

```
In [ ]: import numpy as np
import pandas as pd

import matplotlib.pyplot as plt
import seaborn as sns
```

## Level 1: Data Loading and Basic Analysis

### 1. Load the data into pandas dataframes

```
In [1]: df_sales=pd.read_csv("given_sales.csv")

df_customer=pd.read_csv("given_customers.csv")

df_regions=pd.read_csv("given_regions.csv")

df_products=pd.read_csv("given_products.csv")
```

```
In [5]: df_sales
```

```
Out[5]:
```

	sales_id	product_id	customer_id	sales_amount	date	region_id
0	1	101	201	1000	01-01-2022	1
1	2	102	202	1500	02-01-2022	2
2	3	103	203	800	03-01-2022	1
3	4	104	204	1200	04-01-2022	3
4	5	105	205	2000	05-01-2022	1
...	...	...	...	...	...	...
95	96	196	296	1250	06-04-2022	9
96	97	197	297	850	07-04-2022	10
97	98	198	298	1950	08-04-2022	1
98	99	199	299	1300	09-04-2022	2
99	100	200	300	1750	10-04-2022	3

100 rows x 6 columns

```
In [6]: df_products
```

Out [6]:

	product_id	product_name	category	price
0	101	Product A	Electronics	500
1	102	Product B	Clothing	800
2	103	Product C	Home Decor	300
3	104	Product D	Electronics	1000
4	105	Product E	Beauty	1500
...	...	...	...	...
95	196	Product CR	Sports	850
96	197	Product CS	Clothing	650
97	198	Product CT	Electronics	1400
98	199	Product CU	Home Decor	1900
99	200	Product CV	Beauty	2400

100 rows × 4 columns

In [7]: df\_customer

Out [7]:

	customer_id	customer_name	email	address
0	201	John Doe	john@example.com	123 Main St, Anytown, USA
1	202	Jane Smith	jane@example.com	456 Elm St, Othertown, USA
2	203	Robert Brown	robert@example.com	789 Oak St, Anycity, USA
3	204	Lisa Johnson	lisa@example.com	321 Maple St, Somewhere, USA
4	205	Michael Wilson	michael@example.com	654 Pine St, Nowhere, USA
...	...	...	...	...
95	296	Amelia Foster	amelia@example.com	366 Oak St, Nowhere, USA
96	297	Oliver Reyes	oliver@example.com	586 Maple St, Anytown, USA
97	298	Sophia Gray	sophia@example.com	747 Pine St, Othertown, USA
98	299	Elijah Bryant	elijah@example.com	983 Cedar St, Somewhere, USA
99	300	Harper Watson	harper@example.com	624 Elm St, Nowhere, USA

100 rows × 4 columns

In [8]: df\_regions

Out [8]:

	region_id	region_name
0	1	East Coast
1	2	West Coast
2	3	Midwest
3	4	South
4	5	Northeast
...	...	...
95	96	West Coast
96	97	Midwest
97	98	South
98	99	Northeast
99	100	Southwest

100 rows × 2 columns

In [9]:

```
merged_df = pd.merge(df_sales, df_products, on='product_id')  
  
merged_df=pd.merge(merged_df, df_customer, on='customer_id')  
  
merged_df=pd.merge(merged_df, df_regions, on='region_id')
```

In [12]:

```
merged_df
```

Out [12]:

	sales_id	product_id	customer_id	sales_amount	date	region_id	product_name	category
0	1	101	201	1000	01-01-2022	1	Product A	Electronics
1	3	103	203	800	03-01-2022	1	Product C	Home Decor
2	5	105	205	2000	05-01-2022	1	Product E	Beauty
3	8	108	208	3000	08-01-2022	1	Product H	Home Decor
4	11	111	211	1300	11-01-2022	1	Product K	Sports
...	...	...	...	...	...	...	...	...
95	57	157	257	800	26-02-2022	10	Product BE	Clothing
96	67	167	267	850	08-03-2022	10	Product BO	Clothing
97	77	177	277	850	18-03-2022	10	Product BY	Clothing
98	87	187	287	850	28-03-2022	10	Product CI	Clothing
99	97	197	297	850	07-04-2022	10	Product CS	Clothing

100 rows × 13 columns

2. Display the first 5 rows of the database

In [13]:

```
merged_df.head(5)
```

Out [13]:

	sales_id	product_id	customer_id	sales_amount	date	region_id	product_name	category
0	1	101	201	1000	01-01-2022	1	Product A	Electronics
1	3	103	203	800	03-01-2022	1	Product C	Home Decor
2	5	105	205	2000	05-01-2022	1	Product E	Beauty
3	8	108	208	3000	08-01-2022	1	Product H	Home Decor
4	11	111	211	1300	11-01-2022	1	Product K	Sports

3. Check the shape of the database

In [14]: merged\_df.shape

Out [14]: (100, 13)

4. Display basic statistics (mean, median, min, max, etc.) for Sales

In [15]: merged\_df.describe(include='all')

Out [15]:

	sales_id	product_id	customer_id	sales_amount	date	region_id	product_name
count	100.000000	100.000000	100.000000	100.000000	100	100.000000	100
unique	NaN	NaN	NaN	NaN	100	NaN	98
top	NaN	NaN	NaN	NaN	01-01-2022	NaN	Product CC
freq	NaN	NaN	NaN	NaN	1	NaN	2
mean	50.500000	150.500000	250.500000	1499.000000	NaN	4.780000	NaN
std	29.011492	29.011492	29.011492	510.593831	NaN	2.983388	NaN
min	1.000000	101.000000	201.000000	600.000000	NaN	1.000000	NaN
25%	25.750000	125.750000	225.750000	1100.000000	NaN	2.000000	NaN
50%	50.500000	150.500000	250.500000	1500.000000	NaN	4.000000	NaN
75%	75.250000	175.250000	275.250000	1850.000000	NaN	7.000000	NaN
max	100.000000	200.000000	300.000000	3000.000000	NaN	10.000000	NaN

In [16]: merged\_df.describe()

Out[16]:

	sales_id	product_id	customer_id	sales_amount	region_id	price
count	100.000000	100.000000	100.000000	100.000000	100.000000	100.000000
mean	50.500000	150.500000	250.500000	1499.000000	4.780000	1366.000000
std	29.011492	29.011492	29.011492	510.593831	2.983388	591.68822
min	1.000000	101.000000	201.000000	600.000000	1.000000	300.000000
25%	25.750000	125.750000	225.750000	1100.000000	2.000000	887.500000
50%	50.500000	150.500000	250.500000	1500.000000	4.000000	1350.000000
75%	75.250000	175.250000	275.250000	1850.000000	7.000000	1800.000000
max	100.000000	200.000000	300.000000	3000.000000	10.000000	2500.000000

## 5. Determine the number of unique products sold

In [26]:

```
# From above observation we can see that sales_amount,region_id,price have almost same median
# data is normalized and not skewed
```

In [17]:

```
merged_df.product_name.unique()
```

Out[17]:

```
array(['Product A', 'Product C', 'Product E', 'Product H', 'Product K',
      'Product N', 'Product Q', 'Product T', 'Product BB', 'Product LL',
      'Product AV', 'Product BF', 'Product BP', 'Product BZ',
      'Product CJ', 'Product CT', 'Product B', 'Product F', 'Product I',
      'Product L', 'Product O', 'Product R', 'Product CC', 'Product MM',
      'Product AW', 'Product BG', 'Product BQ', 'Product CA',
      'Product CK', 'Product CU', 'Product D', 'Product G', 'Product J',
      'Product M', 'Product P', 'Product S', 'Product DD', 'Product NN',
      'Product AX', 'Product BH', 'Product BR', 'Product CB',
      'Product CL', 'Product CV', 'Product U', 'Product EE',
      'Product AO', 'Product AY', 'Product BI', 'Product BS',
      'Product CM', 'Product V', 'Product FF', 'Product AP',
      'Product AZ', 'Product BJ', 'Product BT', 'Product CD',
      'Product CN', 'Product W', 'Product GG', 'Product AQ',
      'Product BA', 'Product BK', 'Product BU', 'Product CE',
      'Product CO', 'Product X', 'Product HH', 'Product AR',
      'Product BL', 'Product BV', 'Product CF', 'Product CP',
      'Product Y', 'Product II', 'Product AS', 'Product BC',
      'Product BM', 'Product BW', 'Product CG', 'Product CQ',
      'Product Z', 'Product JJ', 'Product AT', 'Product BD',
      'Product BN', 'Product BX', 'Product CH', 'Product CR',
      'Product AA', 'Product KK', 'Product AU', 'Product BE',
      'Product BO', 'Product BY', 'Product CI', 'Product CS'],
      dtype=object)
```

In [18]:

```
merged_df['product_name'].value_counts()
```

```
Out[18]: Product CC      2
Product BB      2
Product A       1
Product BK      1
Product CF      1
..
Product G       1
Product D       1
Product CU      1
Product CK      1
Product CS      1
Name: product_name, Length: 98, dtype: int64
```

## Level 2: Data Cleaning and Preprocessing

1. Handle missing values by filling them with appropriate values (e.g., mean or median).
2. Convert categorical variables to numerical representation using one-hot encoding or label

encoding. 3. Check for and remove any duplicate rows in the dataset. 4. Normalize the 'Sales' column to a scale between 0 and 1. 5. Identify and remove outliers from the dataset using appropriate techniques.

## Level 2: Data Cleaning and Pre-processing

### 1. Handle missing values by filling them with appropriate values (e.g., mean or median)

```
In [19]: merged_df.isnull().sum()
```

```
Out[19]: sales_id      0
product_id    0
customer_id    0
sales_amount   0
date          0
region_id     0
product_name   0
category       0
price         0
customer_name  0
email         0
address       0
region_name    0
dtype: int64
```

### 2. Convert categorical variables to numerical representation using one-hot encoding or label encoding.

```
In [20]: from sklearn.preprocessing import LabelEncoder

encoder=LabelEncoder()
```

```
In [21]: df1=merged_df.copy()
```

```
In [22]: columns=['product_name','category','region_name']
for col in columns:
    encoded=encoder.fit_transform(df1[col])
    df1[col+'_encoded']=encoded
```

In [200...]

df1.head()

Out[200]:

	sales_id	product_id	customer_id	sales_amount	date	region_id	product_name	category
0	1	101	201	1000	01-01-2022	1	Product A	Electronics
1	3	103	203	800	03-01-2022	1	Product C	Home Decor
2	5	105	205	2000	05-01-2022	1	Product E	Beauty
3	8	108	208	3000	08-01-2022	1	Product H	Home Decor
4	11	111	211	1300	11-01-2022	1	Product K	Sports

3. Check for and remove any duplicate rows in the database

In [24]:

```
duplicate_val=df1.duplicated()
duplicate_val.value_counts()
#no duplicate
#incase duplicate is there we can remove using following
# df.drop_duplicates(inplace=True)
```

Out[24]:

```
False      100
dtype: int64
```

In [25]:

```
df1
```



Out [25]:

	sales_id	product_id	customer_id	sales_amount	date	region_id	product_name	category
0	1	101	201	1000	01-01-2022	1	Product A	Electronics
1	3	103	203	800	03-01-2022	1	Product C	Home Decor
2	5	105	205	2000	05-01-2022	1	Product E	Beauty
3	8	108	208	3000	08-01-2022	1	Product H	Home Decor
4	11	111	211	1300	11-01-2022	1	Product K	Sports
...	...	...	...	...	...	...	...	...
95	57	157	257	800	26-02-2022	10	Product BE	Clothing
96	67	167	267	850	08-03-2022	10	Product BO	Clothing
97	77	177	277	850	18-03-2022	10	Product BY	Clothing
98	87	187	287	850	28-03-2022	10	Product CI	Clothing
99	97	197	297	850	07-04-2022	10	Product CS	Clothing

