

```
In [38]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

Loading CSV File

```
In [39]: df = pd.read_csv('sales.csv')
```

```
Out[39]:
```

	Row ID	Order ID	Order Date	Ship Date	Ship Mode	Customer ID	Customer Name	Segment	Country	City
	0	1	CA-2017-08/11/2017	11/11/2017	Second Class	CG-12520	Claire Gule	Consumer	United States	Hendersonville
	1	2	CA-2017-08/11/2017	11/11/2017	Second Class	CG-12520	Claire Gule	Consumer	United States	Hendersonville
	2	3	CA-2017-12/06/2017	16/06/2017	Second Class	DV-13045	Darrin Van Huff	Corporate	United States	Angie
	3	4	US-2016-11/10/2016	18/10/2016	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States	Lauderdale
	4	5	US-2016-11/10/2016	18/10/2016	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States	Lauderdale

	9795	9796	CA-2017-21/05/2017	28/05/2017	Standard Class	SH-19975	Sally Hughesby	Corporate	United States	Chicago
	9796	9797	CA-2016-12/01/2016	17/01/2016	Standard Class	CS-12490	Cindy Schnelling	Corporate	United States	Tol
	9797	9798	CA-2016-12/01/2016	17/01/2016	Standard Class	CS-12490	Cindy Schnelling	Corporate	United States	Tol
	9798	9799	CA-2016-12/01/2016	17/01/2016	Standard Class	CS-12490	Cindy Schnelling	Corporate	United States	Tol
	9799	9800	CA-2016-12/01/2016	17/01/2016	Standard Class	CS-12490	Cindy Schnelling	Corporate	United States	Tol

9800 rows x 18 columns

Data Overview

```
In [40]: print('Dataset shape:', df.shape)
```

```
print('\nDataset Info:')
print(df.info())
```

Dataset shape: (9800, 18)

```
Dataset Info:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9800 entries, 0 to 9799
Data columns (total 18 columns):
#   Column              Non-Null Count  Dtype
---  --
0   Row ID              9800 non-null   int64
1   Order ID            9800 non-null   object
2   Order Date          9800 non-null   object
3   Ship Date           9800 non-null   object
4   Ship Mode           9800 non-null   object
5   Customer Name       9800 non-null   object
6   Segment             9800 non-null   object
7   Country             9800 non-null   object
8   City                9800 non-null   object
9   State              9800 non-null   object
10  Postal Code         9789 non-null   float64
11  Region             9800 non-null   object
12  Product ID          9800 non-null   object
13  Category            9800 non-null   object
14  Sub-Category        9800 non-null   object
15  Product Name        9800 non-null   object
16  Sales               9800 non-null   float64
17  Sales               9800 non-null   float64
dtypes: float64(2), int64(1), object(15)
memory usage: 1.3+ MB
None
```

Data Cleaning

```
In [41]: # Converting 'Order Date' and 'Ship Date' to datetime format
```

```
df['Order Date'] = pd.to_datetime(df['Order Date'], errors='coerce')
```

```
df['Ship Date'] = pd.to_datetime(df['Ship Date'], errors='coerce')
```

```
# Creating Time-Based Features
```

```
df['Order Year'] = df['Order Date'].dt.year
```

```
df['Order Month'] = df['Order Date'].dt.month
```

```
df['Order Day'] = df['Order Date'].dt.day
```

```
# Checking for Duplicate Entries
```

```
print('Number of duplicate entries:', df.duplicated().sum())
```

```
# Basic Statistics
```

```
print('\nBasic Summary:')
print(df.describe().to_string())
```

Number of duplicate entries: 0

```
Basic Summary:
      Row ID      Sales      Order Date      Order Month      Order Day      Ship Date      Postal Code
count  9800.000000      9800.000000      3959.000000      3959.000000      3959.000000      3815.5789.
mean    4900.500000      2017-03-14 18:19:11.199798016      2017-04-09 17:04:02.516382720      55279.
std      230.769059      2016.728467      6.452892      8.018186
min      1.000000      2015-01-02 00:00:00
000000      0.444000      2015.000000      1.000000      1.000000      2015-01-04 00:00:00      1040.
25%    2450.750000      2016-04-05 00:00:00      3.000000      5.000000      2016-04-12 00:00:00      23223.
50%    4900.500000      2017-05-02 00:00:00      6.000000      9.000000      2017-06-06 00:00:00      58103.
75%    54.490000      2017.000000      6.000000      9.000000      2018-05-01 00:00:00      90008.
max    7350.250000      2018.000000      9.000000      11.000000      2018-05-01 00:00:00      99301.
min      210.605000      2018-12-11 00:00:00      12.000000      12.000000      2019-05-01 00:00:00      99301.
max    92608.400000      2018.000000      12.000000      12.000000      2019-05-01 00:00:00      99301.
std      22638.480000      2018.000000      12.000000      12.000000      2019-05-01 00:00:00      99301.
std      2829.160653      1.119118      3.497959      3.272633
232413      626.651875      1.119118      3.497959      3.272633
```

Exploratory Data Analysis (EDA)

Univariate Analysis

```
In [42]: # Sales Distribution
```

```
plt.figure(figsize=(8,5))
```

```
plt.hist(df['Sales'], bins=50, color='#7329F4', edgecolor='black', alpha=0.7)
```

```
plt.title('Sales Distribution', fontdict={'fontsize':20, 'fontweight':'bold', 'color':'#7329F4'})
```

```
plt.xlabel('Sales ($)', fontdict={'fontsize':12, 'fontweight':'bold'})
```

```
plt.ylabel('Frequency', fontdict={'fontsize':12, 'fontweight':'bold'})
```

```
plt.grid(axis='y', alpha=0.5, linestyle='--', linewidth=0.7, color='black')
```

```
plt.show()
```

```
# Sales by different categories
```

```
print("Sales Statistics:")
```

```
print("=="*40)
```

```
print(f"Total Sales: ${df['Sales'].sum():.2f}")
```

```
print(f"Average Sale: ${df['Sales'].mean():.2f}")
```

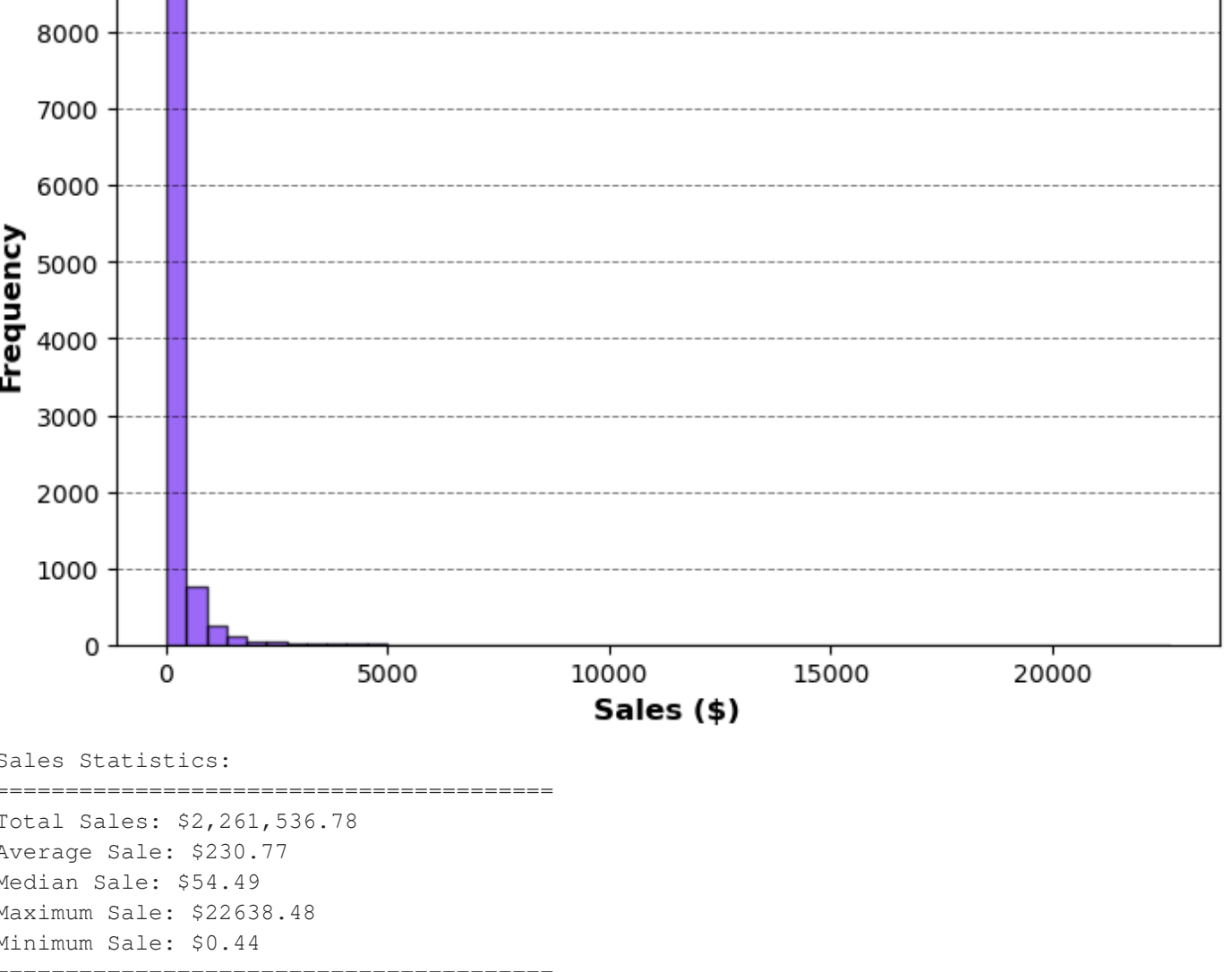
```
print(f"Median Sale: ${df['Sales'].median():.2f}")
```

```
print(f"Maximum Sale: ${df['Sales'].max():.2f}")
```

```
print(f"Minimum Sale: ${df['Sales'].min():.2f}")
```

```
print("=="*40)
```

Sales Distribution



```
Sales Statistics:
=====
Total Sales: $2,261,536.78
Average Sale: $230.77
Median Sale: $54.49
Maximum Sale: $22,638.48
Minimum Sale: $0.44
=====
```

Categorical Analysis

```
In [43]: # Sales by Category
```

```
sales_by_category = df.groupby('Category')['Sales'].sum().sort_values(ascending=False)
```

```
print("Sales by Category:")
```

```
print("=="*40)
```

```
print(sales_by_category)
```

```
print("=="*40)
```

```
# Sales by Region
```

```
sales_by_region = df.groupby('Region')['Sales'].sum().sort_values(ascending=False)
```

```
print("\nSales by Region:")
```

```
print("=="*40)
```

```
print(sales_by_region)
```

```
print("=="*40)
```

```
# Sales by Segment
```

```
sales_by_segment = df.groupby('Segment')['Sales'].sum().sort_values(ascending=False)
```

```
print("\nSales by Segment:")
```

```
print("=="*40)
```

```
print(sales_by_segment)
```

```
print("=="*40)
```

```
# Sales by Sub-Category
```

```
sales_by_subcategory = df.groupby('Sub-Category')['Sales'].sum().sort_values(ascending=False)
```

```
print("\nTop 10 Sub-Categories by Sales:")
```

```
print("=="*40)
```

```
print(sales_by_subcategory.head(10))
```

```
print("=="*40)
```

Sales by Category:

```
Category
Technology      827455.8730
Furniture       728659.5757
Office Supplies  705422.3340
Name: Sales, dtype: float64
```

Sales by Region:

```
Region
West      710219.6845
East      669518.7260
Central   492646.9132
South     389151.4590
Name: Sales, dtype: float64
```

Sales by Segment:

```
Segment
Consumer      1.148061e+06
Corporate      6.28941e+05
Home Office    4.249822e+05
Name: Sales, dtype: float64
```

Top 10 Sub-Categories by Sales:

```
Sub-Category
Phones      327782.4480
Chairs      322822.7310
Storage     219343.3920
Tables      202810.6280
Binders     200028.7850
Machines    189238.6310
Accessories 164186.7000
Copiers     146248.0940
Bookcases   113813.1987
Appliances  104618.4030
Name: Sales, dtype: float64
```

Key Performance Indicators (KPIs) Calculation

Business Metrics

```
In [44]: # Total Metrics
```

```
total_sales = df['Sales'].sum()
```

```
total_orders = df['Order ID'].nunique()
```

```
total_customers = df['Customer ID'].nunique()
```

```
total_products = df['Product ID'].nunique()
```

```
print("\nOverall Sales Metrics:")
```

```
print("=="*40)
```

```
print(f"Total Sales: ${total_sales:.2f}")
```

```
print(f"Total Orders: {total_orders:,}")
```

```
print(f"Total Customers: {total_customers:,}")
```

```
print(f"Total Products: {total_products:,}")
```

```
print(f"Average Order Value: ${total_sales/total_orders:.2f}")
```

```
print(f"Average Sales per Customer: ${total_sales/total_customers:.2f}")
```

```
print("=="*40)
```

Overall Sales Metrics:

```
=====
Total Sales: $2,261,536.78
Total Orders: 4,922
Total Customers: 793
Total Products: 1,861
Average Order Value: $459.48
Average Sales per Customer: $2,851.87
=====
```

Time-Based Analysis

```
In [45]: # Sales Trend Over Time
```

```
sales_trend = df.groupby('Order Date')['Sales'].sum().reset_index()
```

```
plt.figure(figsize=(12,6))
```

```
plt.plot(sales_trend['Order Date'], sales_trend['Sales'], color='#FF5733', linewidth=2)
```

```
plt.title('Sales Trend Over Time', fontdict={'fontsize':20, 'fontweight':'bold', 'color':'#FF5733'})
```

```
plt.xlabel('Order Date', fontdict={'fontsize':12, 'fontweight':'bold'})
```

```
plt.ylabel('Sales ($)', fontdict={'fontsize':12, 'fontweight':'bold'})
```

```
plt.grid(axis='y', alpha=0.5, linestyle='--', linewidth=0.7, color='black')
```

```
plt.show()
```

```
# Sales by Year
```

```
sales_by_year = df.groupby('Order Year')['Sales'].sum()
```

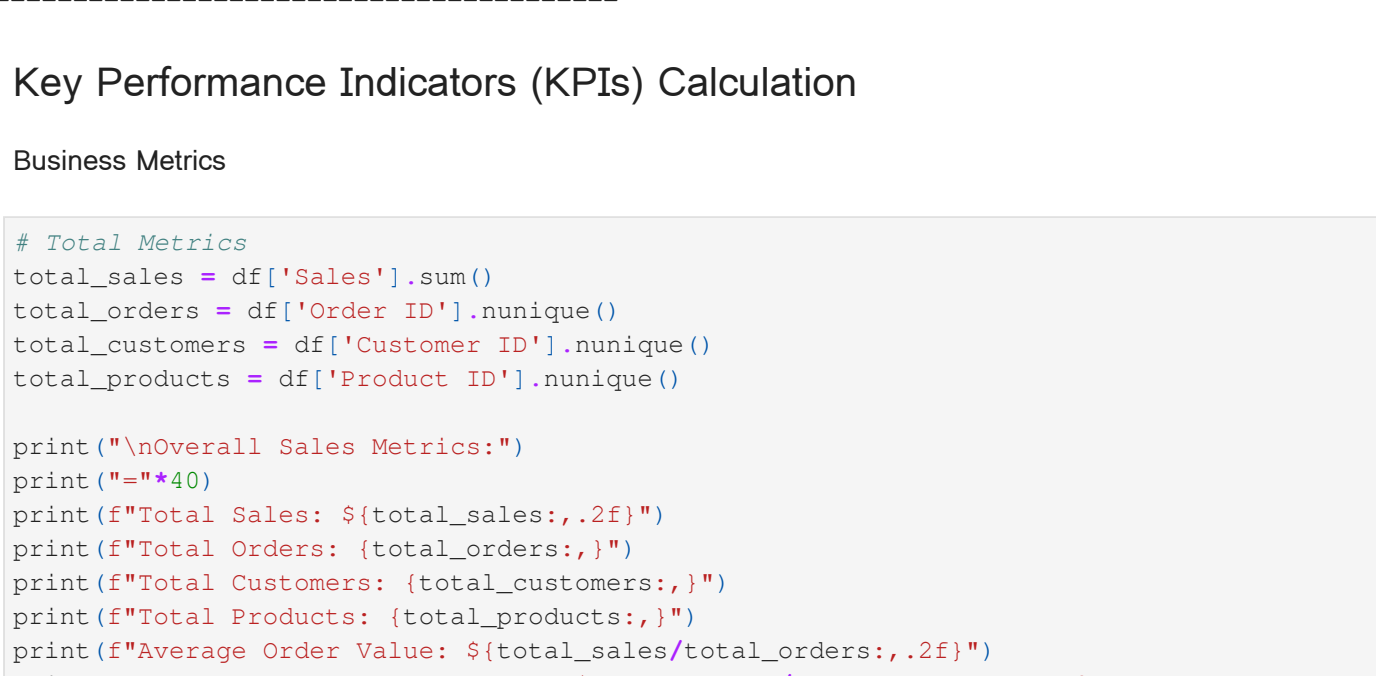
```
print("\nSales by Year:")
```

