

# **WALMART EXPLORATORY DATA ANALYSIS (EDA)**

Prepared for Walmart

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Presented To  
INVENTORY MANAGERS

Presented By  
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## WALMART DATA ANALYSIS

### Case Study

- I am a Data Analyst recently recruited by Walmart which has multiple retail stores nationwide. My task is to help the inventory managers with insights that could help them manage inventories – to match demand and supply.
- I have been assigned to conduct data analysis and develop data-driven recommendations for the inventory manager.
- I have one (1) week to present my analysis and recommendations to the managers.

### About Walmart

Walmart is one of the world's leading multinational retail corporations with a huge turnover. It was founded in the year **1962** and was incorporated in the year **1969** with its headquarters in Bentonville, Arkansas.

Besides being a huge firm, Walmart is still controlled by the Walton family. It operates globally with more than **11,700** retail stores in **28+** countries. Also, it serves its e-retailers in eleven countries.

### Dataset Description

Dataset: <https://www.kaggle.com/datasets/asahu40/walmart-data-analysis-and-forecasting>

Initially, the dataset had 8 columns and 6,435 rows, I feature-engineered the columns to 11 by splitting the 'Date' column into 'Day', 'Month', and 'Year'.

Interestingly, I found a strong correlation between the 'Weekly\_Sales' and the 'Month' columns.

### Description of the Columns

**Store:** *Store number.*

**Date:** *Week of sales.*

**Weekly\_Sales:** *Weekly sales for the given store.*

**Holiday\_Flag:** *Whether the week is a special holiday week or not (1 = holiday week; 0 = non-holiday week).*

**Temperature:** *Average temperature in the region for the given week.*

**Fuel\_Price:** *Cost of fuel in the region for the given week.*

**CPI:** *Consumer price index for the given week.*

**Unemployment:** *Unemployment rate for the region for the given week.*

## Tools Used

Python, Visual Studio Code, GitHub, Microsoft Word

## Issues / Difficulties Encountered

1. The 'Date' column was in a String (Object) datatype format; I needed to convert it to a datetime datatype. I used the .to\_datetime function of the pandas dataframe to convert the date column into a datetime format.

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6435 entries, 0 to 6434
Data columns (total 8 columns):
#   Column          Non-Null Count  Dtype
---  ---
0   Store            6435 non-null   int64
1   Date             6435 non-null   object
2   Weekly_Sales     6435 non-null   float64
3   Holiday_Flag     6435 non-null   int64
4   Temperature      6435 non-null   float64
5   Fuel_Price       6435 non-null   float64
6   CPI              6435 non-null   float64
7   Unemployment     6435 non-null   float64
dtypes: float64(5), int64(2), object(1)
memory usage: 402.3+ KB
```

Fig: Object (String) datatype

```
Store            int64
Date             datetime64[ns]
Weekly_Sales     float64
Holiday_Flag     int64
Temperature      float64
Fuel_Price       float64
CPI              float64
Unemployment     float64
dtype: object
```

Fig: Datetime format

2. Converting the exponential outputs of the 'Weekly\_Sales' column from exponential to integer values by using the .astype(int) function to make it easier to read.

```
▶ ▾
Total_weekly_sales = df.groupby('Store')['Weekly_Sales'].sum()
print("Total Weekly Sales For each store: ", Total_weekly_sales)

[109] ✓ 0.0s

... Total Weekly Sales For each store: Store
1      2.224028e+08
2      2.753824e+08
3      5.758674e+07
4      2.995440e+08
5      4.547569e+07
6      2.237561e+08
7      8.159828e+07
8      1.299512e+08
9      7.778922e+07
10     2.716177e+08
11     1.939628e+08
12     1.442872e+08
13     2.865177e+08
14     2.889999e+08
15     8.913368e+07
16     7.425243e+07
17     1.277821e+08
18     1.551147e+08
19     2.066349e+08
20     3.013978e+08
21     1.081179e+08
22     1.470756e+08
23     1.987506e+08
24     1.940160e+08
...
43     9.056544e+07
44     4.329309e+07
45     1.123953e+08
Name: Weekly_Sales, dtype: float64
```

Fig: Exponential outputs

```
▶ ▾
Average_weekly_sales = df.groupby('Store')['Weekly_Sales'].mean().astype(int)
print("Average Weekly Sales For each store: ", Average_weekly_sales)

[102] ✓ 0.0s

... Average Weekly Sales For each store: Store
1      1555264
2      1925751
3       402704
4      2094712
5       318011
6      1564728
7       570617
8       908749
9       543980
10     1899424
11     1356383
12     1009001
13     2003620
14     2020978
15     623312
16     519247
17     893581
18     1084718
19     1444999
20     2107676
21     756069
22     1028501
23     1389864
24     1356755
...
43     633324
44     302748
45     785981
Name: Weekly_Sales, dtype: int32
```

Fig: Using the .astype(int) function

3. I encountered 'depreciation' warnings, which I ignored by adding the Python script below.

```
import warnings
warnings.filterwarnings("ignore")
```

Fig: Python script to ignore depreciation warnings

### **Findings**

1. Store Number '20' has the highest Weekly Sales across all 45 Walmart Stores.
2. Store Number '33' has the lowest Weekly Sales across all 45 Walmart Stores.
3. The Stores experience more Weekly Sales when it is a non-holiday week (0).
4. Q4 (October - December) Weekly Sales increases are experienced across all stores.

### **Recommendations**

1. Increase inventories during the Q4 - fourth quarter to match the increasing sales.
2. Increase Inventories on days that are marked as non-holiday weeks as compared to holiday weeks.