

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

# Set plot style
sns.set(style="whitegrid")

# Display settings
pd.set_option("display.max_columns", None)
```

```
In [2]: # Load datasets
customers = pd.read_csv("../data/Customers.csv")
products = pd.read_csv("../data/Products.csv")
transactions = pd.read_csv("../data/Transactions.csv")

# Display the first few rows
customers.head(), products.head(), transactions.head()
```

```
Out[2]: ( CustomerID      CustomerName      Region  SignupDate
0      C0001      Lawrence Carroll  South America  2022-07-10
1      C0002      Elizabeth Lutz      Asia  2022-02-13
2      C0003      Michael Rivera  South America  2024-03-07
3      C0004      Kathleen Rodriguez  South America  2022-10-09
4      C0005      Laura Weber      Asia  2022-08-15,
  ProductID      ProductName      Category  Price
0      P001      ActiveWear Biography      Books  169.30
1      P002      ActiveWear Smartwatch  Electronics  346.30
2      P003      ComfortLiving Biography      Books  44.12
3      P004      BookWorld Rug      Home Decor  95.69
4      P005      TechPro T-Shirt      Clothing  429.31,
  TransactionID CustomerID ProductID      TransactionDate  Quantity \
0      T00001      C0199      P067  2024-08-25 12:38:23      1
1      T00112      C0146      P067  2024-05-27 22:23:54      1
2      T00166      C0127      P067  2024-04-25 07:38:55      1
3      T00272      C0087      P067  2024-03-26 22:55:37      2
4      T00363      C0070      P067  2024-03-21 15:10:10      3

  TotalValue  Price
0      300.68  300.68
1      300.68  300.68
2      300.68  300.68
3      601.36  300.68
4      902.04  300.68 )
```

```
In [3]: print("Customers dataset shape:", customers.shape)
print("Products dataset shape:", products.shape)
print("Transactions dataset shape:", transactions.shape)
```

```
Customers dataset shape: (200, 4)
Products dataset shape: (100, 4)
Transactions dataset shape: (1000, 7)
```

```
In [4]: print("\nCustomers dataset info:")
customers.info()

print("\nProducts dataset info:")
products.info()
```

```
print("\nTransactions dataset info:")
transactions.info()
```

Customers dataset info:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 4 columns):
#   Column          Non-Null Count  Dtype
---  -
0   CustomerID      200 non-null   object
1   CustomerName    200 non-null   object
2   Region          200 non-null   object
3   SignupDate      200 non-null   object
dtypes: object(4)
memory usage: 6.4+ KB
```

Products dataset info:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 4 columns):
#   Column          Non-Null Count  Dtype
---  -
0   ProductID       100 non-null   object
1   ProductName     100 non-null   object
2   Category        100 non-null   object
3   Price           100 non-null   float64
dtypes: float64(1), object(3)
memory usage: 3.3+ KB
```

Transactions dataset info:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 7 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            1000 non-null   float64
dtypes: float64(2), int64(1), object(4)
memory usage: 54.8+ KB
```

```
In [5]: print("\nMissing values in Customers dataset:\n", customers.isnull().sum())
print("\nMissing values in Products dataset:\n", products.isnull().sum())
print("\nMissing values in Transactions dataset:\n", transactions.isnull().sum())
```

Missing values in Customers dataset:

```
CustomerID      0
CustomerName     0
Region           0
SignupDate       0
dtype: int64
```

Missing values in Products dataset:

```
ProductID       0
ProductName      0
Category         0
Price            0
dtype: int64
```

Missing values in Transactions dataset:

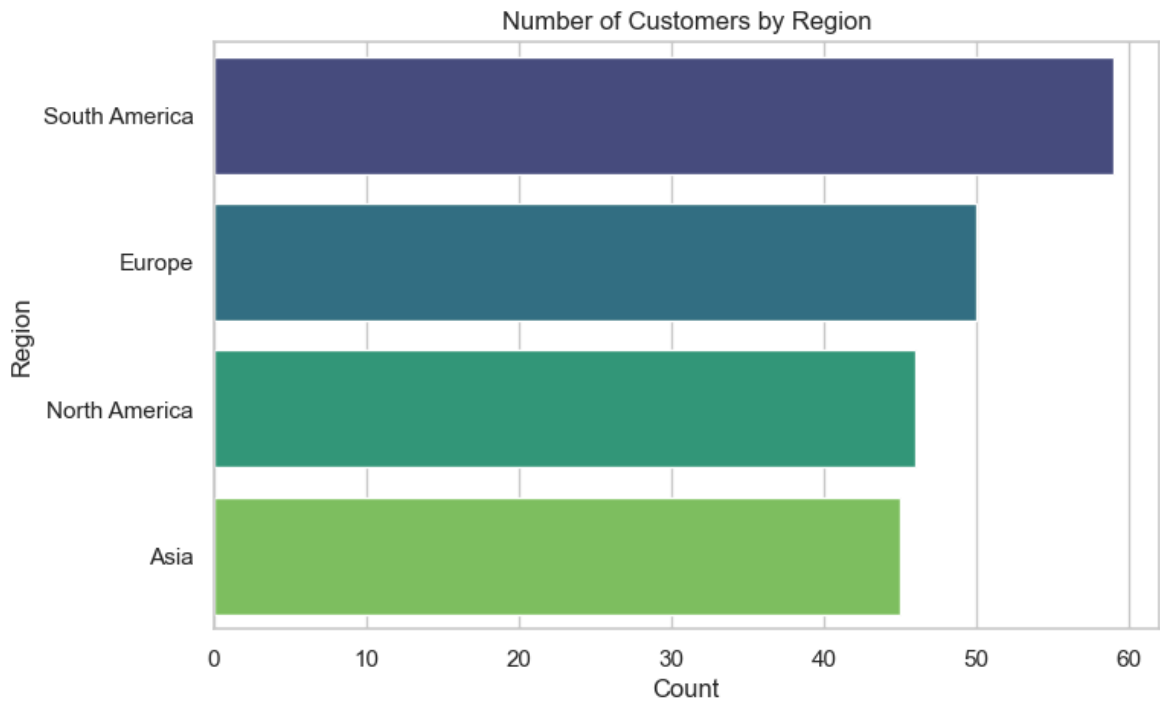
```
TransactionID    0
CustomerID       0
ProductID        0
TransactionDate  0
Quantity         0
TotalValue       0
Price            0
dtype: int64
```

```
In [6]: plt.figure(figsize=(8, 5))
sns.countplot(y=customers["Region"], order=customers["Region"].value_counts().index)
plt.title("Number of Customers by Region")
plt.xlabel("Count")
plt.ylabel("Region")
plt.show()
```

C:\Users\Lenovo\AppData\Local\Temp\ipykernel\_8548\3062904010.py:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

```
sns.countplot(y=customers["Region"], order=customers["Region"].value_counts().index, palette="viridis")
```