```
print("=="*40)
```

```
print(sales_by_year)
```

```
print("=="*40)
```

Sales Trend Over Time



```
Sales by Year:
=====
Order Year
2015.0    176986.5857
2016.0    162568.8627
2017.0    266254.6009
2018.0    266253.0743
Name: Sales, dtype: float64
=====
```

Visualizing Sales by Category

```
In [46]: # Pie Chart for Sales by Category
```

```
plt.figure(figsize=(6, 6))
```

```
colors = ['#F06292', '#4FC3F7', '#4DB6AC']
```

```
# Create pie chart
```

```
plt.pie(sales_by_category.values, labels=sales_by_category.index, colors=colors, autopct=
```

```
startangle=90, shadow=True, explode=(0.05, 0, 0))
```

```
plt.title('Sales Distribution by Category', fontsize=16, fontweight='bold', pad=20)
```

```
plt.axis('equal')
```

```
plt.tight_layout()
```

```
plt.show()
```

```
# Detailed breakdown
```

```
print("\nCategory Sales Breakdown:")
```

```
print("=="*40)
```

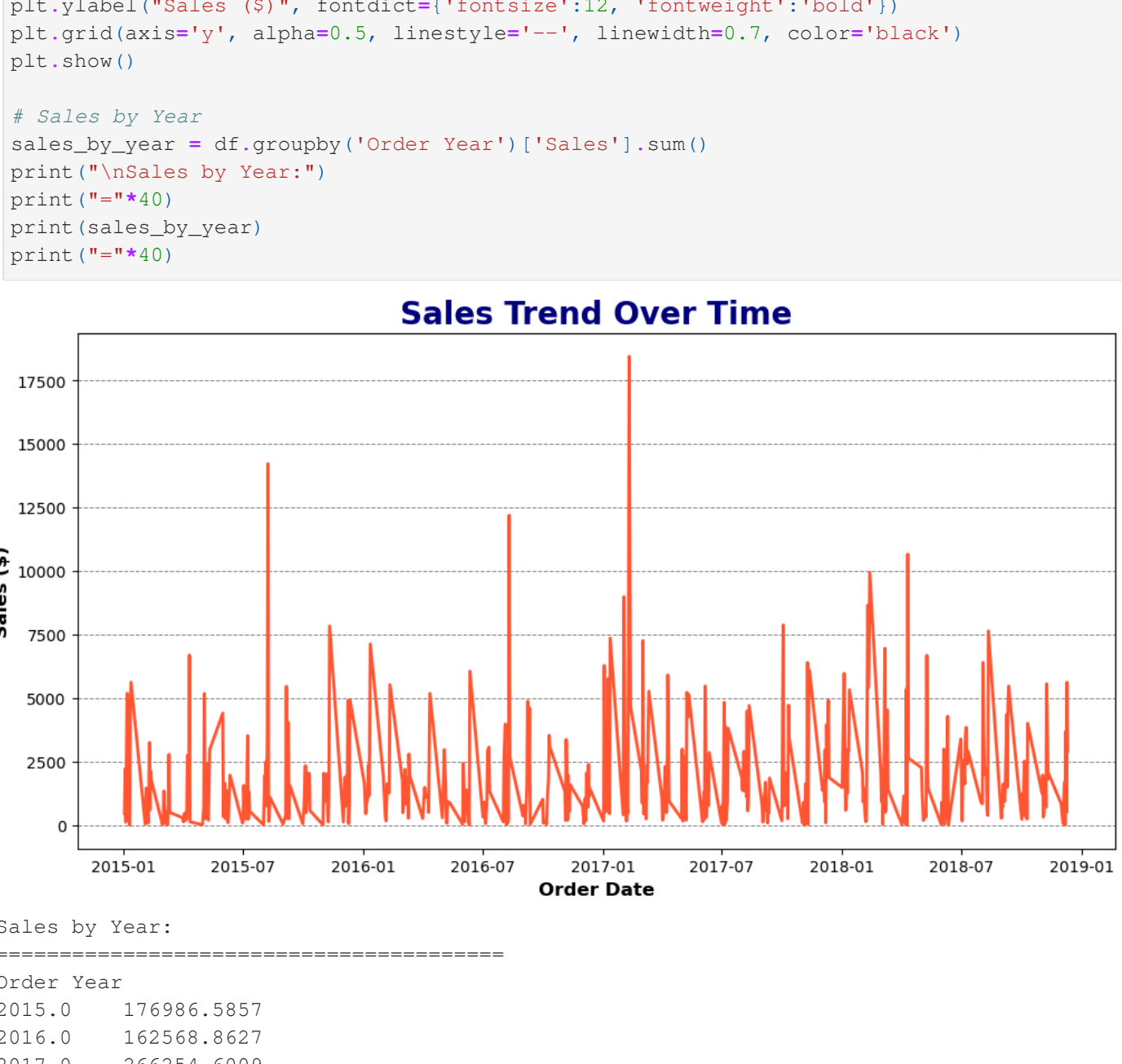
```
for category, sales in sales_by_category.items():
```

