

February 2, 2025

1 Task 1: Exploratory Data Analysis (EDA) and Business Insights

1.0.1 1. Perform EDA on the provided dataset.

1.0.2 2. Derive at least 5 business insights from the EDA.

Write these insights in short point-wise sentences (maximum 100 words per insight).

```
[2]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[3]: # Load the datasets
customers = pd.read_csv('Customers.csv')
products = pd.read_csv('Products.csv')
transactions = pd.read_csv('Transactions.csv')
```

```
[4]: # Merge datasets for comprehensive analysis
df = pd.merge(transactions, customers, on='CustomerID')
df = pd.merge(df, products, on='ProductID')
```

```
[5]: # Basic EDA
print(df.info())
print(df.describe())
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1000 entries, 0 to 999
Data columns (total 13 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   TransactionID         1000 non-null   object
 1   CustomerID            1000 non-null   object
 2   ProductID             1000 non-null   object
 3   TransactionDate       1000 non-null   object
 4   Quantity              1000 non-null   int64
 5   TotalValue            1000 non-null   float64
 6   Price_x               1000 non-null   float64
 7   CustomerName          1000 non-null   object
 8   Region               1000 non-null   object
 9   SignupDate            1000 non-null   object
```

```

10  ProductName      1000 non-null  object
11  Category        1000 non-null  object
12  Price_y         1000 non-null  float64
dtypes: float64(3), int64(1), object(9)
memory usage: 109.4+ KB
None

```

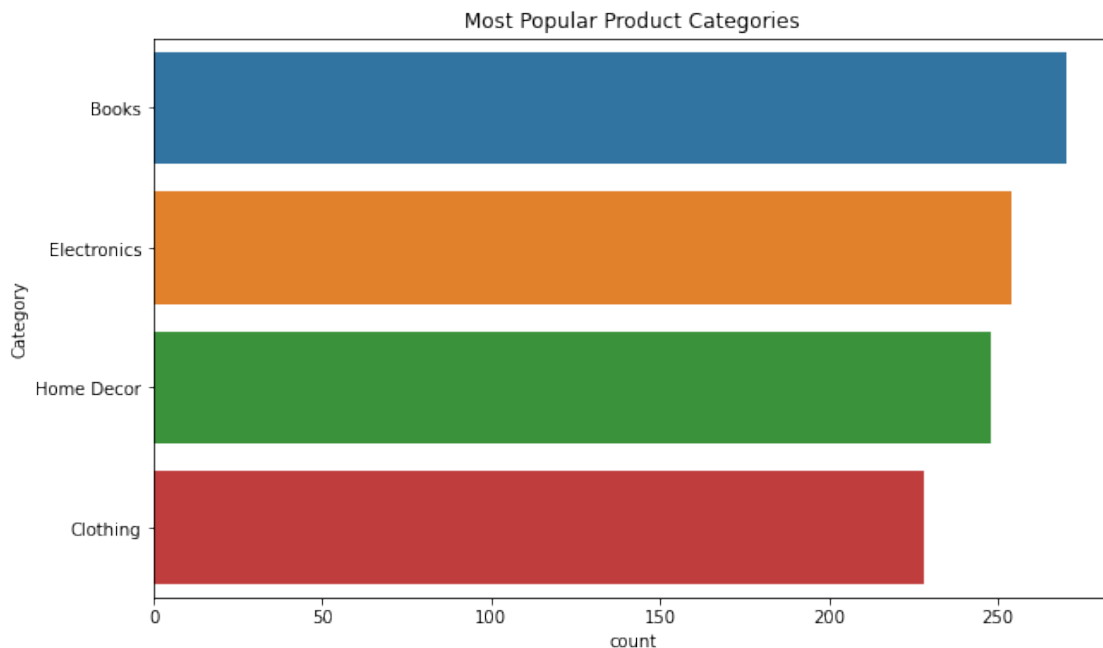
	Quantity	TotalValue	Price_x	Price_y
count	1000.000000	1000.000000	1000.00000	1000.00000
mean	2.537000	689.995560	272.55407	272.55407
std	1.117981	493.144478	140.73639	140.73639
min	1.000000	16.080000	16.08000	16.08000
25%	2.000000	295.295000	147.95000	147.95000
50%	3.000000	588.880000	299.93000	299.93000
75%	4.000000	1011.660000	404.40000	404.40000
max	4.000000	1991.040000	497.76000	497.76000