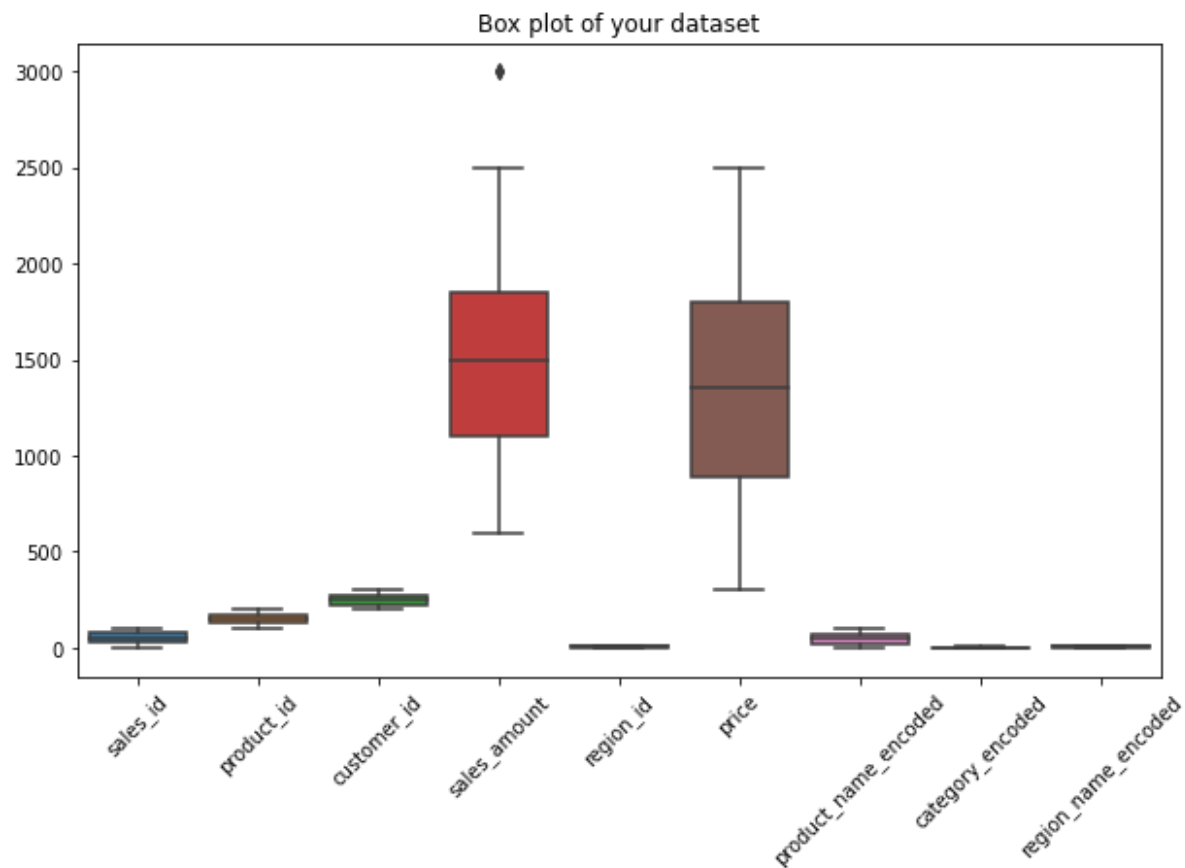
100 rows x 16 columns

4. Normalize the 'Sales' column to a scale between 0 and 1.

5. Identify and remove outliers from the database using appropriate techniques.

In [29]:

```
plt.figure(figsize=(10, 6))
sns.boxplot(data=df1)
plt.title('Box plot of your dataset')
plt.xticks(rotation=45) # Rotate x-axis labels for better readability
plt.show()
```



```
In [30]: # we can see that there is outlier present n the sales amount column
```

```
In [32]: Q1=df1['sales_amount'].quantile(0.25)
Q3=df1['sales_amount'].quantile(0.75)
IQR=Q3-Q1
```

```
In [38]: lower_bound = 0
upper_bound = Q3 + 1.5 * IQR
```

```
In [40]: upper_bound
```

```
Out[40]: 2975.0
```

```
In [45]: cleaned_sales=df1[(df1['sales_amount']>lower_bound )& (df1['sales_amount']<upper_bound)]
cleaned_sales
```

Out [45]:

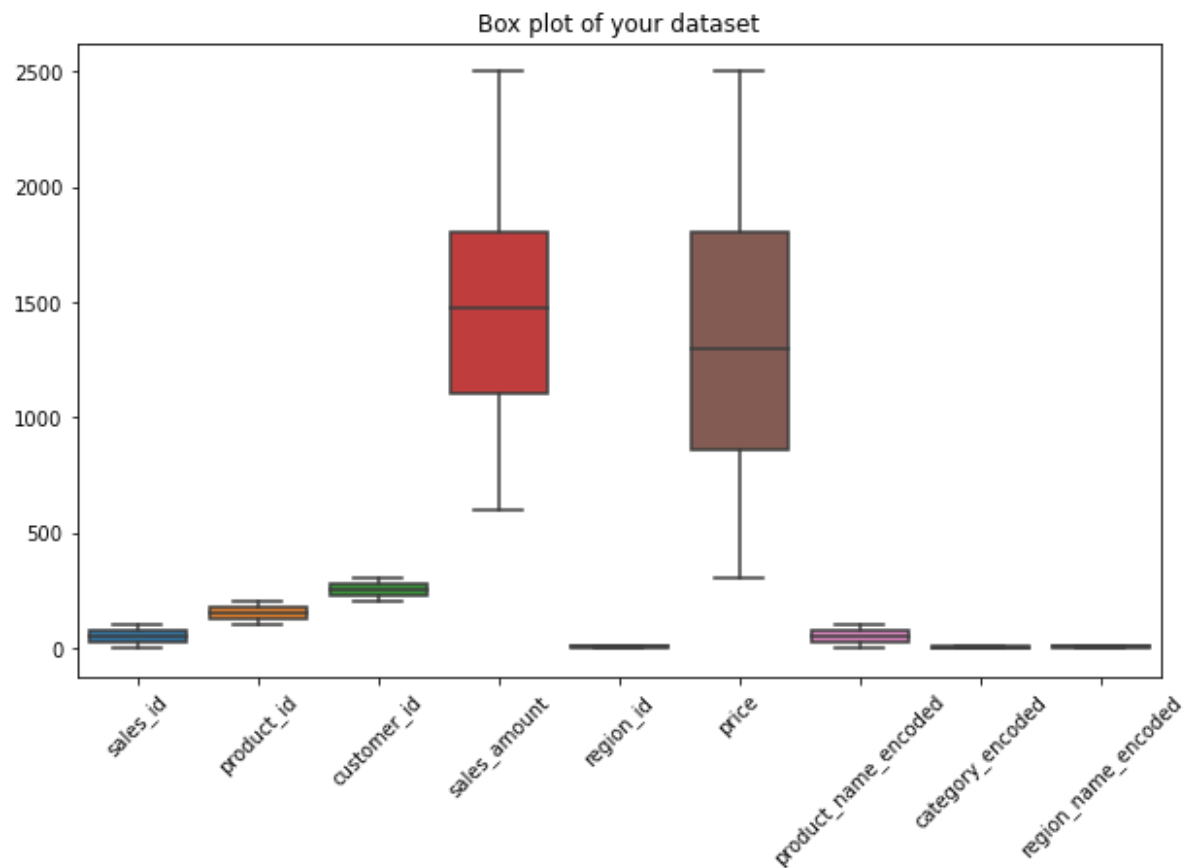
	sales_id	product_id	customer_id	sales_amount	date	region_id	product_name	category	
	0	1	101	201	1000	01-01-2022	1	Product A	Electronics
	1	3	103	203	800	03-01-2022	1	Product C	Home Decor
	2	5	105	205	2000	05-01-2022	1	Product E	Beauty
	4	11	111	211	1300	11-01-2022	1	Product K	Sports
	5	14	114	214	1700	14-01-2022	1	Product N	Home Decor
	...	...	...	...	...	...	...	...	...
	95	57	157	257	800	26-02-2022	10	Product BE	Clothing
	96	67	167	267	850	08-03-2022	10	Product BO	Clothing
	97	77	177	277	850	18-03-2022	10	Product BY	Clothing
	98	87	187	287	850	28-03-2022	10	Product CI	Clothing
	99	97	197	297	850	07-04-2022	10	Product CS	Clothing

98 rows x 16 columns

In [46]:

```
import matplotlib.pyplot as plt
import seaborn as sns
plt.figure(figsize=(10, 6))
sns.boxplot(data=cleaned_sales)
plt.title('Box plot of your dataset')
plt.xticks(rotation=45) # Rotate x-axis labels for better readability
plt.show()

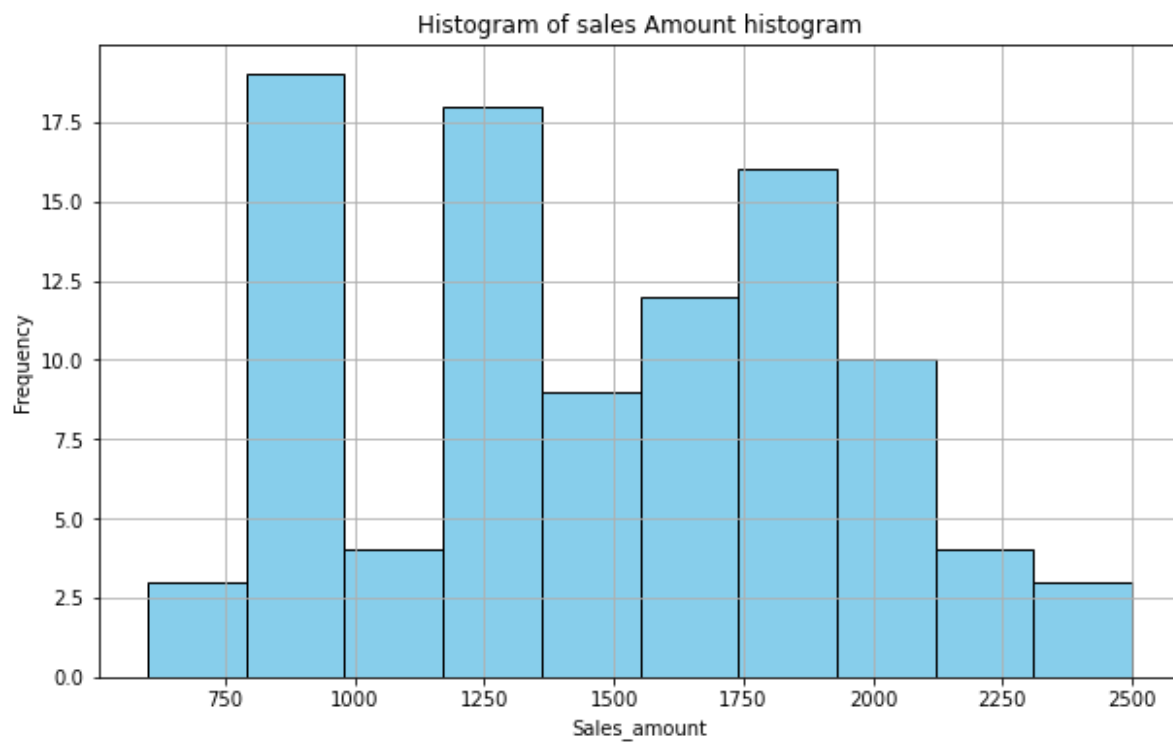
# we can see that two rows has been removed and the two outlier values has been removed
```