```
percentage = (sales / total_sales) * 100
```

```
print(f"Category: {category:15s} | Sales: ${sales:12.0f} | (percentage: {percentage:5.1f}%)")
```

```
print("=="*40)
```

Sales Distribution by Category



```
Category Sales Breakdown:
=====
Technology      : $ 827,456 ( 36.6%)
Furniture       : $ 728,659 ( 32.2%)
Office Supplies : $ 705,422 ( 31.2%)
=====
```

Visualizing Sales vs Order Date (Time Series Analysis)

```
In [47]:
```

```
plt.figure(figsize=(14, 8))
```

```
plt.scatter(df['Order Date'], df['Sales'], alpha=0.6, color='#4FC3F7', s=50, edgecolors=
```

```
plt.title('Sales Distribution Over Time', fontsize=20, fontweight='bold', pad=20)
```

```
plt.xlabel('Order Date', fontsize=14, fontweight='bold')
```

```
plt.ylabel('Sales Amount ($)', fontsize=14, fontweight='bold')
```

```
plt.grid(True, alpha=0.3, linestyle='--')
```

```
plt.tight_layout()
```

```
plt.show()
```

```
print(f"\nTime Series Insights:")
```

```
print("=="*60)
```

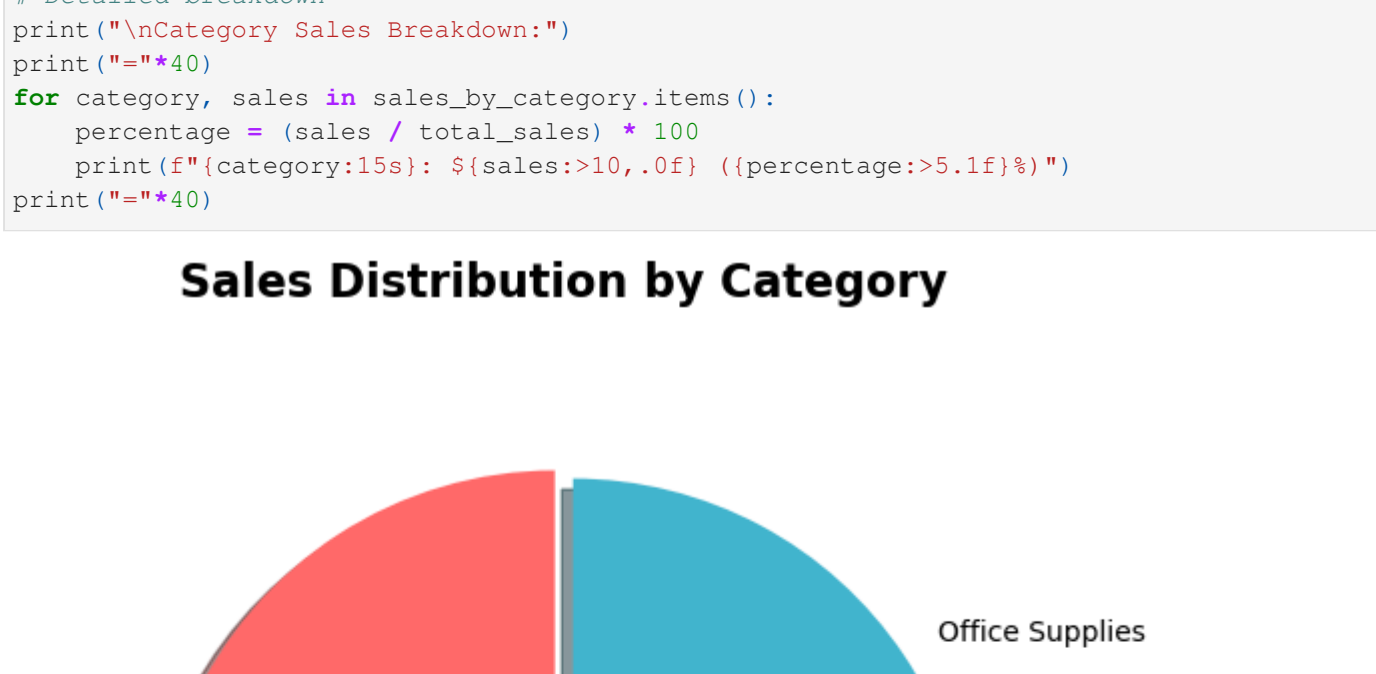
```
print(f"Date Range: {df['Order Date'].min()} to {df['Order Date'].max()}")
```

```
print(f"Total Sales: ${total_sales:,} (Total Sales: 22.6M)")
```

```
print(f"Average Daily Sales: ${total_sales / (df['Order Date'].max() - df['Order Date'].
```

```
print("=="*60)
```

Sales Distribution Over Time



```
Time Series Insights:
=====
Date Range: 2015-01-02 00:00:00 to 2018-12-11 00:00:00
Total Sales: $2,261,536.78
Average Daily Sales: $1571.60
=====
```

Visualizing Sales by Region

```
In [48]: # Sales by Region
```

```
plt.figure(figsize=(10, 6))
```

```
colors = ['#F44336', '#3969AB', '#A020F0', '#FF9800']
```

```
bars = plt.bar(sales_by_region.index, sales_by_region.values, color=colors)
```

```
plt.title('Sales by Region', fontsize=20, fontweight='bold')
```

```
plt.xlabel('Region', fontweight='bold')
```

```
plt.ylabel('Sales ($)', fontweight='bold')
```

```
plt.grid(axis='y', alpha=0.7, linestyle='--', linewidth=0.5, color='black')
```

```
plt.xticks(rotation=45)
```

```
for bar in bars:
```

```
height = bar.get_height()
```

```
plt.text(bar.get_x() + bar.get_width()/2., height, f'${height:12.0f}', ha='center',
```