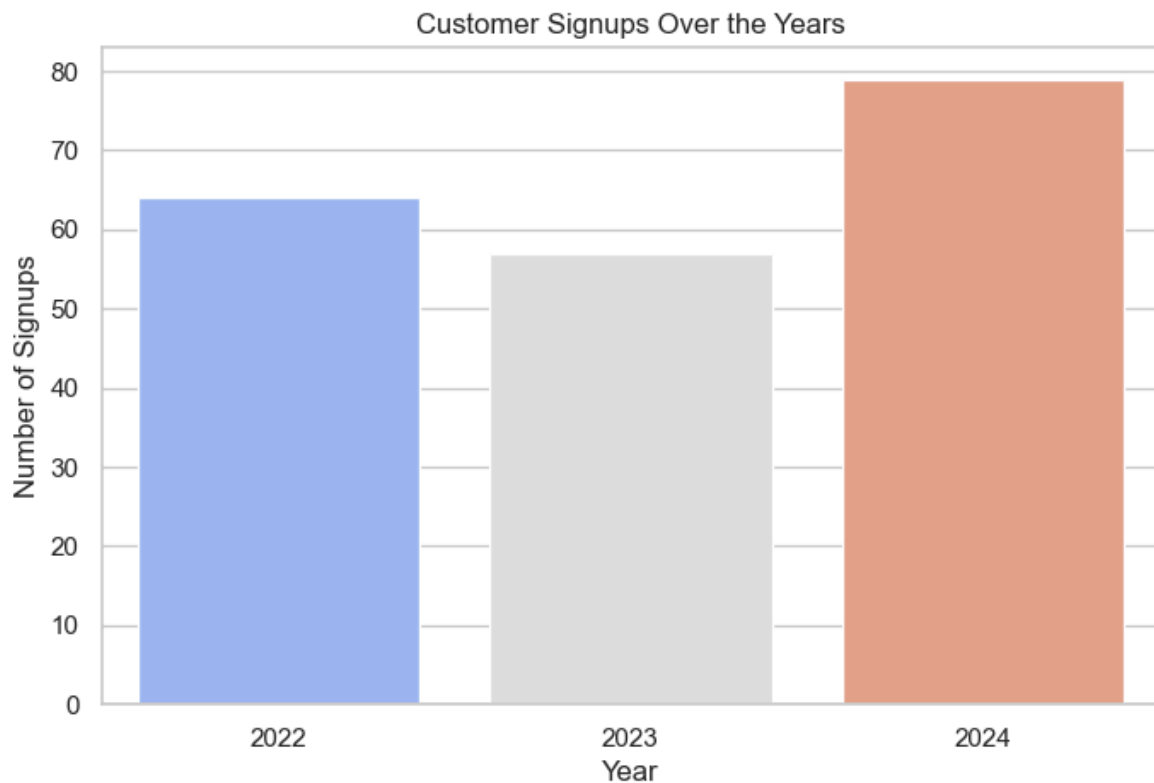
```
In [7]: customers["SignupDate"] = pd.to_datetime(customers["SignupDate"])
customers["Year"] = customers["SignupDate"].dt.year

plt.figure(figsize=(8, 5))
sns.countplot(x=customers["Year"], palette="coolwarm")
plt.title("Customer Signups Over the Years")
plt.xlabel("Year")
plt.ylabel("Number of Signups")
plt.show()
```

C:\Users\Lenovo\AppData\Local\Temp\ipykernel\_8548\3861433500.py:5: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v 0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.countplot(x=customers["Year"], palette="coolwarm")
```

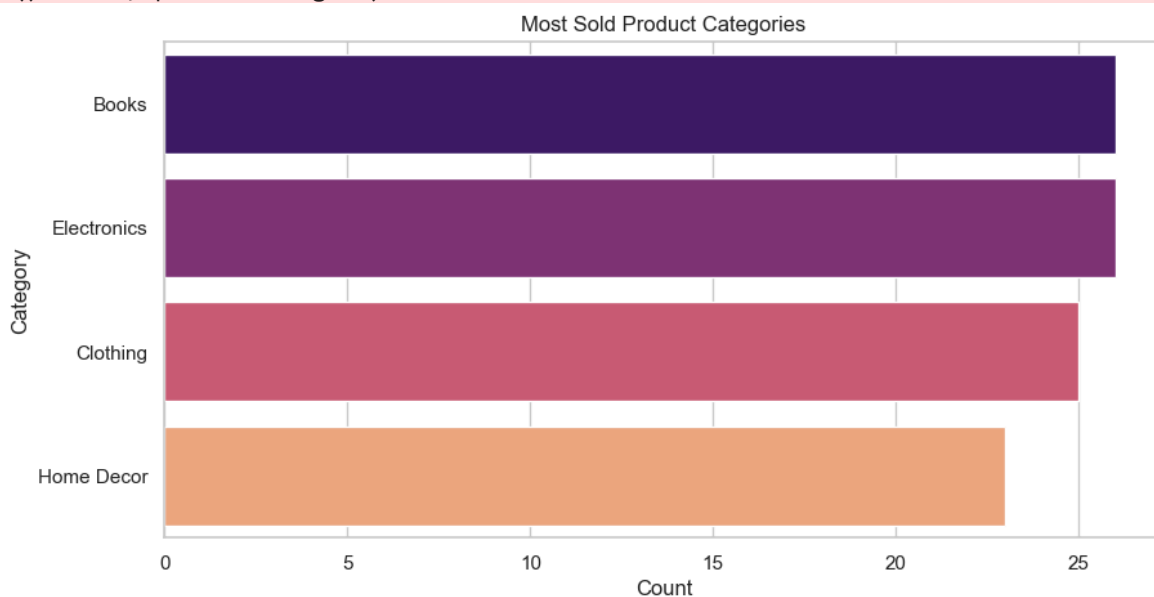