## Level 3: Exploratory Data Analysis (EDA)

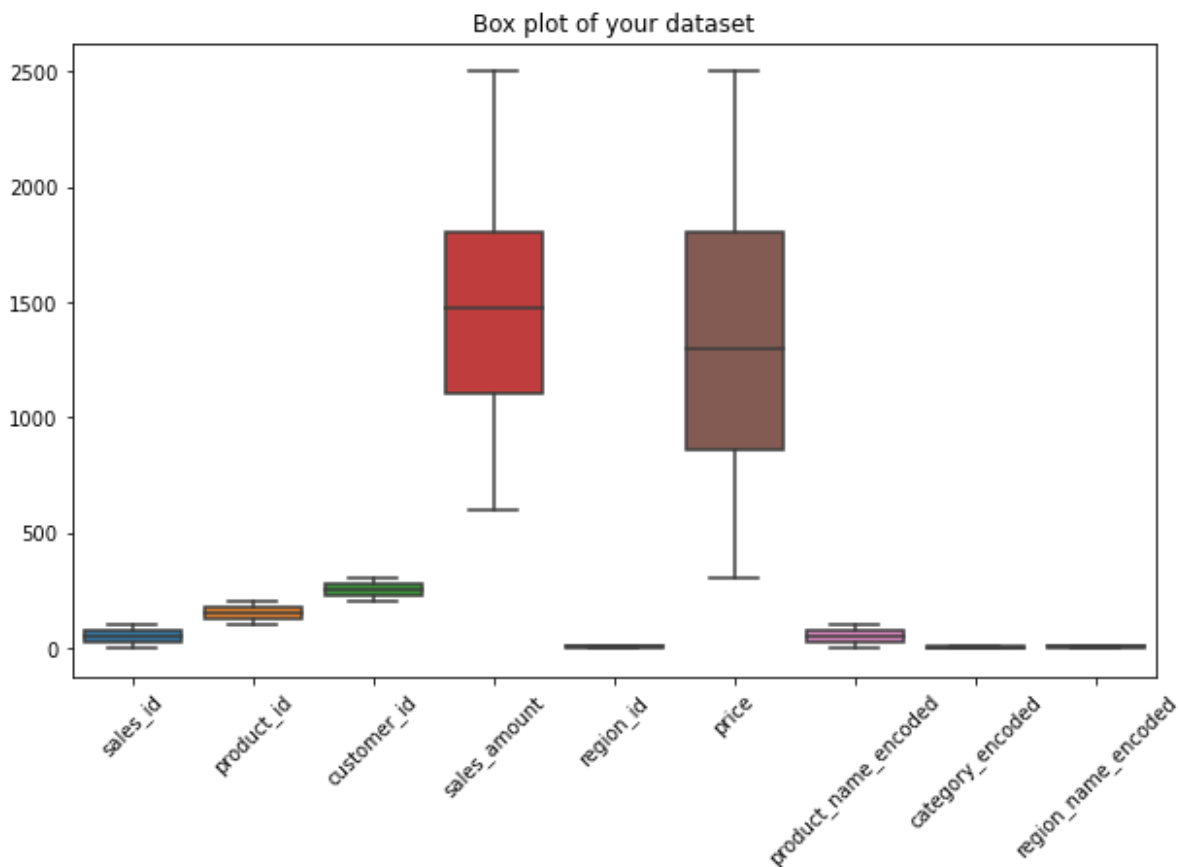
### 1. Visualize the distribution of sales amounts using a histogram

```
In [182... plt.figure(figsize=(10, 6))
plt.hist(df['sales_amount'], bins=10, color='skyblue', edgecolor='black') # Adjust the number of bins
plt.title('Histogram of {}'.format('sales Amount histogram'))
plt.xlabel('Sales_amount')
plt.ylabel('Frequency')
plt.grid(True)
plt.show()
```



2. Create a boxplot to identify any outliers in the sales data.

```
In [183]: import matplotlib.pyplot as plt
import seaborn as sns
plt.figure(figsize=(10, 6))
sns.boxplot(data=cleaned_sales)
plt.title('Box plot of your dataset')
plt.xticks(rotation=45) # Rotate x-axis labels for better readability
plt.show()
```



### 3. Explore the relationship between sales and other variables using scatter plots.

In [186...

```
# Define the required columns for scatter plots
required_columns = ['date', 'region_id', 'product_name', 'category', 'region_name']

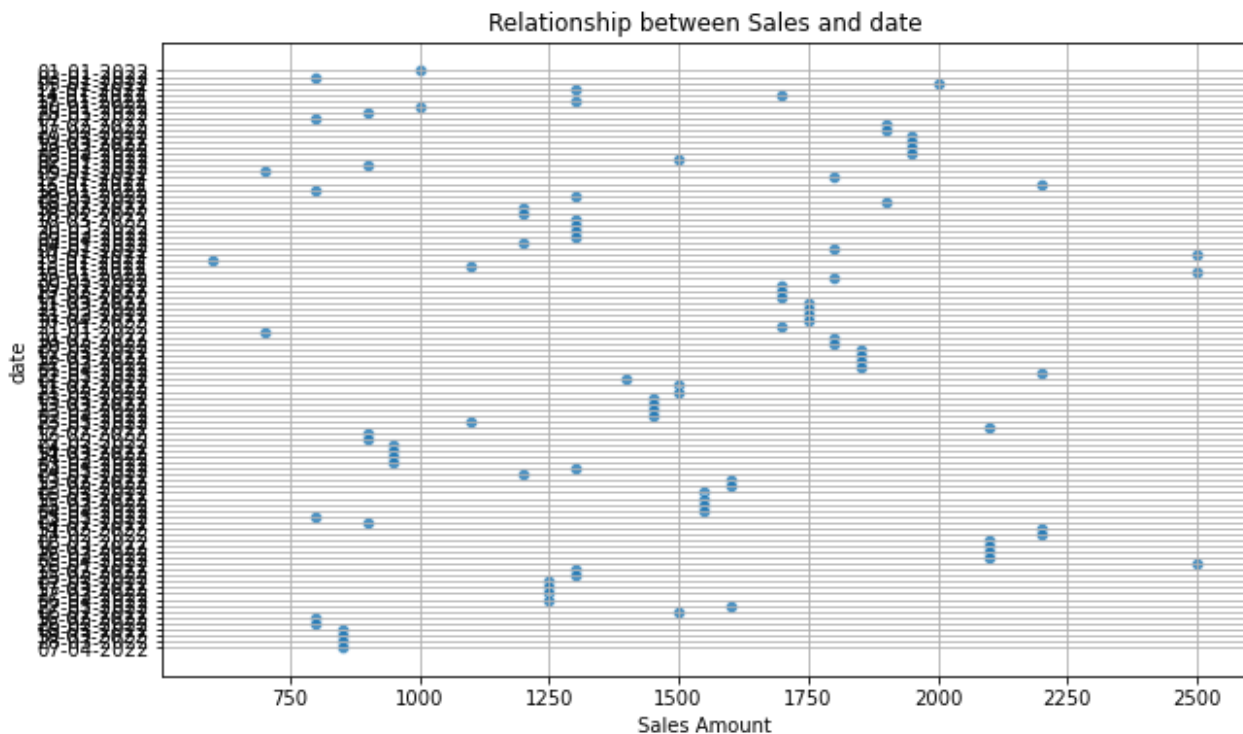
# Create scatter plots for each combination of sales and other variables
for column in required_columns:
    plt.figure(figsize=(10, 6))

    # Scatter plot between sales and the current variable
    sns.scatterplot(data=cleaned_sales, x='sales_amount', y=column, alpha=0.8)

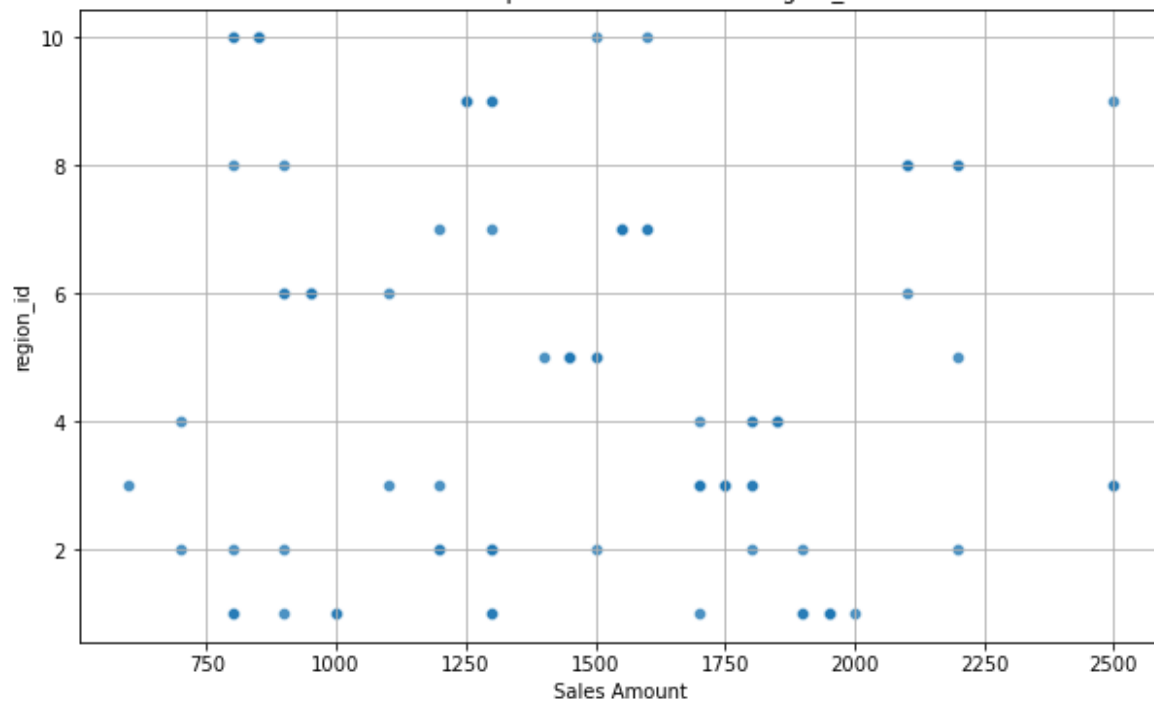
    # Set plot labels and title
    plt.title(f'Relationship between Sales and {column}')
    plt.xlabel('Sales Amount')
    plt.ylabel(column)

    # Show grid
    plt.grid(True)

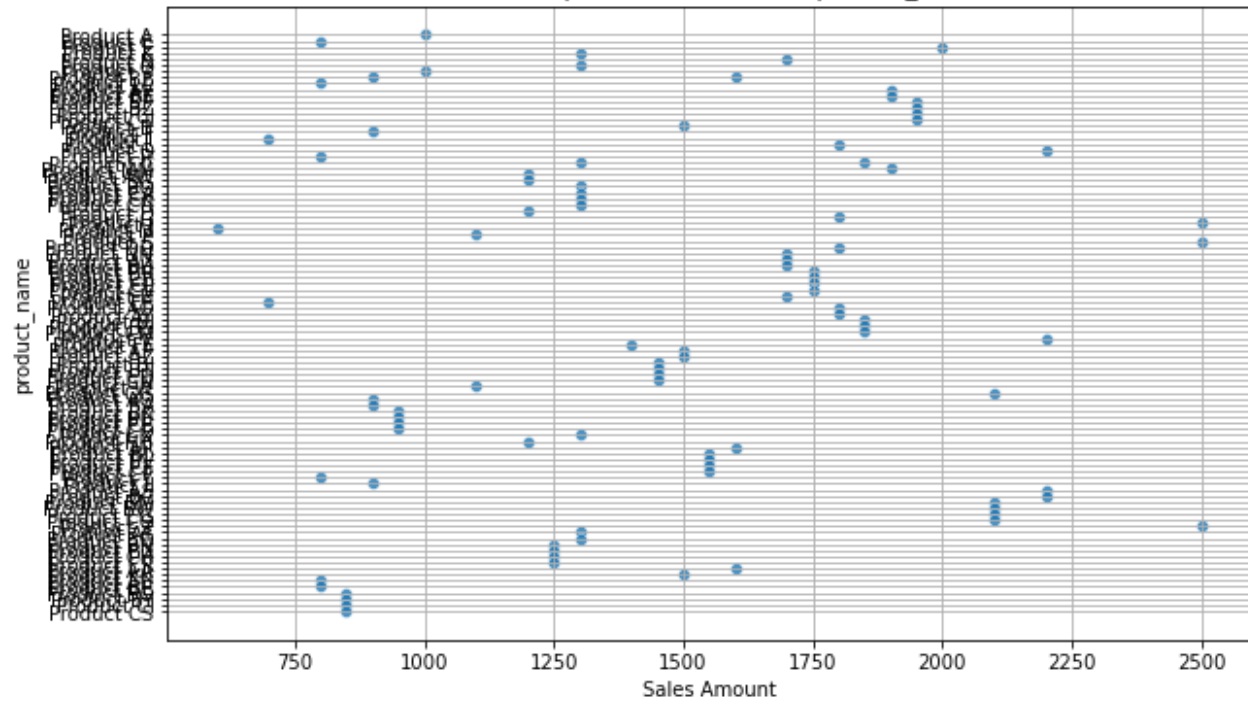
    # Show plot
    plt.show()
```