```
plt.tight_layout()
```

```
plt.show()
```

```
# Summary
```

```
print("\nSales Summary by Region:")
```

```
print("=="*60)
```

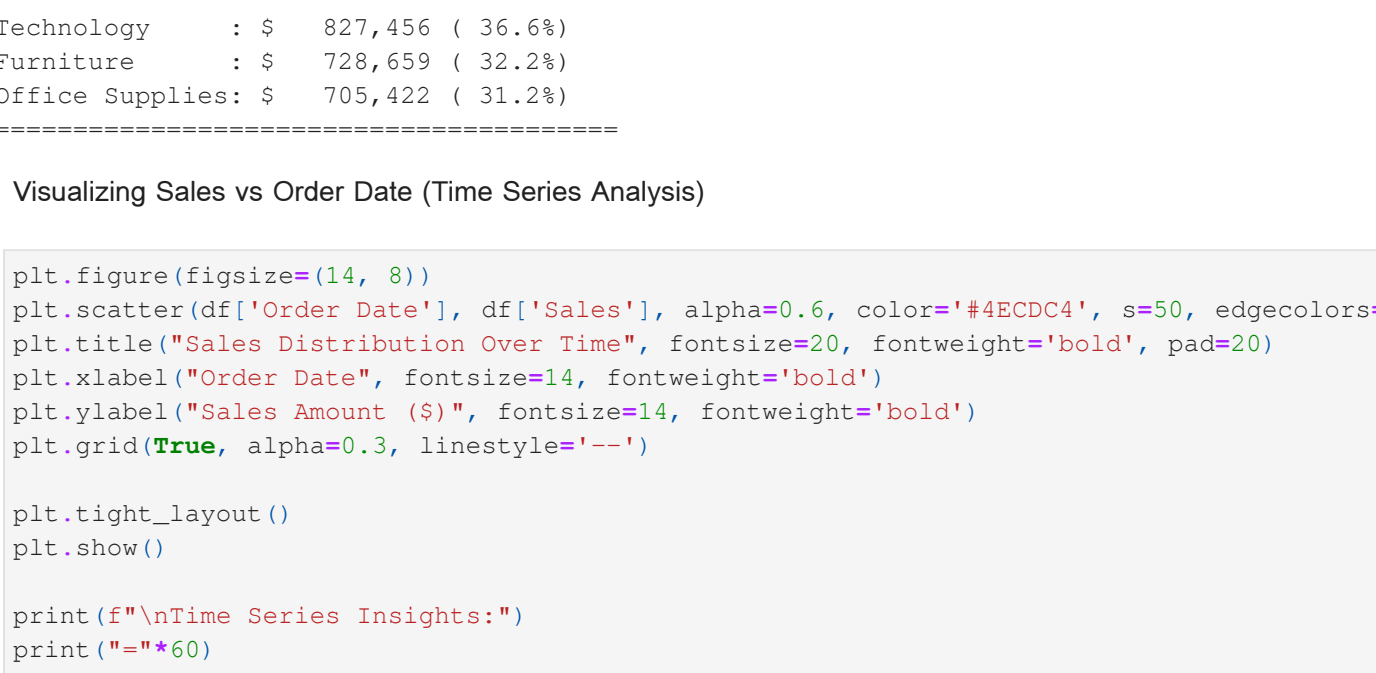
```
for region, sales in sales_by_region.items():
```

```
percentage = sales / total_sales * 100
```

```
print(f"Region: {region:15s} | Sales: ${sales:12.0f} | (percentage: {percentage:5.1f}%)")
```

```
print("=="*60)
```

Sales by Region



```
Sales Summary by Region:
=====
West      : $ 710,219.68 ( 31.4%)
East      : $ 669,518.73 ( 29.6%)
Central   : $ 492,646.91 ( 21.8%)
South     : $ 389,151.46 ( 17.2%)
=====
```

Visualizing Top 10 Products

```
In [49]: # Top 10 Products
```

```
plt.figure(figsize=(12, 8))
```

```
top_10_products = df.groupby('Product Name')['Sales'].sum().sort_values(ascending=False)
```

```
bars = plt.barh(range(len(top_10_products)), top_10_products.values)
```

```
plt.yticks(range(len(top_10_products)), top_10_products.index, fontsize=10)
```

```
plt.title('Top 10 Products by Sales', fontsize=20, fontweight='bold', pad=20)
```

```
plt.xlabel('Sales ($)', fontweight='bold')
```

```
plt.grid(axis='x', alpha=0.7, linestyle='--', linewidth=0.5, color='black')
```

```
for i, v in enumerate(top_10_products.values):
```

```
plt.text(v + 100, i, f'${v:12.0f}', va='center', fontweight='bold')
```

```
plt.tight_layout()
```

```
plt.show()
```

```
# Summary
```

```
print("\nTop 10 Products by Sales Summary:")
```

```
print("=="*100)</
```