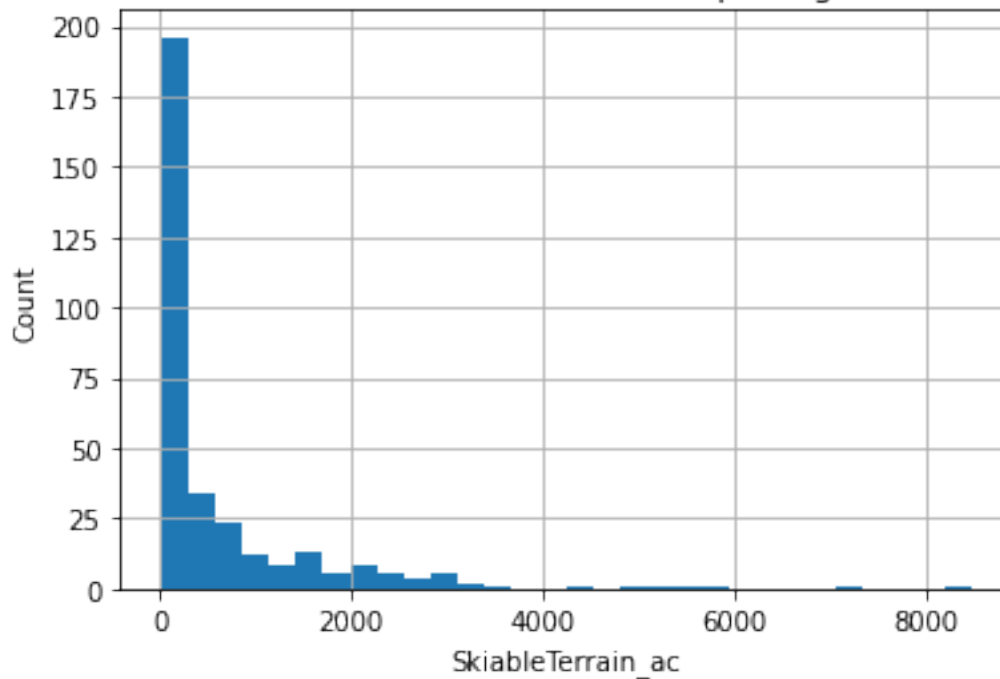
```
[ ]: ski_data.loc[39, 'SkiableTerrain_ac'] = 1819
```

```
[ ]: ski_data.loc[39, 'SkiableTerrain_ac']
```

```
[ ]: 1819.0
```

```
[ ]: ski_data.SkiableTerrain_ac.hist(bins=30)
plt.xlabel('SkiableTerrain_ac')
plt.ylabel('Count')
plt.title('Distribution of skiable area (acres) after replacing erroneous_
↪value');
```

Distribution of skiable area (acres) after replacing erroneous value



```
[ ]: ski_data['Snow Making_ac'][ski_data['Snow Making_ac'] > 1000]
```

```
[ ]: 11    3379.0
     18    1500.0
     Name: Snow Making_ac, dtype: float64
```

```
[ ]: ski_data[ski_data['Snow Making_ac'] > 3000].T
```

```
[ ]:
      Name      Heavenly Mountain Resort
      Region      Sierra Nevada
      state      California
      summit_elev      10067
      vertical_drop      3500
      base_elev      7170
      trams      2
      fastEight      0.0
      fastSixes      2
      fastQuads      7
      quad      1
      triple      5
      double      3
      surface      8
```

total_chairs	28
Runs	97.0
TerrainParks	3.0
LongestRun_mi	5.5
SkiableTerrain_ac	4800.0
Snow Making_ac	3379.0
daysOpenLastYear	155.0
yearsOpen	64.0
averageSnowfall	360.0
AdultWeekday	NaN
AdultWeekend	NaN
projectedDaysOpen	157.0
NightSkiing_ac	NaN

```
[ ]: .6 * 4800
```

```
[ ]: 2880.0
```

```
[ ]: ski_data.fastEight.value_counts()
```

```
[ ]: 0.0    163
     1.0     1
     Name: fastEight, dtype: int64
```

```
[ ]: ski_data.drop(columns='fastEight', inplace=True)
```

```
[ ]: ski_data.loc[ski_data.yearsOpen > 100]
```

```
[ ]:
```