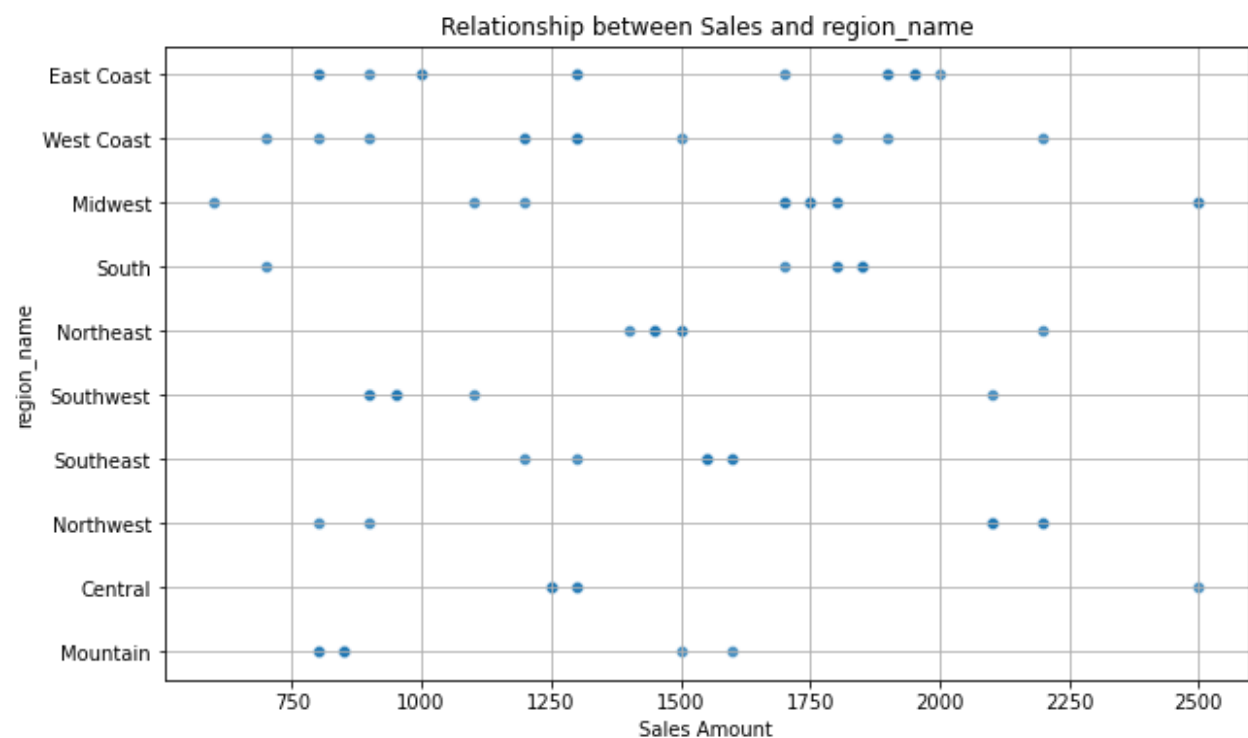
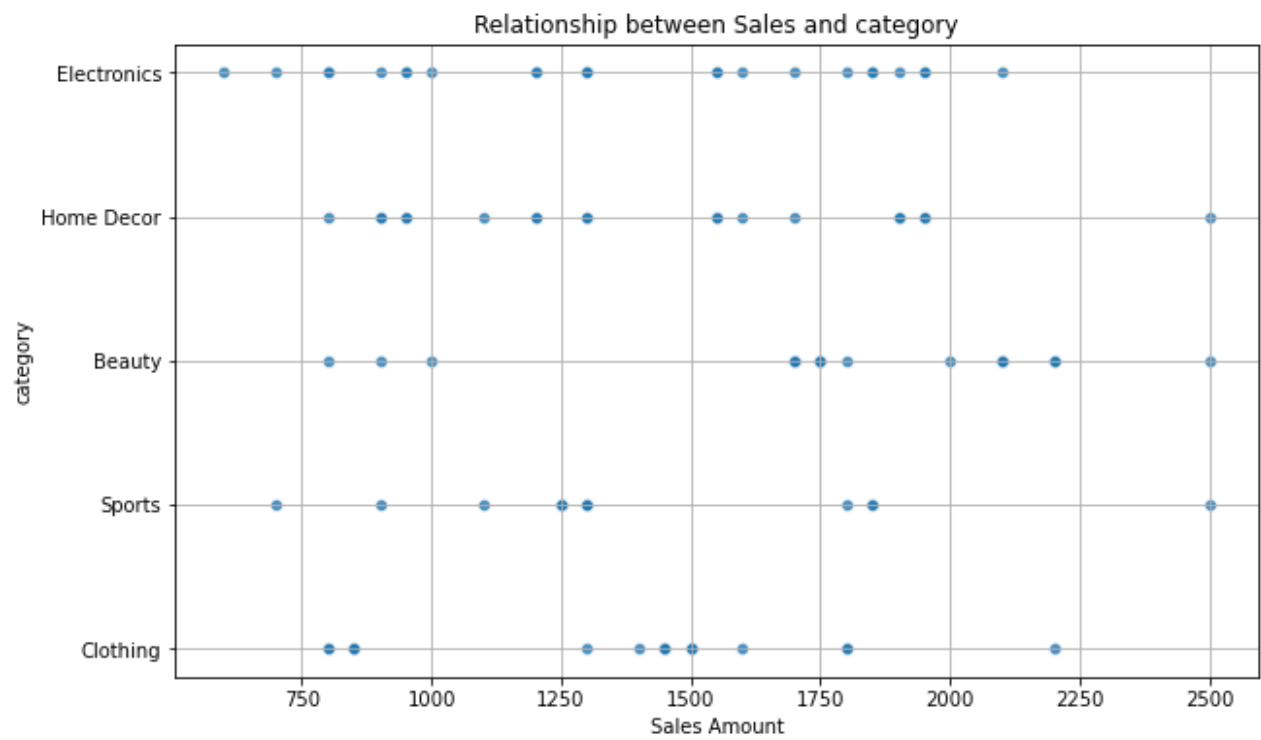


Relationship between Sales and region\_id



Relationship between Sales and product\_name





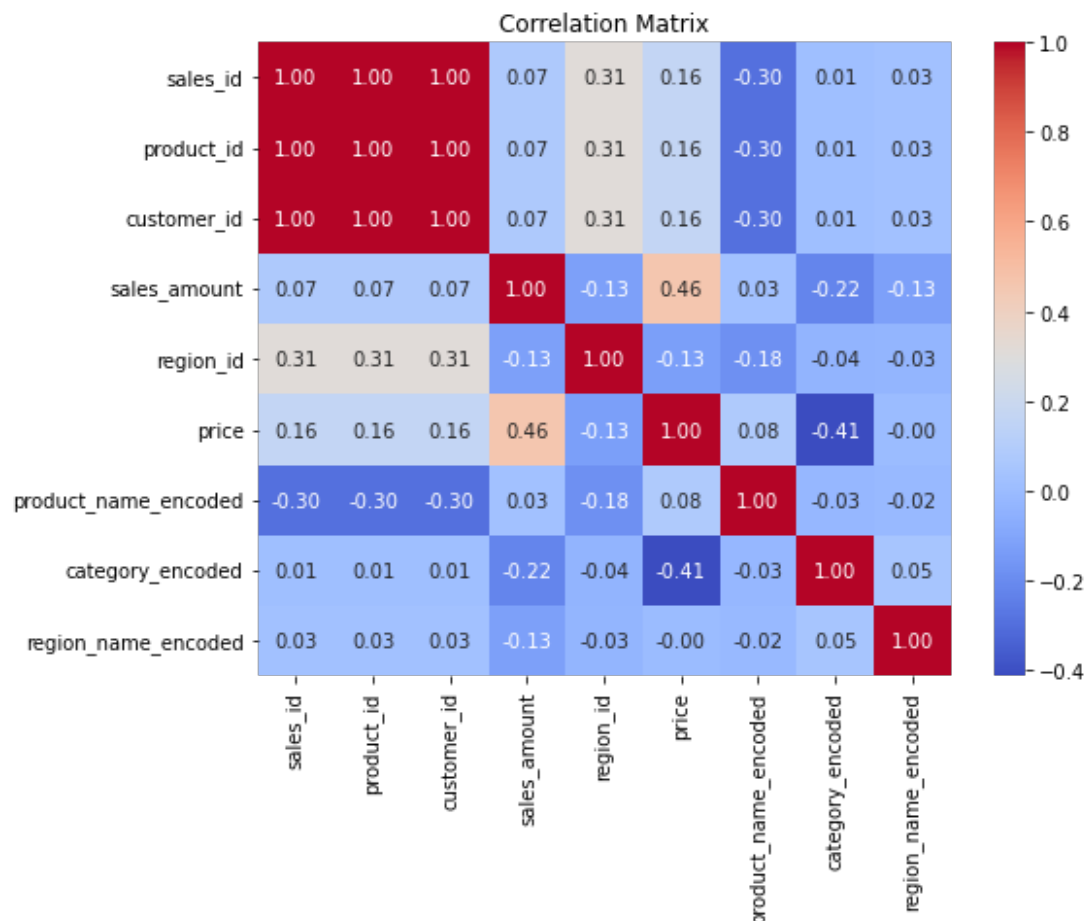
#### 4. Calculate and visualize the correlation matrix between numerical variables.

In [184...]

```
# Calculate the correlation matrix
correlation_matrix = cleaned_sales.corr()

# Visualize the correlation matrix using a heatmap
import seaborn as sns
plt.figure(figsize=(8, 6))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt=".2f")
plt.title('Correlation Matrix')
plt.show()
```





5. Analyze the sales trend over time using line plots or time series plots

## Level 4: Data Aggregation and Grouping

1. Group the sales data by product category and calculate the total sales amount for each category.

```
In [188]: category_sales = cleaned_sales.groupby('category')['sales_amount'].sum()
category_sales
```

```
Out[188]: category
Beauty          36100
Clothing        26900
Electronics     34100
Home Decor      27200
Sports          19600
Name: sales_amount, dtype: int64
```

2. Group the sales data by month and year and calculate the average sales amount for each month.

```
In [190]: sales_data=cleaned_sales.copy()
sales_data['date'] = pd.to_datetime(sales_data['date'])
sales_data['month'] = sales_data['date'].dt.to_period('M')
sales_data['year'] = sales_data['date'].dt.to_period('Y')
monthly_average_sales = sales_data.groupby('month')['sales_amount'].mean()
yearly_average_sales = sales_data.groupby('year')['sales_amount'].mean()
```

[illegible]

[illegible]

[illegible]

```

sales_data['date'] = pd.to_datetime(sales_data['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2032175852.py:3: UserWarning: Pars
ing '27-03-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True
for consistent parsing.
sales_data['date'] = pd.to_datetime(sales_data['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2032175852.py:3: UserWarning: Pars
ing '27-01-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True
for consistent parsing.
sales_data['date'] = pd.to_datetime(sales_data['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2032175852.py:3: UserWarning: Pars
ing '16-02-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True
for consistent parsing.
sales_data['date'] = pd.to_datetime(sales_data['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2032175852.py:3: UserWarning: Pars
ing '26-02-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True
for consistent parsing.
sales_data['date'] = pd.to_datetime(sales_data['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2032175852.py:3: UserWarning: Pars
ing '18-03-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True
for consistent parsing.
sales_data['date'] = pd.to_datetime(sales_data['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2032175852.py:3: UserWarning: Pars
ing '28-03-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True
for consistent parsing.
sales_data['date'] = pd.to_datetime(sales_data['date'])

```

```
In [191]: print(monthly_average_sales, yearly_average_sales)
```

```

month
2022-01    1436.956522
2022-02    1540.000000
2022-03    1415.217391
2022-04    1150.000000
2022-05    1883.333333
2022-06    1437.500000
2022-07    1175.000000
2022-08    1566.666667
2022-09    1412.500000
2022-10    1837.500000
2022-11    1516.666667
2022-12    1516.666667
Freq: M, Name: sales_amount, dtype: float64 year
2022      1468.367347
Freq: A-DEC, Name: sales_amount, dtype: float64

```

### 3. Aggregate the sales data by region and calculate the total sales amount for each region.

```
In [193]: region_sales = sales_data.groupby('region_name')['sales_amount'].sum()
region_sales
```

```

Out[193]: region_name
Central      10100
East Coast   22400
Midwest      23600
Mountain      8100
Northeast    12400
Northwest    14500
South        13400
Southeast    11900
Southwest     8800
West Coast   18700
Name: sales_amount, dtype: int64

```

4. Group the sales data by customer segment and calculate the average sales amount for each segment.

```
In [ ]: sales_data['customer_segment'] = sales_data['customer_name'].apply(lambda x: x.split()[0])
segment_average_sales = sales_data.groupby('customer_segment')['sales_amount'].mean()
```

5. Aggregate the sales data by sales representative and calculate the total sales amount for each representative

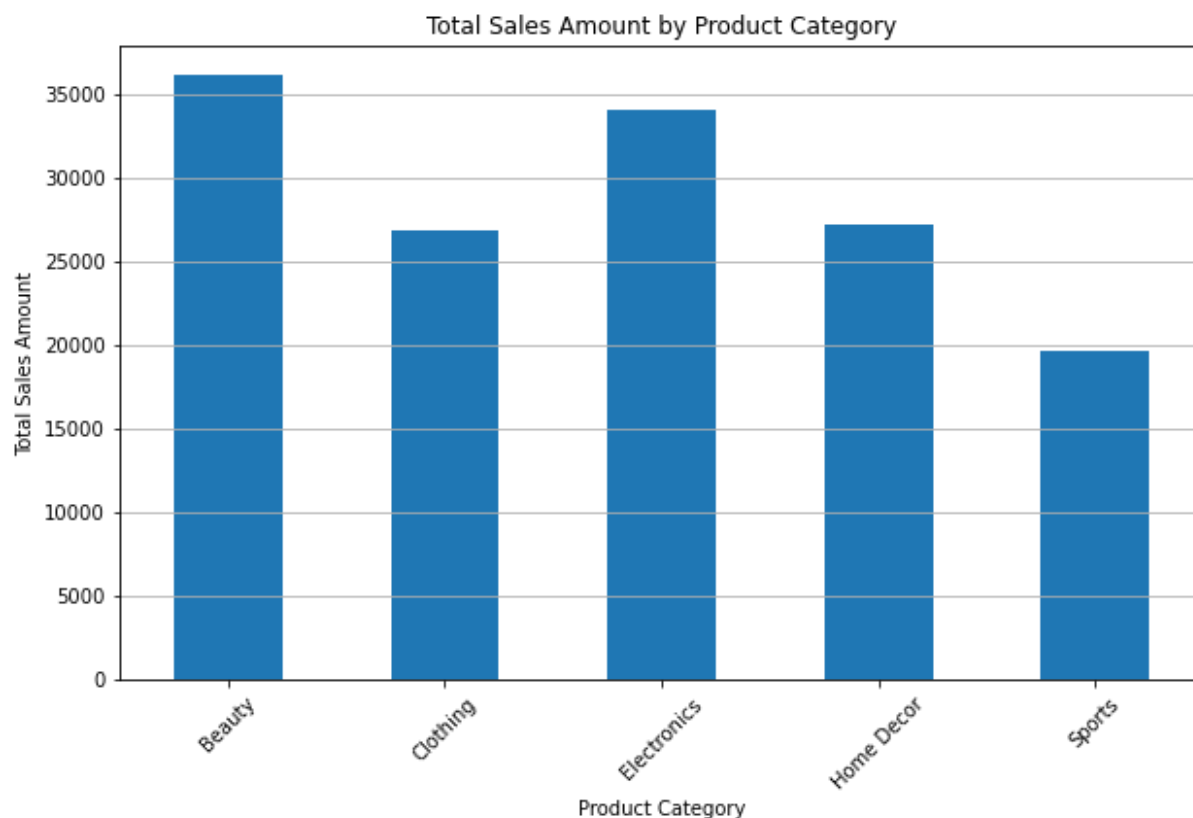
```
In [ ]:
```

## Level 5: Data Visualization with Matplotlib

1. Create a bar chart to visualize the total sales amount by product category.

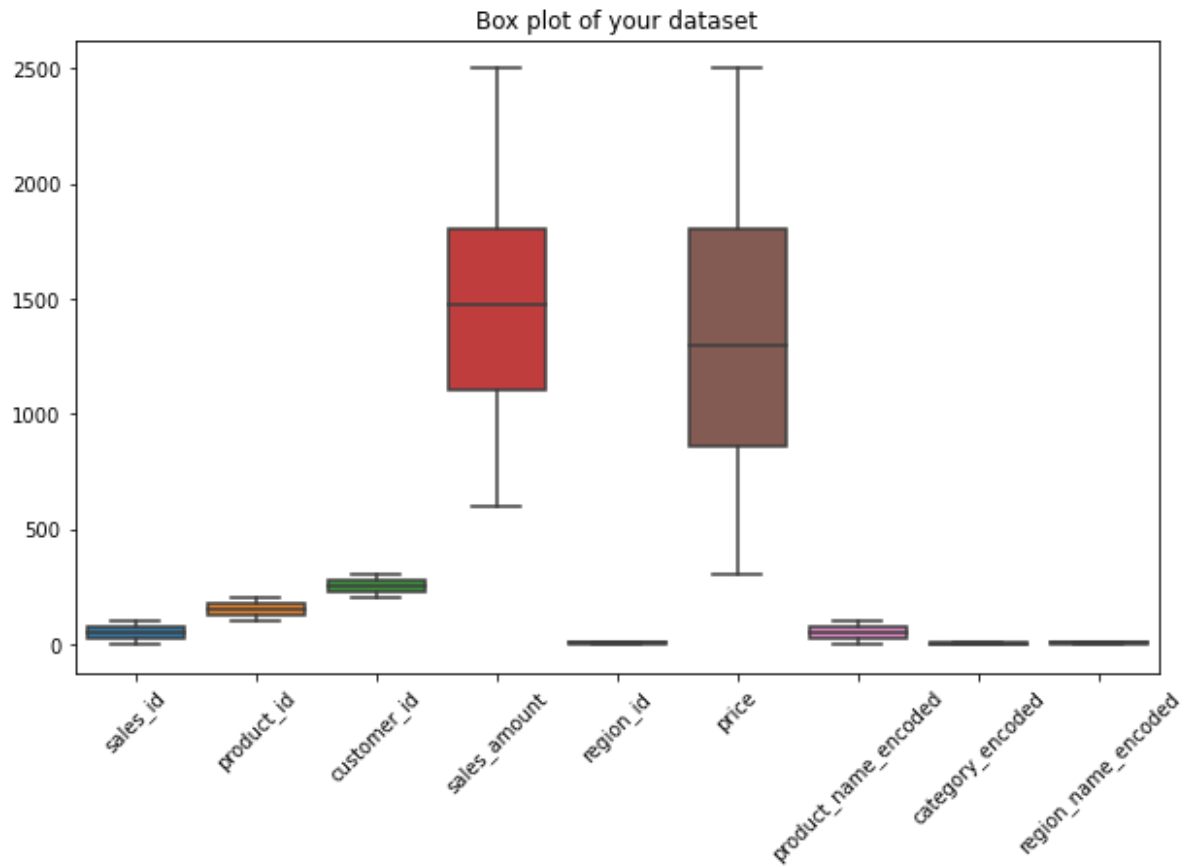
```
In [97]: df=cleaned_sales.copy()
```

```
In [181]: # Create a bar chart to visualize the total sales amount by product category
sales_by_category=df.groupby('category')['sales_amount'].sum()
plt.figure(figsize=(10, 6))
sales_by_category.plot(kind='bar')
plt.title('Total Sales Amount by Product Category')
plt.xlabel('Product Category')
plt.ylabel('Total Sales Amount')
plt.xticks(rotation=45)
plt.grid(axis='y')
plt.show()
```

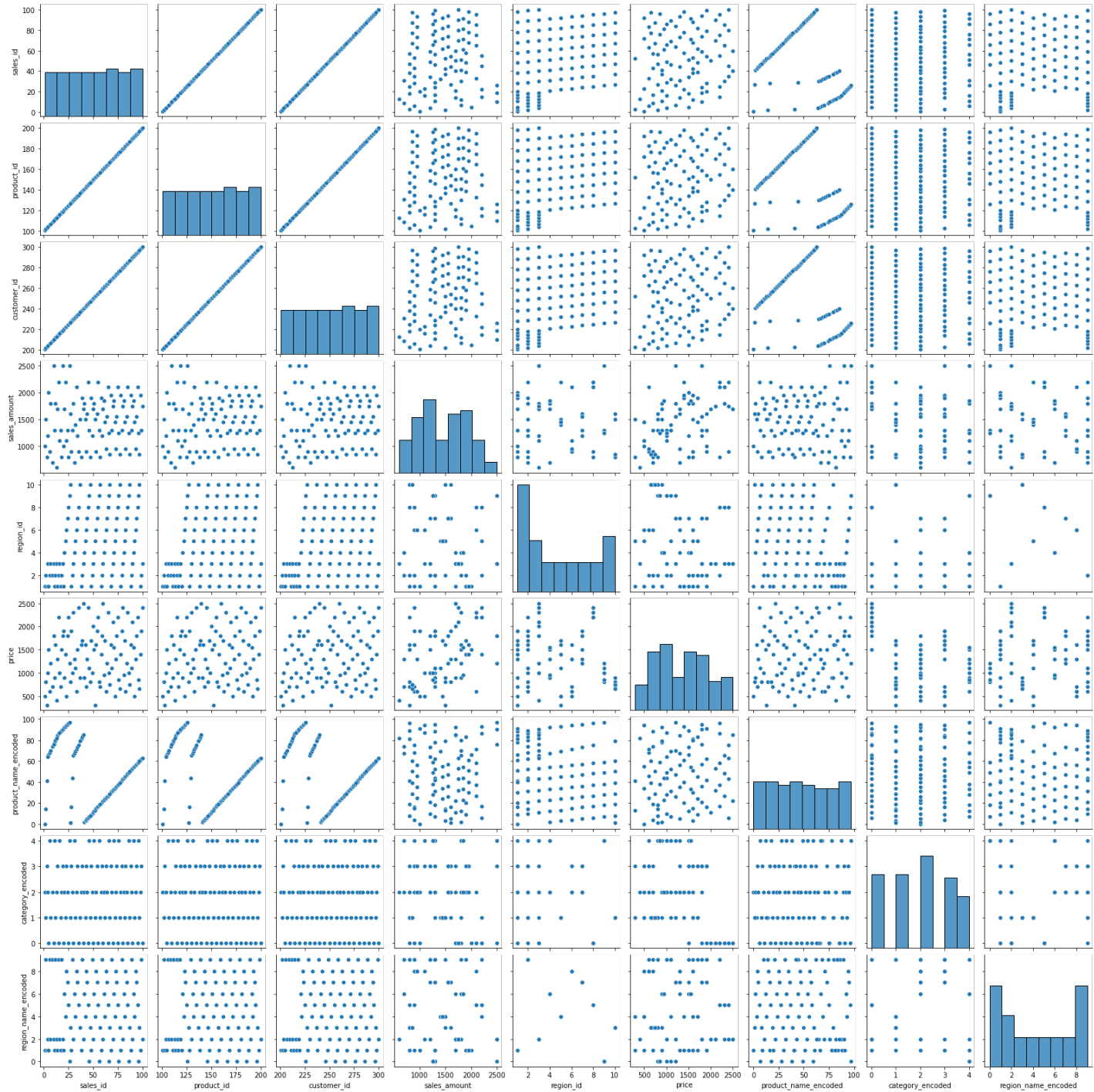




```
In [84]: import matplotlib.pyplot as plt
import seaborn as sns
plt.figure(figsize=(10, 6))
sns.boxplot(data=cleaned_sales)
plt.title('Box plot of your dataset')
plt.xticks(rotation=45) # Rotate x-axis labels for better readability
plt.show()
```



```
In [85]: sns.pairplot(df)
plt.show()
```



## 2. Generate a line plot to show the sales trend over time

```
In [98]: df['date'] = pd.to_datetime(df['date'])
# converted the date column so that we can use for time series analysis
df
```



[illegible]

[illegible]