2 Business Insights

```

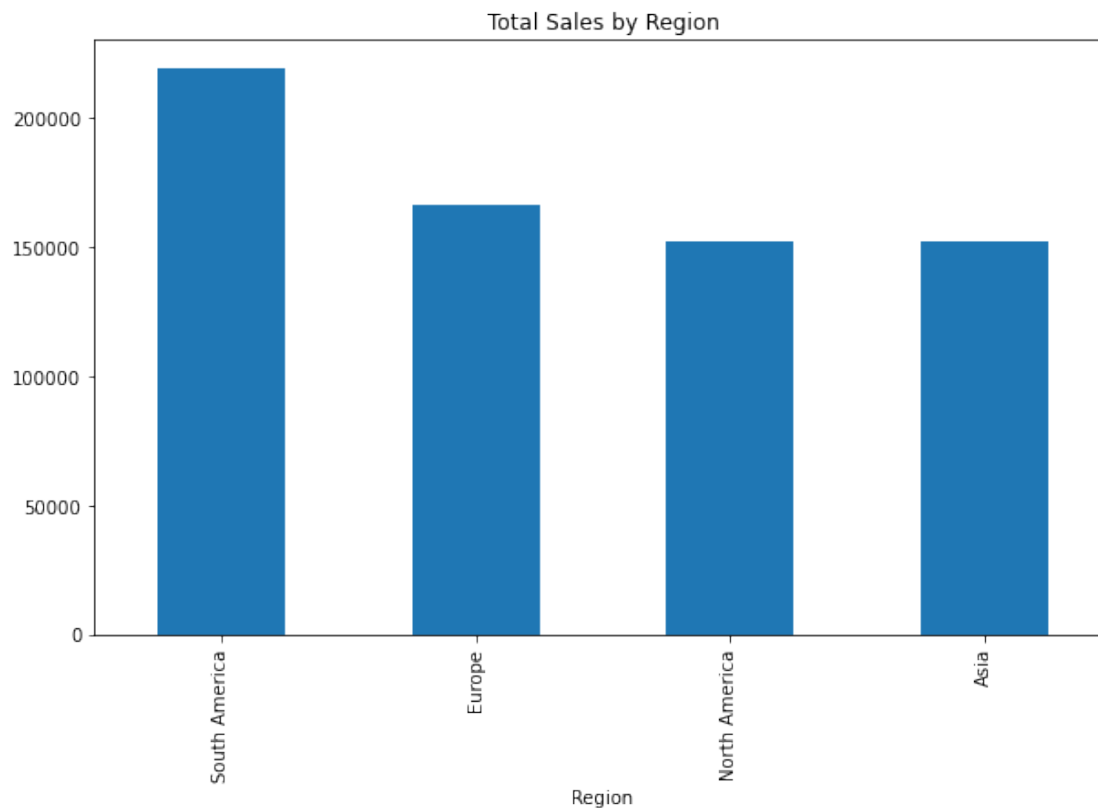
[6]: # Insight 1: Most popular product categories
plt.figure(figsize=(10, 6))
sns.countplot(y='Category', data=df, order=df['Category'].value_counts().index)
plt.title('Most Popular Product Categories')
plt.show()