	Name	Region	state	summit_elev	vertical_drop	\
34	Howelsen Hill	Colorado	Colorado	7136	440	
115	Pine Knob Ski Resort	Michigan	Michigan	1308	300	

	base_elev	trams	fastSixes	fastQuads	quad	...	LongestRun_mi	\
34	6696	0	0	0	0	...	6.0	
115	1009	0	0	0	0	...	1.0	

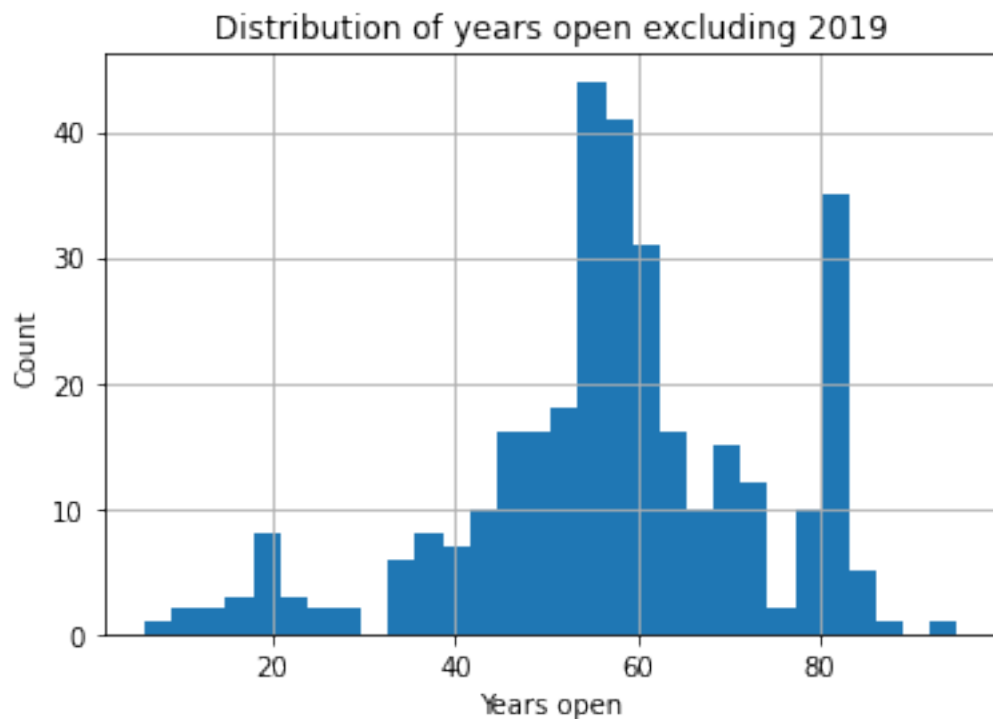
	SkiableTerrain_ac	Snow Making_ac	daysOpenLastYear	yearsOpen	\
34	50.0	25.0	100.0	104.0	
115	80.0	80.0	NaN	2019.0	

	averageSnowfall	AdultWeekday	AdultWeekend	projectedDaysOpen	\
34	150.0	25.0	25.0	100.0	
115	NaN	49.0	57.0	NaN	

