

PRODUCT DEMAND PREDICTION WITH MACHINE LEARNING

PHASE 03: Development Part 1

INTRODUCTION:

Product prediction involves using data and analysis to make informed forecasts about future product trends, demand, or market behavior. This can be done in various industries, including retail, e-commerce, manufacturing, and more

Machine learning techniques, such as regression, time series analysis, and deep learning, are commonly used for product prediction. Additionally, predictive analytics software and tools can assist in automating and streamlining the prediction process.

It's essential to remember that product prediction is an ongoing process, and the accuracy of predictions can improve over time as more data becomes available and the model becomes more refined.

Development model:

Creating a machine learning model for product demand prediction involves several steps, including data preparation, model selection, training, and evaluation. Below is a simple example of how to perform product demand prediction using Python and the scikit-learn library. For a more complex and accurate model, you may need more data and advanced techniques, but this example serves as a starting point.

```

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

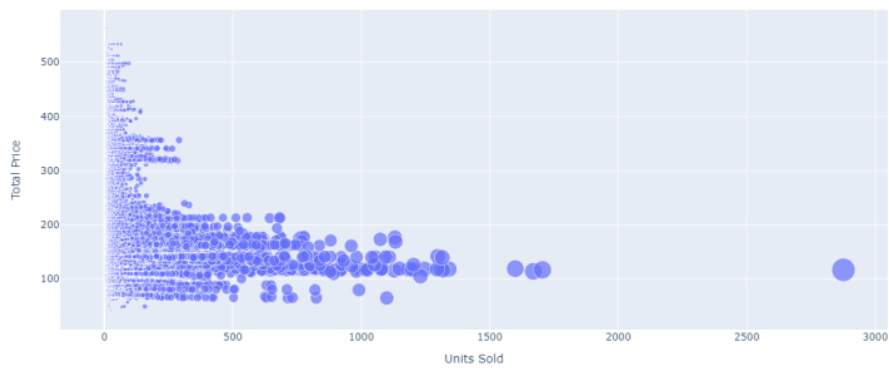
Now let's have a look at whether this dataset contains any null values or not:

```

data.isnull().sum()

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

```



```
print(data.corr())
```

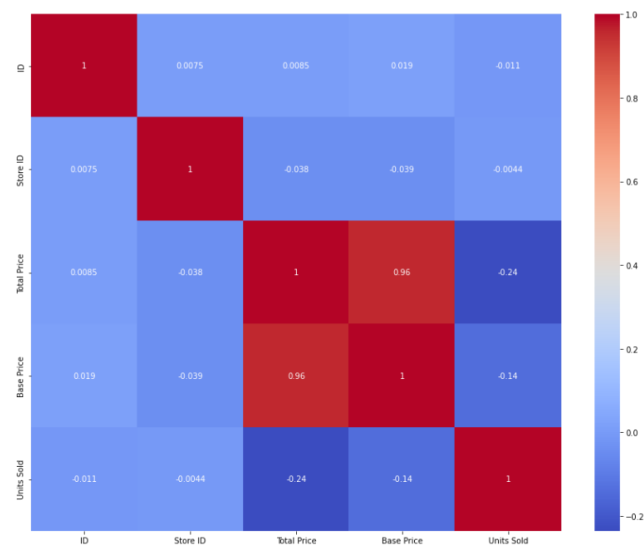
	ID	Store ID	Total Price	Base Price	Units Sold
ID	1.000000	0.007464	0.008473	0.018932	-0.010616
Store ID	0.007464	1.000000	-0.038315	-0.038848	-0.004372
Total Price	0.008473	-0.038315	1.000000	0.958885	-0.235625
Base Price	0.018932	-0.038848	0.958885	1.000000	-0.140032
Units Sold	-0.010616	-0.004372	-0.235625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

*Product Demand prediction with machine learning model prepared for **Decision Tree Regressor** in this phase.*

THANK YOU !