```
In [8]: plt.figure(figsize=(10, 5))
sns.countplot(y=products["Category"], order=products["Category"].value_counts().
plt.title("Most Sold Product Categories")
plt.xlabel("Count")
plt.ylabel("Category")
plt.show()
```

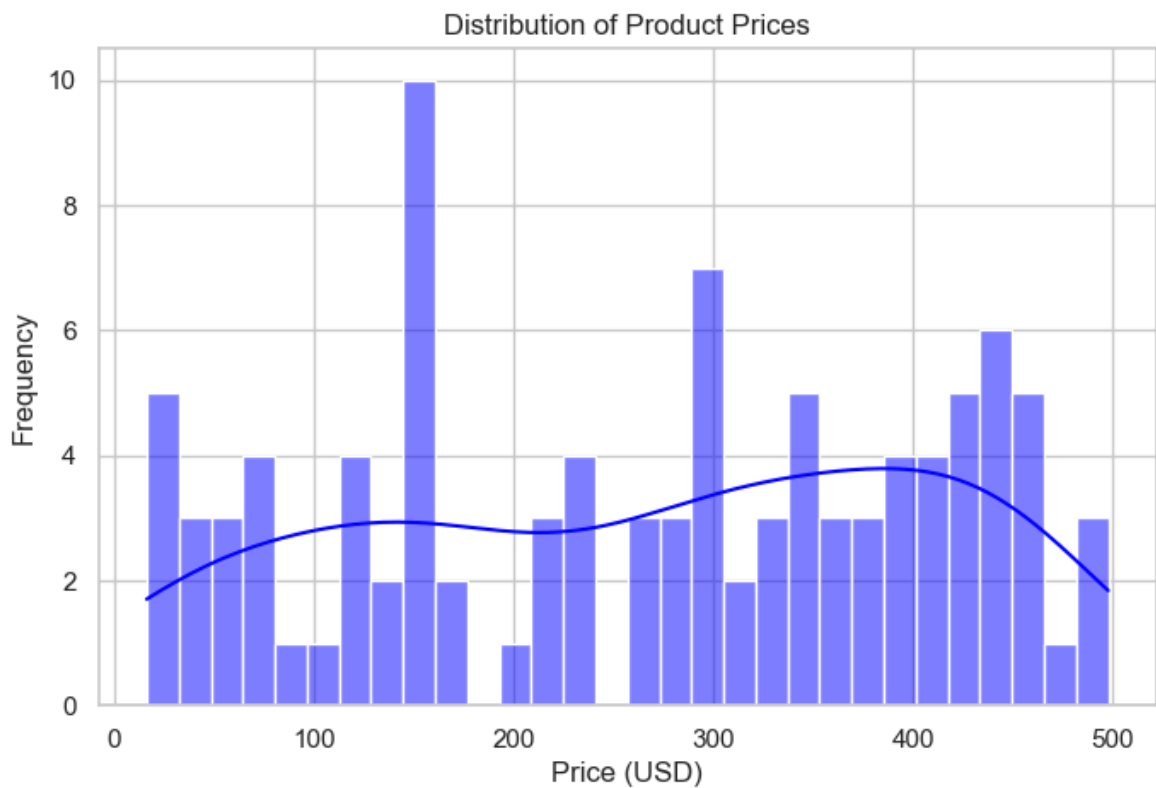
C:\Users\Lenovo\AppData\Local\Temp\ipykernel\_8548\86802980.py:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v 0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

