Session 13: Simulation Modeling II

In this session, we analyze a simulation case that puts everything we learned in the course so far together, including probability, algorithmic thinking and Python programming.

Case 10: Pricing Two Substitutable Products by Simulation

Part I: Simulating Customer Valuations

A firm sells two styles of headphones, which we refer to as model 0 and model 1. Based on a clustering analysis using historic data, the firm estimates that customers will come from three segments (A, B or C), and the valuation (maximum willingness to pay) of customers for the two products can be modelled as normally distributed according to the following parameters.

| Segment | μ_0 | σ_0 | μ_1 | σ_1 | Proportion |
|---------|-------------|------------|---------|------------|------------|
| A | 30 | 30 | 70 | 30 | 0.1 |
| В | 80 | 20 | 20 | 10 | 0.3 |
| C | - 10 | 20 | -10 | 20 | 0.6 |

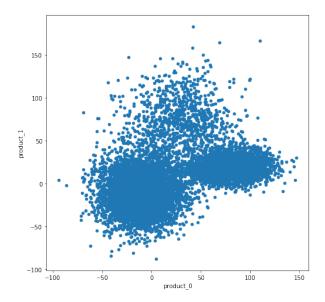
As in the above table, a randomly chosen customer will be from segment A with 10% probability, segment B with 30% probability and segment C with 60% probability. Segment A customers have high valuations for model 1, while segment B customers have high valuations for model 0. Segment C customers, which make up the majority, do not on average value either products.

Generate a pandas DataFrame called "values" representing the simulated valuations of 10,000 randomly chosen customers. Each row represents a customer. There are three columns:

- **segment**: The segment of the customer, being "A", "B" or "C".
- product_0: The customer's maximum willingness to pay for Model 0.
- product_1: The customer's maximum willingness to pay for Model 1.

Once you have completed this part, you should be able to run the following code and obtain similar outputs.

```
[2]: values.head()
```



Part II: Analysis and Optimization

The following code is a modification of the solution to case 9, using Pandas vectorized functions instead of for loops for improved performance.

```
[4]: import numpy as np
    def demand(df,priceVector):
        diff=df[['product_0','product_1']]-priceVector
        demand0=((diff['product_0']>=diff['product_1'])&(diff['product_0']>=0)).sum()
        demand1=((diff['product_0']<diff['product_1'])&(diff['product_1']>=0)).sum()
        return demand0,demand1
    demand(values,[30,50])
(3314, 652)
```

Write a function called "tabluate" which takes as input a DataFrame in the format of the "values" DataFrame from Part I and outputs a DataFrame with the following columns:

- **Price_0**: The price for Model 0.
- **Price_1**: The price for Model 1.
- **Demand_0**: The simulated demand for Model 0 under the above prices.
- **Demand_1**: The simulated demand for Model 1 under the above prices.
- **Revenue**: The total revenue from the two products.

The rows of the DataFrame corresponds to every combination of Price_0 and Price_1 with values from range(0,200,5), which is equivalent to the list [0,5,10,...,195].

```
[6]: result=tabulate(values)
    result.shape

(1600, 5)

[7]: result.head()

    Price_0 Price_1 Demand_0 Demand_1 Revenue
0     0     4622     2431     0
```

| 1 | 0 | 5 | 4744 | 1996 | 9980 |
|---|---|----|------|------|-------|
| 2 | 0 | 10 | 4838 | 1605 | 16050 |
| 3 | 0 | 15 | 4925 | 1284 | 19260 |
| 4 | 0 | 20 | 4989 | 1049 | 20980 |

Using the "result" DataFrame, you can obtain the best revenue found using a number of ways, as below.

```
[8]: result['Revenue'].max()
194215
[9]: result['Revenue'].idxmax()
491
[10]: result.iloc[491,:]
Price_0
                60
Price_1
                55
Demand_0
              2587
Demand_1
               709
Revenue
            194215
Name: 491, dtype: int64
[11]: result.sort_values(by='Revenue',ascending=False).head(1)
     Price_0 Price_1 Demand_0
                                  Demand_1 Revenue
491
          60
                   55
                            2587
                                       709
                                             194215
```

Part III: Obtaining Additional Insights

A. Value of price discrimination

Suppose that the company can observe which segment each customer belongs to, and charge separate prices to each segment. What would be the optimal prices for each segment and what would be the additional revenue from this flexibility?

(**Hint:** Filter the "values" DataFrame by whether the segment is A, B or C, and use the tabulate function to obtain DataFrames "resultA", "resultB", "resultC", which are analogous to the "result" DataFrame from above but are computed using valuations from one segment at a time.)

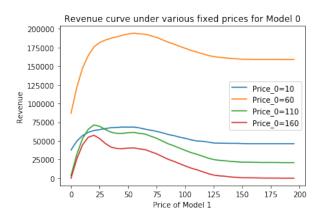
```
[13]: # Best Pricing for Segment A
    Price_0 Price_1 Demand_0 Demand_1
                                           Revenue
451
                   55
         55
                            103
                                      692
                                             43725
[14]: # Best Pricing for Segment B
    Price_0 Price_1 Demand_0 Demand_1 Revenue
          60
                           2465
486
                   30
                                       98
                                            150840
[15]: # Best Pricing for Segment C
    Price_0 Price_1 Demand_0
                                Demand_1 Revenue
123
          15
                   15
                            580
                                      603
                                             17745
[16]:
```

Potential benefit of price differentiation is about \$18095.

B. Fixed Prices for One Product

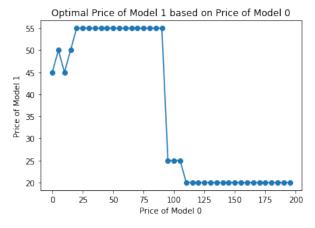
Suppose that the price for Model 0 is fixed and the firm can only alter the price for Model 1, plot the total Revenue as a function of the price for Model 1, when the price for Model 0 is 10, 60, 110, and 160. (Note: you do not have to plot them on the same figure as below.)

[17]:



An alternative analysis based on more advanced Pandas functionality is as follows.

```
[18]: def bestPrice1(df):
          return df.sort_values(by='Revenue',ascending=False)['Price_1'].iloc[0]
      optPrice1=result.groupby('Price_0').apply(bestPrice1)
      optPrice1.head()
Price_0
      45
0
5
      50
10
      45
15
      50
20
dtype: int64
[19]: import matplotlib.pyplot as plt
      optPrice1.plot(marker='o')
      plt.xlabel('Price of Model 0')
      plt.ylabel('Price of Model 1')
      plt.title('Optimal Price of Model 1 based on Price of Model 0')
      plt.show()
```

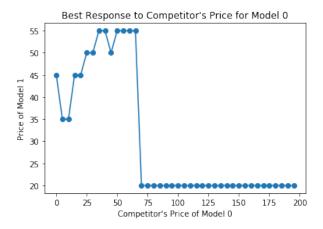


(Optional) C. Competitive Pricing

Suppose now that Model 0 is sold by a competitor, and only revenue from Model 1 counts. Modify the above code to display the optimal price for Model 1 given the competitor's pricing for Model 0. Moreover, plot the optimal attainable revenue given the competitor's pricing.

(**Hint**: You can create a new column in the "result" DataFrame corresponding to the revenue for Model 1 only, and sort by that revenue instead of by total revenue in the above code.)

[21]:



[22]:

