Product Demand Prediction with Machine Learning

A product company plans to offer discounts on its product during the upcoming holiday season. The company wants to find the price at which its product can be a better deal compared to its competitors. For this task, the company provided a dataset of past changes in sales based on price changes. You need to train a model that can predict the demand for the product in the market with different price segments.

The **dataset** that we have for this task contains data about:

- 1. the product id;
- 2. store id:
- 3. total price at which product was sold;
- 4. base price at which product was sold;
- 5. Units sold (quantity demanded);

Product Demand Prediction using Python

Let's start by importing the necessary Python libraries and the dataset we need for the task of product demand prediction:

import pandas as pd

import numpy as np

import plotly.express as px

import seaborn as sns

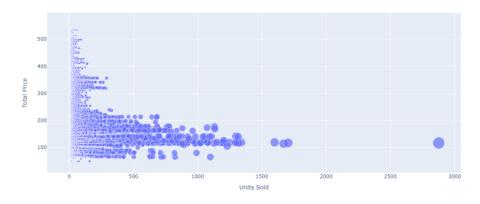
import matplotlib.pyplot as plt

from sklearn.model_selection import train_test_split

from sklearn.tree import DecisionTreeRegressor

data = pd.read_csv("https://raw.githubusercontent.com/amankharwal/Website-data/master/demand.csv")

data.head()



| | ID | Store ID | Total Price | Base Price | Units Sold |
|---|----|----------|-------------|------------|------------|
| 0 | 1 | 8091 | 99.0375 | 111.8625 | 20 |
| 1 | 2 | 8091 | 99.0375 | 99.0375 | 28 |
| 2 | 3 | 8091 | 133.9500 | 133.9500 | 19 |
| 3 | 4 | 8091 | 133.9500 | 133.9500 | 44 |
| 4 | 5 | 8091 | 141.0750 | 141.0750 | 52 |
| | | | | | |

Now let's have a look at whether this dataset contains any null values or not:

data.isnull().sum()

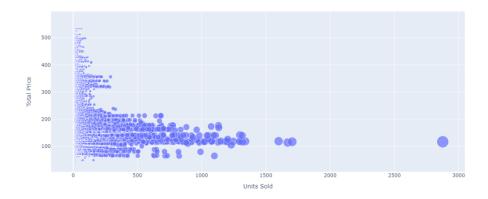
```
ID 0
Store ID 0
Total Price 1
Base Price 0
Units Sold 0
dtype: int64
```

So the dataset has only one missing value in the **Total Price** column, I will remove that entire row for now:data = data.dropna()

Let us now analyze the relationship between the price and the demand for the product. Here I will use a **scatter plot** to see how the demand for the product varies with the price change:

```
1 fig = px.scatter(data, x="Units Sold", y="Total Price", size='Units
Sold')
```

fig.show()



We can see that most of the data points show the sales of the product is increasing as the price is decreasing with some exceptions. Now let's have a look at the correlation between the features of the dataset:

print(data.corr())

```
Store ID
                                 Total Price Base Price
                                                          Units Sold
ID
             1.000000
                       0.007464
                                    0.008473
                                                0.018932
                                                            -0.010616
Store ID
             0.007464
                       1.000000
                                   -0.038315
                                               -0.038848
                                                            -0.004372
Total Price 0.008473 -0.038315
                                    1.000000
                                                0.958885
                                                            -0.235625
             0.018932 -0.038848
Base Price
                                    0.958885
                                                1.000000
                                                            -0.140032
Units Sold -0.010616 -0.004372
                                   -0.235625
                                               -0.140032
                                                             1.000000
```

```
correlations = data.corr(method='pearson')
plt.figure(figsize=(15, 12))
sns.heatmap(correlations, cmap="coolwarm", annot=True)
plt.show()
```



Product Demand Prediction Model

Now let's move to the task of training a machine learning model to predict the demand for the product at different prices. I will choose the **Total Price** and the **Base Price** column as the features to train the model, and the **Units Sold** column as labels for the model:

```
x = data[["Total Price", "Base Price"]]
y = data["Units Sold"]
```

Now let's split the data into training and test sets and use the decision tree regression algorithm to train our model:

```
xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size=0.2,
random_state=42)
from sklearn.tree import DecisionTreeRegressor
model = DecisionTreeRegressor()
model.fit(xtrain, ytrain)
```

Now let's input the features (**Total Price**, **Base Price**) into the model and predict how much quantity can be demanded based on those values:

```
#features = [["Total Price", "Base Price"]]
features = np.array([[133.00, 140.00]])
model.predict(features)
array([27.])
```

Summary

So this is how you can train a machine learning model for the task of product demand prediction using Python. Price is one of the major factors that affect the demand for the product. If a product is not a necessity, only a few people buy the product even if the price increases. I hope you liked this article on product demand prediction with machine learning using Python. Feel free to ask your valuable questions in the comments section below.