

PRODUCT DEMAND PREDICTION WITH MACHINE LEARNING

PROJECT TITLE: PRODUCT DEMAND ANALYSIS

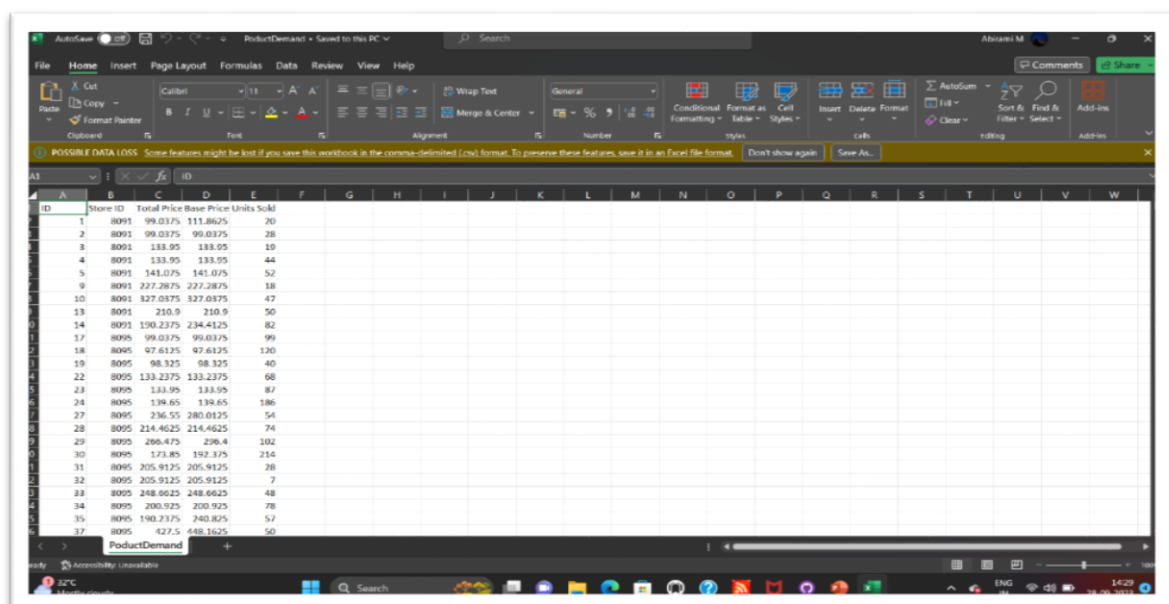
PROBLEM STATEMENT:

Create a machine learning model that forecasts product demand based on historical sales and external factors, helping businesses optimize inventory management and production planning to meet customer needs efficiently.

DATASET: Product demand dataset

link: <https://www.kaggle.com/datasets/chakradharmattapali/product-demand-prediction-with-machine-learning>

some of the screenshot of product demand Exel sheet of dataset are:



ID	Store ID	Total Price	Base Price	Units Sold
1	8091	99.0375	111.8625	20
2	8091	99.0375	99.0375	28
3	8091	133.95	133.95	10
4	8091	133.95	133.95	44
5	8091	141.075	141.075	52
6	8091	227.2875	227.2875	18
7	8091	327.0375	327.0375	47
8	8091	210.9	210.9	50
9	8091	190.2375	234.4125	82
10	8095	99.0375	99.0375	99
11	8095	97.6125	97.6125	130
12	8095	98.325	98.325	40
13	8095	133.2375	133.2375	68
14	8095	133.95	133.95	87
15	8095	136.65	136.65	186
16	8095	236.55	280.0125	54
17	8095	214.4625	214.4625	74
18	8095	296.475	296.4	102
19	8095	173.85	192.375	214
20	8095	205.9125	205.9125	78
21	8095	205.9125	205.9125	7
22	8095	248.6625	248.6625	48
23	8095	200.925	200.925	78
24	8095	190.2375	240.825	57
25	8095	427.5	449.1625	50

POSSIBLE DATA LOSS: Some features might be lost if you save this workbook in the comma-separated (.csv) format. To preserve these features, save it as an Excel file format. Don't show again Save As...

	A	B	C	D
35	8095	190.2375	240.825	57
37	8095	427.5	948.1625	50
38	8095	426.6375	458.1375	63
39	8095	177.4125	177.4125	22
43	8094	87.6375	87.6375	108
43	8094	88.35	88.35	133
44	8094	85.5	85.5	11
45	8094	128.25	180.975	9
47	8094	127.5375	127.5375	19
48	8094	123.975	123.975	33
49	8094	139.85	104.5875	49
50	8094	235.8375	235.8375	82
51	8094	234.4125	234.4125	47
52	8094	235.125	235.125	27
53	8094	227.2875	227.2875	88
54	8094	512.7875	512.7875	48
55	8094	230.9	230.9	90
56	8094	177.4125	177.4125	27
57	8094	177.4125	177.4125	33
58	8094	240.825	240.825	18
59	8094	213.0375	213.0375	72
60	8094	190.95	713.0375	81
61	8094	426.7875	448.1625	11
62	8094	426.7875	448.875	13
63	8094	426.7875	448.1625	28
65	8094	170.2875	170.2875	56

POSSIBLE DATA LOSS: Some features might be lost if you save this workbook in the comma-separated (.csv) format. To preserve these features, save it as an Excel file format. Don't show again Save As...

	A	B	C	D
0127	212809	9991	213.0375	213.0375
0128	212810	9991	213.0375	213.0375
0129	212812	9991	252.9375	252.9375
0130	212813	9991	232.275	232.275
0131	212814	9991	236.55	236.55
0132	212816	9991	355.5375	355.5375
0133	212817	9991	356.9625	356.9625
0134	212819	9991	143.7875	143.7875
0135	212822	9994	87.6375	87.6375
0136	212823	9994	86.925	86.925
0137	212824	9991	86.2125	86.2125
0138	212826	9994	131.1	131.1
0139	212827	9994	130.3875	130.3875
0140	212828	9994	170.2875	170.2875
0141	212832	9994	194.5125	194.5125
0142	212833	9991	327.0375	327.0375
0143	212834	9994	164.825	164.825
0144	212835	9991	205.9125	205.9125
0145	212836	9994	205.9125	205.9125
0146	212837	9994	239.4	239.4
0147	212838	9991	235.8375	235.8375
0148	212839	9994	235.8375	235.8375
0149	212842	9991	357.675	357.675
0150	212843	9994	143.7875	143.7875
0151	212844	9994	234.4125	234.4125

Project steps:

1.problem definition

2.Design Thinking

Step-1: Problem definition:

The problem is to create a machine learning model that forecasts product demand based on historical sales data and external factors. The goal is to help businesses optimize inventory management and production planning to efficiently meet customer needs. This project involves data collection, data preprocessing, feature engineering, model selection, training, and evaluation.

Step-2: Design Thinking:

(1).Data Collection:

Data collection is a systematic process of gathering observations or measurements. Whether you are performing research for business, governmental or academic purposes, data collection allows you to gain first-hand knowledge and original insights into your [research problem](#).

While methods and aims may differ between fields, the overall process of data collection remains largely the same.

Before you begin collecting data, you need to consider:

The aim of the research

The type of data that you will collect

The methods and procedures you will use to collect, store, and process the data.

(2).Data Preprocessing :

Data preprocessing is an important step in the data mining process. It refers to the cleaning, transforming and integration of data in order to make it ready for analysis. The goal of data preprocessing is to improve the quality of the data and to make it more suitable for the specific data mining task.

Some common steps in data preprocessing are:

- (a).Data cleaning
- (b).Data Integration
- (c).Data Transformation
- (d).Data Reduction
- (e).Data Discretization
- (f).Data Normalization

(3).Feature Engineering :

Feature engineering involves creating relevant features from the raw data. For instance:

- Lag features: Include past sales data (e.g., sales from the previous week or month) as features.
- Date-related features: Extract features like day of the week, month, quarter, or year.
- External factors: Incorporate external data such as holidays, economic indicators, or weather forecasts.

(4).Model Selection:

Choose an appropriate machine learning algorithm for your demand forecasting task. Time series models like ARIMA or machine learning models like Random Forest, XGBoost, or LSTM (if you have a significant amount of data) are common choices.

For this example, we'll use a Random Forest regressor.

```
from sklearn.ensemble import RandomForestRegressor  
  
model = RandomForestRegressor(n_estimators=100,  
random_state=42)
```

(5).Model Training:

Train the selected model on your training data.

Example:

```
model.fit(X_train, y_train)
```

(6).Evaluation:

Evaluate your model's performance on the testing dataset using appropriate metrics such as Mean Absolute Error (MAE), Root Mean Square Error (RMSE), or Mean Absolute Percentage Error (MAPE).

Example:

```
from sklearn.metrics import mean_absolute_error
```

```
y_pred = model.predict(X_test)
mae = mean_absolute_error(y_test, y_pred)
print(f"Mean Absolute Error: {mae}")
```