	NightSkiing_ac
34	10.0
115	NaN

[2 rows x 26 columns]

```
[ ]: ski_data['yearsOpen'].loc[ski_data.yearsOpen < 100].hist(bins=30)
plt.xlabel('Years open')
plt.ylabel('Count')
plt.title('Distribution of years open excluding 2019');
```



```
[ ]: ski_data.yearsOpen[ski_data.yearsOpen < 1000].describe()
```

```
[ ]: count      328.000000
     mean       57.695122
     std        16.841182
     min         6.000000
     25%        50.000000
     50%        58.000000
     75%        68.250000
     max       104.000000
     Name: yearsOpen, dtype: float64
```

```
[ ]: ski_data = ski_data[ski_data.yearsOpen < 1000]
```

```
[ ]: state_summary = ski_data.groupby('state').agg(
    resorts_per_state=pd.NamedAgg(column='Name', aggfunc='size'), #could pick
    ↪any column here
    state_total_skiable_area_ac=pd.NamedAgg(column='SkiableTerrain_ac',
    ↪aggfunc='sum'),
    state_total_days_open=pd.NamedAgg(column='daysOpenLastYear', aggfunc='sum'),
    state_total_terrain_parks=pd.NamedAgg(column='TerrainParks', aggfunc='sum'),
    state_total_nightskiing_ac=pd.NamedAgg(column='NightSkiing_ac',
    ↪aggfunc='sum')
).reset_index()
state_summary.head()
```

```
[ ]:
      state  resorts_per_state  state_total_skiable_area_ac \
0      Alaska                3                2280.0
1      Arizona                2                1577.0
2  California               21               25948.0
3    Colorado               22              43682.0
4  Connecticut                5                358.0

      state_total_days_open  state_total_terrain_parks \
0                   345.0                   4.0
1                   237.0                   6.0
2                  2738.0                  81.0
3                  3258.0                  74.0
4                   353.0                  10.0

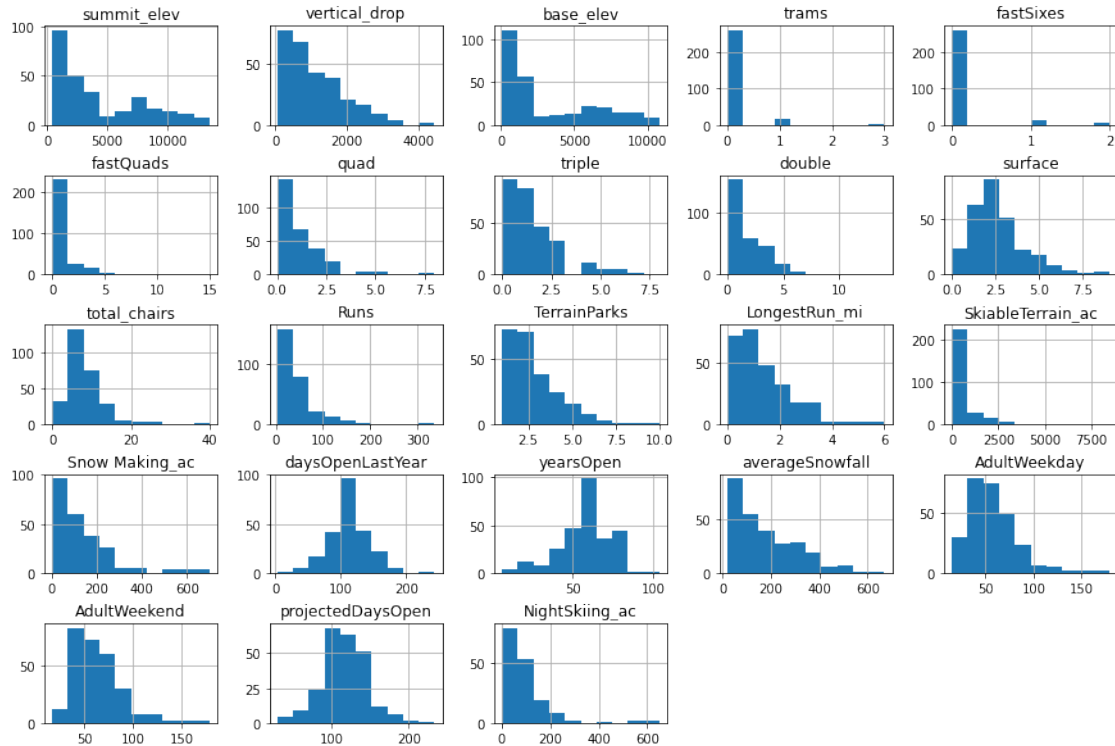
      state_total_nightskiing_ac
0                   580.0
1                    80.0
2                  587.0
3                  428.0
4                  256.0
```

```
[ ]: missing_price = ski_data[['AdultWeekend', 'AdultWeekday']].isnull().sum(axis=1)
missing_price.value_counts()/len(missing_price) * 100
```

```
[ ]: 0    82.317073
      2    14.329268
      1     3.353659
      dtype: float64
```

```
[ ]: ski_data = ski_data[missing_price != 2]
```

```
[ ]: ski_data.hist(figsize=(15, 10))
plt.subplots_adjust(hspace=0.5);
```



```
[ ]: states_url = 'https://simple.wikipedia.org/w/index.php?title=List_of_U.S.
↪_states&oldid=7168473'
usa_states = pd.read_html(states_url)
```

```
[ ]: type(usa_states)
```

```
[ ]: list
```

```
[ ]: len(usa_states)
```

```
[ ]: 1
```

```
[ ]: usa_states = usa_states[0]
usa_states.head()
```

```
[ ]: Name &postal abbs. [1]          Cities \
      Name &postal abbs. [1] Name &postal abbs. [1].1    Capital    Largest[5]
0          Alabama          AL    Montgomery    Birmingham
1          Alaska          AK      Juneau      Anchorage
2          Arizona          AZ      Phoenix      Phoenix
3          Arkansas          AR    Little Rock    Little Rock
4    California          CA    Sacramento    Los Angeles
```

	Established[A]	Population[B] [3]	Total area[4]		Land area[4]	\
	Established[A]	Population[B] [3]	mi2	km2	mi2	
0	Dec 14, 1819	4903185	52420	135767	50645	
1	Jan 3, 1959	731545	665384	1723337	570641	
2	Feb 14, 1912	7278717	113990	295234	113594	
3	Jun 15, 1836	3017804	53179	137732	52035	
4	Sep 9, 1850	39512223	163695	423967	155779	

	Water area[4]		Numberof Reps.	
	km2	mi2	km2	Numberof Reps.
0	131171	1775	4597	7
1	1477953	94743	245384	1
2	294207	396	1026	9
3	134771	1143	2961	4
4	403466	7916	20501	53

```
[ ]: established = usa_states.iloc[:, 4]
```

```
[ ]: established
```

```
[ ]: 0    Dec 14, 1819
      1    Jan 3, 1959
      2    Feb 14, 1912
      3    Jun 15, 1836
      4    Sep 9, 1850
      5    Aug 1, 1876
      6    Jan 9, 1788
      7    Dec 7, 1787
      8    Mar 3, 1845
      9    Jan 2, 1788
     10    Aug 21, 1959
     11    Jul 3, 1890
     12    Dec 3, 1818
     13    Dec 11, 1816
     14    Dec 28, 1846
     15    Jan 29, 1861
     16    Jun 1, 1792
     17    Apr 30, 1812
     18    Mar 15, 1820
     19    Apr 28, 1788
     20    Feb 6, 1788
     21    Jan 26, 1837
     22    May 11, 1858
     23    Dec 10, 1817
     24    Aug 10, 1821
     25    Nov 8, 1889
     26    Mar 1, 1867
```

```

27    Oct 31, 1864
28    Jun 21, 1788
29    Dec 18, 1787
30    Jan 6, 1912
31    Jul 26, 1788
32    Nov 21, 1789
33    Nov 2, 1889
34    Mar 1, 1803
35    Nov 16, 1907
36    Feb 14, 1859
37    Dec 12, 1787
38    May 29, 1790
39    May 23, 1788
40    Nov 2, 1889
41    Jun 1, 1796
42    Dec 29, 1845
43    Jan 4, 1896
44    Mar 4, 1791
45    Jun 25, 1788
46    Nov 11, 1889
47    Jun 20, 1863
48    May 29, 1848
49    Jul 10, 1890

```

Name: (Established[A], Established[A]), dtype: object

```
[ ]: usa_states_sub = usa_states.iloc[:, [0, 5, 6]].copy()
      usa_states_sub.columns = ['state', 'state_population', 'state_area_sq_miles']
      usa_states_sub.head()
```

```
[ ]:
      state  state_population  state_area_sq_miles
0    Alabama          4903185             52420
1     Alaska           731545             665384
2    Arizona          7278717             113990
3   Arkansas          3017804              53179
4  California          3951223             163695
```

```
[ ]: missing_states = set(state_summary.state) - set(usa_states_sub.state)
      missing_states
```

```
[ ]: {'Massachusetts', 'Pennsylvania', 'Rhode Island', 'Virginia'}
```

```
[ ]: usa_states_sub.state[usa_states_sub.state.str.
      ↪contains('Massachusetts|Pennsylvania|Rhode Island|Virginia')]
```

```
[ ]: 20    Massachusetts[C]
      37    Pennsylvania[C]
      38    Rhode Island[D]
```

```

45         Virginia[C]
47         West Virginia
Name: state, dtype: object