[illegible]

```
df['date'] = pd.to_datetime(df['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2777264012.py:1: UserWarning: Parsing '27-03-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True for consistent parsing.
df['date'] = pd.to_datetime(df['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2777264012.py:1: UserWarning: Parsing '27-01-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True for consistent parsing.
df['date'] = pd.to_datetime(df['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2777264012.py:1: UserWarning: Parsing '16-02-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True for consistent parsing.
df['date'] = pd.to_datetime(df['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2777264012.py:1: UserWarning: Parsing '26-02-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True for consistent parsing.
df['date'] = pd.to_datetime(df['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2777264012.py:1: UserWarning: Parsing '18-03-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True for consistent parsing.
df['date'] = pd.to_datetime(df['date'])
C:\Users\piyush thakur\AppData\Local\Temp\ipykernel_32880\2777264012.py:1: UserWarning: Parsing '28-03-2022' in DD/MM/YYYY format. Provide format or specify infer_datetime_format=True for consistent parsing.
df['date'] = pd.to_datetime(df['date'])
```

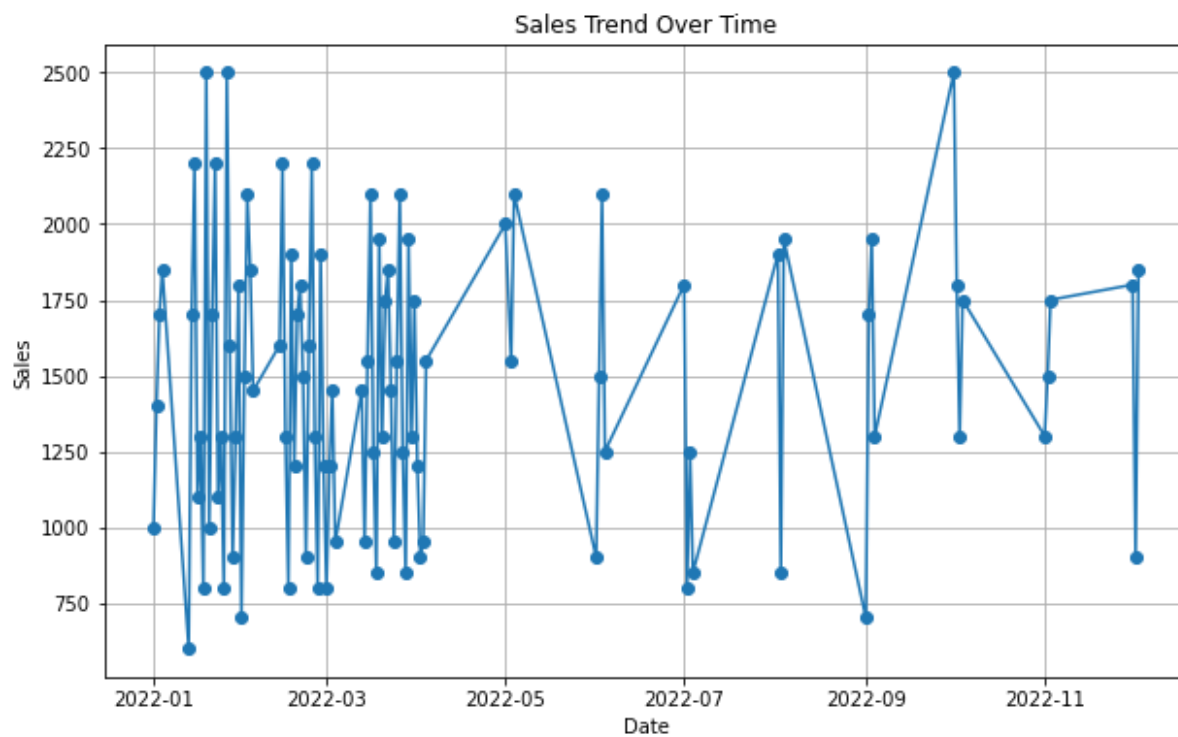
Out [98]:

	sales_id	product_id	customer_id	sales_amount	date	region_id	product_name	category
0	1	101	201	1000	2022-01-01	1	Product A	Electronics
1	3	103	203	800	2022-03-01	1	Product C	Home Decor
2	5	105	205	2000	2022-05-01	1	Product E	Beauty
4	11	111	211	1300	2022-11-01	1	Product K	Sports
5	14	114	214	1700	2022-01-14	1	Product N	Home Decor
...	...	...	...	...	...	...	...	...
95	57	157	257	800	2022-02-26	10	Product BE	Clothing
96	67	167	267	850	2022-08-03	10	Product BO	Clothing
97	77	177	277	850	2022-03-18	10	Product BY	Clothing
98	87	187	287	850	2022-03-28	10	Product CI	Clothing
99	97	197	297	850	2022-07-04	10	Product CS	Clothing

98 rows × 16 columns

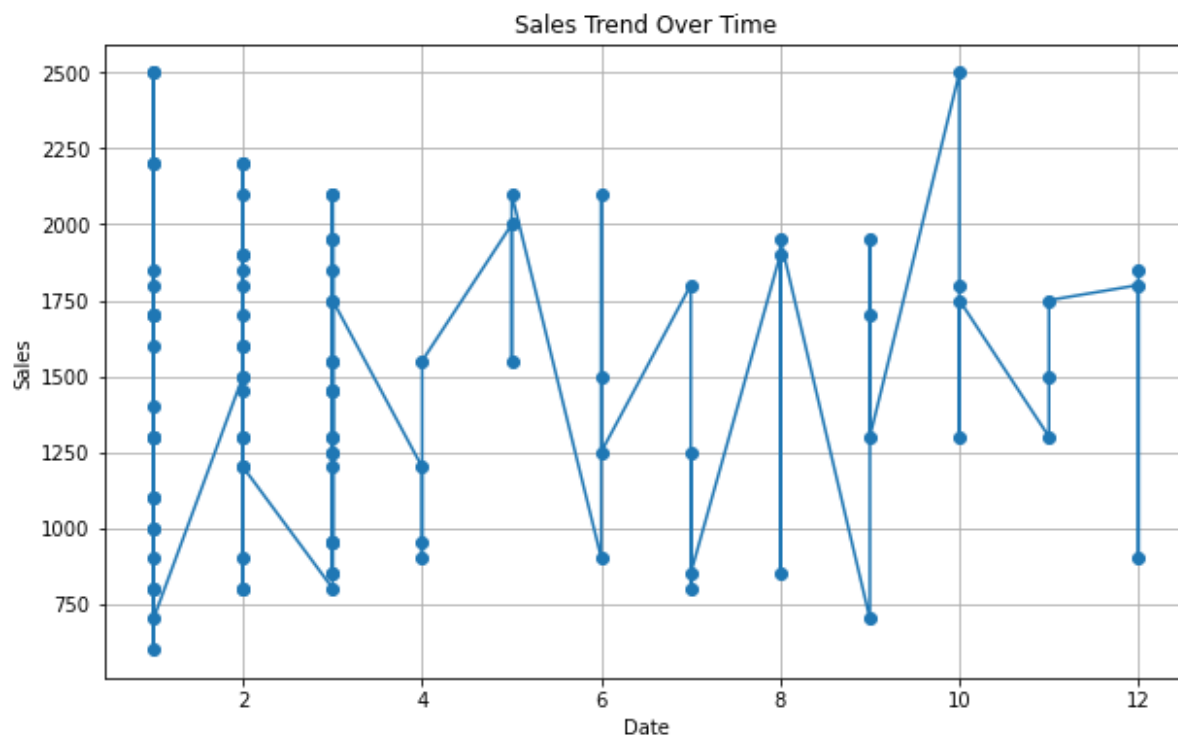
In [104]:

```
#sorted the date data for time series analysis
df_sort=df.sort_values('date')
plt.figure(figsize=(10, 6))
plt.plot(df_sort['date'], df_sort['sales_amount'], marker='o', linestyle='--')
plt.title('Sales Trend Over Time')
plt.xlabel('Date')
plt.ylabel('Sales')
plt.grid(True)
plt.show()
```



In [116...]

```
df_sort=df.sort_values('date')
plt.figure(figsize=(10, 6))
plt.plot(df_sort['month'], df_sort['sales_amount'], marker='o', linestyle='--')
plt.title('Sales Trend Over Time')
plt.xlabel('Date')
plt.ylabel('Sales')
plt.grid(True)
plt.show()
```

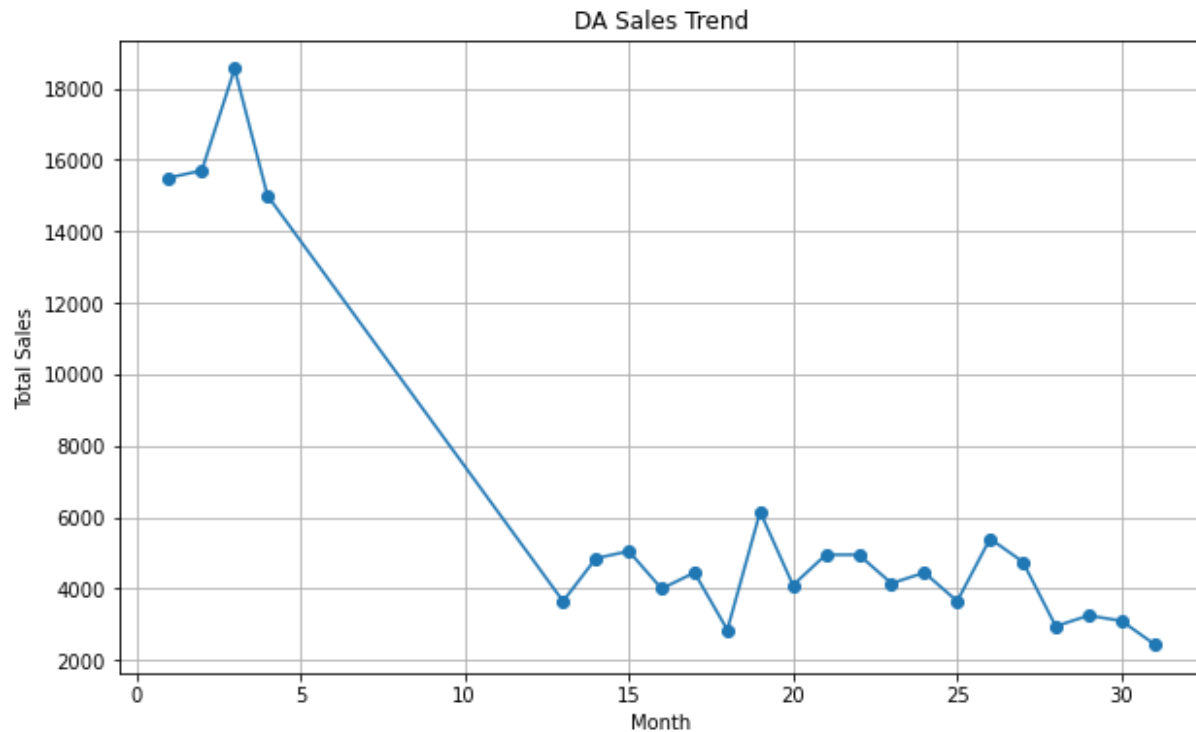


In [119...

```
df_sort['day'] = df_sort['date'].dt.day

daily_sales = df_sort.groupby('day')['sales_amount'].sum()

# Create a line plot for sales trend over months
plt.figure(figsize=(10, 6))
plt.plot(daily_sales.index, daily_sales.values, marker='o', linestyle='--')
plt.title('DA Sales Trend')
plt.xlabel('Month')
plt.ylabel('Total Sales')
plt.grid(True)
plt.show()
```



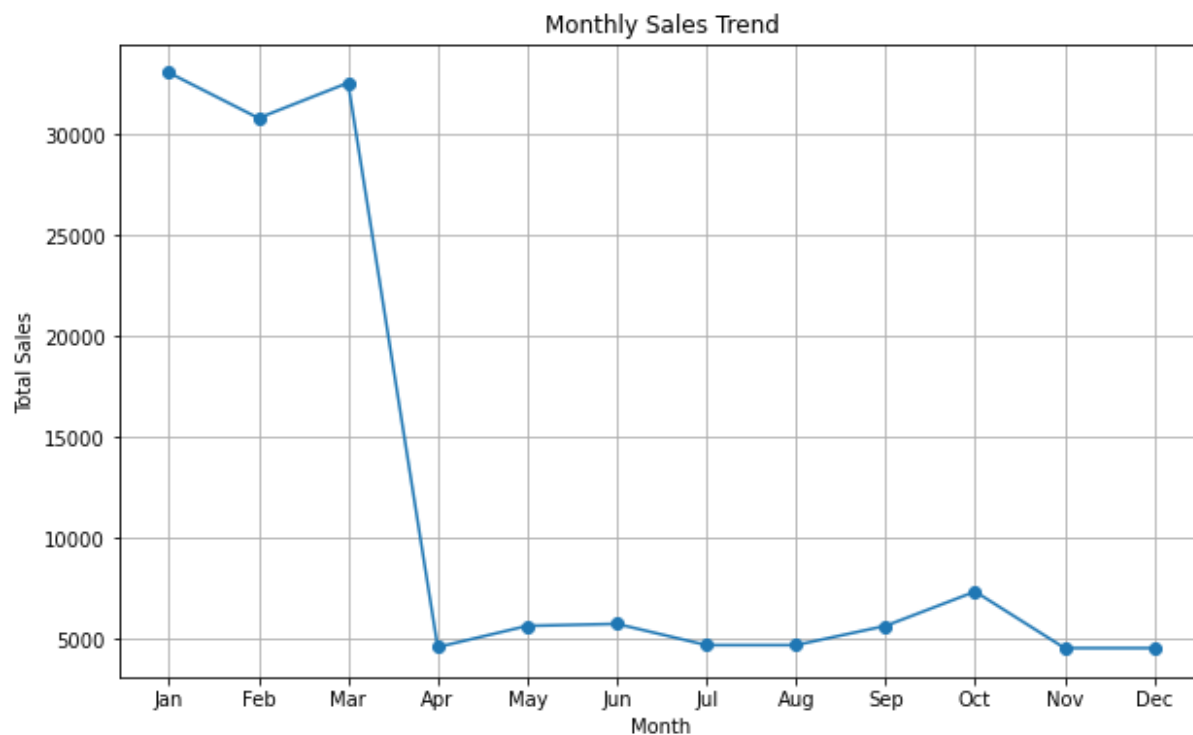
In [117...

```
df_sort['month'] = df_sort['date'].dt.month

monthly_sales = df_sort.groupby('month')['sales_amount'].sum()

monthly_sales
# Create a line plot for sales trend over months
plt.figure(figsize=(10, 6))
plt.plot(monthly_sales.index, monthly_sales.values, marker='o', linestyle='--')
plt.title('Monthly Sales Trend')
plt.xlabel('Month')
plt.ylabel('Total Sales')
plt.xticks(range(1, 13), ['Jan', 'Feb', 'Mar', 'Apr', 'May', 'Jun', 'Jul', 'Aug', 'Sep', 'Oct', 'Nov', 'Dec'])
plt.grid(True)
plt.show()
```

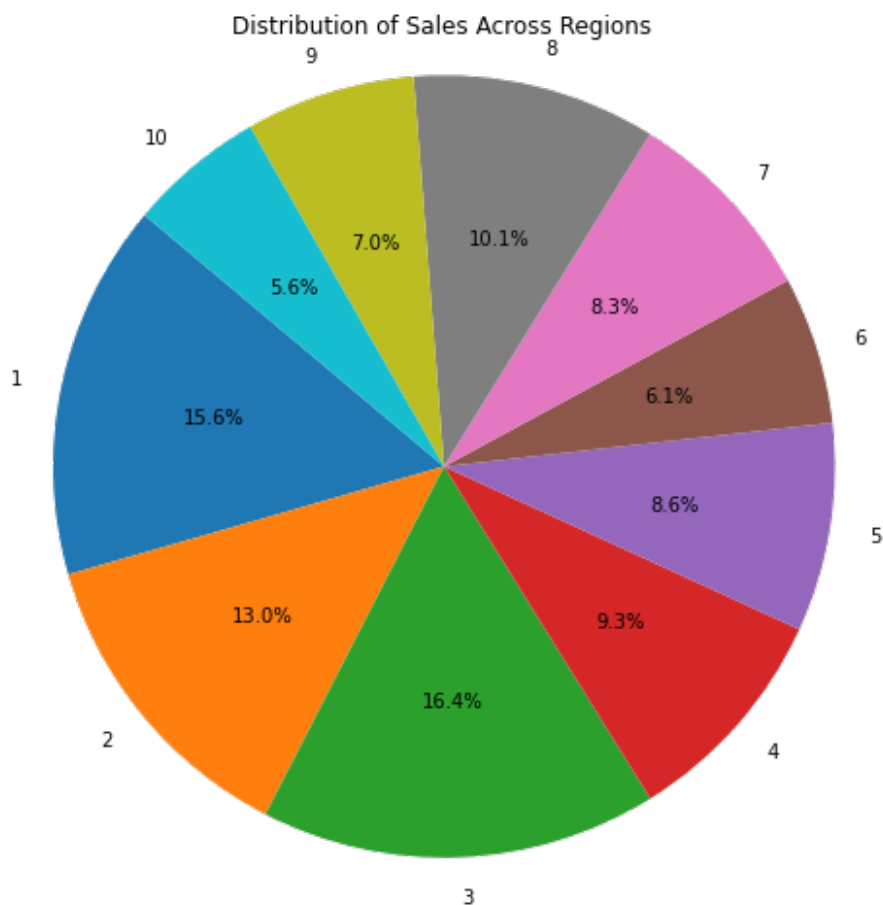




3. Create a pie chart to represent the distribution of sales across different regions.

```
In [122... sales_by_region=df_sort.groupby('region_id')['sales_amount'].sum()

plt.figure(figsize=(8, 8))
plt.pie(sales_by_region, labels=sales_by_region.index, autopct='%1.1f%%', startangle=140)
plt.title('Distribution of Sales Across Regions')
plt.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.
plt.show()
```

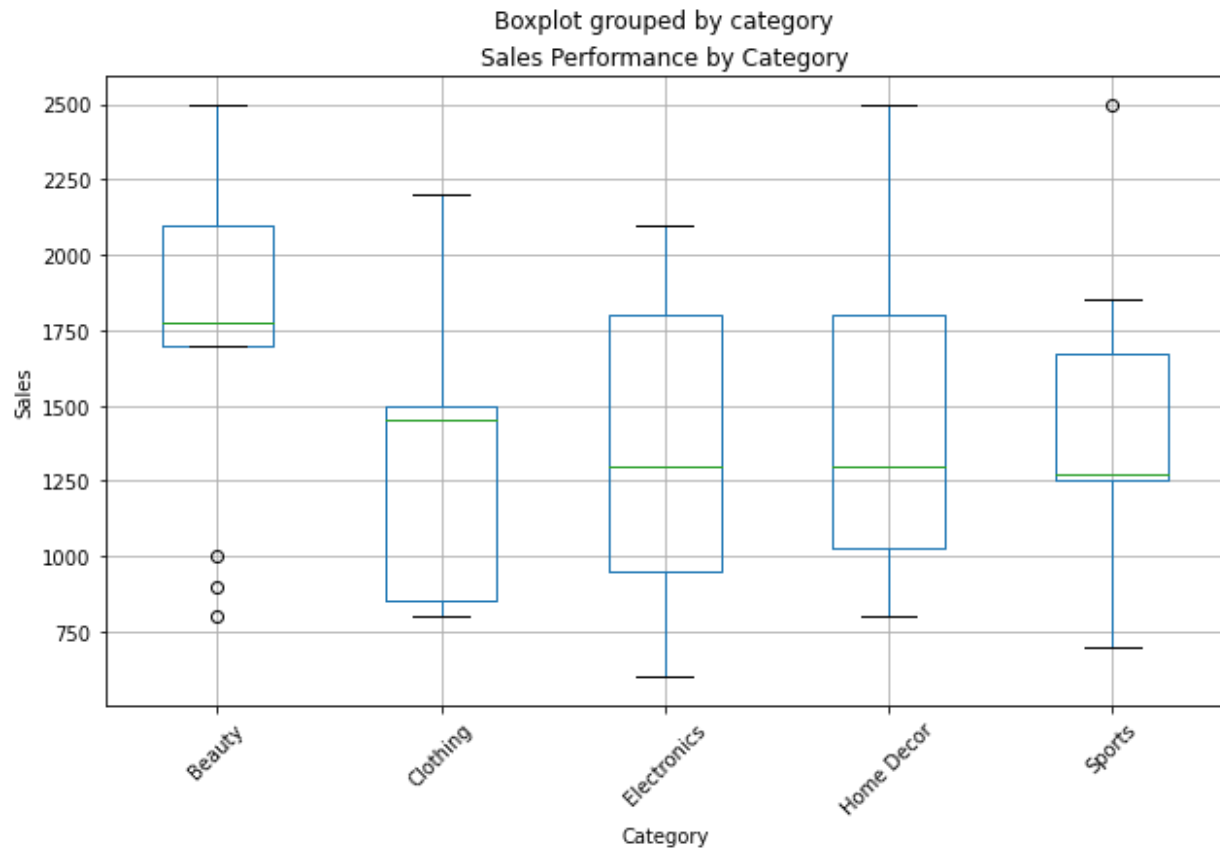




#### 4. Generate a boxplot to compare the sales performance of different customer segments.

```
In [127... plt.figure(figsize=(10, 6))
df.boxplot(column='sales_amount', by='category', figsize=(10, 6))
plt.title('Sales Performance by Category')
plt.xlabel('Category')
plt.ylabel('Sales')
plt.xticks(rotation=45)
plt.grid(True)
plt.show()
```

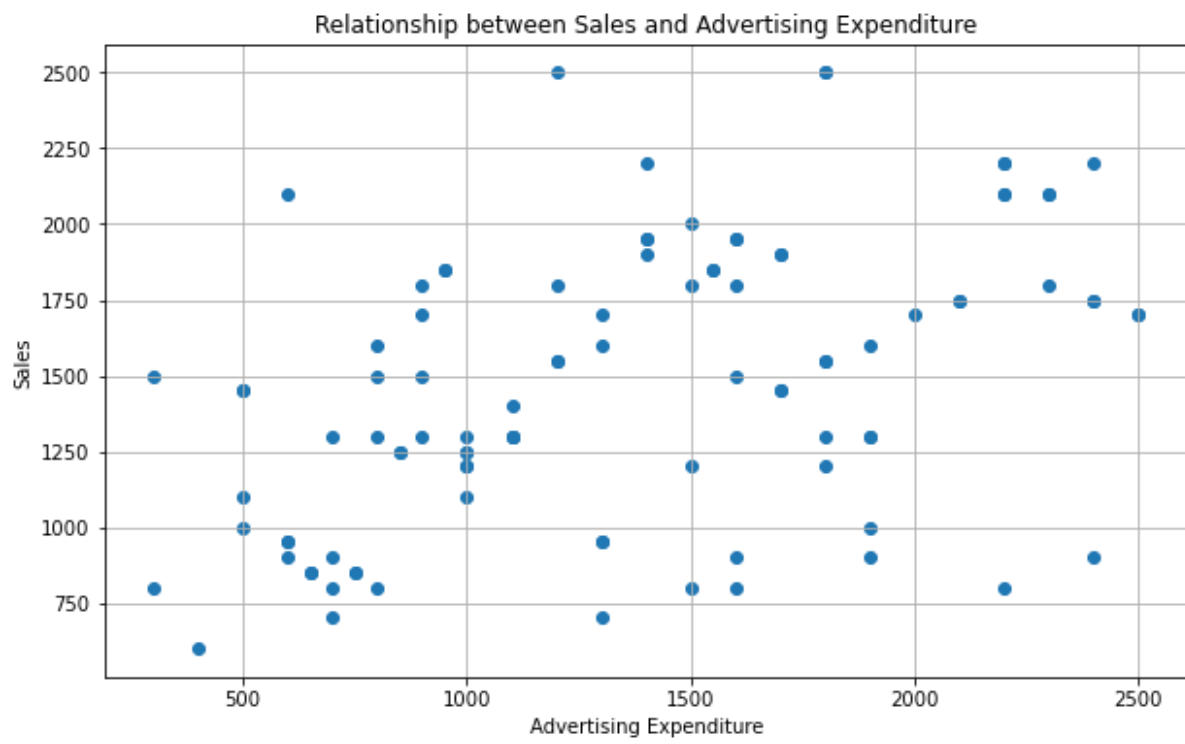
<Figure size 720x432 with 0 Axes>



```
In [128... #Beauty product data is heavily skewed and mostly has high end products but there are some low valued products, whereas Electronics, Home Decor, and sports has average value products.
```

#### 5. Visualize the relationship between sales and advertising expenditure using a scatter plot.

```
In [131... plt.figure(figsize=(10, 6))
plt.scatter(df_sort['price'], df_sort['sales_amount'])
plt.title('Relationship between Sales and Advertising Expenditure')
plt.xlabel('Advertising Expenditure')
plt.ylabel('Sales')
plt.grid(True)
plt.show()
```



In [133... *# there is no some sort of relationship between price and the sales\_amount, but it has weak*

In [132... df

Out [132]:

	sales_id	product_id	customer_id	sales_amount	date	region_id	product_name	category
0	1	101	201	1000	2022-01-01	1	Product A	Electronic
1	3	103	203	800	2022-03-01	1	Product C	Home Decor
2	5	105	205	2000	2022-05-01	1	Product E	Beauty
4	11	111	211	1300	2022-11-01	1	Product K	Sports
5	14	114	214	1700	2022-01-14	1	Product N	Home Decor
...	...	...	...	...	...	...	...	...
95	57	157	257	800	2022-02-26	10	Product BE	Clothing
96	67	167	267	850	2022-08-03	10	Product BO	Clothing
97	77	177	277	850	2022-03-18	10	Product BY	Clothing
98	87	187	287	850	2022-03-28	10	Product CI	Clothing
99	97	197	297	850	2022-07-04	10	Product CS	Clothing

98 rows × 18 columns

## Level 6: Advanced Data Manipulation with Numpy

In [135]:

```
np.mean(df_sort)
```

```
C:\Users\piyush thakur\anaconda3\lib\site-packages\numpy\core\fromnumeric.py:3430: FutureWarning: In a future version, DataFrame.mean(axis=None) will return a scalar mean over the entire DataFrame. To retain the old behavior, use 'frame.mean(axis=0)' or just 'frame.mean()'
    return mean(axis=axis, dtype=dtype, out=out, **kwargs)
C:\Users\piyush thakur\anaconda3\lib\site-packages\numpy\core\fromnumeric.py:3430: FutureWarning: DataFrame.mean and DataFrame.median with numeric_only=None will include datetime64 and datetime64tz columns in a future version.
    return mean(axis=axis, dtype=dtype, out=out, **kwargs)
C:\Users\piyush thakur\anaconda3\lib\site-packages\numpy\core\fromnumeric.py:3430: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.
    return mean(axis=axis, dtype=dtype, out=out, **kwargs)
```

```
Out[135]: sales_id          51.081633
          product_id      151.081633
          customer_id      251.081633
          sales_amount     1468.367347
          region_id         4.775510
          price            1359.183673
          product_name_encoded 47.591837
          category_encoded    1.867347
          region_name_encoded  4.418367
          month              3.918367
          day                12.989796
          dtype: float64
```

## 1. Use Numpy to calculate the mean, median, and standard deviation of the sales data.

```
In [152]: np.median(df.sales_id), np.median(df.product_id), np.median(df.sales_amount), np.median(df.region_id)
```

```
Out[152]: (51.5, 151.5, 1475.0, 4.0, 1300.0)
```

```
In [153]: np.std(df_sort)
```

```
C:\Users\piyush thakur\anaconda3\lib\site-packages\numpy\core\fromnumeric.py:3571: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.
    return std(axis=axis, dtype=dtype, out=out, ddof=ddof, **kwargs)
```

```
Out[153]: sales_id          28.798269
          product_id      28.798269
          customer_id      28.798269
          sales_amount     465.240105
          date              90 days 16:47:50.182474781
          region_id         2.943449
          price            591.191591
          product_name_encoded 28.222092
          category_encoded    1.329706
          region_name_encoded  3.003139
          month              3.186942
          day                10.293147
          dtype: object
```

## 2. Perform element-wise arithmetic operations on the sales data (e.g., addition, subtraction, multiplication).

```
In [160... # Convert the sales data to a NumPy array
sales_array = np.array(df_sort['sales_amount'])

# Example values for operations
value_to_add = 10
value_to_subtract = 5
value_to_multiply = 2

# Perform element-wise arithmetic operations
sales_added = sales_array + value_to_add
sales_subtracted = sales_array - value_to_subtract
sales_multiplied = sales_array * value_to_multiply
```

```
In [161... print(sales_added,sales_subtracted,sales_multiplied)

[1010 1410 1710 1860 610 1710 2210 1110 1310 810 2510 1010 1710 2210
 1110 1310 810 2510 1610 910 1310 1810 710 1510 2110 1860 1460 1610
 2210 1310 810 1910 1210 1710 1810 1510 910 1610 2210 1310 810 1910
 1210 810 1210 1460 960 1460 960 1560 2110 1260 860 1960 1310 1760
 1860 1460 960 1560 2110 1260 860 1960 1310 1760 1210 910 960 1560
 2010 1560 2110 910 1510 2110 1260 1810 810 1260 860 1910 860 1960
 710 1710 1960 1310 2510 1810 1310 1760 1310 1510 1760 1810 910 1860] [ 995 1395 1695 1845
595 1695 2195 1095 1295 795 2495 995 1695 2195
 1095 1295 795 2495 1595 895 1295 1795 695 1495 2095 1845 1445 1595
 2195 1295 795 1895 1195 1695 1795 1495 895 1595 2195 1295 795 1895
 1195 795 1195 1445 945 1445 945 1545 2095 1245 845 1945 1295 1745
 1845 1445 945 1545 2095 1245 845 1945 1295 1745 1195 895 945 1545
 1995 1545 2095 895 1495 2095 1245 1795 795 1245 845 1895 845 1945
 695 1695 1945 1295 2495 1795 1295 1745 1295 1495 1745 1795 895 1845] [2000 2800 3400 3700
1200 3400 4400 2200 2600 1600 5000 2000 3400 4400
 2200 2600 1600 5000 3200 1800 2600 3600 1400 3000 4200 3700 2900 3200
 4400 2600 1600 3800 2400 3400 3600 3000 1800 3200 4400 2600 1600 3800
 2400 1600 2400 2900 1900 2900 1900 3100 4200 2500 1700 3900 2600 3500
 3700 2900 1900 3100 4200 2500 1700 3900 2600 3500 2400 1800 1900 3100
 4000 3100 4200 1800 3000 4200 2500 3600 1600 2500 1700 3800 1700 3900
 1400 3400 3900 2600 5000 3600 2600 3500 2600 3000 3500 3600 1800 3700]
```

### 3. Use Numpy to reshape the sales data into a different dimension.

```
In [163... sales_resaped = sales_array.reshape(1, 1, 1, -1)
sales_resaped
```

```
Out[163]: array([[[[1000, 1400, 1700, 1850, 600, 1700, 2200, 1100, 1300, 800,
2500, 1000, 1700, 2200, 1100, 1300, 800, 2500, 1600, 900,
1300, 1800, 700, 1500, 2100, 1850, 1450, 1600, 2200, 1300,
800, 1900, 1200, 1700, 1800, 1500, 900, 1600, 2200, 1300,
800, 1900, 1200, 800, 1200, 1450, 950, 1450, 950, 1550,
2100, 1250, 850, 1950, 1300, 1750, 1850, 1450, 950, 1550,
2100, 1250, 850, 1950, 1300, 1750, 1200, 900, 950, 1550,
2000, 1550, 2100, 900, 1500, 2100, 1250, 1800, 800, 1250,
850, 1900, 850, 1950, 700, 1700, 1950, 1300, 2500, 1800,
1300, 1750, 1300, 1500, 1750, 1800, 900, 1850]]]], dtype=int64)
```

### 4. Apply broadcasting to perform operations on arrays with different shapes.

```
In [168... four_dim=sales_resaped
one_dim=sales_array

result=four_dim+one_dim
result.shape
```

```
Out[168]: (1, 1, 1, 98)
```

## 5. Use Numpy to perform matrix multiplication on sales data arrays.

```
In [176]: mult_dim=np.dot(four_dim,one_dim)
mult_dim.shape
```

```
Out[176]: (1, 1, 1)
```

## Level 7: Advanced Pandas Queries

### 1. Use Pandas to filter the sales data for a specific time period (e.g., quarter or year).

```
In [198]: df.columns
```

```
Out[198]: Index(['sales_id', 'product_id', 'customer_id', 'sales_amount', 'date',
               'region_id', 'product_name', 'category', 'price', 'customer_name',
               'email', 'address', 'region_name', 'product_name_encoded',
               'category_encoded', 'region_name_encoded', 'month', 'year'],
              dtype='object')
```

```
In [199]: # Extract the quarter and year from the 'date' column
df['quarter'] = df['date'].dt.quarter
df['year'] = df['date'].dt.year

# Filter the sales data for a specific quarter and year
target_quarter = 1
target_year = 2022

filtered_sales_data = df[(df['quarter'] == target_quarter) & (df['year'] == target_year)]

# Display the filtered sales data
print("Filtered Sales Data for Quarter {} of Year {}".format(target_quarter, target_year))
print(filtered_sales_data)
```

Filtered Sales Data for Quarter 1 of Year 2022:

	sales_id	product_id	customer_id	sales_amount	date	region_id	\
0	1	101	201	1000	2022-01-01	1	
1	3	103	203	800	2022-03-01	1	
5	14	114	214	1700	2022-01-14	1	
6	17	117	217	1300	2022-01-17	1	
7	20	120	220	1000	2022-01-20	1	
..	...	...	...	...	...	...	
92	27	127	227	1600	2022-01-27	10	
94	47	147	247	800	2022-02-16	10	
95	57	157	257	800	2022-02-26	10	
97	77	177	277	850	2022-03-18	10	
98	87	187	287	850	2022-03-28	10	

	product_name	category	price	customer_name	email	\
0	Product A	Electronics	500	John Doe	john@example.com	
1	Product C	Home Decor	300	Robert Brown	robert@example.com	
5	Product N	Home Decor	1300	Rachel Hall	rachel@example.com	
6	Product Q	Clothing	700	Michelle Perez	michelle@example.com	
7	Product T	Beauty	1900	Brian Hall	brian@example.com	
..	...	...	...	...	...	
92	Product AA	Clothing	800	Isabella Martinez	isabella@example.com	
94	Product AU	Clothing	800	Abigail King	abigail@example.com	
95	Product BE	Clothing	700	Grace Ramirez	grace@example.com	
97	Product BY	Clothing	650	Sophia Reyes	sophia@example.com	
98	Product CI	Clothing	750	Harper Coleman	harper@example.com	

	address	region_name	product_name_encoded	\
0	123 Main St, Anytown, USA	East Coast	0	
1	789 Oak St, Anycity, USA	East Coast	41	
5	852 Cedar St, Nowhere, USA	East Coast	84	
6	357 Spruce St, Othertown, USA	East Coast	88	
7	159 Elm St, Nowhere, USA	East Coast	91	
..	...	...	...	
92	587 Maple St, Somewhere, USA	Mountain	1	
94	587 Maple St, Somewhere, USA	Mountain	8	
95	586 Maple St, Anytown, USA	Mountain	19	
97	586 Maple St, Anytown, USA	Mountain	39	
98	587 Maple St, Somewhere, USA	Mountain	50	

	category_encoded	region_name_encoded	month	year	quarter
0	2	1	1	2022	1
1	3	1	3	2022	1
5	3	1	1	2022	1
6	1	1	1	2022	1
7	0	1	1	2022	1
..	...	...	...	...	...
92	1	3	1	2022	1
94	1	3	2	2022	1
95	1	3	2	2022	1
97	1	3	3	2022	1
98	1	3	3	2022	1

[66 rows x 19 columns]

## 2. Apply boolean indexing to select rows based on multiple conditions.

```
In [ ]: # done in earlier sections as part of the problem
```

## 3. Use Pandas groupby and aggregate functions to calculate custom metrics.

```
In [ ]: # done in earlier sections as part of the problem
```

**Combine multiple DataFrames using merge or join operations.**

```
In [ ]: # done in earlier sections as part of the problem
```