```



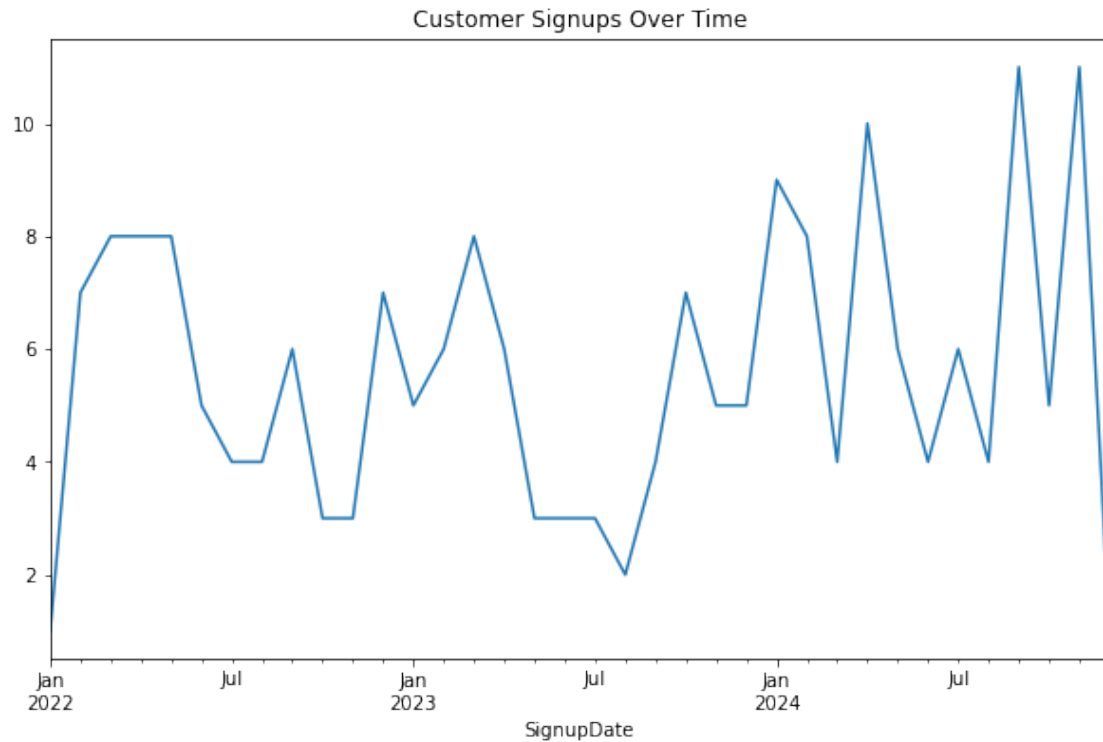
```
[7]: # Insight 2: Total sales by region
region_sales = df.groupby('Region')['TotalValue'].sum().
    ↪sort_values(ascending=False)
region_sales.plot(kind='bar', figsize=(10, 6), title='Total Sales by Region')
```

```
[7]: <AxesSubplot:title={'center':'Total Sales by Region'}, xlabel='Region'>
```



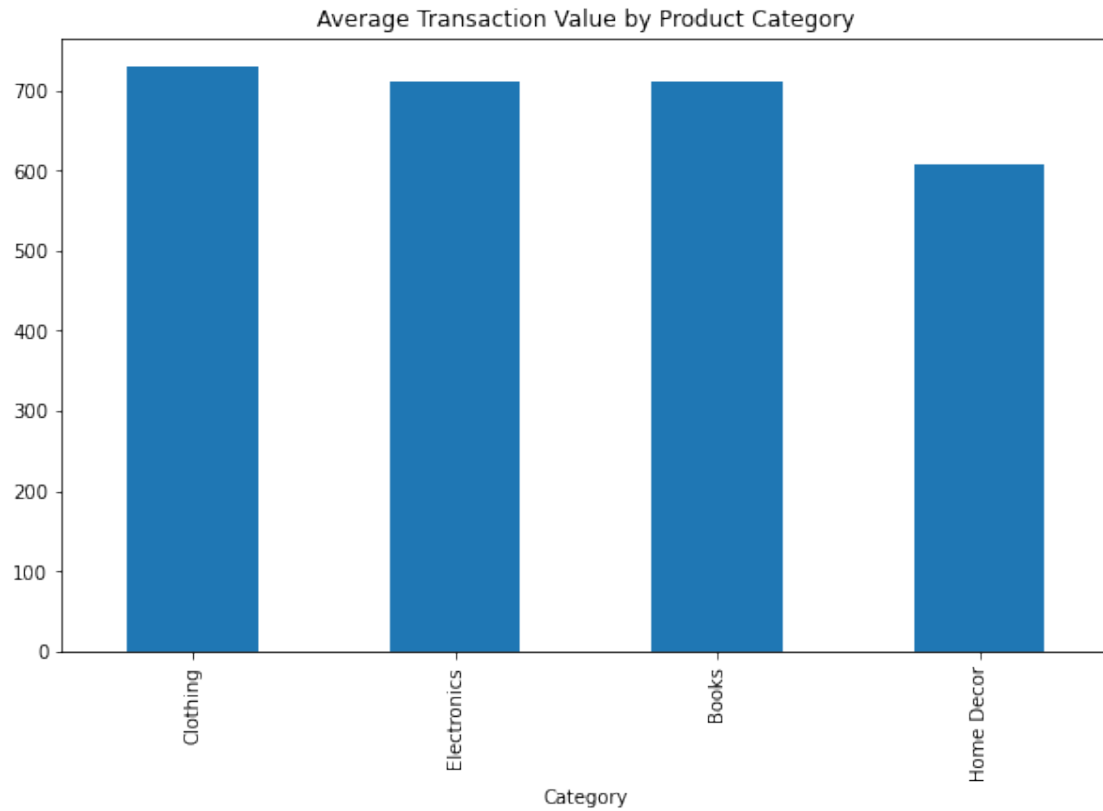
```
[8]: # Insight 3: Customer signups over time
customers['SignupDate'] = pd.to_datetime(customers['SignupDate'])
customers.set_index('SignupDate', inplace=True)
customers.resample('M').size().plot(figsize=(10, 6), title='Customer Signups_
    ↪Over Time')
```

```
[8]: <AxesSubplot:title={'center':'Customer Signups Over Time'}, xlabel='SignupDate'>
```