```

```

[ ]: usa_states_sub.state.replace(to_replace='\[.*\]', value='', regex=True,
    ↪inplace=True)
usa_states_sub.state[usa_states_sub.state.str.
    ↪contains('Massachusetts|Pennsylvania|Rhode Island|Virginia')]

```

```

[ ]: 20    Massachusetts
37    Pennsylvania
38    Rhode Island
45    Virginia
47    West Virginia
Name: state, dtype: object

```

```

[ ]: missing_states = set(state_summary.state) - set(usa_states_sub.state)
missing_states

```

```

[ ]: state_summary = state_summary.merge(usa_states_sub, how='left', on='state')
state_summary.head()

```

```

[ ]:
      state  resorts_per_state  state_total_skiable_area_ac \
0      Alaska                3                2280.0
1      Arizona                2                1577.0
2  California               21             25948.0
3      Colorado               22             43682.0
4  Connecticut                5                358.0

      state_total_days_open  state_total_terrain_parks \
0                345.0                4.0
1                237.0                6.0
2             2738.0             81.0
3             3258.0             74.0
4                353.0             10.0

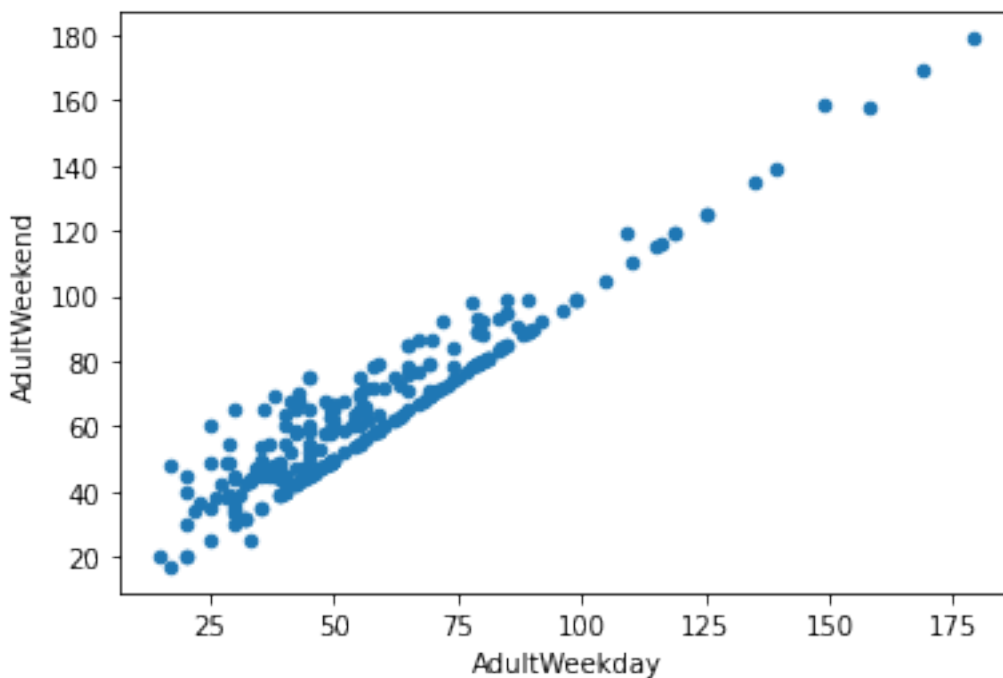
      state_total_nightskiing_ac  state_population  state_area_sq_miles
0                580.0             731545             665384
1                80.0             7278717             113990
2                587.0            39512223             163695
3                428.0            5758736             104094
4                256.0            3565278              5543

```

```

[ ]: ski_data.plot(x='AdultWeekday', y='AdultWeekend', kind='scatter');

```



```
[ ]: ski_data.loc[ski_data.state == 'Montana', ['AdultWeekend', 'AdultWeekday']]
```

```
[ ]:      AdultWeekend  AdultWeekday
141          42.0          42.0
142          63.0          63.0
143          49.0          49.0
144          48.0          48.0
145          46.0          46.0
146          39.0          39.0
147          50.0          50.0
148          67.0          67.0
149          47.0          47.0
150          39.0          39.0
151          81.0          81.0
```

```
[ ]: ski_data[['AdultWeekend', 'AdultWeekday']].isnull().sum()
```

```
[ ]: AdultWeekend    4
     AdultWeekday    7
     dtype: int64
```

```
[ ]: ski_data.drop(columns='AdultWeekday', inplace=True)
     ski_data.dropna(subset=['AdultWeekend'], inplace=True)
```

```
[ ]: ski_data.shape
```

```
[ ]: (277, 25)
```

```
[ ]: missing = pd.concat([ski_data.isnull().sum(axis=1), 100 * ski_data.isnull().
    ↳mean(axis=1)], axis=1)
missing.columns=['count', '%']
missing.sort_values(by='count', ascending=False).head(10)
```

```
[ ]:      count      %
329      5  20.0
62       5  20.0
141      5  20.0
86       5  20.0
74       5  20.0
146      5  20.0
184      4  16.0
108      4  16.0
198      4  16.0
39       4  16.0
```

```
[ ]: missing['%'].unique()
```

```
[ ]: array([ 0.,  4.,  8., 12., 16., 20.])
```

```
[ ]: missing['%'].value_counts()
```

```
[ ]: 0.0      107
4.0       94
8.0       45
12.0      15
16.0      10
20.0       6
Name: %, dtype: int64
```

```
[ ]: ski_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 277 entries, 0 to 329
Data columns (total 25 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Name                   277 non-null   object
1   Region                 277 non-null   object
2   state                  277 non-null   object
3   summit_elev            277 non-null   int64
4   vertical_drop          277 non-null   int64
5   base_elev              277 non-null   int64
6   trams                  277 non-null   int64
```



```

7   fastSixes          277 non-null    int64
8   fastQuads          277 non-null    int64
9   quad              277 non-null    int64
10  triple             277 non-null    int64
11  double             277 non-null    int64
12  surface           277 non-null    int64
13  total_chairs       277 non-null    int64
14  Runs              274 non-null    float64
15  TerrainParks       233 non-null    float64
16  LongestRun_mi      272 non-null    float64
17  SkiableTerrain_ac  275 non-null    float64
18  Snow Making_ac     240 non-null    float64
19  daysOpenLastYear   233 non-null    float64
20  yearsOpen          277 non-null    float64
21  averageSnowfall    268 non-null    float64
22  AdultWeekend       277 non-null    float64
23  projectedDaysOpen  236 non-null    float64
24  NightSkiing_ac     163 non-null    float64
dtypes: float64(11), int64(11), object(3)
memory usage: 56.3+ KB

```

```
[ ]: ski_data.shape
```

```
[ ]: (277, 25)
```

```
[ ]: # save the data to a new csv file
ski_data.to_csv('ski_data_cleaned.csv', index=False)
```

```
[ ]: # save the state_summary separately.
state_summary.to_csv('state_summary.csv', index = False)
```

In the original data, there are 329 rows and 27 columns with information on 277 skiing resorts across the nation. After our observation of the histograms of the numeric features of the resorts, it is obvious that some of them are not very plausible.

To begin with, we can see the data for Skiable Terrain_ac are clustering down below 10,000. To investigate further, we can print out the resorts with the value of Skiable Terrain_ac greater than 10,000. It turns out that there is only one resort, Silverton Mountain, has more than 10,000 acres of skiable terrain. The value for Silverton Mountain is 26819, which is suspiciously high compared to other resorts. By searching “silverton mountain skiable area”, we can find that the real skiable terrain value for Silverton Mountain is 1819 instead of 26819. We can replace the wrong value with the right one using the .loc accessor. We can see that the new plot makes more sense with the value change. For the same reason, we delete one row from the data where yearsOpen is 2019.

The fastEight's plot is also strange. Most of the values are 0 and a lot of values are null, which means this feature won't provide us with little information. We can drop the whole column from our data. For the same reason, we drop the rows with no price data.

We also need to set a target feature. Since the goal of the project is to provide a better pricing strategy, we want to use the value of AdultWeekday or AdultWeekend. By plotting a graph, we can see the relationship between AdultWeekday or AdultWeekend is linear (i.e. the higher the value of AdultWeekday, the higher the value of AdultWeekend.) Therefore, we can drop one of the prices. Since AdultWeekend has the least missing value of the two, we choose to drop the AdultWeekday.

After wrangling with the data, we have 277 rows and 25 columns left. Created in Deepnote