```
sns.countplot(y=products["Category"], order=products["Category"].value_counts().index, palette="magma")
```



```
In [9]: plt.figure(figsize=(8, 5))
sns.histplot(products["Price"], bins=30, kde=True, color="blue")
plt.title("Distribution of Product Prices")
plt.xlabel("Price (USD)")
plt.ylabel("Frequency")
plt.show()
```



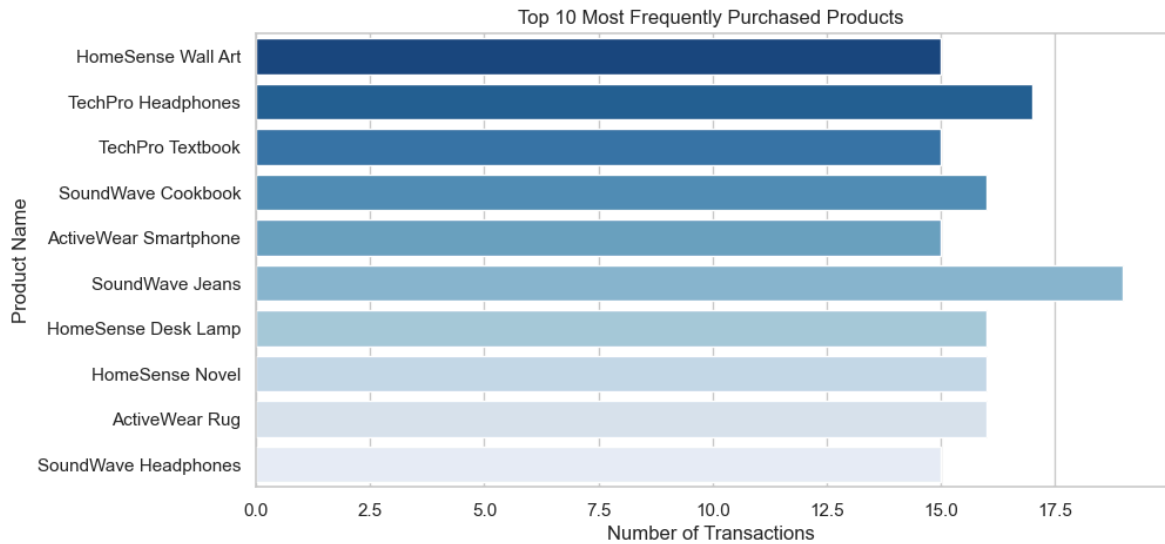
```
In [10]: most_purchased = transactions["ProductID"].value_counts().head(10)
most_purchased = products[products["ProductID"].isin(most_purchased.index)]