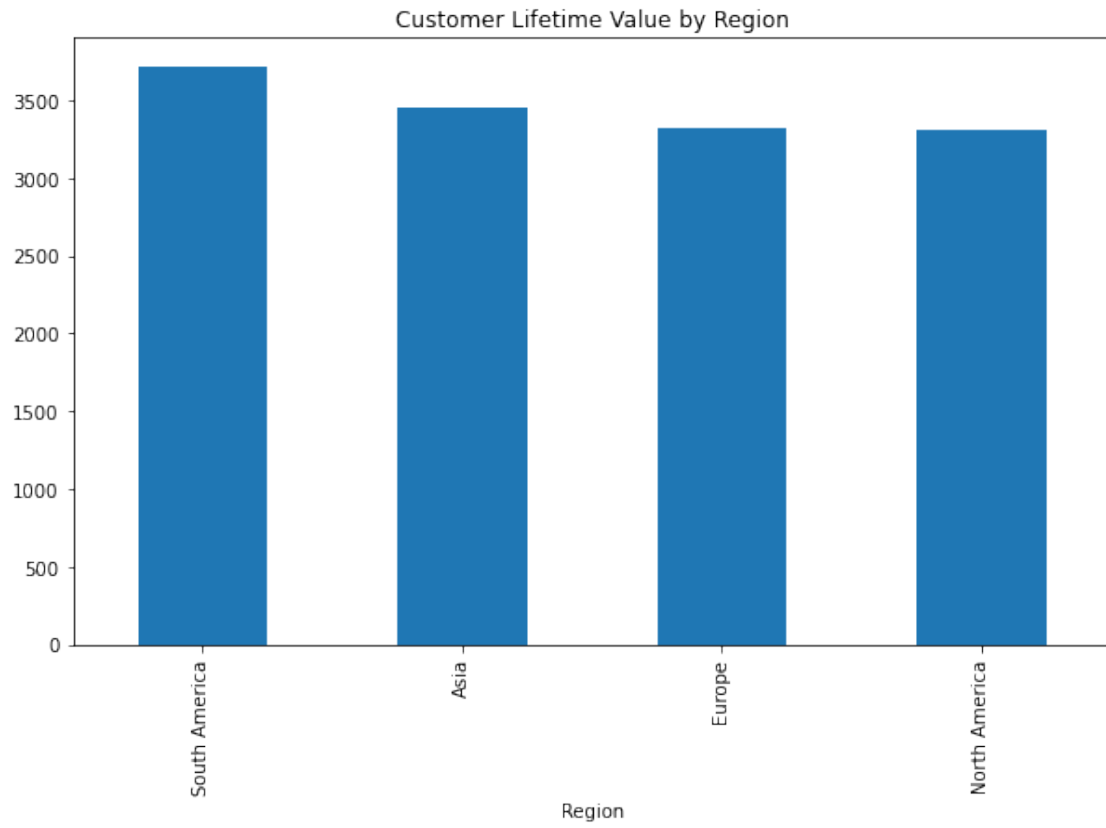
```
[9]: # Insight 4: Average transaction value by product category
avg_transaction_value = df.groupby('Category')['TotalValue'].mean().
    ↪sort_values(ascending=False)
avg_transaction_value.plot(kind='bar', figsize=(10, 6), title='Average_
    ↪Transaction Value by Product Category')
```

```
[9]: <AxesSubplot:title={'center':'Average Transaction Value by Product Category'},
    xlabel='Category'>
```



```
[10]: # Insight 5: Customer lifetime value (CLV) by region
clv = df.groupby('CustomerID')['TotalValue'].sum().reset_index()
clv = pd.merge(clv, customers.reset_index()[['CustomerID', 'Region']],
               on='CustomerID')
clv.groupby('Region')['TotalValue'].mean().sort_values(ascending=False).
    plot(kind='bar', figsize=(10, 6), title='Customer Lifetime Value by Region')
```

```
[10]: <AxesSubplot:title={'center':'Customer Lifetime Value by Region'},
      xlabel='Region'>
```



[]: