```
# Import necessary libraries
import pandas as pd
import seaborn as sns
import numpy as np
import matplotlib.pyplot as plt
from matplotlib import dates
from datetime import datetime
```

Business Understanding

The objective of this analysis is to provide insights into the sales performance of Walmart across different cities and branches, understand the average price of items sold at each branch, and identify focus areas to improve sales for April 2019..



In this Assignment, we focused to answer the following questions:

A. Analyze the performance of sales and revenue at the city and branch level .

B. What is the average price of an item sold at each branch of the city.

C. Analyze the performance of sales and revenue, Month over Month across the Product line, Gender, and Payment Method, and identify the focus areas to get better sales for April 2019.

Data Understanding

The Walmart sales dataset comprises information on sales transactions recorded over a period of time. Key aspects of understanding the dataset include:

- 1. Data Composition:
- The dataset likely consists of rows and columns, with each row representing a unique sales transaction and each column representing a different attribute or feature of the transaction.
- Common columns may include:
- Invoice ID: Unique identifier for each transaction.
- Branch: Identifier for the branch of Walmart where the transaction occurred.
- City: The city where the Walmart branch is located.
- Customer type: Type of customer (e.g., member, non-member).
- Gender: Gender of the customer.
- Product line: Category or type of product purchased. Unit price: Price of a single unit of the product.
- Quantity: Number of units of the product purchased.
- Date: Date of the transaction.
- Time: Time of the transaction.
- Payment: Method of payment used for the transaction.
- Rating: Customer rating or satisfaction score for the transaction.
- 1. Data Quality Issues:
- Missing Values: Check for missing values in any of the columns, which could impact the analysis.
- Data Consistency: Ensure consistency in data formats across columns (e.g., date format, gender categories).
- Outliers: Identify any unusual or extreme values that may skew the analysis.
- Data Errors: Look for any errors or inconsistencies in the data entries (e.g., negative quantities, invalid payment methods).

```
# Load dataset
df = pd.read_excel("Walmart_Sales.xlsx")
df['Date'] = pd.to_datetime(df['Date'])
```

Data Preparation

```
1
     Branch
                     1000 non-null
                                      object
 2
     City
                     1000 non-null
                                      object
 3
     Customer type
                    1000 non-null
                                      object
 4
                     1000 non-null
     Gender
                                      object
 5
     Product line
                     1000 non-null
                                      object
 6
     Unit price
                     1000 non-null
                                      float64
 7
                     1000 non-null
     Quantity
                                      int64
 8
     Date
                     1000 non-null
                                      datetime64[ns]
 9
     Time
                     1000 non-null
                                      object
10
    Payment
                     1000 non-null
                                      object
 11
     Rating
                     1000 non-null
                                      float64
dtypes: datetime64[ns](1), float64(2), int64(1), object(8)
memory usage: 93.9+ KB
# checking for missing values
df.isnull().sum()
                 0
Invoice ID
                  0
Branch
                  0
City
Customer type
                  0
                  0
Gender
Product line
                  0
                  0
Unit price
Quantity
                  0
                  0
Date
Time
                  0
Payment
                  0
                 0
Rating
dtype: int64
# Splitting Date and create new columns (Day, Month, and Year)
df["Day"]= pd.DatetimeIndex(df['Date']).day
df['Month'] = pd.DatetimeIndex(df['Date']).month
df['Year'] = pd.DatetimeIndex(df['Date']).year
df
      Invoice ID Branch
                               City Customer type
                                                    Gender \
0
     750-67-8428
                                            Member
                                                    Female
                       Α
                             Yangon
1
     226-31-3081
                       Α
                          Naypyitaw
                                            Normal
                                                    Female
2
     631-41-3108
                       Α
                             Yangon
                                            Normal
                                                      Male
3
     123-19-1176
                       В
                             Yangon
                                            Member
                                                      Male
4
                       C
     373-73-7910
                             Yangon
                                            Normal
                                                      Male
                                               . . .
995
     233-67-5758
                          Naypyitaw
                                                      Male
                       Α
                                            Normal
996
     303-96-2227
                       Α
                           Mandalay
                                            Normal
                                                    Female
997
     727-02-1313
                       Α
                             Yangon
                                            Member
                                                      Male
                       В
998
     347-56-2442
                             Yangon
                                            Normal
                                                      Male
999
     849-09-3807
                       C
                             Yangon
                                            Member
                                                    Female
```

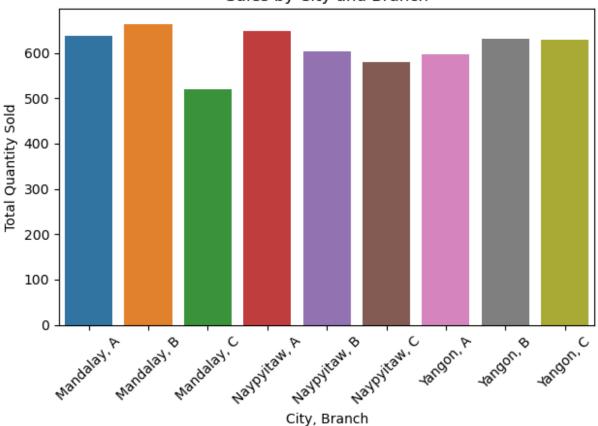
	Product 1	ine U	Init pri	ce Q	uantity	Date	Time
0	Health and bea	uty	74.	69	7	2019-01-05	13:08:00
1	Electronic accessor	ies	15.	28	5	2019-03-08	10:29:00
2	Home and lifestyle			33	7	2019-03-03	13:23:00
3	Health and bea	uty	58.	22	8	2019-01-27	20:33:00
4	Sports and tra	ivel	86.	31	7	2019-02-08	10:37:00
995	Health and bea	uty	40.	35	1	2019-01-29	13:46:00
996	Home and lifestyle			38	10	2019-03-02	17:16:00
997	Food and beverages		31.	84	1	2019-02-09	13:22:00
998	Home and lifest	yle	65.	82	1	2019-02-22	15:33:00
999	Fashion accessor	ies	88.	34	7	2019-02-18	13:28:00
0 1 2 3 4 995 996 997 998 999	Payment Rating Ewallet 9.1 Cash 9.6 Credit card 7.4 Ewallet 8.4 Ewallet 5.3 Ewallet 6.2 Ewallet 4.4 Cash 7.7 Cash 6.6	5 8 3 27 8 8 29 2 9	Month	Year 2019 2019 2019 2019 2019 2019 2019 2019			
[1000 rows x 15 columns]							

A. Analyze the performance of sales and revenue at the city and branch level

```
sales_revenue_city_branch = df.groupby(['City',
'Branch']).agg({'Quantity': 'sum', 'Unit price': 'sum'})
print("Performance of sales and revenue at the city and branch
level:")
print(sales_revenue_city_branch)
```

```
Performance of sales and revenue at the city and branch level:
                  Quantity Unit price
City
          Branch
Mandalay
                       637
                               6349.11
          Α
                       664
                               6623.73
                       519
                               5506.04
Naypyitaw A
                       648
                               5953.55
                       604
                               6298.64
          В
          C
                       579
                               6315.57
Yangon
          Α
                       598
                               6342.88
                               6329.25
          В
                       631
          C
                       630
                               5953.36
# Combine 'City' and 'Branch' into a single column for x-axis
sales revenue city branch.reset index(inplace=True)
sales revenue city branch['City Branch'] =
sales revenue city branch['City'] + ', ' +
sales revenue city branch['Branch']
# Plot
sns.barplot(x='City Branch', y='Quantity',
data=sales_revenue_city_branch)
plt.title('Sales by City and Branch')
plt.xlabel('City, Branch')
plt.ylabel('Total Quantity Sold')
plt.xticks(rotation=45)
plt.tight_layout() # Adjust layout to prevent overlapping labels
plt.show()
```



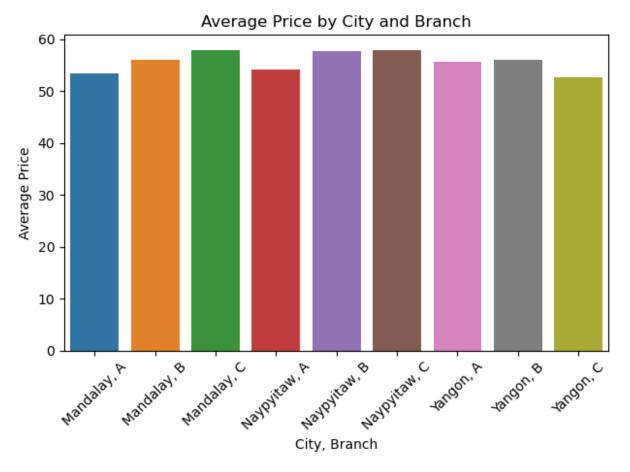


B. What is the average price of an item sold at each branch of the city

```
avg_price_branch_city = df.groupby(['City', 'Branch']).agg({'Unit
price': 'mean'})
print("\nAverage price of an item sold at each branch of the city:")
print(avg price branch city)
Average price of an item sold at each branch of the city:
                  Unit price
City
          Branch
Mandalay
                   53.353866
          В
                   56.133305
          C
                   57.958316
Naypyitaw A
                   54.123182
          В
                   57.785688
                   57.941009
          C
Yangon
          Α
                   55.639298
          В
                   56.011062
          C
                   52.684602
# Combine 'City' and 'Branch' into a single column for x-axis
avg price branch city.reset index(inplace=True)
```

```
avg_price_branch_city['City_Branch'] = avg_price_branch_city['City'] +
', ' + avg_price_branch_city['Branch']

# Plot
sns.barplot(x='City_Branch', y='Unit price',
data=avg_price_branch_city)
plt.title('Average Price by City and Branch')
plt.xlabel('City, Branch')
plt.ylabel('Average Price')
plt.ylabel('Average Price')
plt.xticks(rotation=45)
plt.tight_layout() # Adjust layout to prevent overlapping labels
plt.show()
```



C. Analyze the performance of sales and revenue, Month over Month
across Product line, Gender, and Payment Method
Focus areas to improve sales for April 2019.

C. Analyze the performance of sales and revenue, Month over Month
across Product line, Gender, and Payment Method
Focus areas to improve sales for April 2019

Extract month and year from the 'Date' column

```
df['Month'] = pd.to_datetime(df['Date']).dt.month
df['Year'] = pd.to_datetime(df['Date']).dt.year

# Filter data for April 2019
april_2019_data = df[(df['Month'] == 4) & (df['Year'] == 2019)]

# Group by Product line, Gender, and Payment Method
april_sales_analysis = april_2019_data.groupby(['Product line', 'Gender', 'Payment']).agg({'Quantity': 'sum', 'Unit price': 'sum'})
print("\nAnalysis of sales and revenue for April 2019:")
print(april_sales_analysis)

Analysis of sales and revenue for April 2019:
Empty DataFrame
Columns: [Quantity, Unit price]
Index: []
```

2. App Exploration:

Explore the features and user experience of the Jar app. Identify two aspects that you think could be significantly improved and explain your reasoning behind each suggestion.

Explanation For App Exploration

""" Explore the features and user experience of the Jar app:

The Jar app is designed to simplify expense and savings management by allowing users to allocate funds into different "jars" or categories. Users can set budgets, track spending, and visualize financial goals. However, there are two key areas where the app could significantly improve:

- 1. Streamlined User Onboarding:
 - The current onboarding process may overwhelm new users, potentially leading to drop-offs. To enhance user experience, the app could streamline onboarding by breaking it into smaller, manageable steps. Providing clear instructions and guidance at each stage can help users set up their accounts and navigate the app efficiently. Additionally, integrating interactive tutorials or tooltips can further assist users in understanding the app's features.
- 2. Enhanced Budgeting Features:
 - While the Jar app offers budgeting functionality, there's room for improvement to make it more versatile and personalized. Introducing customizable budget categories would empower users to create specific budgets for various expenses or savings objectives. Moreover, incorporating predictive budgeting, where the app analyzes past spending patterns to propose future budgets, can facilitate proactive financial management. Furthermore, integrating notifications or alerts to notify users when they approach or exceed budget limits can encourage better financial decision-making.

Addressing these aspects will enhance user satisfaction, retention, and overall usability of the Jar app, resulting in a more positive user experience. """

3. Product Optimisation:

The Jar app has an engagement feature called 'Spin to Win'. Right now, if 100 people come to the app each day, only 23 of them try out this spinning game. But, we know that people who spin are more likely to retain on the app and do transactions. Now, we want to get more people to play the game. So, the question is, how can we make sure that at least 50 people out of every 100 who visit the app each day will play 'Spin to Win'? What can we do to get more people interested in spinning the wheel?

Explanation For Product Optimisation:

""" Product Optimization:

The Jar app's 'Spin to Win' feature currently has a participation rate of only 23% out of 100 daily visitors. However, engaging with this feature has shown to increase user retention and transactions. To boost participation to at least 50%, the following strategies can be implemented:

- 1. Enhance Rewards and Incentives:
 - Offer more enticing rewards such as discounts, exclusive deals, loyalty points, or big prizes. Valuable rewards will motivate users to play the 'Spin to Win' game.
- 2. Promotional Campaigns:
 - Launch targeted marketing campaigns via email, push notifications, and social media to raise awareness about the 'Spin to Win' feature. Emphasize the benefits and rewards to attract more users.
- 3. Improve User Interface and Experience:
 - Enhance the visual appeal and interactivity of the 'Spin to Win' feature with smooth spinning mechanisms, attractive graphics, and clear instructions. A seamless user experience will encourage more users to participate.
- 4. Personalization and Customization:
 - Personalize the 'Spin to Win' experience based on user preferences, behavior, and demographics. Offer customized rewards and allow users to tailor their spinning experience.
- 5. Limited-Time Offers and Events:
 - Introduce time-sensitive promotions, themed events, or flash sales associated with the 'Spin to Win' game to create a sense of urgency and exclusivity. Limitedtime offers can drive higher participation rates.
- Social Sharing and Virality:
 - Enable social sharing functionalities to encourage users to share their winnings and experiences on social media platforms. Incentivize sharing and leverage usergenerated content to amplify visibility.

Implementing these strategies will help the Jar app achieve its goal of increasing participation in the 'Spin to Win' game, leading to greater user engagement and retention.