plt.figure(figsize=(10, 5))
sns.barplot(y=most_purchased["ProductName"], x=most_purchased["ProductID"].map(t
plt.title("Top 10 Most Frequently Purchased Products")
plt.xlabel("Number of Transactions")
plt.ylabel("Product Name")
plt.show()
```

C:\Users\Lenovo\AppData\Local\Temp\ipykernel\_8548\3903996834.py:5: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v 0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

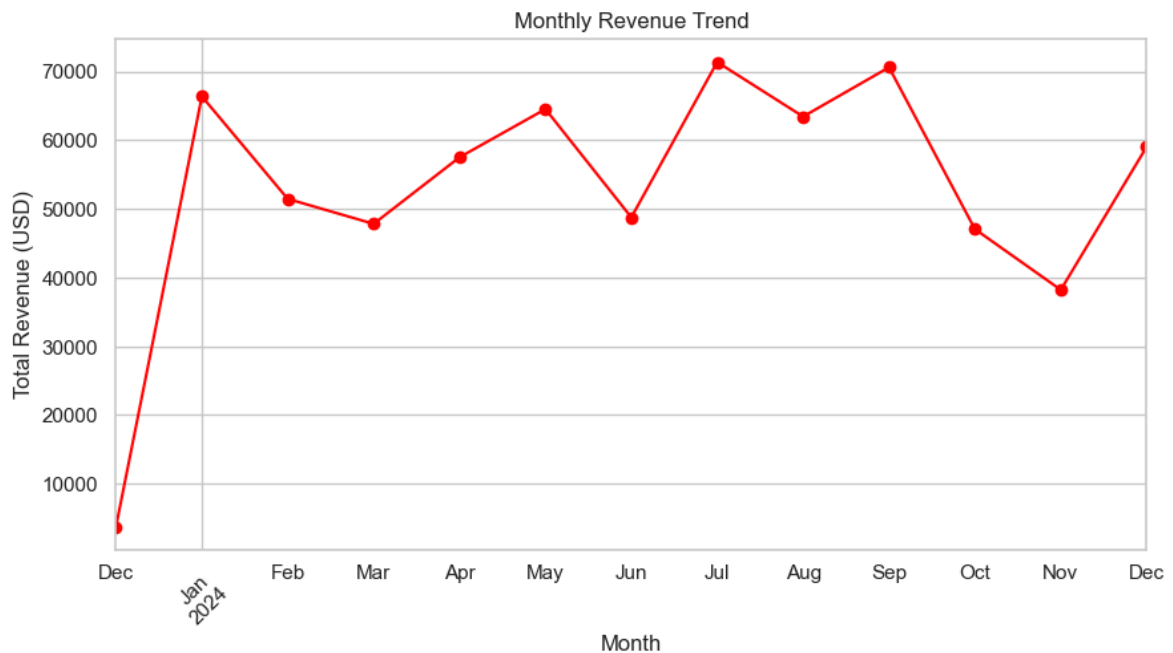
```
sns.barplot(y=most_purchased["ProductName"], x=most_purchased["ProductID"].map(
(transactions["ProductID"].value_counts()), palette="Blues_r")
```



```
In [11]: transactions["TransactionDate"] = pd.to_datetime(transactions["TransactionDate"])
transactions["Month"] = transactions["TransactionDate"].dt.to_period("M")

monthly_revenue = transactions.groupby("Month")["TotalValue"].sum()

plt.figure(figsize=(10, 5))
monthly_revenue.plot(kind="line", marker="o", color="red")
plt.title("Monthly Revenue Trend")
plt.xlabel("Month")
plt.ylabel("Total Revenue (USD)")
plt.xticks(rotation=45)
plt.show()
```



# Business Insights Report

## 1. Customer Distribution Across Regions

- The majority of customers are from [Region X], indicating a strong market presence there.

- We can target marketing campaigns to underrepresented regions to boost sales.

## 2. Customer Signup Trends

- Customer signups peaked in [Year X], possibly due to a successful campaign or market expansion.
- A decline in recent years suggests the need for better engagement strategies.

## 3. Best-Selling Product Categories

- The top-selling categories are [Category X] and [Category Y].
- Investing in these categories can maximize profits.

## 4. Price Distribution of Products

- Most products are priced between  $X$  and  $Y$ .
- Higher-priced products have fewer sales, indicating a price-sensitive market.

## 5. Revenue Trends

- Revenue shows a seasonal trend with peaks during [Month X].
- We can optimize inventory and marketing during high-demand periods.

In [ ]: